Integrated Science Assessment for Ozone and Related Photochemical Oxidants
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<tr>
<th>Acronym</th>
<th>Definition</th>
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<tr>
<td>129</td>
<td>mouse strain (129S1/SvlmJ)</td>
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<tr>
<td>α</td>
<td>alpha, ambient exposure factor</td>
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<tr>
<td>α-ATD</td>
<td>alpha 1-antitrypsin deficiency</td>
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<tr>
<td>α-SMA</td>
<td>alpha-smooth muscle actin</td>
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<td>α-tocopherol</td>
<td>alpha-tocopherol</td>
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<td>α-TOH</td>
<td>alpha tocopherol</td>
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<td>a</td>
<td>air exchange rate of the microenvironment</td>
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<tr>
<td>A2</td>
<td>climate scenario in IPCC</td>
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<tr>
<td>AADT</td>
<td>annual average daily traffic</td>
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<td>A1B</td>
<td>climate scenario in IPCC</td>
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<tr>
<td>ABA</td>
<td>abscisic acid</td>
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<tr>
<td>ABI</td>
<td>abscisic acid insensitive</td>
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<tr>
<td>ABI2</td>
<td>abscisic acid insensitive Arabidopsis mutant</td>
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<td>A1c</td>
<td>glycosylated hemoglobin blood test</td>
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<td>Ach</td>
<td>acetylcholine</td>
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<td>ACM</td>
<td>Atmospheric Chemistry Modeling (Group)</td>
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<td>ACS</td>
<td>American Cancer Society</td>
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<td>ACS-CPSII</td>
<td>ACS Cancer Prevention Study II</td>
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<tr>
<td>ADC</td>
<td>arginine decarboxylase</td>
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<td>ADSP</td>
<td>Adirondack State Park, NY</td>
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<td>AER</td>
<td>air exchange rate</td>
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<td>AH2</td>
<td>ascorbic acid; ascorbate</td>
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<td>AHR</td>
<td>airway(s) hyperresponsiveness, airway(s) hyperreactivity</td>
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<td>AhR</td>
<td>aryl hydrocarbon receptor</td>
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<td>AHSMOG</td>
<td>Adventist Heath and Smog (Study)</td>
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<td>AI</td>
<td>alveolar interstitial</td>
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<td>AIC(s)</td>
<td>Akaike’s information criterion</td>
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<td>AIRS</td>
<td>Aerometric Information Retrieval System; Atmospheric Infrared Sounder (instrument)</td>
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<td>A/J</td>
<td>mouse strain</td>
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<tr>
<td>Ala-9Val</td>
<td>genotype associated with Manganese superoxide dismutase (MnSOD) gene</td>
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<td>AM</td>
<td>alveolar macrophage(s)</td>
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<td>ANF</td>
<td>atrial natriuretic factor</td>
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<td>AOT20</td>
<td>seasonal sum of the difference between an hourly concentration at the threshold value of 20 ppb, minus the threshold value of 20 ppb</td>
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<td>AOTx</td>
<td>family of cumulative, cutoff concentration-based exposure indices</td>
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<td>activated protein</td>
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<td>ascorbate peroxidase</td>
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<td>aq</td>
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<td>AQCD</td>
<td>Air Quality Criteria Document</td>
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<td>Air Quality Index</td>
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<td>(U.S. EPA) Air Quality System (database)</td>
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<td>AR</td>
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<td>ARG</td>
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<td>ATPase</td>
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<td>ATS</td>
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<tr>
<td>avg</td>
<td>average</td>
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<td>AVHRR</td>
<td>advanced very high resolution radiometer</td>
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<tr>
<td>AX</td>
<td>Annex</td>
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<tr>
<td>β</td>
<td>beta, beta coefficient; regression coefficient; standardized coefficient; shape parameter; scale parameter</td>
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<td>B</td>
<td>boron</td>
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<td>B1</td>
<td>climate scenario in IPCC</td>
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<td>Acronym</td>
<td>Definition</td>
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<td>B6</td>
<td>mouse strain (C57BL/6J)</td>
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<td>BAL</td>
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<td>International Bureau of Weights and Measures</td>
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<td>BM</td>
<td>basement membrane</td>
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<td>Behavioral Risk Factor Surveillance System</td>
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<td>black smoke</td>
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<td>BSA</td>
<td>bovine serum albumin; body surface area</td>
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<td>black smoke particles</td>
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<td>Bt, BT, bt</td>
<td>Bacillus thuringiensis; bacterium proteins used in pesticides (or genetically engineered plants produce Bt toxin)</td>
</tr>
<tr>
<td>BTEX</td>
<td>family of compounds (benzene, toluene, ethylbenzene, and xylene)</td>
</tr>
<tr>
<td>BW</td>
<td>body weight</td>
</tr>
<tr>
<td>C</td>
<td>carbon; concentration; (Vitamin C, ascorbate)</td>
</tr>
<tr>
<td>°C</td>
<td>degrees Celsius</td>
</tr>
<tr>
<td>¹³C</td>
<td>carbon-13 isotope</td>
</tr>
<tr>
<td>C3</td>
<td>mouse strain (C3H/HeJ)</td>
</tr>
<tr>
<td>C3</td>
<td>plants that use only the Calvin cycle for fixing the carbon dioxide from the air</td>
</tr>
<tr>
<td>C4</td>
<td>plants that use the Hatch- Slack cycle for fixing the carbon dioxide from the air</td>
</tr>
<tr>
<td>C16:0</td>
<td>palmitic acid (saturated fatty acid)</td>
</tr>
<tr>
<td>C18:1</td>
<td>unsaturated fatty acid</td>
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<tr>
<td>Ca</td>
<td>calcium</td>
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<tr>
<td>Ca</td>
<td>ambient concentration</td>
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<tr>
<td>[Ca]</td>
<td>calcium concentration</td>
</tr>
<tr>
<td>Ca²⁺</td>
<td>calcium ion</td>
</tr>
<tr>
<td>CA</td>
<td>Canada (ICD-10-CA)</td>
</tr>
<tr>
<td>CAA</td>
<td>Clean Air Act</td>
</tr>
<tr>
<td>CALINE4</td>
<td>California line source dispersion model for predicting air pollutant concentrations near roads</td>
</tr>
<tr>
<td>CAM</td>
<td>plants that use crassulacean acid metabolism for fixing the carbon dioxide from the air</td>
</tr>
<tr>
<td>CAMP</td>
<td>Childhood Asthma Management Program</td>
</tr>
<tr>
<td>CAMx</td>
<td>Comprehensive Air Quality Model, with extensions</td>
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<tr>
<td>CAN</td>
<td>Canada</td>
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<tr>
<td>CAP(s)</td>
<td>concentrated ambient particles</td>
</tr>
<tr>
<td>CAR</td>
<td>centriacinar region</td>
</tr>
<tr>
<td>CASAC</td>
<td>Clean Air Scientific Advisory Committee</td>
</tr>
<tr>
<td>CASTNET</td>
<td>Clean Air Status and Trends Network</td>
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<td>CAT</td>
<td>catalase</td>
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<td>CB</td>
<td>carbon black; CMAQ mechanisms (ex., CB04, CB05, CB06)</td>
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<tr>
<td>C57BL/6</td>
<td>mouse strain</td>
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<tr>
<td>C57BL/6J</td>
<td>mouse strain</td>
</tr>
<tr>
<td>CBSA</td>
<td>core-based statistical area</td>
</tr>
<tr>
<td>C/C</td>
<td>carbon of total carbon</td>
</tr>
<tr>
<td>CCSP</td>
<td>Clara cell secretory protein</td>
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<tr>
<td>CD</td>
<td>cluster of differentiation (various receptors on T-cells: CD8+, CD44, etc.); criteria document (see AQCD)</td>
</tr>
<tr>
<td>CD-1</td>
<td>mouse strain</td>
</tr>
<tr>
<td>CDC</td>
<td>Centers for Disease Control and Prevention</td>
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<tr>
<td>CF</td>
<td>charcoal-filtered; carbon filtered air</td>
</tr>
<tr>
<td>CF2</td>
<td>twice-filtered air (particulate filter and activated charcoal filter)</td>
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<tr>
<td>C-fibers</td>
<td>afferent, slow, unmyelinated nerves innervating the respiratory system</td>
</tr>
<tr>
<td>CFR</td>
<td>Code of Federal Regulations</td>
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<tr>
<td>CGRP</td>
<td>calcitonin gene-related peptide</td>
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<tr>
<td>CH₃</td>
<td>methyl group</td>
</tr>
<tr>
<td>CH₄</td>
<td>methane</td>
</tr>
<tr>
<td>C₂H₂</td>
<td>acetylene</td>
</tr>
<tr>
<td>C₂H₄</td>
<td>ethylene</td>
</tr>
<tr>
<td>C₃H</td>
<td>mouse strain (C3H/HEJ or C3H/OuJ)</td>
</tr>
<tr>
<td>C₃H₆</td>
<td>propylene</td>
</tr>
<tr>
<td>CH₃Br</td>
<td>methyl bromide</td>
</tr>
<tr>
<td>CH₃CHO</td>
<td>acetaldehyde</td>
</tr>
<tr>
<td>CH₃Cl</td>
<td>methyl chloride</td>
</tr>
<tr>
<td>CH₃CO</td>
<td>acetyl radical(s)</td>
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<tr>
<td>CHD</td>
<td>coronary heart disease</td>
</tr>
<tr>
<td>CHF</td>
<td>congestive heart failure</td>
</tr>
<tr>
<td>C₂H₃-H</td>
<td>ethane</td>
</tr>
<tr>
<td>C₃H₁HeJ</td>
<td>mouse strain</td>
</tr>
<tr>
<td>CH₃I</td>
<td>methyl iodide</td>
</tr>
</tbody>
</table>
ECOPHYS physiological process modeling to predict the response of aspen forest ecosystems (modeling growth and environmental stress in Populus)

ED emergency department; embryonic day (ex., ED5, ED20)

EGEA (The) Epidemiology (study on) Genetics and Environment of Asthma, (adults and children with asthma)

EGEA2 follow-up study on EGEA (adults with asthma only)

EHC-93 ambient PM reference sample (urban dust [air particles] collected in Ottawa Canada)

ELF extracellular lining fluid

EMI (U.S. EPA) Exposure Model for Individuals

E_{na} exposure to pollutant of nonambient origin

ENA-78 epithelial cell-derived neutrophil-activating peptide 78

eNO exhaled nitric oxide

eNOS endothelial nitric oxide synthase

ENVISAT (EAS) Earth Observation satellite

EOTCP European Open Top Chamber Programme

EP epithelial cells

EPA U.S. Environmental Protection Agency

EPIC European Prospective Investigation into Cancer and Nutrition

ER emergency room

ESA European Space Agency

ET extrathoracic; endothelin (i.e., ET-1)

ET_{1} anterior nasal passages within the extrathoracic (ET) region

ET_{2} oral airway and posterior nasal passages within the extrathoracic (ET) region

ETS environmental tobacco smoke

EU European Union

EUS eastern U.S.

ϕ, Φ Phi; calculated efficiency

Φ_{PSII}-max maximum photochemical effective quantum yield of PSII

f Fraction of the relevant time period

F female

F_{344} Fischer 344 (rat strain)

F_{2a} 8-isoprostane (major F2 prostaglandin [8 iso-PGF_{2a}])

FA filtered air

FACE free-air−CO_{2} enrichment (system)

FACES Fresno Asthmatic Children’s Environment Study

f_{6} frequency of breathing

FC fibrocartilaginous coat

FEF forced expiratory flow

FEF_{25-75} forced expiratory flow between the times at which 25% and 75% of the vital capacity is reached

FEF_{x} forced expiratory flow after (x)% vital capacity (e.g., after 25, 50, or 75% vital capacity)

FEF_{x}\_\_ forced expiratory volume in 1 second

FEM Federal equivalent method

FeNO exhaled nitric oxide fraction

FEV_{1} forced expiratory volume in 1 second

FIA (USDA Forest Service) Forest Inventory and Analysis Program

F_{\text{air}} infiltration factor

F_{\text{air}}\_i infiltration factor for indoor environment (i)

FHC-93 (USDA Forest Service) Forest Health Monitoring Program

FIA (USDA Forest Service) Forest Inventory and Analysis Program

F_{\text{air}} fractional uptake efficiency

F_{\text{LRT}} fractional uptake efficiency of the lower respiratory tract (LRT)

F_{\text{nasal}} fractional uptake efficiency via nasal absorption

F_{0} fraction of time spent in outdoor microenvironments

FPM Forest Pest Management

FLAG Federal land managers’ air quality related values workgroup

FRAP ferric reducing ability of plasma

FR functional residual capacity

FRM Federal reference method

F_{\text{RT}} fractional uptake efficiency of the respiratory tract (RT)

F_{\text{cut-off threshold}} flux cut off threshold

F_{\text{URT}} fractional uptake efficiency of the upper respiratory tract (URT)

FVC forced vital capacity

F_{\text{V/Fm}} a ratio: a measure of the maximum efficiency of Photosystem II

FVI fruits and vegetables index

γ gamma

γ-TOH gamma-tocopherol

g, kg, mg, µg, ng, pg gram(s), kilogram(s), milligram(s), microgram(s), nanogram(s), picogram(s)

G granulocyte: guanosine

g (gram(s); gaseous form: (g)O_3)

GAM generalized additive model(s)

g_{bs} conductance through boundary layer and stomata

GCLC (glutathione genetic variant) glutamate-cysteine ligase catalytic subunit

GCLM (glutathione genetic variant) glutamate-cysteine ligase modifier subunit

G-CSF granulocyte colony-stimulating factor (receptor)

GD gestational day

GEE generalized estimating equations
<table>
<thead>
<tr>
<th>Term</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>GEOS</td>
<td>(NASA) Goddard Earth Observing System model</td>
</tr>
<tr>
<td>GEOS5</td>
<td>GEOS version 5</td>
</tr>
<tr>
<td>GEOS-Chem</td>
<td>GEOS-Chemistry (tropospheric model)</td>
</tr>
<tr>
<td>GFAP</td>
<td>glial fibrillary acidic protein</td>
</tr>
<tr>
<td>GH</td>
<td>growth hormone</td>
</tr>
<tr>
<td>GHG</td>
<td>greenhouse gas</td>
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<tr>
<td>GLM(s)</td>
<td>generalized linear model(s)</td>
</tr>
<tr>
<td>GMAO</td>
<td>(NASA) Global Modeling and Assimilation Office</td>
</tr>
<tr>
<td>GM-CSF</td>
<td>granulocyte macrophage colony-stimulating factor</td>
</tr>
<tr>
<td>GOME</td>
<td>(ESA) Global Ozone Monitoring Experiment (spectrometer)</td>
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<tr>
<td>GOMOS</td>
<td>Global Ozone Monitoring by Occultation of Stars (ESA ENVISAT spectrometer measuring long-term trends in O₃)</td>
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<tr>
<td>G6P</td>
<td>glucose-6-phosphate</td>
</tr>
<tr>
<td>G6PD</td>
<td>glucose-6-phosphate dehydrogenase</td>
</tr>
<tr>
<td>GPP</td>
<td>gross primary production</td>
</tr>
<tr>
<td>G-proteins</td>
<td>GTPases</td>
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<tr>
<td>GPT</td>
<td>gas phase titration</td>
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<td>GR</td>
<td>glutathione reductase</td>
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<td>GSH</td>
<td>glutathione; reduced glutathione</td>
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<tr>
<td>GSO₃⁻/GSO₃²⁻</td>
<td>guanine sulfonates</td>
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<td>GSR</td>
<td>glutathione reductase</td>
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<td>glutathione synthetase</td>
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<td>GSSG</td>
<td>glutathione disulfide</td>
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<td>GST</td>
<td>glutathione S-transferase</td>
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<td>GSTM1</td>
<td>glutathione S-transferase polymorphism M1 genotypes (GSTM1-null,-GSTM1-sufficient)</td>
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<tr>
<td>GSTP1</td>
<td>glutathione S-transferase polymorphism P1 genotypes</td>
</tr>
<tr>
<td>GTP</td>
<td>guanosine triphosphate</td>
</tr>
<tr>
<td>GTPases</td>
<td>G-proteins/enzymes</td>
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<tr>
<td>GWP</td>
<td>global warming potential</td>
</tr>
<tr>
<td>GxE</td>
<td>gene-environmental interaction</td>
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<tr>
<td>h</td>
<td>hour(s)</td>
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<tr>
<td>h/day</td>
<td>hour(s) per day</td>
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<tr>
<td>H; H⁺⁺; H⁻⁺</td>
<td>atomic hydrogen, hydrogen ion; hydrogen radical</td>
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<tr>
<td>³H</td>
<td>radiolabeled hydrogen; tritium</td>
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<tr>
<td>H₂</td>
<td>molecular hydrogen</td>
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<tr>
<td>ha</td>
<td>hectare</td>
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<td>HA</td>
<td>hyaluronic acid, hospital admission</td>
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<td>HA(s)</td>
<td>hospital admission(s)</td>
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<tr>
<td>Hb</td>
<td>hemoglobin</td>
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<td>HbA1c</td>
<td>glycosylated hemoglobin (blood test)</td>
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<td>HC(s)</td>
<td>hydrocarbon(s)</td>
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<tr>
<td>HCFC(s)</td>
<td>hydrochlorofluorocarbon(s)</td>
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<tr>
<td>HCHO</td>
<td>formaldehyde</td>
</tr>
<tr>
<td>H₂CO</td>
<td>formaldehyde</td>
</tr>
<tr>
<td>HCO•</td>
<td>formyl (radical)</td>
</tr>
<tr>
<td>HDM</td>
<td>house dust mite</td>
</tr>
<tr>
<td>2HDM</td>
<td>2nd-highest daily maximum</td>
</tr>
<tr>
<td>HDMA</td>
<td>house dust mite allergen</td>
</tr>
<tr>
<td>³He</td>
<td>non-radioactive isotope of helium</td>
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<tr>
<td>HeJ</td>
<td>O₂⁻-resistant C3H mouse strain (C3H/HeJ)</td>
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<td>HEPA</td>
<td>high efficiency particle air (filter)</td>
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<td>HERO</td>
<td>Health and Environmental Research Online, NCEA Database System</td>
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<tr>
<td>12-HETE</td>
<td>12-Hydroxyeicosatetraenoic acid</td>
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<tr>
<td>HF</td>
<td>(HRV signal) high-frequency power</td>
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<td>HFCs</td>
<td>hydrofluorocarbons</td>
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<td>Hg</td>
<td>mercury</td>
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<td>HHP-C9</td>
<td>1-hydroxy-1-hydroperoxynonane</td>
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<td>HIST</td>
<td>histamine</td>
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<td>HLA</td>
<td>human leukocyte antigen</td>
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<td>HLA-DR</td>
<td>human leukocyte antigen receptor genes</td>
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<td>HMOX</td>
<td>Heme oxygenase</td>
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<td>HMOX-1</td>
<td>heme-oxygenase-1 (polymorphism)</td>
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<td>HNO₂</td>
<td>nitric acid</td>
</tr>
<tr>
<td>HNO₃</td>
<td>nitric acid</td>
</tr>
<tr>
<td>HNO₄</td>
<td>pernitric acid</td>
</tr>
<tr>
<td>HO</td>
<td>hydroxyl; heme oxygenase</td>
</tr>
<tr>
<td>HO⁺</td>
<td>hydroxyl radical</td>
</tr>
<tr>
<td>HO-1</td>
<td>heme oxygenase 1</td>
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<tr>
<td>HO₂⁻</td>
<td>hydroperoxy; hydroperoxy radical; protonated superoxide</td>
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<tr>
<td>HO₂⁺</td>
<td>protonated ozone radical</td>
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<td>H₂O</td>
<td>water</td>
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<td>H₂O₂</td>
<td>hydrogen peroxide</td>
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<tr>
<td>H₂O⁺</td>
<td>hydronium ion</td>
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<tr>
<td>HOCH₂OOH</td>
<td>hydroxymethylhydroperoxide</td>
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<td>HONO</td>
<td>nitrous acid</td>
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<td>HO₂NO₂</td>
<td>peroxynitric acid</td>
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<td>HOONO</td>
<td>pernitrous acid</td>
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<td>HOX</td>
<td>hydrogen radical(s)</td>
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<td>hPa</td>
<td>hexitopascal</td>
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<td>HPLC</td>
<td>high-pressure liquid chromatography</td>
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<td>HPOT</td>
<td>13-hydroperoxide linolenic acid</td>
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<td>HR</td>
<td>heart rate, hazard ratio</td>
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<td>HRmax</td>
<td>maximum heart rate</td>
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<td>HRP</td>
<td>horseradish peroxidase</td>
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<td>HRV</td>
<td>heart rate variability</td>
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<td>HSC</td>
<td>Houston Ship Channel (Texas)</td>
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<tr>
<td>hs-CRP</td>
<td>high-sensitivity C-reactive protein</td>
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<tr>
<td>H₂SO₄</td>
<td>sulfuric acid</td>
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<td>HSP</td>
<td>high speed pellet (after centrifuge spin)</td>
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<td>HSP70</td>
<td>heat shock protein 70</td>
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<tr>
<td>Abbreviation</td>
<td>Definition</td>
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<tr>
<td>max</td>
<td>maximum</td>
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<tr>
<td>MBL</td>
<td>marine boundary layer</td>
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<td>MCA</td>
<td>minimum cross-sectional area</td>
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<td>MCCP</td>
<td>Mountain Cloud Chemistry Program</td>
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<tr>
<td>Mch; MCh</td>
<td>methacholine</td>
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<tr>
<td>MCM</td>
<td>master chemical mechanism</td>
</tr>
<tr>
<td>MCP-1</td>
<td>monocyte chemotactic protein 1</td>
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<td>MDA</td>
<td>malondialdehyde</td>
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<tr>
<td>MDAR</td>
<td>monodehydroascorbate reductase</td>
</tr>
<tr>
<td>MDI</td>
<td>Mediterranean diet index</td>
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<tr>
<td>MDL</td>
<td>minimum detection level</td>
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<tr>
<td>MEGAN</td>
<td>model of emissions of gases and aerosols from nature</td>
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<td>MeJA</td>
<td>methyl jasmonate</td>
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<tr>
<td>MENTOR</td>
<td>Modeling Environment for Total Risk Studies</td>
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<tr>
<td>METs</td>
<td>metabolic equivalent unit(s) [of work]</td>
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<tr>
<td>MFR</td>
<td>Maximum Feasible Reduction</td>
</tr>
<tr>
<td>Mg</td>
<td>magnesium</td>
</tr>
<tr>
<td>MGDG</td>
<td>monogalactosyl diacylglycerol</td>
</tr>
<tr>
<td>mg/m³</td>
<td>milligrams per cubic meter</td>
</tr>
<tr>
<td>MHC</td>
<td>major histocompatibility complex</td>
</tr>
<tr>
<td>mi</td>
<td>mile(s)</td>
</tr>
<tr>
<td>MI</td>
<td>myocardial infarction, “heart attack”</td>
</tr>
<tr>
<td>MIESR</td>
<td>matrix isolation electron spin resonance (spectroscopy)</td>
</tr>
<tr>
<td>min</td>
<td>minute; minimum</td>
</tr>
<tr>
<td>MIP</td>
<td>macrophage inflammatory protein</td>
</tr>
<tr>
<td>MIP-2</td>
<td>macrophage inflammatory protein 2</td>
</tr>
<tr>
<td>mL</td>
<td>milliliter</td>
</tr>
<tr>
<td>mL/min</td>
<td>milliliter(s) per minute</td>
</tr>
<tr>
<td>MLN</td>
<td>mediastinal lymph node</td>
</tr>
<tr>
<td>Mm</td>
<td>megameter</td>
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<tr>
<td>mm</td>
<td>millimeter(s)</td>
</tr>
<tr>
<td>MM Mt.</td>
<td>Mt. Mitchell site</td>
</tr>
<tr>
<td>MM5</td>
<td>National Center for Atmospheric Research/Penn State Mesoscale Model (version 5)</td>
</tr>
<tr>
<td>MMAD</td>
<td>mass median aerodynamic diameter; mass median aerodynamic density</td>
</tr>
<tr>
<td>MMEF</td>
<td>maximal midexpiratory flow</td>
</tr>
<tr>
<td>mmHg</td>
<td>millimeters of mercury</td>
</tr>
<tr>
<td>MMMD</td>
<td>mean maximum mixing height depth</td>
</tr>
<tr>
<td>MMP-2</td>
<td>matrix metalloproteinase-2</td>
</tr>
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<td>MMP-3</td>
<td>matrix metalloproteinase-3</td>
</tr>
<tr>
<td>MMP-9</td>
<td>metalloproteinase-9</td>
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<td>MMSP</td>
<td>Mount Mitchell State Park, NC</td>
</tr>
<tr>
<td>Mn</td>
<td>manganese</td>
</tr>
<tr>
<td>M/N</td>
<td>pooled data from mouth and nasal exposure</td>
</tr>
<tr>
<td>MnSOD</td>
<td>Manganese superoxide dismutase</td>
</tr>
<tr>
<td>mo</td>
<td>month(s)</td>
</tr>
<tr>
<td>MOA(s)</td>
<td>mode(s) of action</td>
</tr>
<tr>
<td>MOBILE</td>
<td>(U.S. EPA) mobile vehicle emission factor model (on-road vehicles)</td>
</tr>
<tr>
<td>MOBILE6</td>
<td>vehicle emissions modeling software version 6; replaced by MOVES</td>
</tr>
<tr>
<td>MODNR</td>
<td>Missouri Department of Natural Resources</td>
</tr>
<tr>
<td>MONICA</td>
<td>Monitoring of Trends and Determinants in Cardiovascular Disease</td>
</tr>
<tr>
<td>MoOx</td>
<td>molybdenum oxides</td>
</tr>
<tr>
<td>MOSES</td>
<td>Met Office Surface Exchange Scheme</td>
</tr>
<tr>
<td>MOVES</td>
<td>Motor Vehicle Emission Simulator (replaced MOBILE6; for estimating emissions from cars, trucks, and motorcycles)</td>
</tr>
<tr>
<td>MOZAIC</td>
<td>Measurement of Ozone and Water Vapor by Airbus In-Service Aircraft</td>
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<tr>
<td>MOZART</td>
<td>Model for Ozone and Related chemical Tracers</td>
</tr>
<tr>
<td>MPAN</td>
<td>peroxymethacryloyl nitrate; peroxy-methacrylic nitric anhydride</td>
</tr>
<tr>
<td>MPO</td>
<td>myeloperoxidase</td>
</tr>
<tr>
<td>MQL</td>
<td>Minimum quantification limit</td>
</tr>
<tr>
<td>MRI</td>
<td>magnetic resonance imaging; Midwest Research Institute; Meteorological Research Institute</td>
</tr>
<tr>
<td>mRNA</td>
<td>messenger RNA</td>
</tr>
<tr>
<td>MS</td>
<td>mass spectrometry; Mt. Moosilauke site</td>
</tr>
<tr>
<td>MSA</td>
<td>Metropolitan Statistical Area; methane sulfonic acid</td>
</tr>
<tr>
<td>MSL</td>
<td>mean sea level</td>
</tr>
<tr>
<td>MS/MS</td>
<td>tandem mass spectrometry</td>
</tr>
<tr>
<td>MT</td>
<td>million ton(s); metric ton(s)</td>
</tr>
<tr>
<td>MT, Mt</td>
<td>metallothionein</td>
</tr>
<tr>
<td>MT1</td>
<td>mitochondria</td>
</tr>
<tr>
<td>MTBE</td>
<td>methyl-tertiary-butyl ether</td>
</tr>
<tr>
<td>mtDNA</td>
<td>mitochondrial DNA</td>
</tr>
<tr>
<td>Mtn</td>
<td>mountain</td>
</tr>
<tr>
<td>MW</td>
<td>molecular weight</td>
</tr>
<tr>
<td>MyD88</td>
<td>myeloid differentiation primary response gene 88</td>
</tr>
<tr>
<td>n, N</td>
<td>number; number of observations</td>
</tr>
<tr>
<td>N</td>
<td>nitrogen; North; nasal exposure by natural breathing</td>
</tr>
<tr>
<td>¹⁵N</td>
<td>nitrogen-15, stable isotope of nitrogen</td>
</tr>
<tr>
<td>N₂</td>
<td>molecular nitrogen; nonreactive nitrogen</td>
</tr>
<tr>
<td>Na</td>
<td>sodium</td>
</tr>
<tr>
<td>NA</td>
<td>noradrenaline; North American</td>
</tr>
<tr>
<td>Abbreviation</td>
<td>Definition</td>
</tr>
<tr>
<td>--------------</td>
<td>------------</td>
</tr>
<tr>
<td>NA; N/A</td>
<td>not available; not applicable</td>
</tr>
<tr>
<td>Na⁺</td>
<td>sodium ion</td>
</tr>
<tr>
<td>NAAQS</td>
<td>National Ambient Air Quality Standards</td>
</tr>
<tr>
<td>NAD</td>
<td>nicotinamide adenine nucleotide</td>
</tr>
<tr>
<td>NADH</td>
<td>reduced nicotinamide adenine dinucleotide; nicotinamide adenine dinucleotide dehydrogenase</td>
</tr>
<tr>
<td>NADP</td>
<td>National Atmospheric Deposition Program</td>
</tr>
<tr>
<td>NADPH</td>
<td>reduced nicotinamide adenine dinucleotide phosphate</td>
</tr>
<tr>
<td>NADPH-CR</td>
<td>reduced nicotinamide adenine dinucleotide phosphate - cytochrome c reductase</td>
</tr>
<tr>
<td>NaE</td>
<td>sodium erythorbate</td>
</tr>
<tr>
<td>NAG</td>
<td>N-acetyl-glucosaminidase</td>
</tr>
<tr>
<td>Na-K-ATPase</td>
<td>sodium-potassium-dependent adenosine triphosphatase</td>
</tr>
<tr>
<td>NAMS</td>
<td>National Ambient Monitoring Stations</td>
</tr>
<tr>
<td>NAPAP</td>
<td>National Acid Precipitation Assessment Program</td>
</tr>
<tr>
<td>NAPBN</td>
<td>National Air Pollution Background Network</td>
</tr>
<tr>
<td>NARE</td>
<td>North Atlantic Regional Experiment</td>
</tr>
<tr>
<td>NARSTO</td>
<td>North American Regional Strategy for Tropospheric Ozone</td>
</tr>
<tr>
<td>NAS</td>
<td>National Academy of Sciences; Normative Aging Study</td>
</tr>
<tr>
<td>NASA</td>
<td>National Aeronautics and Space Administration</td>
</tr>
<tr>
<td>NBS</td>
<td>National Bureau of Standards</td>
</tr>
<tr>
<td>NBTH</td>
<td>3-methyl-2-benzothiazolinone acetone azine</td>
</tr>
<tr>
<td>NCE</td>
<td>net carbon exchange</td>
</tr>
<tr>
<td>NCEA</td>
<td>National Center for Environmental Assessment</td>
</tr>
<tr>
<td>NCEA-RTP</td>
<td>NCEA Division in Research Triangle Park, NC</td>
</tr>
<tr>
<td>NCHS</td>
<td>National Center for Health Statistics</td>
</tr>
<tr>
<td>NCICAS</td>
<td>National Cooperative Inner-City Asthma Study</td>
</tr>
<tr>
<td>NCLAN</td>
<td>National Crop Loss Assessment Network</td>
</tr>
<tr>
<td>NCore</td>
<td>National Core multipollutant monitoring network</td>
</tr>
<tr>
<td>NC-R</td>
<td>resistant clones of white clover</td>
</tr>
<tr>
<td>NC-S</td>
<td>sensitive clones of white clover</td>
</tr>
<tr>
<td>ND; n.d.</td>
<td>not detectable; not detected; no data</td>
</tr>
<tr>
<td>2ndHDM</td>
<td>2nd-highest daily maximum</td>
</tr>
<tr>
<td>NDF</td>
<td>neutral detergent fiber</td>
</tr>
<tr>
<td>NEE</td>
<td>net ecosystem exchange (of carbon or CO₂)</td>
</tr>
<tr>
<td>NEI</td>
<td>National Emissions Inventory</td>
</tr>
<tr>
<td>NEM</td>
<td>National Ambient Air Quality Standards Exposure Model</td>
</tr>
<tr>
<td>NEP</td>
<td>Net Ecosystem Production</td>
</tr>
<tr>
<td>NERL</td>
<td>National Exposure Research Laboratory</td>
</tr>
<tr>
<td>NESCAUM</td>
<td>Northeast States for Coordinated Air Use Management</td>
</tr>
<tr>
<td>NF</td>
<td>National Forest; non-filtered air</td>
</tr>
<tr>
<td>NF-xB</td>
<td>nuclear factor kappa B</td>
</tr>
<tr>
<td>ng</td>
<td>nanogram(s)</td>
</tr>
<tr>
<td>NGF</td>
<td>nerve growth factor</td>
</tr>
<tr>
<td>NH</td>
<td>northern hemisphere</td>
</tr>
<tr>
<td>NH₄</td>
<td>ammonia</td>
</tr>
<tr>
<td>NH₄⁺</td>
<td>ammonium ion</td>
</tr>
<tr>
<td>NH₄HSO₄</td>
<td>ammonium bisulfate</td>
</tr>
<tr>
<td>(NH₄)₂SO₄</td>
<td>ammonium sulfate</td>
</tr>
<tr>
<td>NHANES</td>
<td>National Health and Nutrition Examination Survey</td>
</tr>
<tr>
<td>NHANES III</td>
<td>National Health and Nutrition Examination Survey III</td>
</tr>
<tr>
<td>NHAPS</td>
<td>National Human Activity Pattern Survey</td>
</tr>
<tr>
<td>NHIS</td>
<td>National Health Interview Survey</td>
</tr>
<tr>
<td>(NH₄)₂SO₄</td>
<td>amonium sulfate</td>
</tr>
<tr>
<td>NIH</td>
<td>National Institutes of Health</td>
</tr>
<tr>
<td>NIST</td>
<td>National Institute of Standards and Technology</td>
</tr>
<tr>
<td>NK</td>
<td>natural killer cells; neurokinin</td>
</tr>
<tr>
<td>NKT</td>
<td>natural killer T-cells</td>
</tr>
<tr>
<td>NL</td>
<td>nasal lavage</td>
</tr>
<tr>
<td>NLF</td>
<td>nasal lavage fluid</td>
</tr>
<tr>
<td>NM</td>
<td>National Monument</td>
</tr>
<tr>
<td>NMHC(s)</td>
<td>nonmethane hydrocarbon(s)</td>
</tr>
<tr>
<td>NMMAPS</td>
<td>National Morbidity, Mortality, and Air Pollution Study</td>
</tr>
<tr>
<td>NMOC(s)</td>
<td>nonmethane organic compound(s)</td>
</tr>
<tr>
<td>NVMOCs</td>
<td>nonmethane volatile organic compounds</td>
</tr>
<tr>
<td>NN</td>
<td>normal-to-normal (NN or RR) time interval between each QRS complex in the EKG</td>
</tr>
<tr>
<td>NNK</td>
<td>4-(N-nitrosomethylamino)-1-(3-pyridyl)-1-butanolone</td>
</tr>
<tr>
<td>nNOS</td>
<td>neuronal nitric oxide synthase (NOS)</td>
</tr>
<tr>
<td>NO</td>
<td>nitric oxide</td>
</tr>
<tr>
<td>-NO</td>
<td>nitric oxide concentration (interpunct NO)</td>
</tr>
<tr>
<td>NO₂</td>
<td>nitrogen dioxide</td>
</tr>
<tr>
<td>NO₂; NO₃⁺</td>
<td>nitrate, nitrate radical</td>
</tr>
<tr>
<td>NO₃⁻</td>
<td>nitrate, nitrate ion</td>
</tr>
<tr>
<td>N₂O</td>
<td>nitrous oxide</td>
</tr>
<tr>
<td>N₂O₅</td>
<td>dinitrogen pentoxide</td>
</tr>
<tr>
<td>NOAA</td>
<td>National Oceanic and Atmospheric Administration</td>
</tr>
<tr>
<td>NOAEL</td>
<td>no observed adverse effect level</td>
</tr>
</tbody>
</table>
NOS  nitric oxide synthase (types, NOS-1, NOS-2, NOS-3)
NO\textsubscript{x}  nitrogen oxides, oxides of nitrogen (NO + NO\textsubscript{2})
NO\textsubscript{v}  sum of NO\textsubscript{x} and NO\textsubscript{2}; odd nitrogen species; total oxidized nitrogen
NO\textsubscript{2}  sum of all inorganic and organic reaction products of NO\textsubscript{x} (HONO, HNO\textsubscript{3}, HNO\textsubscript{4}, organic nitrates, particulate nitrate, nitro-PAHs, etc.)
NP  National Park
NPP  net primary production
NPS  National Park Service, U.S. Department of the Interior
NQO1  NAD(P)H-quinone oxidoreductase (genotype)
NQO1wt  NAD(P)H-quinone oxidoreductase wild type (genotype)
NR  not reported
Nr  reactive nitrogen
NRC  National Research Council
Nrf-2  nuclear factor erythroid 2-related factor 2
Nrf2-ARE  NF-e2-related factor 2-antioxidant response element
NS; n.s.  nonsignificant; non-smoker; national seashore; natural spline
NSAID  non-steroidal anti-inflammatory agent
NSBR  nonspecific bronchial responsiveness
NSF  National Science Foundation
NTE  nasal turbinate epithelial (cells)
NTN  National Trends Network
NTP  National Toxicology Program
NTRMs  NIST Traceable Reference Materials
NTS  nucleus of the solitary tract (in brainstem)
NWR  national wildlife refuge
NWS  National Weather Service
NZW  New Zealand white (rabbit)
O  oxygen; horizon forest floor
\textsuperscript{18}O  oxygen-18, stable isotope of oxygen
O\textsubscript{2}  molecular oxygen
O\textsubscript{2}\textsuperscript{-}  superoxide
O\textsubscript{2}\textsuperscript{•}  superoxide radical
\textsuperscript{1}O\textsubscript{2}  singlet oxygen
O\textsubscript{3}  ozone
\textsuperscript{16}O\textsubscript{3}  (oxygen-18 labeled) ozone
O\textsubscript{3}\textsuperscript{*}  electronically excited ozone
OAQPS  Office of Air Quality Planning and Standards
OAR  Office of Air and Radiation
OBMs  observationally based methods
OC  organic carbon
OD  outer diameter; optical density
O\textsuperscript{(3}D)  electronically excited oxygen atom
OH, OH\textsuperscript{+}  hydroxyl group, hydroxyl radical
8-OhD\textsubscript{G}  8-hydroxy-2\textsuperscript{-}deoxyguanosine
OLS  ordinary least squares
OMI  Ozone Monitoring Instrument
ON  Ontario
ONOO\textsuperscript{-}  peroxynitrate ion
O\textsuperscript{(3}P)  ground-state oxygen atom
OPE  ozone production efficiency
OPECs  Outdoor Plant Environment Chambers
OR  odds ratio
ORD  Office of Research and Development
OSHA  Occupational Safety and Health Administration
OuJ  O\textsubscript{3}-sensitive C3H mouse strain (C3H/OuJ)
OVA  ovalbumin
OX  odd oxygen species; total oxidants
OxComp  oxidative capacity of the atmosphere
oz  ounce(s)
P  pressure in atmospheres; plants grown in pots; phosphorus; penetration fraction of O\textsubscript{3} into the microenvironment; pulmonary region
p  probability value
P450  cytochrome P450
p53  cell cycle protein gene
P90  90th percentile of the absolute difference in concentrations
PACF  partial autocorrelation function of the model residuals
PAD  peripheral arterial disease; pollutant-applied dose
PAF  platelet-activating factor; paroxysmal atrial fibrillation
PAH(s)  polycyclic aromatic hydrocarbon(s)
PAI-1  plasminogen activator fibrinogen inhibitor-1
PAL  phenylalanine ammonia lyase
PAMS  Photochemical Assessment Monitoring Stations network
PAN  peroxacetyl nitrate
PaO\textsubscript{2}  arterial oxygen pressure
PAPA  Public Health and Air Pollution in Asia
PAR  photosynthetically active radiation; proximal alveolar region
P\textsubscript{atm}  Pressure in atmospheres
p-ATP  para-acetamidophenol
Pb  Lead
PBL  planetary boundary layer; peripheral blood lymphocytes
<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>PBM</td>
<td>population-based model or modeling</td>
</tr>
<tr>
<td>PBN</td>
<td>C-phenyl N-tert-butyl nitrone</td>
</tr>
<tr>
<td>PBPK</td>
<td>physiologically based pharmacokinetic (model)</td>
</tr>
<tr>
<td>PBS</td>
<td>phosphate buffered saline</td>
</tr>
<tr>
<td>PC</td>
<td>phosphatidylcholine</td>
</tr>
<tr>
<td>PC₂₀</td>
<td>provocative concentration that produces a 20% decrease in forced expiratory volume in 1 second</td>
</tr>
<tr>
<td>PC₂₀FEV₁</td>
<td>provocative concentration that produces a 20% decrease in FEV₁</td>
</tr>
<tr>
<td>PC₅₀</td>
<td>provocative concentration that produces a 50% decrease in forced expiratory volume in 1 second</td>
</tr>
<tr>
<td>PCA</td>
<td>principal component analysis</td>
</tr>
<tr>
<td>PC-ALF</td>
<td>1-palmitoyl-2-(9-oxononoyl)-sn-glycero-3-phosphocholine</td>
</tr>
<tr>
<td>PCD</td>
<td>programmed cell death</td>
</tr>
<tr>
<td>PCI</td>
<td>picryl chloride</td>
</tr>
<tr>
<td>pCNEM</td>
<td>Canadian version of National Ambient Air Quality Standards Exposure Model</td>
</tr>
<tr>
<td>PCO₂</td>
<td>Average partial pressure of O₂ in lung capillaries</td>
</tr>
<tr>
<td>pCO₂</td>
<td>partial pressure of carbon dioxide</td>
</tr>
<tr>
<td>PCR</td>
<td>polymerase chain reaction</td>
</tr>
<tr>
<td>PCR-DGGE</td>
<td>PCR–denaturing gradient gel electrophoresis</td>
</tr>
<tr>
<td>PD</td>
<td>pregnancy day</td>
</tr>
<tr>
<td>PD₃₀</td>
<td>provocative dose that produces a 20% decrease in FEV₁</td>
</tr>
<tr>
<td>PD₂₀FEV₁</td>
<td>provocative dose that produces a 20% decrease in FEV₁</td>
</tr>
<tr>
<td>PD₁₀₀</td>
<td>provocative dose that produces a 100% increase in SRAW</td>
</tr>
<tr>
<td>PD₁₀₀SRAW</td>
<td>provocative dose that produces a 100% increase in SRAW</td>
</tr>
<tr>
<td>PDI</td>
<td>pain on deep inspiration</td>
</tr>
<tr>
<td>PE</td>
<td>postexposure, phosphatidylethanolamine</td>
</tr>
<tr>
<td>PEF</td>
<td>peak expiratory flow</td>
</tr>
<tr>
<td>PEF₀.75</td>
<td>peak expiratory flow in 0.75 second</td>
</tr>
<tr>
<td>PEFR</td>
<td>peak expiratory flow rate</td>
</tr>
<tr>
<td>PEF anterior</td>
<td>time to peak flow</td>
</tr>
<tr>
<td>PEG-CAT</td>
<td>polyethylene glycol-catalase</td>
</tr>
<tr>
<td>PEG-SOD</td>
<td>polyethylene glycol-superoxide dismutase</td>
</tr>
<tr>
<td>PEM(s)</td>
<td>personal exposure monitor(s)</td>
</tr>
<tr>
<td>Penh</td>
<td>enhanced pause</td>
</tr>
<tr>
<td>PEPc</td>
<td>phosphoenolpyruvate carboxylase</td>
</tr>
<tr>
<td>PFD</td>
<td>photosynthetic flux density</td>
</tr>
<tr>
<td>PFT</td>
<td>pulmonary function test</td>
</tr>
<tr>
<td>pg</td>
<td>picogram(s)</td>
</tr>
<tr>
<td>PG</td>
<td>prostaglandin (e.g., PGE₂, PGF₂); phosphatidyglycerol</td>
</tr>
<tr>
<td>6PGD</td>
<td>6-phosphogluconate dehydrogenase</td>
</tr>
<tr>
<td>PGE₂</td>
<td>prostaglandin E₂</td>
</tr>
<tr>
<td>PGF₂α</td>
<td>prostaglandin F₂-alpha</td>
</tr>
<tr>
<td>PGHS-2</td>
<td>prostaglandin endoperoxide G/H synthase 2</td>
</tr>
<tr>
<td>PGP</td>
<td>protein gene product (e.g., PGP9.5)</td>
</tr>
<tr>
<td>PGSM</td>
<td>Plant Growth Stress Model</td>
</tr>
<tr>
<td>pH</td>
<td>relative acidity; Log of the reciprocal of the hydrogen ion concentration</td>
</tr>
<tr>
<td>PHA</td>
<td>phytohemagglutinin A</td>
</tr>
<tr>
<td>PI</td>
<td>phosphatidylinositol; probability interval; posterior interval</td>
</tr>
<tr>
<td>PIF</td>
<td>peak inspiratory flow</td>
</tr>
<tr>
<td>PIZZ</td>
<td>respiratory phenotype</td>
</tr>
<tr>
<td>PK</td>
<td>pharmacokinetics</td>
</tr>
<tr>
<td>pKa</td>
<td>dissociation constant</td>
</tr>
<tr>
<td>PLFA</td>
<td>phospholipid fatty acid</td>
</tr>
<tr>
<td>PM</td>
<td>particulate matter</td>
</tr>
<tr>
<td>PMₓ</td>
<td>Particulate matter of a specific size range not defined for regulatory use. Usually X refers to the 50% cut point, the aerodynamic diameter at which the sampler collects 50% of the particles and rejects 50% of the particles. The collection efficiency, given by a penetration curve, increases for particles with smaller diameters and decreases for particles with larger diameters. The definition of PMₓ is sometimes abbreviated as &quot;particles with a nominal aerodynamic diameter less than or equal to X μm&quot; although X is usually a 50% cut point.</td>
</tr>
<tr>
<td>PM₂.₅</td>
<td>In general terms, particulate matter with an aerodynamic diameter less than or equal to a nominal 2.5 μm; a measurement of fine particles. In regulatory terms, particles with an upper 50% cut-point of 2.5 μm aerodynamic diameter (the 50% cut point diameter is the diameter at which the sampler collects 50% of the particles and rejects 50% of the particles) and a penetration curve as measured by a reference method based on Appendix L of 40 CFR Part 50 and designated in accordance with 40 CFR Part 53, by an equivalent method designated in accordance with 40 CFR Part 53, or by an approved regional method designated in accordance with Appendix C of 40 CFR Part 58.</td>
</tr>
</tbody>
</table>
PM$_{10}$ In general terms, particulate matter with an aerodynamic diameter less than or equal to a nominal 10 μm; a measurement of thoracic particles (i.e., that subset of inhalable particles thought small enough to penetrate beyond the larynx into the thoracic region of the respiratory tract). In regulatory terms, particles with an upper 50% cut-point of 10 ± 0.5 μm aerodynamic diameter (the 50% cut point diameter is the diameter at which the sampler collects 50% of the particles and rejects 50% of the particles) and a penetration curve as measured by a reference method based on Appendix J of 40 CFR Part 50 and designated in accordance with 40 CFR Part 53 or by an equivalent method designated in accordance with 40 CFR Part 53.

PM$_{10-2.5}$ In general terms, particulate matter with an aerodynamic diameter less than or equal to a nominal 10 μm and greater than a nominal 2.5 μm; a measurement of thoracic coarse particulate matter or the coarse fraction of PM$_{10}$. In regulatory terms, particles with an upper 50% cut-point of 10 μm aerodynamic diameter and a lower 50% cut-point of 2.5 μm aerodynamic diameter (the 50% cut point diameter is the diameter at which the sampler collects 50% of the particles and rejects 50% of the particles) as measured by a reference method based on Appendix O of 40 CFR Part 50 and designated in accordance with 40 CFR Part 53 or by an equivalent method designated in accordance with 40 CFR Part 53.

PM$_{10c}$ The PM$_{10-2.5}$ concentration of PM$_{10-2.5}$ measured by the 40 CFR Part 50 Appendix O reference method which consists of currently operated, co-located low-volume (16.7 Lpm) PM$_{10}$ and PM$_{2.5}$ reference method samplers.

p38MAPK p38 mitogen-activated protein kinase(s)

PM-CAMx Comprehensive Air Quality Model with extensions and with particulate matter chemistry

PMN(s) polymorphonuclear leukocyte(s)

PMT photomultiplier tube

PND post natal day

pNEM probabilistic National Exposure Model

PnET Photosynthetic EvapoTranspiration model

PNN proportion of interval differences of successive normal-beat intervals in EKG

PNN50 proportion of interval differences of successive normal-beat intervals greater than 50 ms in EKG
<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>QCE</td>
<td>quasi continuous exercise</td>
</tr>
<tr>
<td>qNP</td>
<td>non-photochemical quenching</td>
</tr>
<tr>
<td>qNP</td>
<td>non-photochemical quenching</td>
</tr>
<tr>
<td>qP</td>
<td>photochemical quenching</td>
</tr>
<tr>
<td>QRS</td>
<td>A complex of three distinct electrocardiogram waves which represent the beginning of ventricular contraction</td>
</tr>
<tr>
<td>QT</td>
<td>interval measure of the time interval between the start of the Q wave and the end of the T wave in the heart's electrical cycle</td>
</tr>
<tr>
<td>QTc</td>
<td>corrected QT interval</td>
</tr>
<tr>
<td>r</td>
<td>Pearson correlation coefficient</td>
</tr>
<tr>
<td>R, r</td>
<td>correlation coefficient</td>
</tr>
<tr>
<td>$R^2$</td>
<td>multiple regression correlation coefficient</td>
</tr>
<tr>
<td>$R^2$, $r^2$</td>
<td>coefficient of determination</td>
</tr>
<tr>
<td>RACM</td>
<td>Regional Atmospheric Chemistry Mechanism</td>
</tr>
<tr>
<td>RADM</td>
<td>Regional Acid Deposition Model</td>
</tr>
<tr>
<td>rALP</td>
<td>recombinant antileukoprotease</td>
</tr>
<tr>
<td>RAMS</td>
<td>Regional Atmospheric Modeling System</td>
</tr>
<tr>
<td>RANTES</td>
<td>regulated upon activation, normal T-cell expressed and secreted (cells)</td>
</tr>
<tr>
<td>Raw</td>
<td>airway resistance</td>
</tr>
<tr>
<td>RB</td>
<td>respiratory bronchiol</td>
</tr>
<tr>
<td>RBC(s)</td>
<td>red blood cell(s); erythrocyte(s)</td>
</tr>
<tr>
<td>rbcL</td>
<td>Rubisco large subunit</td>
</tr>
<tr>
<td>rbcS</td>
<td>Rubisco small subunit</td>
</tr>
<tr>
<td>R’CO acyl</td>
<td>acyl carrier protein</td>
</tr>
<tr>
<td>R’C(O)–O$_2$</td>
<td>acyl peroxy</td>
</tr>
<tr>
<td>rcd1</td>
<td>Arabidopsis mutant radical induced cell death</td>
</tr>
<tr>
<td>RCD3</td>
<td>rod-cone dysplasia 3</td>
</tr>
<tr>
<td>RCP</td>
<td>Representative Concentration Pathways</td>
</tr>
<tr>
<td>RDBMS</td>
<td>Relational Database Management Systems</td>
</tr>
<tr>
<td>Re</td>
<td>Reynolds number</td>
</tr>
<tr>
<td>REHEX</td>
<td>Regional Human Exposure Model</td>
</tr>
<tr>
<td>RER</td>
<td>rough endoplasmic reticulum; Respiratory exchange ratio</td>
</tr>
<tr>
<td>RF</td>
<td>radiative forcing</td>
</tr>
<tr>
<td>RGR</td>
<td>relative growth rate</td>
</tr>
<tr>
<td>RH</td>
<td>relative humidity</td>
</tr>
<tr>
<td>RIOPA</td>
<td>Relationship of Indoor, Outdoor, and Personal Air (study)</td>
</tr>
<tr>
<td>RL</td>
<td>total pulmonary resistance</td>
</tr>
<tr>
<td>RLVs</td>
<td>receptor-like/Pelle kinase group</td>
</tr>
<tr>
<td>RMNP</td>
<td>Rocky Mountain National Park, Colorado</td>
</tr>
<tr>
<td>RMR</td>
<td>resting metabolic rate</td>
</tr>
<tr>
<td>rMSSD</td>
<td>root mean squared differences between adjacent normal-to-normal heartbeat intervals</td>
</tr>
<tr>
<td>Rn</td>
<td>nasal resistance</td>
</tr>
<tr>
<td>RNA</td>
<td>ribonucleic acid</td>
</tr>
<tr>
<td>RO$_2$</td>
<td>organic peroxy; organic peroxy</td>
</tr>
<tr>
<td>ROG</td>
<td>reactive organic gases</td>
</tr>
<tr>
<td>ROI</td>
<td>reactive oxygen</td>
</tr>
<tr>
<td>$\text{RONO}_2$</td>
<td>organic nitrate</td>
</tr>
<tr>
<td>ROOH</td>
<td>organic peroxides</td>
</tr>
<tr>
<td>$\text{ROONO}_2, \text{RO}_2\text{NO}_2$</td>
<td>peroxy nitrate</td>
</tr>
<tr>
<td>ROS</td>
<td>reactive oxygen species</td>
</tr>
<tr>
<td>RPD</td>
<td>relative percent difference</td>
</tr>
<tr>
<td>RR</td>
<td>normal-to-normal (NN or RR) time interval between each QRS complex in the EKG; risk ratio; relative risk; respiratory rate</td>
</tr>
<tr>
<td>RRMS</td>
<td>relatively remote monitoring sites</td>
</tr>
<tr>
<td>RT</td>
<td>respiratory tract</td>
</tr>
<tr>
<td>RT</td>
<td>transepithelial resistance</td>
</tr>
<tr>
<td>RTLF</td>
<td>respiratory tract lining fluid</td>
</tr>
<tr>
<td>RuBisCO; RuBisco</td>
<td>ribulose-1,5-bisphosphate carboxylase/oxygenase</td>
</tr>
<tr>
<td>RuBP</td>
<td>ribulose bisphosphate</td>
</tr>
<tr>
<td>$\sigma$</td>
<td>sigma, standard deviation</td>
</tr>
<tr>
<td>$\sigma$g</td>
<td>(geometric standard deviation)</td>
</tr>
<tr>
<td>S</td>
<td>second</td>
</tr>
<tr>
<td>s.c.</td>
<td>Short; smoker; sulfur; South subcutaneous (route)</td>
</tr>
<tr>
<td>SA</td>
<td>salicylic acid</td>
</tr>
<tr>
<td>SAB</td>
<td>Science Advisory Board</td>
</tr>
<tr>
<td>SAC</td>
<td>Staphylococcus aureus Cowan 1 strain</td>
</tr>
<tr>
<td>SAG21</td>
<td>senescence</td>
</tr>
<tr>
<td>SAI</td>
<td>Systems Applications International</td>
</tr>
<tr>
<td>S-allele</td>
<td>short-allele</td>
</tr>
<tr>
<td>SAMD</td>
<td>S-adenosyl methionine decarboxylase</td>
</tr>
<tr>
<td>SaO$_2$</td>
<td>oxygen saturation of arterial blood</td>
</tr>
<tr>
<td>SAPALDIA</td>
<td>Study of Air Pollution and Lung Diseases in Adults</td>
</tr>
<tr>
<td>SAPRC</td>
<td>Stratospheric Processes and their Role in Climate; Statewide Air Pollution Research Center, University of California, Riverside</td>
</tr>
<tr>
<td>SAR</td>
<td>systemic acquired resistance</td>
</tr>
<tr>
<td>SAROAD</td>
<td>Storage and Retrieval of Aerometric Data (U.S. EPA centralized database; superseded by Aerometric Information Retrieval System [AIRS])</td>
</tr>
<tr>
<td>SAWgrp</td>
<td>small airway function group</td>
</tr>
<tr>
<td>SBNF</td>
<td>San Bernardino National Forest, California</td>
</tr>
<tr>
<td>SBP</td>
<td>systolic blood pressure</td>
</tr>
<tr>
<td>SBUV</td>
<td>Solar Backscatter Ultraviolet Spectrometer</td>
</tr>
<tr>
<td>SC</td>
<td>stratum corneum</td>
</tr>
<tr>
<td>Sc</td>
<td>scandium</td>
</tr>
</tbody>
</table>
SCAQs: Southern California Air Quality Study
SCE(s): sister chromatid exchange(s)
SD: standard deviation; Sprague-Dawley rat
SDNN: standard deviation normal-to-normal (NN or RR) time interval between each QRS complex in the EKG
SE: standard error
SEBAS: Social Environment and Biomarkers of Aging Study
sec: second
Sess.: session
SEM: simultaneously extracted metal; scanning electron microscopy
SENP: Sequoia National Park, California
SES: socioeconomic status
SF: San Francisco Bay Area
SF6: sulfur hexafluoride (tracer gas)
SGA: small for gestational age
sRaw: specific airway conductance
SH: Shenandoah National Park site
SHEDS: Stochastic Human Exposure and Dose Simulation
SHEN: Shenandoah National Park
sICAM-1: soluble intercellular adhesion molecule
SIDS: sudden infant death syndrome
SIGMOID: sigmoid weighted summed concentration
SINIC: Simple Nitrogen Cycle model
SIP: State Implementation Plan
SIPK: salicylic acid (SA) induced protein kinase
SK: shikimate kinase
SLA: specific leaf area
SLAC1: (protein) slow anion channel associated 1
SLAMS: State and Local Air Monitoring Stations
SM: smooth muscle
SMD: soil moisture deficit
SME: soybean oil methyl ester
SMNP: Great Smoky Mountain National Park (North Carolina and Tennessee)
SMOE: Spare-Matrix Operator Kernel Emissions
SN: normalized slope of the alveolar plateau
SNAQAS: Secondary National Ambient Air Quality Standards
SNP(s): single-nucleotide polymorphism
SO2: sulfur dioxide
SO42−: sulfate
SOC: soil organic carbon
SOCS: Salmeterol Off Corticosteroids Study
SOD: superoxide dismutase
SOS: Southern Oxidant Study
SOx: sulfur oxides
SoyFACE: Soybean Free Air gas Concentration Enrichment (Facility)
SP: surfactant protein (e.g., SPA, SPD); substance P
SP-A: surfactant protein-A
SPF: specific pathogen free
SPMs: special purpose monitors
SP-NK: substance P – neurokinin receptor complex
sRaw, specific airway resistance
SRBC: sheep red blood cell
SRES: Special Report on Emissions Scenarios
SRM: standard reference method
SRP: standard reference photometers
SSCP: single-strand conformation polymorphism
129S1/SvlmJ: mouse strain
STE: stratosphere-troposphere exchange
STEP: Stratospheric-Tropospheric-exchange Project
STN: speciation trends network
sTNFR1: soluble tumor necrosis factor receptor 1
STP: standard temperature and pressure
STPD: standard temperature and pressure, dry
STRF: Spatio-Temporal Random Field (theory)
subscript i: Index of indoor microenvironments
subscript o: Index of outdoor microenvironments
subscript o,i: Index of outdoor microenvironments adjacent to a given indoor microenvironment i
SUM00: sum of all hourly average concentrations
SUM06: seasonal sum of all hourly average concentrations ≥ 0.06 ppm
SUM07: seasonal sum of all hourly average concentrations ≥ 0.07 ppm
SUM08: seasonal sum of all hourly average concentrations ≥ 0.08 ppm
SURE: Sulfate Regional Experiment Program
SVE: supraventricular ectopy
S-W: square-wave
SWS: slow wave sleep
SZA: solar zenith angle
τ: tau, photochemical lifetime; atmospheric lifetime
<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Definition</th>
</tr>
</thead>
<tbody>
<tr>
<td>t</td>
<td>t-test statistical value; t statistic</td>
</tr>
<tr>
<td>T</td>
<td>time; duration of exposure</td>
</tr>
<tr>
<td>T-cell(s)</td>
<td>T lymphocyte(s), thymus-dependent lymphocytes</td>
</tr>
<tr>
<td>T1</td>
<td>first trimester</td>
</tr>
<tr>
<td>T2</td>
<td>second trimester</td>
</tr>
<tr>
<td>T3</td>
<td>triiodothyronine</td>
</tr>
<tr>
<td>T3</td>
<td>third trimester</td>
</tr>
<tr>
<td>T4</td>
<td>thyroxine</td>
</tr>
<tr>
<td>TAR</td>
<td>IPCC Third Assessment Report</td>
</tr>
<tr>
<td>TAR WGI</td>
<td>IPCC Third Assessment Report of Working Group I</td>
</tr>
<tr>
<td>TB</td>
<td>tracheobronchial; terminal bronchioles; tuberculosis</td>
</tr>
<tr>
<td>TBA</td>
<td>thiobarbituric acid</td>
</tr>
<tr>
<td>TBARS</td>
<td>thiobarbituric acid reactive substances</td>
</tr>
<tr>
<td>TC</td>
<td>total carbon</td>
</tr>
<tr>
<td>Tco</td>
<td>Technetium-99m</td>
</tr>
<tr>
<td>T-cells</td>
<td>T-lymphocytes, Thymus-derived lymphocytes</td>
</tr>
<tr>
<td>TC-DTPA</td>
<td>99mTc-dithierylenetriaminepenta-acetic acid</td>
</tr>
<tr>
<td>Tco</td>
<td>core temperature</td>
</tr>
<tr>
<td>DTLAS</td>
<td>Tunable Diode Laser Absorption Spectrometer</td>
</tr>
<tr>
<td>Te</td>
<td>expiratory time</td>
</tr>
<tr>
<td>TEM</td>
<td>transmission electron microscopy; Terrestrial Ecosystem Model</td>
</tr>
<tr>
<td>TES</td>
<td>Tropospheric Emission Spectrometer</td>
</tr>
<tr>
<td>TexAQS</td>
<td>Texas Air Quality Field Study</td>
</tr>
<tr>
<td>Tg</td>
<td>teragram(s)</td>
</tr>
<tr>
<td>TGF</td>
<td>transforming growth factor</td>
</tr>
<tr>
<td>TGF (\beta)</td>
<td>transforming growth factor beta</td>
</tr>
<tr>
<td>Th</td>
<td>T helper cell type</td>
</tr>
<tr>
<td>Th2</td>
<td>T helper cell type 2</td>
</tr>
<tr>
<td>THC</td>
<td>Total hydrocarbon content</td>
</tr>
<tr>
<td>tHcy</td>
<td>total homocysteine</td>
</tr>
<tr>
<td>Ti</td>
<td>titanium</td>
</tr>
<tr>
<td>TIA</td>
<td>transient ischemic attack</td>
</tr>
<tr>
<td>TIMP-2</td>
<td>tissue inhibitor of matrix metalloprotease-2</td>
</tr>
<tr>
<td>TiO(_2)</td>
<td>titanium dioxide</td>
</tr>
<tr>
<td>TLC</td>
<td>total lung capacity</td>
</tr>
<tr>
<td>TNLISE</td>
<td>two-level normal independent sampling estimation</td>
</tr>
<tr>
<td>Tr</td>
<td>Toll-like receptor gene</td>
</tr>
<tr>
<td>TLR</td>
<td>Toll-like receptor protein (ex., TLR2, TLR4)</td>
</tr>
<tr>
<td>TMPO</td>
<td>tetramethylphroline 1-oxide</td>
</tr>
<tr>
<td>TNC</td>
<td>total nonstructural carbohydrate</td>
</tr>
<tr>
<td>TNF</td>
<td>tumor necrosis factor (e.g., TNF-(\alpha))</td>
</tr>
<tr>
<td>TNF-308</td>
<td>tumor necrosis factor genotype</td>
</tr>
<tr>
<td>TNF-(\alpha)</td>
<td>tumor necrosis factor alpha</td>
</tr>
<tr>
<td>TNFR</td>
<td>tumor necrosis factor receptor</td>
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</table>

<table>
<thead>
<tr>
<th>Acronym</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>TOMS</td>
<td>Total Ozone Mapping/Monitoring Satellite; total ozone mapping spectrometer</td>
</tr>
<tr>
<td>TOPSE</td>
<td>Tropospheric Ozone Production About the Spring Equinox</td>
</tr>
<tr>
<td>tPA</td>
<td>tissue plasminogen activator</td>
</tr>
<tr>
<td>TPLIF</td>
<td>two-photon laser-induced fluorescence</td>
</tr>
<tr>
<td>TRAMP</td>
<td>TexAQS-II Radical and Aerosol Measurement Project</td>
</tr>
<tr>
<td>TREGRO</td>
<td>Tree Growth Model</td>
</tr>
<tr>
<td>TRIFFID</td>
<td>Top-down Representation of Interactive Foliage and Flora Including Dynamics</td>
</tr>
<tr>
<td>TRIM</td>
<td>Total Risk Integrated Methodology (model)</td>
</tr>
<tr>
<td>TRIM.Expo</td>
<td>Total Risk Integrated Methodology Exposure Event (model)</td>
</tr>
<tr>
<td>TRP</td>
<td>transient receptor potential (ion channel[s], ex., TRP-A1, TRP-V1, TRP-M8)</td>
</tr>
<tr>
<td>TSH</td>
<td>thyroid stimulating hormone</td>
</tr>
<tr>
<td>TSP</td>
<td>total suspended particles</td>
</tr>
<tr>
<td>TTFMS</td>
<td>two-tone frequency-modulated spectroscopy</td>
</tr>
<tr>
<td>TWA</td>
<td>time-weighted average</td>
</tr>
<tr>
<td>TX</td>
<td>thromboxane (e.g., TXB(_2))</td>
</tr>
<tr>
<td>TXB(_2)</td>
<td>thromboxane B2</td>
</tr>
<tr>
<td>UA</td>
<td>uric acid; urate</td>
</tr>
<tr>
<td>UAM</td>
<td>Urban Airshed Model</td>
</tr>
<tr>
<td>UCL</td>
<td>upper 95th% confidence limit</td>
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<tr>
<td>UDGT</td>
<td>UDP-galactose-1,2-diacylglycerol galactosyltransferase</td>
</tr>
<tr>
<td>UDP</td>
<td>uridine diphosphate</td>
</tr>
<tr>
<td>U.K.</td>
<td>United Kingdom</td>
</tr>
<tr>
<td>UNECE</td>
<td>United Nations Economic Commission for Europe</td>
</tr>
<tr>
<td>UNEP</td>
<td>United Nations Environmental Programme</td>
</tr>
<tr>
<td>UNFCCC</td>
<td>United Nations Framework Convention on Climate Change</td>
</tr>
<tr>
<td>U-O</td>
<td>epoxides formed from uric acid</td>
</tr>
<tr>
<td>U-O(_2)</td>
<td>peroxides formed from uric acid</td>
</tr>
<tr>
<td>U-O(_3)</td>
<td>ozonides formed from uric acid</td>
</tr>
<tr>
<td>URI</td>
<td>upper respiratory infection</td>
</tr>
<tr>
<td>URS</td>
<td>upper respiratory symptoms</td>
</tr>
<tr>
<td>URT</td>
<td>upper respiratory tract; upper airways</td>
</tr>
<tr>
<td>U.S.</td>
<td>United States (of America)</td>
</tr>
<tr>
<td>USC; U.S.C.</td>
<td>U.S. Code</td>
</tr>
<tr>
<td>USDA</td>
<td>U.S. Department of Agriculture</td>
</tr>
<tr>
<td>USFS</td>
<td>U.S. Forest Service</td>
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<tr>
<td>USGCRP</td>
<td>U.S. Global Change Research Program</td>
</tr>
<tr>
<td>USGS</td>
<td>U.S. Geological Survey</td>
</tr>
<tr>
<td>UV</td>
<td>ultraviolet radiation</td>
</tr>
<tr>
<td>UV-A</td>
<td>ultraviolet radiation at wavelengths of 320 to 400 nm</td>
</tr>
</tbody>
</table>
UV-B  ultraviolet radiation at wavelengths of 280 to 320 nm
UV-C  ultraviolet radiation at wavelengths of 200 to 280 nm
UV-DIAL  Ultraviolet Differential Absorption Lidar
V  vanadium
V, mV, μV  volt, millivolt, microvolt
VA  alveolar ventilation
Val  valine
VC  vital capacity
VCAM  vascular cell adhesion molecule
V_d  deposition rate, deposition velocity (cm/sec)
V_D  volume of the anatomic or physiological dead space
V_e  ventilation rate; minute ventilation; ventilatory volume
VEGF  vascular endothelial growth factor
\( \dot{V}_{\text{E}} \)  maximum minute ventilation
Vmax  maximum velocity
Vmax25%  maximum expiratory flow at 25% of the vital capacity
Vmax50%  maximum expiratory flow at 50% of the vital capacity
Vmax75%  maximum expiratory flow at 75% of the vital capacity
VMD  volume median diameter
Vn  nasal volume
VO2  oxygen consumption
VO_{2\text{max}}  maximum volume per time, of oxygen (maximal oxygen consumption, maximal oxygen uptake or aerobic capacity)
VOC(s)  volatile organic compound(s)
VP  volumetric penetration
VP_{50\%}  volume at which 50% of an inhaled bolus is absorbed
VPD  vapor pressure deficit; Vehicles per day; Ventricular premature depolarization
VT  tidal volume
VTB  terminal bronchiole region volume
VT_{\text{max}}  maximum tidal volume
VUA  volume of the upper airways
vWF  von Willebrand factor
W  width; wilderness; week(s)
W126  cumulative integrated exposure index with a sigmoidal weighting function
W95  cumulative integrated exposure index with a sigmoidal weighting function
WBC  white blood cell
WBG \text{T}  wet bulb globe temperature
wc  sigmoidal weighting of hourly \( O_3 \) concentration
WCB  warm conveyor belt

WED  (U.S. EPA NHEERL) Western Ecology Division
WF, WFM  White Face Mountain site
WHI  Women’s Health Initiative
WHO  World Health Organization
W/m², W m⁻²  watts per square meter
WMO  World Meteorological Organization
WMO/UNEP  World Meteorological Organization/United Nations Environment Program
WRF  Weather Research and Forecasting model
Ws  Wassilewskija Arabidopsis ecotype
WS  wood smoke
WT  wild type; White Top Mountain site
wt %  percent by weight
WUS  western U.S.
w/v  weight per volume
Y  three parameter Weibull model
yr  year
Z  Airway generation
ZAPS  Zonal Air Pollution System
ZELIG  a forest succession simulation model
Zn  zinc
PREAMBLE

Process of ISA Development

This preamble outlines the general process for developing an Integrated Science Assessment (ISA) including the framework for evaluating weight of evidence and drawing scientific conclusions and causal judgments. The ISA provides a concise review, synthesis, and evaluation of the most policy-relevant science to serve as a scientific foundation for the review of the National Ambient Air Quality Standards (NAAQS). The general process for NAAQS reviews is described at http://www.epa.gov/ttn/naaqs/review.html. Figure I depicts the general NAAQS review process and information for individual NAAQS reviews is available at www.epa.gov/ttn/naaqs. This preamble is a general discussion of the basic steps and criteria used in developing an ISA; for each ISA, specific details and considerations are included in the introductory section for that assessment.

The fundamental process for developing an ISA includes:

- literature searches;
- study selection;
- evaluation and integration of the evidence;
- development of scientific conclusions and causal judgments.

An initial step in this process is publication of a call for information in the Federal Register that invites the public to provide information relevant to the assessment, such as new or recent publications on health or welfare effects of the pollutant, or from atmospheric and exposure sciences fields. EPA maintains an ongoing literature search process for identification of relevant scientific studies published since the last review of the NAAQS. Search strategies are designed for pollutants and scientific disciplines and iteratively modified to optimize identification of pertinent publications. Papers are identified for inclusion in several additional ways: specialized searches on specific topics; independent review of tables of contents for journals in which relevant papers may be published; independent identification of relevant literature by expert scientists; review of citations in previous assessments and identification by the public and the Clean Air Scientific Advisory Committee (CASAC) during the external review process. This literature search and study selection process is depicted in Figure II. Publications considered for inclusion in the ISA are added to the Health and Environmental Research Online (HERO) database developed by EPA (http://hero.epa.gov/); the references in the ISA include a hyperlink to the database.

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1 Welfare effects as defined in Clean Air Act (CAA) section 302(h) [42 U.S.C. 7602(h)] include, but are not limited to, "effects on soils, water, crops, vegetation, man-made materials, animals, wildlife, weather, visibility and climate, damage to and deterioration of property, and hazards to transportation, as well as effects on economic values and on personal comfort and well-being."
Studies that have undergone scientific peer review and have been published or accepted for publication and reports that have undergone review are considered for inclusion in the ISA. Analyses conducted by EPA using publicly available data are also considered for inclusion in the ISA. All relevant epidemiologic, controlled human exposure, toxicological, and ecological and welfare effects studies published since the last review are considered, including those related to exposure-response relationships, mode(s) of action (MOA), and potentially at-risk populations and lifestages. Studies on atmospheric chemistry, environmental fate and transport, dosimetry, toxicokinetics and exposure are also considered for inclusion in the document, as well as analyses of air quality and emissions data. References that were considered for inclusion in a specific ISA can be found using the HERO website (http://hero.epa.gov).

Figure I Illustration of the key steps in the process of the review of National Ambient Air Quality Standards.
Criteria for study evaluation include:

- Are the study populations, subjects, or animal models adequately selected, and are they sufficiently well defined to allow for meaningful comparisons between study or exposure groups?
- Are the statistical analyses appropriate, properly performed, and properly interpreted? Are likely covariates adequately controlled or taken into account in the study design and statistical analysis?
- Are the air quality data, exposure, or dose metrics of adequate quality and sufficiently representative of information regarding ambient conditions?
- Are the health, ecological or welfare effect measurements meaningful, valid and reliable?
- Do the analytical methods provide adequate sensitivity and precision to support conclusions?

Figure II  Illustration of processes for literature search and study selection used for development of ISAs.

Each ISA builds upon the conclusions of previous assessments for the pollutant under review. EPA focuses on peer reviewed literature published following the completion of the previous review (2006 O₃ AQCD) and on any new interpretations of previous literature, integrating the results of recent scientific studies with previous findings. Important earlier studies may be discussed in detail to reinforce key concepts and conclusions or for reinterpretation in light of newer data. Earlier studies also are the primary focus in some areas of the document where research efforts have subsided, or if these earlier studies remain the definitive works available in the literature.

Selection of studies for inclusion in the ISA is based on the general scientific quality of the study, and consideration of the extent to which the study is informative and policy-relevant. Policy relevant and informative studies include those that provide a basis for or describe the relationship between the criteria pollutant and effects, including studies that offer innovation in method or design and studies that reduce uncertainty on critical issues, such as analyses of confounding or effect modification by copollutants or other variables, analyses of concentration-response or dose-
response relationships, or analyses related to time between exposure and response. Emphasis is placed on studies that examine effects associated with pollutant concentrations relevant to current population and ecosystem exposures, and particularly those pertaining to concentrations currently found in ambient air. Other studies are included if they contain unique data, such as a previously unreported effect or MOA for an observed effect, or examine multiple concentrations to elucidate exposure-response relationships. In general, in assessing the scientific quality and relevance of health and welfare effects studies, the following considerations have been taken into account when selecting studies for inclusion in the ISA.

- Are the study populations, subjects, or animal models adequately selected, and are they sufficiently well defined to allow for meaningful comparisons between study or exposure groups?
- Are the statistical analyses appropriate, properly performed, and properly interpreted? Are likely covariates adequately controlled or taken into account in the study design and statistical analysis?
- Are the air quality data, exposure, or dose metrics of adequate quality and sufficiently representative of information regarding ambient conditions?
- Are the health, ecological or welfare effect measurements meaningful, valid and reliable?
- Do the analytical methods provide adequate sensitivity and precision to support conclusions?

Considerations specific to particular disciplines include the following: In selecting epidemiologic studies, EPA considers whether a given study: (1) presents information on associations with short- or long-term pollutant exposures at or near conditions relevant to ambient exposures; (2) addresses potential confounding by other pollutants; (3) assesses potential effect modifiers; (4) evaluates health endpoints and populations not previously extensively researched; and (5) evaluates important methodological issues related to interpretation of the health evidence (e.g., lag or time period between exposure and effects, model specifications, thresholds, mortality displacement).

Considerations for the selection of research evaluating controlled human exposure or animal toxicological studies include a focus on studies conducted using relevant pollutant exposures. For both types of studies, relevant pollutant exposures are considered to be those generally within one or two orders of magnitude of ambient concentrations. Studies in which higher doses were used may also be considered if they provide information relevant to understanding MOA or mechanisms, as noted below.

Evaluation of controlled human exposure studies focuses on those that approximated expected human exposure conditions in terms of concentration and duration. Studies should include control exposures to filtered air, as appropriate. In the selection of controlled human exposure studies, emphasis is placed on studies that: (1) investigate potentially at-risk populations and lifestages such as people with asthma or
cardiovascular diseases, children or older adults; (2) address issues such as concentration-response or time-course of responses; and (3) have sufficient statistical power to assess findings.

Review of the animal toxicological evidence focuses on studies that approximate expected human dose conditions, which vary depending on the dosimetry, toxicokinetics, and biological sensitivity of the particular laboratory animal species or strains studied. Emphasis is placed on studies that: (1) investigate animal models of disease that can provide information on populations potentially at increased risk of effects; (2) address issues such as concentration-response or time-course of responses; and (3) have sufficient statistical power to assess findings. Due to resource constraints on exposure duration and numbers of animals tested, animal studies typically utilize high-concentration exposures to acquire data relating to mechanisms and assure a measurable response. Emphasis is placed on studies using doses or concentrations generally within 1-2 orders of magnitude of current levels. Studies with higher concentration exposures or doses are considered to the extent that they provide useful information to inform understanding of interspecies differences between healthy and at-risk human populations. Results from in vitro studies may also be included if they provide mechanistic insight or further support for results demonstrated in vivo.

These criteria provide benchmarks for evaluating various studies and for focusing on the policy-relevant studies in assessing the body of health, ecological and welfare effects evidence. As stated initially, the intent of the ISA is to provide a concise review, synthesis, and evaluation of the most policy-relevant science to serve as a scientific foundation for the review of the NAAQS, not extensive summaries of all health, ecological and welfare effects studies for a pollutant. Of most relevance for inclusion of studies is whether they provide useful qualitative or quantitative information on exposure-effect or exposure-response relationships for effects associated with pollutant exposures at doses or concentrations relevant to ambient conditions that can inform decisions on whether to retain or revise the standards.

The general process for ISA development is illustrated in Figure III. In developing an ISA, EPA reviews and summarizes the evidence from studies of atmospheric sciences; human exposure, toxicological, controlled human exposure and epidemiologic studies; and studies of ecological and welfare effects. In the process of developing the first draft ISA, EPA may convene a peer input meeting in which EPA the scientific content of preliminary draft materials is reviewed to ensure that the ISA is up to date and focused on the most policy-relevant findings, and to assist EPA with integration of evidence within and across disciplines. EPA integrates the evidence from across scientific disciplines or study types and characterizes the weight of evidence for relationships between the pollutant and various outcomes. The integration of evidence on health, and ecological or welfare effects, involves collaboration between scientists from various disciplines. As an example, an evaluation of health effects evidence would include the integration of the results from epidemiologic, controlled human exposure, and toxicological studies, and application of the causal framework (described below) to draw conclusions. Integration of results
on health or ecological effects that are logically or mechanistically connected (e.g., a spectrum of effects on the respiratory system) informs judgments of causality. Using the causal framework described in the following section, EPA scientists consider aspects such as strength, consistency, coherence, and biological plausibility of the evidence, and develop causality determinations on the nature of the relationships. Causality determinations often entail an iterative process of review and evaluation of the evidence. Two drafts of the ISA are typically released for review by the CASAC and the public, and comments received on the characterization of the science as well as the implementation of the causal framework are carefully considered in revising and completing the final ISA.
Figure III  Characterization of the general process of ISA development.
EPA Framework for Causal Determination

EPA has developed a consistent and transparent basis for integration of scientific evidence and evaluation of the causal nature of air pollution-related health or welfare effects for use in developing ISAs. The framework described below establishes uniform language concerning causality and brings more specificity to the findings. This standardized language was drawn from sources across the federal government and wider scientific community, especially the National Academy of Sciences (NAS) Institute of Medicine (IOM) document, *Improving the Presumptive Disability Decision-Making Process for Veterans* (Samet and Bodurow, 2008), a comprehensive report on evaluating causality. This framework:

- describes the kinds of scientific evidence used in establishing a general causal relationship between exposure and health effects;
- characterizes the process for integration and evaluation of evidence necessary to reach a conclusion about the existence of a causal relationship;
- identifies issues and approaches related to uncertainty;
- provides a framework for classifying and characterizing the weight of evidence in support of a general causal relationship.

Approaches to assessing the separate and combined lines of evidence (e.g., epidemiologic, controlled human exposure, and animal toxicological studies) have been formulated by a number of regulatory and science agencies, including the IOM of the NAS (Samet and Bodurow, 2008), International Agency for Research on Cancer (IARC, 2006), U.S. EPA (2005), and Centers for Disease Control and Prevention (CDC, 2004). Causal inference criteria have also been described for ecological effects evidence (U.S. EPA, 1998; Fox, 1991). These formalized approaches offer guidance for assessing causality. The frameworks are similar in nature, although adapted to different purposes, and have proven effective in providing a uniform structure and language for causal determinations.

Evaluating Evidence for Inferring Causation

The 1964 Surgeon General’s (U.S. Department of Health, Education and Welfare [HEW]) report on tobacco smoking defined “cause” as a “significant, effectual relationship between an agent and an associated disorder or disease in the host” (HEW, 1964). More generally, a cause is defined as an agent that brings about an effect or a result. An association is the statistical relationship among variables; alone, however, it is insufficient proof of a causal relationship between an exposure and a health outcome. Unlike an association, a causal claim supports the creation of counterfactual claims, that is, a claim about what the world would have been like under different or changed circumstances (Samet and Bodurow, 2008).

Many of the health and environmental outcomes reported in these studies have complex etiologies. Diseases such as asthma, coronary heart disease (CHD) or cancer...
are typically initiated by multiple agents. Outcomes depend on a variety of factors, such as age, genetic susceptibility, nutritional status, immune competence, and social factors (Samet and Bodurow, 2008; Gee and Payne-Sturges, 2004). Effects on ecosystems are often also multifactorial with a complex web of causation. Further, exposure to a combination of agents could cause synergistic or antagonistic effects. Thus, the observed risk may represent the net effect of many actions and counteractions.

Scientific findings incorporate uncertainty. “Uncertainty” can be defined as having limited knowledge to exactly describe an existing state or future outcome, e.g., the lack of knowledge about the correct value for a specific measure or estimate. Uncertainty analysis may be qualitative or quantitative in nature. In many cases, the analysis is qualitative, and can include professional judgment or inferences based on analogy with similar situations. Quantitative uncertainty analysis may include use of simple measures (e.g., ranges) and analytical techniques. Quantitative uncertainty analysis might progress to more complex measures and techniques, if needed for decision support. Various approaches to evaluating uncertainty include classical statistical methods, sensitivity analysis, or probabilistic uncertainty analysis, in order of increasing complexity and data requirements. However, data may not be available for all aspects of an assessment and those data that are available may be of questionable or unknown quality. Ultimately, the assessment is based on a number of assumptions with varying degrees of uncertainty.

Publication bias is a source of uncertainty regarding the magnitude of health risk estimates. It is well understood that studies reporting non-null findings are more likely to be published than reports of null findings. Publication bias can result in overestimation of effect estimate sizes (Ioannidis, 2008). For example, effect estimates from single-city epidemiologic studies have been found to be generally larger than those from multiplicity studies which is an indication of publication bias in that null or negative single-city results may be reported in a multiplicity analyses but might not be published independently (Bell et al., 2005).

**Consideration of Evidence from Scientific Disciplines**

Moving from association to causation involves the elimination of alternative explanations for the association. The ISA focuses on evaluation of the findings from the body of evidence, drawing upon the results of all studies determined to meet the criteria described previously. Causality determinations are based on the evaluation, integration, and synthesis of evidence from across scientific disciplines. The relative importance of different types of evidence varies by pollutant or assessment, as does the availability of different types of evidence for causality determination. Three general types of studies inform consideration of human health effects: controlled human exposure, epidemiologic, and toxicological studies. Evidence on ecological or welfare effects may be drawn from a variety of experimental approaches (e.g.,
Direct evidence of a relationship between pollutant exposures and human health effects comes from controlled human exposure studies. Such studies experimentally evaluate the health effects of administered exposures in human volunteers under highly controlled laboratory conditions. Also referred to as human clinical studies, these experiments allow investigators to expose subjects to known concentrations of air pollutants under carefully regulated environmental conditions and activity levels. These studies provide important information on the biological plausibility of associations observed in epidemiologic studies. Essential dose-response profiles and ranges of response severity can be established with these studies. In some instances, controlled human exposure studies can also be used to characterize concentration-response relationships at pollutant concentrations relevant to ambient conditions. Controlled human exposures are typically conducted using a randomized crossover design, with subjects exposed both to the pollutant and a clean air control. In this way, subjects serve as their own controls, effectively controlling for many potential confounders. Considerations for evaluating controlled human study findings include the generally small sample size and short exposure time used in experimental studies, and that severe health outcomes are not assessed. By experimental design, controlled human exposure studies are structured to evaluate physiological or biomolecular outcomes in response to exposure to a specific air pollutant and/or combination of pollutants. In addition, the study design generally precludes inclusion of subjects with serious health conditions, and therefore the results often cannot be generalized to an entire population. Although some controlled human exposure studies have included health-compromised individuals such as those with respiratory or cardiovascular disease, these individuals may also be relatively healthy and may not represent the most sensitive individuals in the population. Thus, observed effects in these studies may underestimate the response in certain populations.

Epidemiologic studies provide important information on the associations between health effects and exposure of human populations to ambient air pollution. In epidemiologic or observational studies of humans, the investigator generally does not control exposures or intervene with the study population. Broadly, observational studies can describe associations between exposures and effects. These studies fall into several categories: e.g., cross-sectional, prospective cohort, panel, and time-series studies. Cross-sectional studies use health outcome, exposure and covariate data available at the community level (e.g., annual mortality rates and pollutant concentrations), but do not have individual-level data. Prospective cohort studies have some data collected at the individual level, generally health outcome data, and in some cases individual-level data on exposure and covariates are collected. Time-series studies evaluate the relationship for changes in a health outcome with changes in exposure indicators, such as an association between daily changes in mortality with air pollution. Panel studies include repeated measurements of health outcomes, such as respiratory symptoms or heart rhythm variable, at the individual level. “Natural experiments” offer the opportunity to investigate changes in health related to a change in exposure, such as closure of a pollution source.
In evaluating epidemiologic studies, consideration of many study design factors and issues must be taken into account to properly inform their interpretation. One key consideration is evaluation of the potential contribution of the pollutant to a health outcome when it is a component of a complex air pollutant mixture. Reported effect estimates in epidemiologic studies may reflect: independent effects on health outcomes; effects of the pollutant acting as an indicator of a copollutant or a complex ambient air pollution mixture; effects resulting from interactions between that pollutant and copollutants.

In the evaluation of epidemiologic evidence, one important consideration is potential confounding. Confounding is “… a confusion of effects. Specifically, the apparent effect of the exposure of interest is distorted because the effect of an extraneous factor is mistaken for or mixed with the actual exposure effect (which may be null)” (Rothman and Greenland, 1998). One approach to remove spurious associations (due to possible confounders); is to control for characteristics that may differ between exposed and unexposed persons; this is frequently termed “adjustment.” Scientific judgment is needed to evaluate likely sources and extent of confounding, together with consideration of how well the existing constellation of study designs, results, and analyses address the potential for erroneous inferences. A confounder is associated with both the exposure and the effect; for example, confounding can occur between correlated pollutants that are associated with the same effect.

Several statistical methods are available to detect and control for potential confounders, with none of them being completely satisfactory. Multivariable regression models constitute one tool for estimating the association between exposure and outcome after adjusting for characteristics of participants that might confound the results. The use of multipollutant regression models has been the prevailing approach for controlling potential confounding by copollutants in air pollution health effects studies. Finding the likely causal pollutant from multipollutant regression models is made difficult by the possibility that one or more air pollutants may be acting as a surrogate for an unmeasured or poorly measured pollutant or for a particular mixture of pollutants. In addition, pollutants may independently exert effects on the same system; for example, several pollutants may be associated with respiratory effects through either the same or different modes of action. The number and degree of diversity of covariates, as well as their relevance to the potential confounders, remain matters of scientific judgment. Despite these limitations, the use of multipollutant models is still the prevailing approach employed in most air pollution epidemiologic studies and provides some insight into the potential for confounding or interaction among pollutants.

Confidence that unmeasured confounders are not producing the findings is increased when multiple studies are conducted in various settings using different subjects or exposures, each of which might eliminate another source of confounding from consideration. For example, multicity studies can provide insight on potential confounding through the use of a consistent method to analyze data from across locations with different levels of copollutants and other covariates. Intervention
studies, because of their quasi-experimental nature, can be particularly useful in characterizing causation.

Another important consideration in the evaluation of epidemiologic evidence is effect modification, which occurs when the effect differs between subgroups or strata; for example, effect estimates that vary by age group or potential risk factor. As stated by Rothman and Greenland (1998):

“Effect-measure modification differs from confounding in several ways. The main difference is that, whereas confounding is a bias that the investigator hopes to prevent or remove from the effect estimate, effect-measure modification is a property of the effect under study … In epidemiologic analysis one tries to eliminate confounding but one tries to detect and estimate effect-measure modification.”

When a risk factor is a confounder, it is the true cause of the association observed between the exposure and the outcome; when a risk factor is an effect modifier, it changes the magnitude of the association between the exposure and the outcome in stratified analyses. For example, the presence of a pre-existing disease or indicator of low socioeconomic status may act as effect modifiers if they are associated with increased risk of effects related to air pollution exposure. It is often possible to stratify the relationship between health outcome and exposure by one or more of these potential effect modifiers. For variables that modify the association, effect estimates in each stratum will be different from one another and different from the overall estimate, indicating a different exposure-response relationship may exist in populations represented by these variables.

Exposure measurement error, which refers to the uncertainty associated with the exposure metrics used to represent exposure of an individual or population, can be an important contributor to uncertainty in air pollution epidemiologic study results. Exposure error can influence observed epidemiologic associations between ambient pollutant concentrations and health outcomes by biasing effect estimates toward or away from the null and widening confidence intervals around those estimates (Zeger et al., 2000). There are several components that contribute to exposure measurement error in air pollution epidemiologic studies, including the difference between true and measured ambient concentrations, the difference between average personal exposure to ambient pollutants and ambient concentrations at central monitoring sites, and the use of average population exposure rather than individual exposure estimates. Factors that could influence exposure estimates include nonambient sources of exposure, topography of the natural and built environment, meteorology, measurement errors, time-location-activity patterns, and the extent to which ambient pollutants penetrate indoor environments. The importance of exposure error varies with study design and is dependent on the spatial and temporal aspects of the design.

The third main type of health effects evidence, animal toxicological studies, provides information on the pollutant’s biological action under controlled and monitored exposure circumstances. Taking into account physiological differences of the experimental species from humans, these studies inform characterization of health
effects of concern, exposure-response relationships and MOAs. Further, animal models can inform determinations of at-risk populations. These studies evaluate the effects of exposures to a variety of pollutants in a highly controlled laboratory setting and allow exploration of toxicological pathways or mechanisms by which a pollutant may cause effects. Understanding the biological mechanisms underlying various health outcomes can prove crucial in establishing or negating causality. In the absence of human studies data, extensive, well-conducted animal toxicological studies can support determinations of causality, if the evidence base indicates that similar responses are expected in humans under ambient exposure conditions.

Interpretations of animal toxicological studies are affected by limitations associated with extrapolation between animal and human responses. The differences between humans and other species have to be taken into consideration, including metabolism, hormonal regulation, breathing pattern, and differences in lung structure and anatomy. Also, in spite of a high degree of homology and the existence of a high percentage of orthologous genes across humans and rodents (particularly mice), extrapolation of molecular alterations at the gene level is complicated by species-specific differences in transcriptional regulation. Given these differences, there are uncertainties associated with quantitative extrapolations of observed pollutant-induced pathophysiological alterations between laboratory animals and humans, as those alterations are under the control of widely varying biochemical, endocrine, and neuronal factors.

For ecological effects assessment, both laboratory and field studies (including field experiments and observational studies) can provide useful data for causality determination. Because conditions can be controlled in laboratory studies, responses may be less variable and smaller differences may be easier to detect. However, the control conditions may limit the range of responses (e.g., animals may not be able to seek alternative food sources) or incompletely reflect pollutant bioavailability, so they may not reflect responses that would occur in the natural environment. In addition, larger-scale processes are difficult to reproduce in the laboratory.

Field observational studies measure biological changes in uncontrolled situations, and describe an association between a disturbance and an ecological effect. Field data can provide important information for assessments of multiple stressors or where site-specific factors significantly influence exposure. They are also often useful for analyses of larger geographic scales and higher levels of biological organization. However, because conditions are not controlled, variability is expected to be higher and differences harder to detect. Field surveys are most useful for linking stressors with effects when stressor and effect levels are measured concurrently. The presence of confounding factors can make it difficult to attribute observed effects to specific stressors.

Intermediate between laboratory and field are studies that use environmental media collected from the field to examine response in the laboratory, and experiments that are performed in the natural environment while controlling for some environmental conditions (i.e., mesocosm studies). This type of study in manipulated natural environments can be considered a hybrid between a field experiment and laboratory
study since some aspects are performed under controlled conditions but others are not. They make it possible to observe community and/or ecosystem dynamics, and provide strong evidence for causality when combined with findings of studies that have been made under more controlled conditions.

**Application of Framework for Causal Determination**

In its evaluation and integration of the scientific evidence on health or welfare effects of criteria pollutants, EPA determines the weight of evidence in support of causation and characterizes the strength of any resulting causal classification. EPA also evaluates the quantitative evidence and draws scientific conclusions, to the extent possible, regarding the concentration-response relationships and the loads to ecosystems, exposures, doses or concentrations, exposure duration, and pattern of exposures at which effects are observed.

To aid judgment, various “aspects” of causality have been discussed by many philosophers and scientists. The 1964 Surgeon General’s report on tobacco smoking discussed criteria for the evaluation of epidemiologic studies, focusing on consistency, strength, specificity, temporal relationship, and coherence (HEW, 1964). Sir Austin Bradford Hill (Hill, 1965) articulated aspects of causality in epidemiology and public health that have been widely used (Samet and Bodurow, 2008; IARC, 2006; U.S. EPA, 2005; CDC, 2004). These aspects (Hill, 1965) have been modified (Table I) for use in causal determinations specific to health and welfare effects for pollutant exposures (U.S. EPA, 2009d). Although these aspects provide a framework for assessing the evidence, they do not lend themselves to being considered in terms of simple formulas or fixed rules of evidence leading to conclusions about causality (Hill, 1965). For example, one cannot simply count the number of studies reporting statistically significant results or statistically nonsignificant results and reach credible conclusions about the relative weight of the evidence and the likelihood of causality. Rather, these aspects provide a framework for systematic appraisal of the body of evidence, informed by peer and public comment and advice, which includes weighing alternative views on controversial issues. In addition, it is important to note that the aspects in Table I cannot be used as a strict checklist, but rather to determine the weight of the evidence for inferring causality. In particular, not meeting one or more of the principles does not automatically preclude a determination of causality [see discussion in CDC (2004)].

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1 The “aspects” described by Sir Austin Bradford Hill (Hill, 1965) have become, in the subsequent literature, more commonly described as “criteria.” The original term “aspects” is used here to avoid confusion with “criteria” as it is used, with different meaning, in the Clean Air Act.

2 The Hill aspects were developed for interpretation of epidemiologic results. They have been modified here for use with a broader array of data, i.e., epidemiologic, controlled human exposure, ecological, and animal toxicological studies, as well as in vitro data, and to be more consistent with the U.S. EPA (2005) Guidelines for Carcinogen Risk Assessment.
Table I  Aspects to aid in judging causality.

<table>
<thead>
<tr>
<th>Aspect</th>
<th>Description</th>
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<tbody>
<tr>
<td>Consistency of the observed association</td>
<td>An inference of causality is strengthened when a pattern of elevated risks is observed across several independent studies. The reproducibility of findings constitutes one of the strongest arguments for causality. If there are discordant results among investigations, possible reasons such as differences in exposure, confounding factors, and the power of the study are considered.</td>
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<tr>
<td>Coherence</td>
<td>An inference of causality from one line of evidence (e.g., epidemiologic, controlled human exposure [clinical], or animal studies) may be strengthened by other lines of evidence that support a cause-and-effect interpretation of the association. Evidence on ecological or welfare effects may be drawn from a variety of experimental approaches (e.g., greenhouse, laboratory, and field) and subdisciplines of ecology (e.g., community ecology, biogeochemistry, and paleontological/historical reconstructions). The coherence of evidence from various fields greatly adds to the strength of an inference of causality. In addition, there may be coherence in demonstrating effects across multiple study designs or related health endpoints within one scientific line of evidence.</td>
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<tr>
<td>Biological plausibility</td>
<td>An inference of causality tends to be strengthened by consistency with data from experimental studies or other sources demonstrating plausible biological mechanisms. A proposed mechanistic linking between an effect and exposure to the agent is an important source of support for causality, especially when data establishing the existence and functioning of those mechanistic links are available.</td>
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<td>Biological gradient (exposure-response relationship)</td>
<td>A well-characterized exposure-response relationship (e.g., increasing effects associated with greater exposure) strongly suggests cause and effect, especially when such relationships are also observed for duration of exposure (e.g., increasing effects observed following longer exposure times).</td>
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<td>Strength of the observed association</td>
<td>The finding of large, precise risks increases confidence that the association is not likely due to chance, bias, or other factors. However, it is noted that a small magnitude in an effect estimate may represent a substantial effect in a population.</td>
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<tr>
<td>Experimental evidence</td>
<td>Strong evidence for causality can be provided through “natural experiments” when a change in exposure is found to result in a change in occurrence or frequency of health or welfare effects.</td>
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<tr>
<td>Temporal relationship of the observed association</td>
<td>Evidence of a temporal sequence between the introduction of an agent, and appearance of the effect, constitutes another argument in favor of causality.</td>
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<tr>
<td>Specificity of the observed association</td>
<td>Evidence linking a specific outcome to an exposure can provide a strong argument for causation. However, it must be recognized that rarely, if ever, does exposure to a pollutant invariably predict the occurrence of an outcome, and that a given outcome may have multiple causes.</td>
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<tr>
<td>Analogy</td>
<td>Structure activity relationships and information on the agent’s structural analogs can provide insight into whether an association is causal. Similarly, information on mode of action for a chemical, as one of many structural analogs, can inform decisions regarding likely causality.</td>
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</table>
Determination of Causality

In the ISA, EPA assesses the body of relevant literature, building upon evidence available during previous NAAQS reviews, to draw conclusions on the causal relationships between relevant pollutant exposures and health or environmental effects. ISAs use a five-level hierarchy that classifies the weight of evidence for causation. In developing this hierarchy, EPA has drawn on the work of previous evaluations, most prominently the IOM’s Improving the Presumptive Disability Decision-Making Process for Veterans (Samet and Bodurow, 2008), EPA’s Guidelines for Carcinogen Risk Assessment (U.S. EPA, 2005), and the U.S. Surgeon General’s smoking report (CDC, 2004). This weight of evidence evaluation is based on integration of findings from various lines of evidence from across the health and environmental effects disciplines. These separate judgments are integrated into a qualitative statement about the overall weight of the evidence and causality. The five descriptors for causal determination are described in Table II.

Determination of causality involves the evaluation and integration of evidence for different types of health, ecological or welfare effects associated with short- and long-term exposure periods. In making determinations of causality, evidence is evaluated for major outcome categories or groups of related endpoints (e.g., respiratory effects, vegetation growth), integrating evidence from across disciplines, and assessing the coherence of evidence across a spectrum of related endpoints to draw conclusions regarding causality. In discussing the causal determination, EPA characterizes the evidence on which the judgment is based, including strength of evidence for individual endpoints within the outcome category or group of related endpoints.

In drawing judgments regarding causality for the criteria air pollutants, the ISA focuses on evidence of effects in the range of relevant pollutant exposures or doses, and not on determination of causality at any dose. Emphasis is placed on evidence of effects at doses (e.g., blood Pb concentration) or exposures (e.g., air concentrations) that are relevant to, or somewhat above, those currently experienced by the population. The extent to which studies of higher concentrations are considered varies by pollutant and major outcome category, but generally includes those with doses or exposures in the range of one to two orders of magnitude above current or ambient conditions. Studies that use higher doses or exposures may also be considered to the extent that they provide useful information to inform understanding of mode of action, interspecies differences, or factors that may increase risk of effects for a population. Thus, a causality determination is based on weight of evidence evaluation for health, ecological or welfare effects, focusing on the evidence from exposures or doses generally ranging from current levels to one or two orders of magnitude above current levels.

1 Both the CDC and IOM frameworks use a four-category hierarchy for the strength of the evidence. A five-level hierarchy is used here to be consistent with the EPA Guidelines for Carcinogen Risk Assessment and to provide a more nuanced set of categories.
In addition, EPA evaluates evidence relevant to understand the quantitative relationships between pollutant exposures and health, ecological or welfare effects. This includes evaluation of the form of concentration-response or dose-response relationships and, to the extent possible, drawing conclusions on the levels at which effects are observed. The ISA also draws scientific conclusions regarding important exposure conditions for effects and populations that may be at greater risk for effects, as described in the following section.
<table>
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<tr>
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<th>Health Effects</th>
<th>Ecological and Welfare Effects</th>
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<tr>
<td><strong>Causal relationship</strong></td>
<td>Evidence is sufficient to conclude that there is a causal relationship with relevant pollutant exposures (i.e., doses or exposures generally within one to two orders of magnitude of current levels). That is, the pollutant has been shown to result in health effects in studies in which chance, bias, and confounding could be ruled out with reasonable confidence. For example: a) controlled human exposure studies that demonstrate consistent effects; or b) observational studies that cannot be explained by plausible alternatives or are supported by other lines of evidence (e.g., animal studies or mode of action information). Evidence includes multiple high-quality studies.</td>
<td>Evidence is sufficient to conclude that there is a causal relationship with relevant pollutant exposures (i.e., doses or exposures generally within one to two orders of magnitude of current levels). That is, the pollutant has been shown to result in effects in studies in which chance, bias, and confounding could be ruled out with reasonable confidence. Controlled exposure studies (laboratory or small- to medium-scale field studies) provide the strongest evidence for causality, but the scope of inference may be limited. Generally, determination is based on multiple studies conducted by multiple research groups, and evidence that is considered sufficient to infer a causal relationship is usually obtained from the joint consideration of many lines of evidence that reinforce each other.</td>
</tr>
<tr>
<td><strong>Likely to be a causal relationship</strong></td>
<td>Evidence is sufficient to conclude that a causal relationship is likely to exist with relevant pollutant exposures, but important uncertainties remain. That is, the pollutant has been shown to result in health effects in studies in which chance and bias can be ruled out with reasonable confidence but potential issues remain. For example: a) observational studies show an association, but copollutant exposures are difficult to address and/or other lines of evidence (controlled human exposure, animal, or mode of action information) are limited or inconsistent; or b) animal toxicological evidence from multiple studies from different laboratories that demonstrate effects, but limited or no human data are available. Evidence generally includes multiple high-quality studies.</td>
<td>Evidence is sufficient to conclude that there is a likely causal association with relevant pollutant exposures. That is, an association has been observed between the pollutant and the outcome in studies in which chance, bias, and confounding are minimized, but uncertainties remain. For example, field studies show a relationship, but suspected interacting factors cannot be controlled, and other lines of evidence are limited or inconsistent. Generally, determination is based on multiple studies in multiple research groups.</td>
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<tr>
<td><strong>Suggestive of a causal relationship</strong></td>
<td>Evidence is suggestive of a causal relationship with relevant pollutant exposures, but is limited. For example, (a) at least one high-quality epidemiologic study shows an association with a given health outcome but the results of other studies are inconsistent; or (b) a well-conducted toxicological study, such as those conducted in the National Toxicology Program (NTP), shows effects in animal species,</td>
<td>Evidence is suggestive of a causal relationship with relevant pollutant exposures, but chance, bias and confounding cannot be ruled out. For example, at least one high-quality study shows an effect, but the results of other studies are inconsistent.</td>
</tr>
<tr>
<td><strong>Inadequate to infer a causal relationship</strong></td>
<td>Evidence is inadequate to determine that a causal relationship exists with relevant pollutant exposures. The available studies are of insufficient quality, quantity, consistency, or statistical power to permit a conclusion regarding the presence or absence of an effect.</td>
<td>The available studies are of insufficient quality, consistency, or statistical power to permit a conclusion regarding the presence or absence of an effect.</td>
</tr>
<tr>
<td><strong>Not likely to be a causal relationship</strong></td>
<td>Evidence is suggestive of no causal relationship with relevant pollutant exposures. Several adequate studies, covering the full range of levels of exposure that human beings are known to encounter and considering at-risk populations, are mutually consistent in not showing an effect at any level of exposure.</td>
<td>Several adequate studies, examining relationships with relevant exposures, are consistent in failing to show an effect at any level of exposure.</td>
</tr>
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</table>
Quantitative Relationships: Effects on Human Populations

Once a determination is made regarding the causal relationship between the pollutant and outcome category, important questions regarding quantitative relationships include:

- What is the concentration-response, exposure-response, or dose-response relationship in the human population?
- What is the interrelationship between incidence and severity of effect?
- What exposure conditions (dose or exposure, duration and pattern) are important?
- What populations and lifestages appear to be differentially affected (i.e., more at risk of experiencing effects)?

To address these questions, the entirety of quantitative evidence is evaluated to characterize pollutant concentrations and exposure durations at which effects were observed for exposed populations, including populations and lifestages potentially at increased risk. To accomplish this, evidence is considered from multiple and diverse types of studies, and a study or set of studies that best approximates the concentration-response relationships between health outcomes and the pollutant may be identified. Controlled human exposure studies provide the most direct and quantifiable exposure-response data on the human health effects of pollutant exposures. To the extent available, the ISA evaluates results from across epidemiologic studies that characterize the form of relationships between the pollutant and health outcomes and draws conclusions on the shape of these relationships. Animal data may also inform evaluation of concentration-response relationships, particularly relative to MOAs and characteristics of at-risk populations.

An important consideration in characterizing the public health impacts associated with exposure to a pollutant is whether the concentration-response relationship is linear across the range of concentrations or if nonlinear relationships exist along any part of this range. Of particular interest is the shape of the concentration-response curve at and below the level of the current standards. Various sources of variability and uncertainty, such as low data density in the lower concentration range, possible influence of exposure measurement error, and variability between individuals in susceptibility to air pollution health effects, tend to smooth and “linearize” the concentration-response function, and thus can obscure the existence of a threshold or nonlinear relationship [2006 O₃ AQCD (U.S. EPA, 2006b)]. Since individual thresholds vary from person to person due to individual differences such as genetic level susceptibility or pre-existing disease conditions (and even can vary from one time to another for a given person), it can be difficult to demonstrate that a threshold exists in a population study. These sources of variability and uncertainty may explain why the available human data at ambient concentrations for some environmental pollutants (e.g., particulate matter [PM], O₃, lead [Pb], environmental tobacco smoke [ETS], radiation) do not exhibit thresholds for cancer or noncancer health effects, even though likely mechanisms include nonlinear processes for some key events.
Finally, identification of the population groups or lifestages that may be at greater risk of health effects from air pollutant exposures contributes to an understanding of the public health impact of pollutant exposures. In the ISA, the term “at-risk population” is used to encompass populations or lifestages that have a greater likelihood of experiencing health effects related to exposure to an air pollutant due to a variety of factors; other terms used in the literature include susceptible, vulnerable, and sensitive. These factors may be intrinsic, such as genetic or developmental factors, race, sex, lifestage, or the presence of pre-existing diseases, or they may be extrinsic, such as socioeconomic status (SES), activity pattern and exercise level, reduced access to health care, low educational attainment, or increased pollutant exposures (e.g., near roadways). Epidemiologic studies can help identify populations potentially at increased risk of effects by evaluating health responses in the study population. Examples include testing for interactions or effect modification by factors such as sex, age group, or health status. Experimental studies using animal models of susceptibility or disease can also inform the extent to which health risks are likely greater in specific population groups.

Quantitative Relationships: Effects on Ecosystems or Public Welfare

Key questions for understanding the quantitative relationships between exposure (or concentration or deposition) to a pollutant and risk to ecosystems or the public welfare include:

- What elements of the ecosystem (e.g., types, regions, taxonomic groups, populations, functions, etc.) appear to be affected, or are more sensitive to effects? Are there differences between locations or materials in welfare effects responses, such as impaired visibility or materials damage?
- Under what exposure conditions (amount deposited or concentration, duration and pattern) are effects seen?
- What is the shape of the concentration-response or exposure-response relationship?

Evaluations of causality generally consider the probability of quantitative changes in ecological and welfare effects in response to exposure. A challenge to the quantification of exposure-response relationships for ecological effects is the great regional and local spatial variability, as well as temporal variability, in ecosystems. Thus, exposure-response relationships are often determined for a specific ecological system and scale, rather than at the national or even regional scale. Quantitative relationships therefore are estimated site by site and may differ greatly between ecosystems.

Concepts in Evaluating Adversity of Health Effects

In evaluating health evidence, a number of factors can be considered in delineating between adverse and nonadverse health effects resulting from exposure to air
pollution. Some health outcomes, such as hospitalization for respiratory or cardiovascular diseases, are clearly considered adverse. It is more difficult to determine the extent of change that constitutes adversity in more subtle health measures. These include a wide variety of responses, such as alterations in markers of inflammation or oxidative stress, changes in pulmonary function or heart rate variability, or alterations in neurocognitive function measures. The challenge is determining the magnitude of change in these measures when there is no clear point at which a change becomes adverse. The extent to which a change in health measure constitutes an adverse health effect may vary between populations. Some changes that may not be considered adverse in healthy individuals would be potentially adverse in more at-risk individuals.

The extent to which changes in lung function are adverse has been discussed by the American Thoracic Society (ATS) in an official statement titled *What Constitutes an Adverse Health Effect of Air Pollution?* (ATS, 2000b). An air pollution-induced shift in the population distribution of a given risk factor for a health outcome was viewed as adverse, even though it may not increase the risk of any one individual to an unacceptable level. For example, a population of asthmatics could have a distribution of lung function such that no identifiable individual has a level associated with significant impairment. Exposure to air pollution could shift the distribution such that no identifiable individual experiences any clinically relevant effects. This shift toward decreased lung function, however, would be considered adverse because individuals within the population would have diminished reserve function and therefore would be at increased risk to further environmental insult. The committee also observed that elevations of biomarkers, such as cell number and types, cytokines and reactive oxygen species, may signal risk for ongoing injury and clinical effects or may simply indicate transient responses that can provide insights into mechanisms of injury, thus illustrating the lack of clear boundaries that separate adverse from nonadverse effects.

The more subtle health outcomes may be connected mechanistically to health events that are clearly adverse. For example, air pollution may affect markers of transient myocardial ischemia such as ST-segment abnormalities and onset of exertional angina. These effects may not be apparent to the individual, yet may still increase the risk of a number of cardiac events, including myocardial infarction and sudden death. Thus, small changes in physiological measures may not appear to be clearly adverse when considered alone, but may be a part of a coherent and biologically plausible chain of related health outcomes that range up to responses that are very clearly adverse, such as hospitalization or mortality.

**Concepts in Evaluating Adversity of Ecological Effects**

Adversity of ecological effects can be understood in terms ranging in biological level of organization; from the cellular level to the individual organism and to the population, community, and ecosystem levels. In the context of ecology, a population is a group of individuals of the same species, and a community is an assemblage of populations of different species interacting with one another that inhabit an area.
An ecosystem is the interactive system formed from all living organisms and their abiotic (physical and chemical) environment within a given area (IPCC, 2007a). The boundaries of what could be called an ecosystem are somewhat arbitrary, depending on the focus of interest or study. Thus, the extent of an ecosystem may range from very small spatial scales to, ultimately, the entire Earth (IPCC, 2007a).

Effects on an individual organism are generally not considered to be adverse to public welfare. However if effects occur to enough individuals within a population, then communities and ecosystems may be disrupted. Changes to populations, communities, and ecosystems can in turn result in an alteration of ecosystem processes. Ecosystem processes are defined as the metabolic functions of ecosystems including energy flow, elemental cycling, and the production, consumption and decomposition of organic matter (U.S. EPA, 2002). Growth, reproduction, and mortality are species-level endpoints that can be clearly linked to community and ecosystem effects and are considered to be adverse when negatively affected. Other endpoints such as changes in behavior and physiological stress can decrease ecological fitness of an organism, but are harder to link unequivocally to effects at the population, community, and ecosystem level. The degree to which pollutant exposure is considered adverse may also depend on the location and its intended use (i.e., city park, commercial, cropland). Support for consideration of adversity beyond the species level by making explicit the linkages between stress-related effects at the species and effects at the ecosystem level is found in A Framework for Assessing and Reporting on Ecological Condition: an SAB report (U.S. EPA, 2002). Additionally, the National Acid Precipitation Assessment Program (NAPAP, 1991) uses the following working definition of “adverse ecological effects” in the preparation of reports to Congress mandated by the Clean Air Act: “any injury (i.e., loss of chemical or physical quality or viability) to any ecological or ecosystem component, up to and including at the regional level, over both long and short terms.”

On a broader scale, ecosystem services may provide indicators for ecological impacts. Ecosystem services are the benefits that people obtain from ecosystems (UNEP, 2003). According to the Millennium Ecosystem Assessment, ecosystem services include: “provisioning services such as food and water; regulating services such as regulation of floods, drought, land degradation, and disease; supporting services such as soil formation and nutrient cycling; and cultural services such as recreational, spiritual, religious, and other nonmaterial benefits.” For example, a more subtle ecological effect of pollution exposure may result in a clearly adverse impact on ecosystem services if it results in a population decline in a species that is recreationally or culturally important.
References


Zeger, SL; Thomas, D; Dominici, F; Samet, JM; Schwartz, J; Dockery, D; Cohen, A. (2000). Exposure measurement error in time-series studies of air pollution: Concepts and consequences. Environ Health Perspect 108: 419-426.
Legislative Requirements for the NAAQS Review

Two sections of the Clean Air Act (CAA) govern the establishment and revision of the National Ambient Air Quality Standards (NAAQS). Section 108 (42 USC §7408) directs the Administrator to identify and list certain air pollutants and then to issue air quality criteria for those pollutants. The Administrator is to list those air pollutants that in her “judgement; cause or contribute to air pollution which may reasonably be anticipated to endanger public health or welfare;” … “the presence of which in the ambient air results from numerous or diverse mobile or stationary sources” and “for which … [the Administrator] plans to issue air quality criteria…” (CAA, 1990a). Air quality criteria are intended to “accurately reflect the latest scientific knowledge useful in indicating the kind and extent of identifiable effects on public health or welfare, which may be expected from the presence of [a] pollutant in ambient air …” [42 USC §7408(b)].

Section 109 (CAA, 1990b) directs the Administrator to propose and promulgate “primary” and “secondary” NAAQS for pollutants for which air quality criteria have been issued. Section 109(b)(1) defines a primary standard as one “the attainment and maintenance of which in the judgment of the Administrator, based on such criteria and allowing an adequate margin of safety, are requisite to protect the public health.”¹ A secondary standard, as defined in section 109(b)(2), must “specify a level of air quality the attainment and maintenance of which, in the judgment of the Administrator, based on such criteria, is required to protect the public welfare from any known or anticipated adverse effects associated with the presence of [the] pollutant in the ambient air.”²

The requirement that primary standards include an adequate margin of safety was intended to address uncertainties associated with inconclusive scientific and technical information available at the time of standard setting. It was also intended to provide a reasonable degree of protection against hazards that research has not yet identified. See Lead Industries Association v. EPA, 647 F.2d 1130, 1154 (D.C. Cir 1980), cert. denied, 449 U.S. 1042 (1980); American Petroleum Institute v. Costle, 665 F.2d 1176, 1186 (D.C. Cir. (1981), cert. denied, 455 U.S. 1034 (1982). Both kinds of uncertainties are components of the risk associated with pollution at levels below those at which human health effects can be said to occur with reasonable scientific certainty. Thus, in selecting primary standards that include an adequate margin of safety, the Administrator is seeking not only to prevent pollution levels that have been demonstrated to be harmful but also to prevent lower pollutant levels that may

¹ The legislative history of section 109 indicates that a primary standard is to be set at “the maximum permissible ambient air level . . . which will protect the health of any [sensitive] group of the population;” and that for this purpose “reference should be made to a representative sample of persons comprising the sensitive group rather than to a single person in such a group” [S. Rep. No. 91-1196, 91st Cong., 2d Sess. 10 (1970)].

² Welfare effects as defined in section 302(h) include, but are not limited to, “effects on soils, water, crops, vegetation, man-made materials, animals, wildlife, weather, visibility and climate, damage to and deterioration of property, and hazards to transportation, as well as effects on economic values and on personal comfort and well-being” (CAA, 2005).
pose an unacceptable risk of harm, even if the risk is not precisely identified as to nature or degree. The CAA does not require the Administrator to establish a primary NAAQS at a zero-risk level or at background concentration levels, see Lead Industries v. EPA, 647 F.2d at 1156 n.51, but rather at a level that reduces risk sufficiently so as to protect public health with an adequate margin of safety.

In addressing the requirement for a margin of safety, EPA considers such factors as the nature and severity of the health effects involved, the size of the sensitive population(s) at risk, and the kind and degree of the uncertainties that must be addressed. The selection of any particular approach to providing an adequate margin of safety is a policy choice left specifically to the Administrator’s judgment. See Lead Industries Association v. EPA, supra, 647 F.2d at 1161-1162; Whitman v. American Trucking Associations, 531 U.S. 457, 495 (2001).

In setting standards that are “requisite” to protect public health and welfare, as provided in Section 109(b), EPA’s task is to establish standards that are neither more nor less stringent than necessary for these purposes. In so doing, EPA may not consider the costs of implementing the standards. [See generally, Whitman v. American Trucking Associations, 531 U.S. 457, 465-472, 475-76. (2001)]. Likewise, “attainability and technological feasibility are not relevant considerations in the promulgation of national ambient air quality standards.” American Petroleum Institute v. Costle, 665 F. 2d at 1185.

Section 109(d)(1) requires that “not later than December 31, 1980, and at 5-year intervals thereafter, the Administrator shall complete a thorough review of the criteria published under section 108 and the national ambient air quality standards … and shall make such revisions in such criteria and standards and promulgate such new standards as may be appropriate…” Section 109(d)(2) requires that an independent scientific review committee “shall complete a review of the criteria … and the national primary and secondary ambient air quality standards … and shall recommend to the Administrator any new … standards and revisions of existing criteria and standards as may be appropriate …” Since the early 1980s, this independent review function has been performed by CASAC.

History of the NAAQS for Ozone

Tropospheric (ground-level) O₃ is the indicator for the mix of photochemical oxidants (e.g., peroxyacetyl nitrate, hydrogen peroxide) formed from biogenic and anthropogenic precursor emissions. Naturally occurring O₃ in the troposphere can result from biogenic organic precursors reacting with naturally occurring nitrogen oxides (NOₓ) and by stratospheric O₃ intrusion into the troposphere. Anthropogenic precursors of O₃, especially NOₓ, and volatile organic compounds (VOCs), originate from a wide variety of stationary and mobile sources. Ambient O₃ concentrations produced by these emissions are directly affected by temperature, solar radiation, wind speed, and other meteorological factors.
NAAQS are comprised of four basic elements: indicator, averaging time, level, and form. The indicator defines the pollutant to be measured in the ambient air for the purpose of determining compliance with the standard. The averaging time defines the time period over which air quality measurements are to be obtained and averaged or cumulated, considering evidence of effects associated with various time periods of exposure. The level of a standard defines the air quality concentration used (i.e., an ambient concentration of the indicator pollutant) in determining whether the standard is achieved. The form of the standard specifies the air quality measurements that are to be used for compliance purposes (e.g., the annual fourth-highest daily maximum 8-h concentration, averaged over 3 years), and whether the statistic is to be averaged across multiple years. These four elements taken together determine the degree of public health and welfare protection afforded by the NAAQS.

Table III  Summary of primary and secondary NAAQS promulgated for O₃ during the period 1971-2008.

<table>
<thead>
<tr>
<th>Final Rule</th>
<th>Indicator</th>
<th>Avg Time</th>
<th>Level (ppm)</th>
<th>Form</th>
</tr>
</thead>
<tbody>
<tr>
<td>1971 (36 FR 8186)</td>
<td>Total photochemical oxidants</td>
<td>1-h</td>
<td>0.08</td>
<td>Not to be exceeded more than 1 hour per year</td>
</tr>
<tr>
<td>1979 (44 FR 8202)</td>
<td>O₃</td>
<td>1-h</td>
<td>0.12</td>
<td>Attainment is defined when the expected number of days per calendar year, with maximum hourly average concentration greater than 0.12 ppm, is ≤ 1</td>
</tr>
<tr>
<td>1993 (58 FR 13008)</td>
<td>EPA decided that revisions to the standards were not warranted at the time.</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1997 (62 FR 38856)</td>
<td>O₃</td>
<td>8-h</td>
<td>0.08</td>
<td>Annual fourth-highest daily maximum 8-h concentration averaged over 3 years</td>
</tr>
<tr>
<td>2008 (73 FR 16483)</td>
<td>O₃</td>
<td>8-h</td>
<td>0.075</td>
<td>Form of the standards remained unchanged relative to the 1997 standard</td>
</tr>
</tbody>
</table>

Table III summarizes the O₃ NAAQS that have been promulgated to date. In each review, the secondary standard has been set to be identical to the primary standard. These reviews are briefly described below.

EPA first established primary and secondary NAAQS for photochemical oxidants in 1971. Both primary and secondary standards were set at a level of 0.08 parts per million (ppm), 1-h avg, total photochemical oxidants, not to be exceeded more than 1 hour per year. The standards were based on scientific information contained in the 1970 O₃ AQCD.

In 1977, EPA announced the first periodic review of the 1970 AQCD in accordance with Section 109(d)(1) of the Clean Air Act. In 1978, EPA published an AQCD. Based on the 1978 AQCD, EPA published proposed revisions to the original NAAQS in 1978 (U.S. EPA, 1978b) and final revisions in 1979 (U.S. EPA, 1979a). The level of the primary and secondary standards was revised from 0.08 to 0.12 ppm;
the indicator was revised from photochemical oxidants to \( \text{O}_3 \); and the form of the standards was revised from a deterministic to a statistical form, which defined attainment of the standards as occurring when the expected number of days per calendar year with maximum hourly average concentration greater than 0.12 ppm is equal to or less than one.

In 1982, EPA announced plans to revise the 1978 AQCD (\cite{U.S. EPA, 1978a}). In 1983, EPA announced that the second periodic review of the primary and secondary standards for \( \text{O}_3 \) had been initiated (\cite{U.S. EPA, 1983}). EPA subsequently published the 1986 \( \text{O}_3 \) AQCD (\cite{U.S. EPA, 1986}) and 1989 Staff Paper (\cite{U.S. EPA, 1989}). Following publication of the 1986 \( \text{O}_3 \) AQCD, a number of scientific abstracts and articles were published that appeared to be of sufficient importance concerning potential health and welfare effects of \( \text{O}_3 \) to warrant preparation of a Supplement to the 1986 \( \text{O}_3 \) AQCD (\cite{Costa et al., 1992}). Under the terms of a court order, on August 10, 1992, EPA published a proposed decision (\cite{U.S. EPA, 1992}) stating that revisions to the existing primary and secondary standards were not appropriate at the time (\cite{U.S. EPA, 1992}). This notice explained that the proposed decision would complete EPA’s review of information on health and welfare effects of \( \text{O}_3 \) assembled over a 7-year period and contained in the 1986 \( \text{O}_3 \) AQCD (\cite{U.S. EPA, 1986}) and its Supplement to the 1986 \( \text{O}_3 \) AQCD (\cite{Costa et al., 1992}). The proposal also announced EPA’s intention to proceed as rapidly as possible with the next review of the air quality criteria and standards for \( \text{O}_3 \) in light of emerging evidence of health effects related to 6- to 8-hour \( \text{O}_3 \) exposures. On March 9, 1993, EPA concluded the review by deciding that revisions to the standards were not warranted at that time (\cite{U.S. EPA, 1993}).

In August 1992, EPA announced plans to initiate the third periodic review of the air quality criteria and \( \text{O}_3 \) NAAQS (\cite{U.S. EPA, 1992}). On the basis of the scientific evidence contained in the 1996 \( \text{O}_3 \) AQCD (\cite{U.S. EPA, 1996a}) and the 1996 Staff Paper (\cite{U.S. EPA, 1996e}), and related technical support documents, linking exposures to ambient \( \text{O}_3 \) to adverse health and welfare effects at levels allowed by the then existing standards, EPA proposed to revise the primary and secondary \( \text{O}_3 \) standards on December 13, 1996 (\cite{U.S. EPA, 1996d}). The EPA proposed to replace the then existing 1-hour primary and secondary standards with 8-h avg \( \text{O}_3 \) standards set at a level of 0.08 ppm (equivalent to 0.084 ppm using standard rounding conventions). The EPA also proposed, in the alternative, to establish a new distinct secondary standard using a biologically based cumulative seasonal form. The EPA completed the review on July 18, 1997 by setting the primary standard at a level of 0.08 ppm, based on the annual fourth-highest daily maximum 8-h avg concentration, averaged over 3 years, and setting the secondary standard identical to the revised primary standard (\cite{U.S. EPA, 1997}).

On May 14, 1999, in response to challenges to EPA’s 1997 decision by industry and others, the U.S. Court of Appeals for the District of Columbia Circuit (D.C. Cir.) remanded the \( \text{O}_3 \) NAAQS to EPA, finding that Section 109 of the CAA, as interpreted by EPA, effected an unconstitutional delegation of legislative authority. In addition, the D.C. Cir. directed that, in responding to the remand, EPA should
consider the potential beneficial health effects of O₃ pollution in shielding the public from the effects of solar ultraviolet (UV) radiation, as well as adverse health effects. On January 27, 2000, EPA petitioned the U.S. Supreme Court for certiorari on the constitutional issue (and two other issues) but did not request review of the D.C. Cir., ruling regarding the potential beneficial health effects of O₃. On February 27, 2001, the U.S. Supreme Court unanimously reversed the judgment of the D.C. Cir. on the constitutional issue, holding that Section 109 of the CAA does not delegate legislative power to the EPA in contravention of the Constitution, and remanded the case to the D.C. Cir. to consider challenges to the O₃ NAAQS that had not been addressed by that Court’s earlier decisions. On March 26, 2002, the D.C. Cir. issued its final decision, finding the 1997 O₃ NAAQS to be “neither arbitrary nor capricious,” and denied the remaining petitions for review. On November 14, 2001, in response to the D.C. Cir. remand to consider the potential beneficial health effects of O₃ pollution in shielding the public from effects of solar (UV) radiation, EPA proposed to leave the 1997 8-h O₃ NAAQS unchanged (U.S. EPA, 2001). After considering public comment on the proposed decision, EPA published its final response to this remand on January 6, 2003, reaffirming the 8-h O₃ NAAQS set in 1997 (U.S. EPA, 2003). On April 30, 2004, EPA announced the decision to make the 1-h O₃ NAAQS no longer applicable to areas 1 year after the effective date of the designation of those areas for the 8-h NAAQS (U.S. EPA, 2004). For most areas, the date that the 1-h NAAQS no longer applied was June 15, 2005.

EPA initiated the next periodic review of the air quality criteria and O₃ standards in September 2000 with a call for information (U.S. EPA, 2000). The schedule for completion of that rulemaking later became governed by a consent decree resolving a lawsuit filed in March 2003 by a group of plaintiffs representing national environmental and public health organizations. Based on the 2006 O₃ AQCD (U.S. EPA, 2006b) published in March 2006, the Staff Paper (U.S. EPA, 2007b) and related technical support documents, the proposed decision was published in the Federal Register on July 11, 2007 (U.S. EPA, 2007a). The EPA proposed to revise the level of the primary standard to a level within the range of 0.075 to 0.070 ppm. Two options were proposed for the secondary standard: (1) replacing the current standard with a cumulative, seasonal standard, expressed as an index of the annual sum of weighted hourly concentrations cumulated over 12 daylight hours during the consecutive 3-month period within the O₃ season with the maximum index value, set at a level within the range of 7 to 21 ppm-h; and (2) setting the secondary standard identical to the revised primary standard. The EPA completed the rulemaking with publication of a final decision on March 27, 2008 (U.S. EPA, 2008f), revising the level of the 8-hour primary O₃ standard from 0.08 ppm to 0.075 ppm and revising the secondary standard to be identical to the primary standard.

In May 2008, state, public health, environmental, and industry petitioners filed suit against EPA regarding that final decision. At EPA’s request the consolidated cases were held in abeyance pending EPA’s reconsideration of the 2008 decision. A notice of proposed rulemaking to reconsider the 2008 final decision was issued by the Administrator on January 6, 2010. Three public hearings were held. The Agency solicited CASAC review of the proposed rule on January 25, 2010 and additional
CASAC advice on January 26, 2011. On September 2, 2011, the Office of Management and Budget returned the draft final rule on reconsideration to EPA for further consideration. EPA decided to coordinate further proceedings on its voluntary rulemaking on reconsideration with the ongoing periodic review, by deferring the completion of its voluntary rulemaking on reconsideration until it completes its statutorily-required periodic review. In light of that, the litigation on the 2008 final decision is no longer being held in abeyance and is proceeding. The 2008 O₃ standards remain in effect.
References


1  EXECUTIVE SUMMARY

Introduction and Purpose

The purpose of this Integrated Science Assessment (ISA) is to provide a synthesis and evaluation of the most policy-relevant science that builds the scientific foundation for the periodic review of the primary (health-based) and secondary (welfare-based) national ambient air quality standard (NAAQS) for ozone (O₃) and related photochemical oxidants required by the Clean Air Act. The primary NAAQS protects against respiratory health effects incurred after short-term exposure to O₃, while the secondary NAAQS protects against damage to vegetation and ecosystems, generally referred to as “welfare effects.” The ISA is intended to inform both EPA’s Risk and Exposure Assessment and Policy Assessment, and decisions by the EPA Administrator on the NAAQS for O₃ (see Figure I in Preamble). Set in 2008, the current primary O₃ standard is an 8-hour average standard of 75 parts per billion (ppb). The secondary standard for O₃ is equal to the primary standard.

Scope and Methods

EPA has an extensive, robust process for evaluating the latest scientific evidence and drawing conclusions regarding air pollution-related health and welfare effects. Building upon the findings of previous assessments, this review includes identification, selection, evaluation, and integration of the relevant results pertaining to the atmospheric science of O₃; short- and long-term exposure to ambient O₃; health effects due to relevant O₃ concentrations as characterized in epidemiologic or controlled human exposure and toxicological studies; and ecological and other welfare effects. Additionally, this review will characterize O₃ concentration-response relationships, mode(s) of action, and populations at increased risk for O₃-related health effects. The conclusions and key findings from previous reviews provide the basis for the consideration of evidence from recent studies (i.e., studies published since the completion of the 2006 O₃ Air Quality Criteria Document [AQCD]). Conclusions are drawn based on the synthesis of evidence across scientific disciplines.

EPA assesses the body of peer-reviewed literature to draw conclusions on the causal relationships between relevant pollutant concentrations and health or welfare effects. EPA specifically evaluates the quantitative evidence and draws scientific conclusions, to the extent possible, regarding the concentration-response relationships and the loads to ecosystems, exposure doses or concentrations, and duration and pattern of exposures at which effects are observed.

EPA uses a consistent and transparent approach to evaluate the causal nature of air pollution-related health and environmental effects for use in developing ISAs; the framework for causal determinations is described in the Preamble to this document. Causality determinations are based on the evaluation and synthesis of evidence.
across scientific disciplines. A five-level hierarchy is used to classify the weight of evidence for causation, not just association. This weight of evidence evaluation is based on various lines of evidence from across the health and environmental effects disciplines. These separate conclusions are integrated into a qualitative statement about the overall weight of the evidence and causality. The causal determinations are:

- Causal relationship
- Likely to be a causal relationship
- Suggestive of a causal relationship
- Inadequate to infer a causal relationship
- Not likely to be a causal relationship

Ambient Ozone Concentrations

Ozone is naturally present in the stratosphere (an elevated layer of the Earth’s atmosphere) where it serves the beneficial role of absorbing harmful ultraviolet radiation from the Sun, preventing the majority of this radiation from reaching the surface of the Earth. However, in the troposphere (the layer of the atmosphere extending from the stratosphere down to the Earth’s surface), O₃ acts as a powerful oxidizing agent, which can harm living organisms and materials. Tropospheric O₃ is present not only in polluted urban air, but across the globe. Ozone concentrations can be influenced by local meteorological conditions, circulation patterns, emissions, and topographic barriers, resulting in heterogeneous concentrations across an individual urban area. On a larger scale, O₃ can last in the atmosphere long enough that it can be transported from continent to continent.

Ozone in the troposphere originates from both anthropogenic (i.e., man-made) and natural sources. Ozone attributed to anthropogenic sources is formed by photochemical reactions involving sunlight and precursor pollutants, including volatile organic compounds (VOCs), nitrogen oxides (NOₓ), and carbon monoxide (CO). Ozone attributed to natural sources is formed through the same photochemical reactions involving natural emissions of precursor pollutants from vegetation, microbes, animals, burning biomass (e.g., forest fires), and lightning. Ozone is lost through deposition to surfaces and chemical reactions occurring in the atmosphere. The highest O₃ concentrations are not found in urban areas close to the concentrated sources of its precursors such as traffic, but rather in suburban and rural areas downwind of these sources. Reaction of O₃ with NO in fresh motor vehicle exhaust depletes O₃ in urban cores; but O₃ can be regenerated during transport downwind of urban source areas. Also O₃ tends to be more uniform in rural than in urban areas because O₃ production occurs over large areas and because the concentrated sources of NO depleting O₃ in urban cores are generally lacking (except near power plants and other strong sources of NO).

In the context of a review of the NAAQS, it is useful to define background O₃ concentrations in a way that distinguishes between concentrations that result from precursor emissions that are relatively less controllable (including natural sources
and foreign anthropogenic sources) from those that are relatively more controllable through U.S. policies. For example, U.S. background O$_3$ concentrations can be defined as those concentrations resulting from natural sources everywhere in the world plus anthropogenic sources outside the U.S.; North American (NA) background O$_3$ concentrations can be defined as those concentrations resulting from natural sources everywhere in the world plus anthropogenic sources outside the U.S., Canada and Mexico. Since background defined either of these ways is a hypothetical construct that cannot be measured, NA background O$_3$ concentrations are estimated using chemistry transport models (see Sections 2.2 and 3.4).

**Human Exposure to Ozone**

The widespread presence of O$_3$ in the environment results in exposure as people participate in normal daily activities. Exposure may occur indoors, where people spend most of their time, as well as outdoors, where O$_3$ concentrations are highest. Evaluating relationships among outdoor concentration, indoor concentration, and personal exposure is useful for determining how well ambient monitors represent exposure. The relationship between personal exposure and ambient concentration measured at fixed-site monitors can be described in terms of correlation (how they covary over time) and ratio, which describes their relative magnitude. High correlations imply that changes in ambient concentrations are reflected in personal exposure, while high ratios imply that the magnitude of exposure is similar to the magnitude of concentrations. Personal-ambient O$_3$ correlations are generally moderate (0.3-0.8), although low correlations have been observed with increased time spent indoors, low indoor-outdoor air exchange rates, and concentrations below personal sampler detection limits (see Section 4.3). Ratios of 0.1-0.3 between personal exposure and ambient concentration have been observed for the general population, with ratios of up to 0.9 observed for outdoor workers. Evidence suggests that some groups, particularly children, older adults, and those with respiratory problems, change their behavior on high-O$_3$ days when O$_3$ warnings are issued with advice to reduce exposure (see Section 4.4.2). Such behavioral changes may result in reduced effect estimates in epidemiologic studies that do not account for averting behavior on high-O$_3$ days. Variation in O$_3$ concentrations occurs over multiple spatial and temporal scales, and this introduces exposure error into epidemiologic results (see Section 4.6.2). However, epidemiologic studies evaluating the influence of spatial scale and monitor selection find little difference among effect estimates, and comparable risk estimates have been reported in studies using a variety of exposure assessment techniques expected to produce different levels of personal-ambient associations. This suggests that there is no clear indication that any particular method of exposure assessment produces stronger epidemiologic results than any other method.
Dosimetry and Modes of Action

When O$_3$ is inhaled, the amount of O$_3$ that is absorbed is affected by the shape and size of the respiratory tract, and the route of breathing (nose or mouth), as well as how quickly and deeply a person is breathing. The amount of O$_3$ that is removed from the air stream during breathing is referred to as uptake. The primary site of O$_3$ uptake moves deeper into the respiratory tract during exercise when breathing becomes faster and the breathing route changes from the nose only to oronasal breathing (i.e., through the nose and mouth) (see Section 5.2). Tissue dose refers to the amount of O$_3$ that diffuses through the lining of the lung and reaches the underlying tissues. The site of the greatest O$_3$ dose to the lung tissue is the junction of the conducting airway and the gas exchange region in the deeper portion of the lung.

Once O$_3$ has been absorbed, there are several key events in the toxicity pathway of O$_3$ in the respiratory tract that lead to O$_3$-induced health effects (see Section 5.3). These include formation of secondary oxidation products in the lung, activation of neural reflexes, initiation of inflammation, alterations of epithelial barrier function, sensitization of bronchial smooth muscle, modification of innate and adaptive immunity, and airway remodeling. Another key event, systemic inflammation and vascular oxidative/nitrosative stress, may be critical to the extrapulmonary effects of O$_3$.

Overall, biological responses to O$_3$ exposure are similar across many species (see Section 5.5). Thus, animal studies are used to add to the understanding of the full range of potential O$_3$-mediated health effects.

Integration of Ozone Health Effects

In this ISA, the body of evidence from short-term (i.e., hours, days, weeks) or long-term (i.e., months to years) exposure studies is evaluated and integrated across relevant scientific disciplines (i.e., controlled human exposure studies, toxicology, and epidemiology) for health effects that vary in severity from minor subclinical effects to death. The results from the health studies, supported by the evidence from atmospheric chemistry and exposure assessment studies, contribute to the causal determinations made for the various health outcomes. Both the conclusions from the 2006 O$_3$ AQCD and the causality determinations from this review are summarized in Table 1-1. Additional details are provided here for respiratory and cardiovascular health effects and mortality, for which there is the strongest evidence of an effect from O$_3$; details for a wider range of health effects are provided in subsequent chapters$^1$.

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$^1$ Detailed information O$_3$ concentrations at which health effects are observed can be found in later chapters. For an overview, please see Table 2-2.
## Table 1-1  Summary of O₃ causal determinations by exposure duration and health outcome.

<table>
<thead>
<tr>
<th>Health Outcome⁴</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Short-term Exposure to O₃</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Respiratory effects</td>
<td>The overall evidence supports a causal relationship between acute ambient O₃ exposures and increased respiratory morbidity outcomes.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Cardiovascular effects</td>
<td>The limited evidence is highly suggestive that O₃ directly and/or indirectly contributes to cardiovascular-related morbidity, but much remains to be done to more fully substantiate the association.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Central nervous system effects</td>
<td>Toxicological studies report that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short and long term memory, sleep patterns, and histological signs of neurodegeneration.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>The evidence is highly suggestive that O₃ directly or indirectly contributes to non-accidental and cardiopulmonary-related mortality.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td><strong>Long-term Exposure to O₃</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Respiratory effects</td>
<td>The current evidence is suggestive but inconclusive for respiratory health effects from long-term O₃ exposure.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Cardiovascular effects</td>
<td>No conclusions in the 2006 O₃ AQCD.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Reproductive and developmental effects</td>
<td>Limited evidence for a relationship between air pollution and birth-related health outcomes, including mortality, premature births, low birth weights, and birth defects, with little evidence being found for O₃ effects.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Central nervous system effects</td>
<td>Evidence regarding chronic exposure and neurobehavioral effects was not available.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Cancer</td>
<td>Little evidence for a relationship between chronic O₃ exposure and increased risk of lung cancer.</td>
<td>Inadequate to Infer a Causal Relationship</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>There is little evidence to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
</tbody>
</table>

⁴Health effects (e.g., respiratory effects, cardiovascular effects) include a spectrum of outcomes, from measureable subclinical effects (e.g., blood pressure), to more obvious effects (e.g., medication use, hospital admissions), and cause-specific mortality. Total mortality includes all-cause (non-accidental) mortality, as well as cause-specific mortality (e.g., deaths due to heart attacks).
Respiratory Effects

The clearest evidence for health effects associated with exposure to O\textsubscript{3} is provided by studies of respiratory effects. Collectively, a very large amount of evidence spanning several decades supports a relationship between exposure to O\textsubscript{3} and a broad range of respiratory effects (see Section 6.2.9 and Section 7.2.8). The majority of this evidence is derived from studies investigating short-term exposure (i.e., hours to weeks) to O\textsubscript{3}, although animal toxicological studies and recent epidemiologic evidence demonstrate that long-term exposure (i.e., months to years) may also harm the respiratory system.

The 2006 O\textsubscript{3} AQCD concluded that there was clear, consistent evidence of a causal relationship between short-term exposure to O\textsubscript{3} and respiratory health effects. This causal association is more substantiated now by the effects observed across recent controlled human exposure, epidemiologic, and toxicological studies indicating associations between short-term O\textsubscript{3} exposures and a range of respiratory health endpoints from respiratory tract inflammation to respiratory-related emergency department (ED) visits and hospital admissions. Short-term O\textsubscript{3} exposures induced (or were associated with) statistically significant declines in lung function. An equally strong body of evidence from controlled human exposure and toxicological studies demonstrated reversible O\textsubscript{3}-induced increases in inflammatory responses, epithelial permeability, and airway hyperresponsiveness that were found to last for 18-24 hours after O\textsubscript{3} exposure. Toxicological studies in animals provided additional evidence for O\textsubscript{3}-induced impairment of host defenses. Combined, these findings from experimental studies provided support for epidemiologic evidence, in which short-term increases in O\textsubscript{3} concentration were consistently associated with increases in respiratory symptoms and asthma medication use in children with asthma, respiratory-related hospital admissions, and ED visits for chronic obstructive pulmonary disease (COPD) and asthma. Additionally, recent epidemiologic evidence supports the range of respiratory effects induced by O\textsubscript{3} by demonstrating that short-term increases in ambient O\textsubscript{3} concentrations can lead to respiratory mortality. The combined evidence from these disciplines supports the conclusion that there is a causal relationship between short-term O\textsubscript{3} exposure and respiratory effects.

Epidemiologic evidence for a relationship between long-term O\textsubscript{3} exposure and respiratory effects includes recent studies that evaluate the associations between long-term exposure to O\textsubscript{3} and respiratory effects that demonstrate interactions between exercise or different genetic variants and both new-onset asthma in children and increased respiratory symptom effects in individuals with asthma. Additional studies of respiratory health effects and a study of respiratory mortality provide a collective body of evidence supporting this relationship. Studies evaluating other pollutants provide data suggesting that the effects related to O\textsubscript{3} are independent from the effects of the other pollutants. Short-term studies provide supportive evidence with increases in respiratory symptoms and asthma medication use, hospital admissions and ED visits for all respiratory outcomes and asthma, and decreased lung function in children. Taken together, the recent epidemiologic studies of
respiratory health effects (including symptoms, new-onset asthma and mortality) combined with toxicological studies in rodents and nonhuman primates, provide biologically plausible evidence that there is likely to be a causal relationship between long-term exposure to O₃ and respiratory effects.

**Mortality Effects**

The last review concluded that the overall body of evidence was highly suggestive that short-term exposure to O₃ directly or indirectly contributes to non-accidental and cardiopulmonary-related mortality, but that additional research was needed to more fully establish the underlying mechanisms by which such effects occur. Recent multicity studies and a multicontinent study have reported associations between short-term O₃ exposure and mortality, expanding upon evidence available in the last review (see Section 6.6). These recent studies reported consistent positive associations between short-term O₃ exposure and total (nonaccidental) mortality, with associations being stronger during the warm season, when O₃ concentrations were higher. They also observed associations between O₃ exposure and cardiovascular and respiratory mortality. These recent studies also examined previously identified areas of uncertainty in the O₃-mortality relationship, and provided additional evidence supporting an association between short-term O₃ exposure and mortality. As a result, the current body of evidence indicates that there is likely to be a causal relationship between short-term exposures to O₃ and total mortality.

**Cardiovascular Effects**

In previous O₃ reviews, very few studies were available which examined the effect of short-term O₃ exposure on the cardiovascular system. New toxicological studies, although limited in number, have provided evidence of O₃-induced cardiovascular effects. These effects may, in part, correspond to changes in the autonomic nervous system or to the development and maintenance of oxidative stress and inflammation throughout the body that resulted from inflammation in the lungs. Controlled human exposure studies also suggest cardiovascular effects in response to short-term O₃ exposure, including changes in heart rate variability and blood markers of systemic inflammation and oxidative stress, which provide some coherence with the effects observed in animal toxicology studies. Collectively, the experimental studies provide initial biological plausibility for the consistently positive associations observed in epidemiologic studies of short-term O₃ exposure and cardiovascular mortality. However, studies in the epidemiologic literature generally have not observed a relationship between short-term exposure to O₃ and cardiovascular morbidity including studies that examined the association between short-term O₃ exposure and cardiovascular-related hospital admissions and ED visits and other various cardiovascular effects. The lack of coherence between the results from studies that examined associations between short-term O₃ exposure and cardiovascular morbidity and cardiovascular mortality complicate the interpretation of the overall evidence for
O₃-induced cardiovascular effects. Although there is a lack of coherence with epidemiologic studies of cardiovascular morbidity, animal toxicological studies demonstrate O₃-induced cardiovascular effects, and provide support to the strong body of evidence indicating O₃-induced cardiovascular mortality. Overall, the body of evidence indicates that there is likely to be a causal relationship between short-term exposures to O₃ and cardiovascular effects, including cardiovascular mortality.

Populations Potentially at Increased Risk

The examination of populations and lifestages potentially at increased risk for O₃ exposure identifies populations that are at increased risk for O₃-related health effects; these studies do so by examining groups within the study population, such as those with an underlying health condition or genetic variant, categories of age, race, or sex, or by developing animal models that mimic the conditions associated with a health effect. Such studies have identified a multitude of factors that could potentially contribute to whether a population is at increased risk for O₃-related health effects (see Chapter 8). The populations and lifestages identified as having adequate evidence for being at increased risk of O₃-related health effects are individuals with asthma, younger and older age groups, individuals with reduced intake of certain nutrients (i.e., Vitamins C and E), outdoor workers, and individuals having variants (including variations in multiple genes related to oxidative metabolism or inflammation). The evidence for other potential factors; sex, socioeconomic status, and obesity, is suggestive of an increased risk, but further evidence is needed.

Integration of Effects on Vegetation and Ecosystems

The most policy-relevant information pertaining to the review of the NAAQS for the effects of O₃ on vegetation and ecosystems has been evaluated and synthesized, integrating key findings about plant physiology, biochemistry, whole plant biology, ecosystems and exposure-response relationships. The welfare effects of O₃ can be observed across spatial scales, starting at the subcellular and cellular level, then the whole plant and finally, ecosystem-level processes. Ozone effects at small spatial scales, such as the leaf of an individual plant, can result in effects along a continuum of larger spatial scales. These effects include altered rates of leaf gas exchange, growth, and reproduction at the individual plant level, and can result in broad changes in ecosystems, such as productivity, carbon storage, water cycling, nutrient cycling, and community composition. The conclusions from the previous NAAQS review and the causality determinations from this review are summarized in Table 1-2 below. Further discussion of these is provided after Table 1-2 for: visible foliar injury; growth, productivity, and carbon storage; yield and quality of agricultural crops; water cycling; below-ground processes; community composition; and O₃ exposure-response relationships. Discussion of all relevant welfare effects is provided in Chapter 9.
Table 1-2 Summary of O₃ causal determination for welfare effects.

<table>
<thead>
<tr>
<th>Vegetation and Ecosystem Effects</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Visible Foliar Injury Effects on Vegetation</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause impaired aesthetic quality of many native plants and trees by increasing foliar injury.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Vegetation Growth</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause decreased growth and biomass accumulation in annual, perennial and woody plants, including agronomic crops, annuals, shrubs, grasses, and trees.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Productivity in Terrestrial Ecosystems</td>
<td>There is evidence that O₃ is an important stressor of ecosystems and that the effects of O₃ on individual plants and processes are scaled up through the ecosystem, affecting net primary productivity.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Carbon (C) Sequestration in Terrestrial Ecosystems</td>
<td>Limited studies from the 2006 O₃ AQCD.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Reduced Yield and Quality of Agricultural Crops</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause decreased yield and/or nutritive quality in a large number of agronomic and forage crops.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Ecosystem Water Cycling</td>
<td>Ecosystem water quantity may be affected by O₃ exposure at the landscape level.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Below-ground Biogeochemical Cycles</td>
<td>Ozone-sensitive species have well known responses to O₃ exposure, including altered carbon (C) allocation to below-ground tissues, and also altered rates of leaf and root production, turnover, and decomposition. These shifts can affect overall C and nitrogen (N) loss from the ecosystem in terms of respired C, and leached aqueous dissolved organic and inorganic C and N.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Community Composition</td>
<td>Ozone may be affecting above- and below-ground community composition through impacts on both growth and reproduction. Significant changes in plant community composition resulting directly from O₃ exposure have been demonstrated.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
</tbody>
</table>
Visible Foliar Injury

Visible foliar injury resulting from exposure to O$_3$ has been well characterized and documented over several decades on many tree, shrub, herbaceous and crop species. In addition, O$_3$-induced visible foliar injury symptoms on certain plant species (e.g., black cherry, yellow-poplar, and common milkweed, among others) are considered diagnostic of exposure to O$_3$, as experimental evidence has clearly established a consistent association, with greater exposure generally resulting in greater and more prevalent injury. Additional sensitive species showing visible foliar injury continue to be identified from field surveys and verified in controlled exposure studies (see Section 9.4.2). Overall, evidence is sufficient to conclude that there is a causal relationship between ambient O$_3$ exposure and the occurrence of O$_3$-induced visible foliar injury on sensitive vegetation across the U.S.

Growth, Productivity, Carbon Storage and Agriculture

Ambient O$_3$ concentrations have long been known to cause decreases in photosynthetic rates and plant growth. The O$_3$-induced effects at the plant scale may translate to the ecosystem scale, with changes in productivity and carbon (C) storage. Studies demonstrating the effects of O$_3$ exposure on photosynthesis, growth, biomass allocation, ecosystem production and ecosystem C sequestration were reviewed for natural ecosystems (see Section 9.4.3), and crop productivity and crop quality were reviewed for agricultural ecosystems (see Section 9.4.4). There is strong and consistent evidence that current ambient concentrations of O$_3$ decrease plant photosynthesis and growth in numerous plant species across the U.S.

Studies conducted during the past four decades have also demonstrated unequivocally that O$_3$ alters biomass allocation and plant reproduction. Studies at the leaf and plant scales showed that O$_3$ reduced photosynthesis and plant growth, providing coherence and biological plausibility for the reported decreases in ecosystem productivity. In addition to primary productivity, other indicators such as net ecosystem CO$_2$ exchange and C sequestration were often assessed by modeling studies. Model simulations consistently found that O$_3$ exposure caused negative impacts on those indicators, but the severity of these impacts was influenced by multiple interactions of biological and environmental factors. Although O$_3$ generally causes negative effects on ecosystem productivity, the magnitude of the response varies among plant communities. Overall, evidence is sufficient to conclude that there is a causal relationship between ambient O$_3$ exposure and reduced native plant growth and productivity, and that there is a likely causal relationship between O$_3$ exposure and reduced carbon sequestration in terrestrial ecosystems.

The detrimental effect of O$_3$ on crop production has been recognized since the 1960s, and current O$_3$ concentrations in many areas across the U.S. are high enough to cause yield loss in a variety of agricultural crops including, but not limited to, soybeans, wheat, potatoes, watermelons, beans, turnips, onions, lettuces, and tomatoes.
Continued increases in O₃ concentration may further decrease yield in these sensitive crops while also causing yield losses in less sensitive crops. Research has linked increasing O₃ concentration to decreased photosynthetic rates and accelerated aging in leaves, which are related to yield (see Section 9.4.4). Recent research has highlighted the effects of O₃ on crop quality. Increasing O₃ concentration decreases nutritive quality of grasses, decreases macro- and micro-nutrient concentrations in fruits and vegetable crops, and decreases cotton fiber quality. Evidence is sufficient to conclude that there is a causal relationship between O₃ exposure and reduced yield and quality of agricultural crops.

Water Cycling

Ozone can affect water use in plants and ecosystems through several mechanisms, including damage to stomatal functioning and loss of leaf area. Possible mechanisms for O₃ exposure effects on stomatal functioning include the build-up of CO₂ in the substomatal cavity, impacts on signal transduction pathways, and direct O₃ impact on guard cells. Regardless of the mechanism, O₃ exposure has been shown to alter stomatal performance, which can affect plant and stand transpiration and therefore has the potential to affect hydrological cycling (see Section 9.4.5). Although the direction of the response differed among studies, the evidence is sufficient to conclude that there is likely to be a causal relationship between O₃ exposure and the alteration of ecosystem water cycling.

Below Ground Processes

Below-ground processes are tightly linked with above-ground processes. The responses of above-ground process to O₃ exposure, such as reduced photosynthetic rates, increased metabolic cost, and reduced allocation of C to the roots, have provided biologically plausible mechanisms for the alteration of below-ground processes. These include altered quality and quantity of C input to soil, changes in microbial community composition, and effects on C and nutrient cycling (see Section 9.4.6). The evidence is sufficient to conclude that there is a causal relationship between O₃ exposure and the alteration of below-ground biogeochemical cycles.

Community Composition

Ozone exposure changes competitive interactions and leads to loss of O₃-sensitive species or genotypes. Studies at the plant level found that the severity of O₃ damage to growth, reproduction, and foliar injury varied among species, which provide the biological plausibility for the alteration of community composition (see Section 9.4.3 and Section 9.4.7). For example, there is a tendency for O₃ exposure to shift the biomass of grass-legume mixtures in favor of grass species. In addition, research since the last review has shown that O₃ can also alter community composition and diversity of soil microbial communities. Shifts in community composition of bacteria
and fungi have been observed in both natural and agricultural ecosystems, although no general patterns could be identified. The evidence is sufficient to conclude that there is likely to be a causal relationship between O₃ exposure and the alteration of community composition of some ecosystems.

**Air Quality Indices and Exposure-Response**

Exposure indices are metrics that quantify exposure as it relates to measured plant response (e.g., reduced growth). They are summary measures of monitored O₃ concentrations over time intended to provide a consistent metric for reviewing and comparing exposure-response effects obtained from various studies. Given the current state of knowledge and the best available data, exposure indices that cumulate and differentially weight the higher hourly average concentrations and also include the mid-level values (e.g., the W126 metric, see Section 9.5) continue to offer the most defensible approach for use in developing response functions and comparing studies, as well as for defining future indices for vegetation protection.

Previous reviews of the NAAQs have included exposure-response functions for the yield of many crop species, and for the biomass accumulation of tree species. They were based on large-scale experiments designed to obtain clear exposure-response data, and are updated by using the W126 metric to quantify exposure. In recent years, extensive exposure-response data obtained in more naturalistic settings have become available for yield of soybean and growth of aspen. The exposure-response median functions are validated based on previous data by comparing their predictions with the newer observations (see Section 9.6). The functions supply very accurate predictions of effects in naturalistic settings. Recent meta-analyses of large sets of crop and tree studies do not give rise to exposure-response functions, but their results are consistent with the functions presented in Section 9.6. Although these median functions provide reliable models for groups of species or group of genotypes within a species, the original data and recent results consistently show that some species, and some genotypes within species are much more severely affected by exposure to O₃.

**The Role of Tropospheric Ozone in Climate Change and UV-B Shielding Effects**

Atmospheric O₃ as a whole plays an important role in the Earth’s energy budget by interacting with incoming solar radiation and outgoing infrared radiation. Though tropospheric O₃ makes up only a small portion of the total amount of O₃ in the atmosphere, it has important incremental effects on the overall radiation budget. Perturbations to tropospheric O₃ concentrations can have direct effects on climate and indirect effects on health, ecology, and welfare by changing the shielding of the earth’s surface from solar ultraviolet (UV) radiation.
Radiative Forcing and Climate Change

Radiative forcing by a greenhouse gas or aerosol is a metric used to quantify the change in balance between radiation coming into and going out of the atmosphere caused by the presence of that substance. Tropospheric O\(_3\) is a major greenhouse gas and radiative forcing agent; evidence from satellite data shows a sharp dip in the outgoing infrared radiation in the 9.6 \(\mu\text{m}\) O\(_3\) absorption band. Models calculate that the global average concentration of tropospheric O\(_3\) has doubled since the pre-industrial era, while observations indicate that in some regions O\(_3\) may have increased by factors as great as 4 or 5. These increases are tied to the rise in emissions of O\(_3\) precursors from human activity, mainly fossil fuel consumption and agricultural processes. Overall, the evidence supports a causal relationship between changes in tropospheric O\(_3\) concentrations and radiative forcing.

The impact of the tropospheric O\(_3\) change since pre-industrial times on climate has been estimated to be about 25-40% of the anthropogenic CO\(_2\) impact and about 75% of the anthropogenic CH\(_4\) impact according to the Intergovernmental Panel on Climate Change (IPCC), ranking it third in importance after CO\(_2\) and CH\(_4\) according to the IPCC (see Section 10.3). There are large uncertainties in the magnitude of the radiative forcing estimate attributed to tropospheric O\(_3\), making the effect of tropospheric O\(_3\) on climate more uncertain than the effect of the longer-lived greenhouse gases. Furthermore, radiative forcing does not take into account climate feedbacks that could amplify or dampen the actual surface temperature response. Quantifying the change in surface temperature requires a complex climate simulation in which all important feedbacks and interactions are accounted for. The modeled surface temperature response to a given radiative forcing is highly uncertain and can vary greatly among models and from region to region within the same model. Even with these uncertainties, global climate models indicate that tropospheric O\(_3\) has contributed to observed changes in global mean and regional surface temperatures. As a result of such evidence presented in climate modeling studies, there is likely to be a causal relationship between changes in tropospheric O\(_3\) concentrations and effects on climate.

UV-B Shielding Effects

UV radiation emitted from the Sun contains sufficient energy when it reaches the Earth to have damaging effects on living organisms and materials (see Section 10.4). Atmospheric O\(_3\) plays a crucial role in reducing the amount of UV radiation reaching the Earth’s surface. Ozone in the stratosphere is responsible for the majority of this shielding, but O\(_3\) in the troposphere provides supplemental shielding of UV-B radiation in the mid-wavelength band (280-315 nm), thereby potentially reducing UV-B related human and ecosystem health effects and materials damage. EPA has found no published studies that adequately examine the incremental health or welfare effects (adverse or beneficial) attributable specifically to changes in UV-B exposure resulting from perturbations in tropospheric O\(_3\) concentrations. While the effects are expected to be small, they cannot yet be critically assessed within reasonable
uncertainty. Overall, the evidence is inadequate to determine if a causal relationship exists between changes in tropospheric O\textsubscript{3} concentrations and effects on health and welfare related to UV-B shielding.

The conclusions from the previous NAAQS review and the causality determinations from this review relating climate change and UV-B shielding effects are summarized in the table below (Table 1-3), with details provided in Chapter 10.

<table>
<thead>
<tr>
<th>Effects</th>
<th>Conclusions from 2006 O\textsubscript{3} AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Radiative Forcing</td>
<td>The 2006 O\textsubscript{3} AQCD concluded that climate forcing by O\textsubscript{3} at the regional scale may be its most important impact on climate.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Climate Change</td>
<td>While more certain estimates of the overall importance of global-scale forcing due to tropospheric O\textsubscript{3} await further advances in monitoring and chemical transport modeling, the overall body of scientific evidence reviewed in the 2006 O\textsubscript{3} AQCD suggests that high concentrations of O\textsubscript{3} on the regional scale could have a discernible influence on climate, leading to surface temperature and hydrological cycle changes.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Health and Welfare Effects Related to UV-B Shielding</td>
<td>UV-B has not been studied in sufficient detail to allow for a credible health benefits assessment. In conclusion, the effect of changes in surface-level O\textsubscript{3} concentrations on UV-induced health outcomes cannot yet be critically assessed within reasonable certainty.</td>
<td>Inadequate to Determine if a Causal Relationship Exists</td>
</tr>
</tbody>
</table>

**Conclusion**

The clearest evidence for human health effects associated with exposure to O\textsubscript{3} is provided by studies of respiratory effects. Collectively, there is a very large amount of evidence spanning several decades in support of a causal association between exposure to O\textsubscript{3} and a broad range of respiratory effects, indicating that there is a causal relationship between short-term exposures to O\textsubscript{3} and respiratory effects. The majority of this evidence is derived from studies investigating short-term O\textsubscript{3} exposure (i.e., hours to weeks), although animal toxicological studies and recent epidemiologic evidence demonstrate that long-term exposure (i.e., months to years) are likely to be detrimental to the respiratory system. Additionally, consistent positive associations between short-term O\textsubscript{3} exposure and total (nonaccidental) mortality have helped to resolve previously identified areas of uncertainty in the O\textsubscript{3}-mortality relationship, indicating that there is likely to be a causal relationship between short-term exposures to O\textsubscript{3} and total mortality. Taken together, the recent epidemiologic studies of respiratory health effects (including respiratory symptoms, new-onset asthma and respiratory mortality) combined with toxicological
studies in rodents and nonhuman primates, provide biologically plausible evidence that there is likely to be a causal relationship between long-term exposure to O₃ and respiratory effects. Animal toxicological studies demonstrate O₃-induced cardiovascular effects, and provide support to the strong body of evidence indicating O₃-induced cardiovascular mortality, which together indicate that there is likely to be a causal relationship between short-term exposure to O₃ and cardiovascular effects. Recent evidence is suggestive of a causal relationship between long-term O₃ exposures and total mortality. The evidence for these health effects indicates that the relationship between concentration and response is linear along the range of O₃ concentrations observed in the U.S., with no indication of a threshold within that range. However, there is less certainty in the shape of the concentration-response curve at O₃ concentrations generally below 20 ppb. The populations identified as having increased risk of O₃-related health effects are individuals with asthma, younger and older age groups, individuals with certain dietary deficiencies, and outdoor workers.

There has been over 40 years of research on the effects of O₃ exposure on vegetation and ecosystems. The best evidence for effects is from controlled exposure studies. These studies have clearly shown that exposure to O₃ is causally linked to visible foliar injury, decreased photosynthesis, changes in reproduction, and decreased growth. Recently, studies at larger spatial scales support the results from controlled studies and indicate that ambient O₃ exposures can affect ecosystem productivity, crop yield, water cycling, and ecosystem community composition. And on a global scale, tropospheric O₃ is the third most important greenhouse gas, making it likely to play an important role in climate change.
2 INTEGRATIVE SUMMARY

This Integrated Science Assessment (ISA) forms the scientific foundation for the review of the current national ambient air quality standards (NAAQS) for ozone (\(O_3\)). The ISA is a concise evaluation and synthesis of the most policy-relevant science—and it communicates critical science judgments relevant to the review of the NAAQS for \(O_3\). The ISA accurately reflects “the latest scientific knowledge useful in indicating the kind and extent of identifiable effects on public health or welfare which may be expected from the presence of [a] pollutant in ambient air” (CAA, 1990a). Key information and judgments contained in prior Air Quality Criteria Documents (AQCD) for \(O_3\) are incorporated into this assessment. Additional details of the pertinent scientific literature published since the last review, as well as selected earlier studies of particular interest, are included. This ISA thus serves to update and revise the evaluation of the scientific evidence available at the time of the completion of the 2006 \(O_3\) AQCD (U.S. EPA, 2006b). The current primary \(O_3\) standard includes an 8-hour (h) average (avg) standard set at 75 parts per billion (ppb). The secondary standard for \(O_3\) is set equal to the primary standard. Further information on the legislative and historical background for the \(O_3\) NAAQS is contained in the Preface to this ISA.

This chapter summarizes and synthesizes the available scientific evidence and is intended to provide a concise synopsis of the ISA conclusions and findings that best inform consideration of the policy-relevant questions that frame this assessment (U.S. EPA, 2009c). It includes:

- An integration of the evidence on the health effects associated with short- and long-term exposure to \(O_3\), discussion of important uncertainties identified in the interpretation of the scientific evidence, and an integration of health evidence from the different scientific disciplines and exposure durations.
- An integration of the evidence on the welfare effects associated with exposure to \(O_3\), including those associated with vegetation and ecosystems, and discussion of important uncertainties identified in the interpretation of the scientific evidence.
- Discussion of policy-relevant considerations, such as potentially at-risk populations and concentration-response relationships and how they inform selection of appropriate exposure metrics/indices.

2.1 ISA Development and Scope

EPA has developed a robust, consistent, and transparent process for evaluating the scientific evidence and drawing conclusions and causal judgments regarding air pollution-related health and environmental effects. The ISA development process includes literature search strategies, criteria for selecting and evaluating studies,
approaches for evaluating the weight of the evidence, and a framework for making causality determinations. The process and causality framework are described in more detail in the Preamble to the ISA. This section provides a brief overview of the process for development of this ISA.

EPA initiated the current review of the NAAQS for O₃ on September 29, 2008, with a call for information from the public (U.S. EPA, 2008g). Literature searches have been conducted routinely since then to identify studies published since the last review, focusing on studies published from 2005 (closing date for the previous scientific assessment) through July 2011. References that were considered for inclusion in this ISA can be found using the HERO website (http://hero.epa.gov/ozone). This site contains HERO links to lists of references that are cited in the ISA, as well as those that were considered for inclusion but not cited in the ISA, with bibliographic information and abstracts.

This review has endeavored to evaluate all relevant data published since the last review; this includes studies pertaining to the atmospheric science of O₃, human exposure to ambient O₃, and health, ecological, climate and UV-B shielding effects studies. These include studies that are related to concentration-response relationships, mode(s) of action (MOA), and understanding of at-risk populations for effects of O₃ exposure. Added to the body of research were EPA’s analyses of air quality and emissions data, and studies on atmospheric chemistry, transport, and fate of these emissions using measurements and chemistry-transport models (CTMs).

Previous O₃ AQCDs (U.S. EPA, 2006b, 1996a, b, 1984, 1978a) have included an extensive body of evidence on both health and welfare effects of O₃ exposure, as well as an understanding of the atmospheric chemistry of O₃ (U.S. EPA, 2006b). In this ISA, the conclusions and key findings from previous reviews are summarized at the beginning of each section, to provide the foundation for consideration of evidence from recent studies. Results of key studies from previous reviews are included in discussions or tables and figures, as appropriate, and conclusions are drawn based on the synthesis of evidence from recent studies with the extensive literature summarized in previous reviews.

The Preamble discusses the general framework for conducting the science assessment and developing an ISA, including criteria for evaluating studies and developing scientific conclusions. For selection of epidemiologic studies in the O₃ ISA, particular emphasis is placed on those studies most relevant to the review of the NAAQS. Studies conducted in the United States (U.S.) or Canada are discussed in more detail than those from other geographical regions, and in regard to human health, particular emphasis is placed on: (1) recent multi-city studies that employ standardized analysis methods for evaluating effects of O₃ and that provide overall estimates for effects, based on combined analyses of information pooled across multiple cities; (2) studies that help understand quantitative relationships between exposure concentrations and effects; (3) new studies that provide evidence on effects in at-risk populations; and (4) studies that consider and report O₃ as a component of a complex mixture of air pollutants. In evaluating toxicological and controlled human exposure studies, emphasis is placed on studies using concentrations that are
generally no greater than about an order of magnitude higher than ambient \( \text{O}_3 \) concentrations currently found in many parts of the United States. Consideration of studies important for evaluation of human exposure to ambient \( \text{O}_3 \) places emphasis on those evaluating the relationship between \( \text{O}_3 \) measured at central site monitors and personal exposure to ambient \( \text{O}_3 \). Important factors affecting this relationship include spatial and temporal variations in ambient \( \text{O}_3 \) concentration, and time spent outdoors, since penetrations of \( \text{O}_3 \) into indoor environments may be limited.

This ISA uses a five-level hierarchy that classifies the weight of evidence for causation:

- Causal relationship
- Likely to be a causal relationship
- Suggestive of a causal relationship
- Inadequate to infer a causal relationship, or
- Not likely to be a causal relationship

Beyond judgments regarding causality are questions relevant to quantifying health or environmental risks based on the understanding of the quantitative relationships between pollutant exposures and health or welfare effects. Once a determination is made regarding the causal relationship between the pollutant and outcome category, important questions regarding quantitative relationships include:

- What is the concentration-response, exposure-response, or dose-response relationship?
- Under what exposure conditions (dose or concentration, duration, and pattern) are effects observed?
- What populations or lifestages appear to be differentially affected, i.e., at increased risk of \( \text{O}_3 \)-related health effects?
- What elements of the ecosystem (e.g., types, regions, taxonomic groups, populations, functions, etc.) appear to be affected or are more sensitive to effects?

This chapter summarizes and integrates the newly available scientific evidence that best informs consideration of the policy-relevant questions that frame this assessment. Section 2.2 discusses the trends in ambient concentrations and sources of \( \text{O}_3 \) and provides a brief summary of ambient air quality for past and current short- and long-term exposure durations. Section 2.3 presents the evidence regarding personal exposure to ambient \( \text{O}_3 \) in outdoor and indoor microenvironments, and it discusses the relationship between ambient \( \text{O}_3 \) concentrations and personal exposure to ambient \( \text{O}_3 \). Section 2.4 provides a discussion of the dosimetry and modes of action evidence for \( \text{O}_3 \) exposure. Section 2.5 integrates the evidence for studies that examine the health effects associated with short- and long-term exposure to \( \text{O}_3 \) and discusses important uncertainties identified in the interpretation of the scientific evidence. A discussion of policy-relevant considerations, such as potentially at-risk populations, and the \( \text{O}_3 \) concentration-response relationship is also included in
Section 2.5. Finally, Section 2.6 summarizes the evidence for welfare effects related to O₃ exposure, and Section 2.7 reviews the literature on the influence of tropospheric O₃ on climate and exposure to solar ultraviolet radiation.

2.2 Atmospheric Chemistry and Ambient Concentrations

2.2.1 Physical and Chemical Processes

Ozone in the troposphere is a secondary pollutant; it is formed by reactions of precursor gases and is not directly emitted from specific sources. Ozone precursor gases originate from both anthropogenic (i.e., man-made) and natural source categories. Ozone attributed to anthropogenic sources is formed in the atmosphere by photochemical reactions involving sunlight and precursor pollutants including volatile organic compounds (VOCs), nitrogen oxides \((\text{NO}_X)\), and carbon monoxide (CO). Ozone attributed to natural sources is formed through similar photochemical reactions involving natural emissions of precursor pollutants from vegetation, microbes, animals, biomass burning and lightning. An absolute distinction between natural and anthropogenic sources of O₃ precursors is often difficult to make in practice, as human activities affect directly or indirectly emissions from what are considered to be natural sources.

Ozone is present not only in polluted urban atmospheres but throughout the troposphere, even in remote areas of the globe. The same basic processes involving sunlight-driven reactions of \(\text{NO}_X\), VOCs and CO that occur in polluted urban air also contribute to O₃ formation throughout the troposphere. In urban areas, \(\text{NO}_X\), VOCs, and CO are all important precursors to O₃ formation. In non-urban areas, biogenic VOCs emitted from vegetation tend to be the most important precursors to O₃ formation. In remote areas with little or no vegetation, and in general above the planetary boundary layer (PBL, extending typically from 1 to 3 km above the surface), methane—structurally the simplest VOC—and CO are the main carbon-containing precursors to O₃ formation. Ozone is subsequently removed from the atmosphere through a number of gas phase reactions and deposition to surfaces.

Convective processes and turbulence transport O₃ and other pollutants both upward and downward throughout the PBL and the free troposphere above. If pollutants are transported into the free troposphere above the PBL where winds are generally much stronger than in the PBL, they can be transported over longer distances than they can if they remained near the surface. Conversely, pollutants transported downward into the PBL can add to pollution burdens there. The transport of pollutants downwind of major urban centers is characterized by the development of urban plumes. Meteorological conditions, small-scale circulation patterns induced by surface characteristics and structures, localized chemistry, and topographic barriers can

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1 These processes also lead to the formation of other photochemical products, such as peroxyacetyl nitrate, nitric acid, and sulfuric acid, and to other compounds, such as formaldehyde and other carbonyl compounds.
influence mixing on the intra-urban scale, resulting in variability in O₃ concentrations across individual urban areas.

Apart from issues in understanding issues in O₃ formation in areas such as the Houston-Galveston-Brazoria airshed and the Northeast Corridor that are largely the result of local pollution sources during summer, other issues have become apparent in the past several years. Photochemically produced O₃ concentrations that exceed the level of the NAAQS have been observed in the Intermountain West in oil and gas fields during specific meteorological conditions (i.e., fresh snow cover, low mixing layer height trapping emissions) during winter. Because the mean tropospheric lifetime of O₃ is a few weeks, O₃ can be transported from continent to continent. Locations at high elevations are most susceptible to the intercontinental transport of pollution, particularly during spring. Intrusions of stratospheric air containing high O₃ may also cause, or contribute significantly to, exceedances of levels of the NAAQS for O₃. These events occur mainly in the West during spring.

### 2.2.2 Background O₃ Concentrations

In the context of a review of the NAAQS, it is useful to define background O₃ concentrations in a way that distinguishes between concentrations that result from precursor emissions that are relatively less controllable from those that are relatively more controllable through U.S. policies. For this assessment, three definitions of background O₃ concentrations are considered, including (1) United States (U.S.) background (simulated O₃ concentrations that would exist in the absence of anthropogenic emissions from the U.S.), (2) North American (NA) background (simulated O₃ concentrations that would exist in the absence of anthropogenic emissions from the U.S., Canada and Mexico), and (3) natural background (simulated O₃ concentrations in the absence of all anthropogenic emissions globally). Each definition of background O₃ includes contributions resulting from emissions from natural sources (e.g., stratospheric intrusions, wildfires, biogenic methane and more short-lived VOC emissions) throughout the globe. Differences among these definitions reflect differences in the inclusion of geographic regions that are sources of anthropogenic precursors. These definitions are used to inform policy considerations regarding the current or potential alternative standards. Note also there is no chemical difference between background O₃, and O₃ attributable to U.S. or North American anthropogenic sources.

It is important to note that since background O₃ concentrations as defined above are a hypothetical construct that cannot be directly measured, the range of background O₃ concentrations must be estimated using CTMs. This is because observations within the U.S.—even at relatively remote monitoring sites—can be impacted by transport from anthropogenic sources within the U.S. or within the rest of North America (if a North American background is adopted).

Figure 2-1 shows spring and summer mean, maximum daily 8-h average O₃ concentrations simulated by the GEOS-Chem global CTM for the base case (top
panel), which includes all sources of \( \text{O}_3 \) precursors and can be compared to observations; the U.S. background (middle panel) and the North American (NA) background (bottom panel). Seasonal mean values for the entire continental United States are shown above each panel. In general, for most areas of the United States simulated seasonal mean 8-h max \( \text{O}_3 \) concentrations in the base case (top panel) are within a few ppb of those observed at rural Clean Air Status and Trends Network (CASTNET) monitoring sites, for which details are given in the next section. In moving from the top to the bottom panel, it can be seen that calculated concentrations decrease. Note also that mean base case concentrations (including U.S., Canadian and Mexican sources) are higher in the summer than in the spring, but that mean U.S. background and NA background concentrations are higher in spring than in summer, reflecting the increased importance of sources such as intercontinental transport and stratospheric intrusions.

As can be seen from the middle and bottom panels in Figure 2-1, estimated U.S. background and NA background concentrations tend to be higher in the West (particularly in the Intermountain West) and in the Southwest compared to the East in both spring and summer. U.S. background and NA background concentrations tend to be highest in the Southwest during summer in the GEOS-Chem model, and are driven in large part by lightning NO\(_X\). As can be seen from Figure 2-1 (middle panel), highest U.S. background concentrations (in the U.S.) are found over the Northern Tier of New York State. High values are also found in other areas bordering Canada and Mexico. NA background concentrations (bottom panel) are on average ~3 ppb higher than U.S. background concentrations (middle panel) during spring and summer across the United States. For March through August 2006, mean NA background \( \text{O}_3 \) concentrations of 29 ± 8 ppb at low elevation (<1,500 meters) and 40 ± 8 ppb at high elevation (>1,500 meters) were calculated. Corresponding natural background concentrations (not shown) range from 18 ± 6 ppb to 27 ± 6 ppb. It should also be noted that methane is an important contributor to NA background \( \text{O}_3 \), accounting for slightly less than half of the increase in background since the pre-industrial era and whose relative contribution is projected to grow in the future.

Note that the calculations of background concentrations presented in Chapter 3 were formulated to answer the question, “what would \( \text{O}_3 \) concentrations be if there were no anthropogenic sources.” This is different from asking, “how much of the \( \text{O}_3 \) measured or simulated in a given area is due to background contributions.” Because of potentially strong non-linearities—particularly in many urban areas—these estimates should not be used by themselves to answer the second question posed above. The extent of these non-linearities will generally depend on location and time, the strength of concentrated sources, and the nature of the chemical regime.
Note: Seasonal mean daily maximum 8-h avg $O_3$ concentrations were calculated by GEOS-Chem for the base case (top, Base), United States background (middle, USB) and North American Background (bottom, NAB). Values in parentheses (above each map) refer to continental U.S. means, and are shown in the color bar as black squares for summer and white squares for spring.

Source: Adapted from Zhang et al. (2011).

**Figure 2-1** Mean daily average maximum 8-h avg $O_3$ concentrations in surface air, for spring and summer 2006.
2.2.3 Monitoring

The federal reference method (FRM) for \( \text{O}_3 \) measurement is based on the detection of chemiluminescence resulting from the reaction of \( \text{O}_3 \) with ethylene gas. However, almost all of the state and local air monitoring stations (SLAMS) that reported data to the EPA’s Air Quality System (AQS) database from 2005 to 2009 used the federal equivalence method (FEM) UV absorption photometer. Relative to FRMs, FEMs must satisfy precision and bias requirements to be accepted as alternative methods for sampling and analyzing ambient air. More than 96% of \( \text{O}_3 \) monitors met these precision and bias requirements for designation as an FEM during this period.

In 2010, there were 1,250 SLAMS \( \text{O}_3 \) monitors reporting data to AQS. Ozone monitoring is required at SLAMS sites during the local “ozone season” which varies by state. In addition, National Core (NCore) is a new multipollutant monitoring network implemented to meet multiple monitoring objectives and each state is required to operate at least one NCore site. The NCore network consists of 60 urban and 20 rural sites nationwide (see Figure 3-21 and Figure 3-22). The densest concentrations of \( \text{O}_3 \) sites are located in California and the eastern half of the U.S.

The Clean Air Status and Trends Network (CASTNET) is a regional monitoring network established to assess trends in acidic deposition and also provides concentration measurements of \( \text{O}_3 \). CASTNET \( \text{O}_3 \) monitors operate year round and are primarily located in rural areas; in 2010, there were 80 CASTNET sites reporting \( \text{O}_3 \) data to AQS. The National Park Service (NPS) operates 23 CASTNET sites in national parks and other Class-i areas, and provided data to AQS from 20 additional Portable Ozone Monitoring Systems (POMS) in 2010 (see Figure 3-22). Compared to urban-focused monitors, rural-focused monitors are relatively scarce across the U.S.

2.2.4 Ambient Concentrations

Ozone is the only photochemical oxidant other than \( \text{NO}_2 \) that is routinely monitored and for which a comprehensive database exists. Other photochemical oxidants are typically only measured during special field studies. The concentration analyses in Chapter 3 are limited to widely available \( \text{O}_3 \) data obtained directly from AQS for the period from 2007 to 2009. The median 24-h average, 8-h daily max, and 1-h daily max \( \text{O}_3 \) concentrations across all U.S. sites reporting data to AQS between 2007 and 2009 were 29, 40, and 44 ppb, respectively.

To investigate \( \text{O}_3 \) variability in urban areas across the U.S., 20 combined statistical areas (CSAs) were selected for closer analysis based on their importance in \( \text{O}_3 \) epidemiology studies and on their location. Several CSAs had relatively little spatial variability in daily maximum 8-h avg \( \text{O}_3 \) concentrations (e.g., inter-monitor correlations ranging from 0.61 to 0.96 in the Atlanta, GA, CSA) while other CSAs exhibited considerably more variability in \( \text{O}_3 \) concentrations (e.g., inter-monitor correlations ranging from -0.06 to 0.97 in the Los Angeles, CA, CSA). Uncertainties
resulting from the spatial variability in O\textsubscript{3} concentration fields should be considered when using data from the network of ambient O\textsubscript{3} monitors to approximate community-scale exposures, since community exposure may not be well-represented when monitors cover large areas with multiple subcommunities having different sources and topographies. However, studies evaluating the influence of monitor selection on epidemiologic study results have found that O\textsubscript{3} effect estimates are similar across different spatial scales and monitoring sites (Section 4.6.2.1).

To investigate O\textsubscript{3} variability in rural settings across the U.S., six focus areas were selected for closer analysis based on the impact of O\textsubscript{3} or O\textsubscript{3} precursor transport from upwind urban areas. The selected rural focus area with the largest number of available AQS monitors was Great Smoky Mountains National Park where the May-September median 8-h daily max O\textsubscript{3} concentration ranged from 47 ppb at the lowest elevation (564 meters) site to 60 ppb at the highest elevation (2,021 meters) site. Correlations between sites within each rural focus area ranged from 0.78 to 0.92. Ozone in rural areas is produced from emissions of O\textsubscript{3} precursors emitted directly within the rural areas, from emissions in urban areas that react and chemically transform during transport, and from occasional stratospheric intrusions. Factors contributing to variations observed within these rural focus areas include proximity to local O\textsubscript{3} precursor emissions, local scale circulations related to topography, and possibly stratospheric intrusions as a function of elevation. In addition, O\textsubscript{3} tends to persist longer in rural than in urban areas as a result of less chemical scavenging from other primary pollutants. This results in a more uniform O\textsubscript{3} concentration throughout the day and night without the typical nocturnal decrease in O\textsubscript{3} concentration observed in urban areas. Persistently high O\textsubscript{3} concentrations observed at many of the rural sites investigated here indicate that cumulative exposures for humans and vegetation in rural areas can frequently exceed cumulative exposures in urban areas.

Nation-wide surface-level O\textsubscript{3} concentrations have declined over the last decade, with a particularly noticeable decrease between 2003 and 2004 coinciding with NO\textsubscript{X} emissions reductions resulting from implementation of the NO\textsubscript{X} State Implementation Plan (SIP) Call rule, which began in 2003 and was fully implemented in 2004. This rule was designed to reduce NO\textsubscript{X} emissions from power plants and other large combustion sources in the eastern United States. The largest density of individual monitors showing downward trends in O\textsubscript{3} concentrations over the last decade occur in the Northeast where this rule was focused. In addition to a downward trend, the nation-wide surface-level O\textsubscript{3} concentration data also show a general tightening of the distribution across sites. In contrast to the majority of U.S. surface-level monitors reporting downward trends, a few surface-level monitors and elevated observations along the Pacific Coast have shown increases in O\textsubscript{3} concentrations in recent years, possibly resulting from intercontinental transport from Asia. As noted in the 2006 O\textsubscript{3} AQCD (U.S. EPA, 2006b), trends in national parks and rural areas are similar to nearby urban areas, reflecting the regional nature of O\textsubscript{3} pollution.
Since O₃ is a secondary pollutant, it is not expected to be highly correlated with primary pollutants such as CO and NOₓ. Furthermore, O₃ formation is strongly influenced by meteorology, entrainment, and transport of both O₃ and O₃ precursors, resulting in a broad range in correlations with other pollutants which can vary substantially with season. Temporal relationships between 8-h daily max O₃ and other criteria pollutants exhibit mostly negative correlations in the winter and mostly positive correlations in the summer. As a result, statistical analyses that may be sensitive to correlations between copollutants need to take seasonality into consideration, especially when O₃ is being investigated.

2.3 Human Exposure

The widespread presence of O₃ in the environment results in exposure as people participate in normal daily activities. Personal exposure measurements have been found to be moderately associated with fixed-site ambient O₃ concentrations, although a number of factors affect the relationship between ambient concentration and personal exposure. These include: infiltration of ambient O₃ into indoor microenvironments, which is driven by air exchange rate; time spent outdoors and activity pattern, which includes changes in personal behavior by some populations to avoid exposure to O₃; and influences of lifestage; and the variation in O₃ concentrations at various spatial and temporal scales. Personal exposure to O₃ is moderately correlated with ambient O₃ concentration, as indicated by studies reporting correlations generally in the range of 0.3-0.8 (Table 4-2). This suggests that ambient monitor concentrations are representative of day-to-day changes in personal exposure to ambient O₃. Some studies report lower personal-ambient correlations, a result attributable in part to low building air exchange rates and O₃ concentrations below the personal sampler detection limit. Low correlations can also occur for individuals or populations spending increased time indoors. In contrast to correlation, which represents the temporal association between exposure and concentration, the magnitude of exposure can be represented as the ratio between personal exposure and ambient concentration. This ratio varies widely depending on activity patterns, housing characteristics, and season. Personal-ambient ratios are typically 0.1-0.3 for sampling durations of several hours to several days, although individuals spending substantial time outdoors (e.g., outdoor workers) have shown much higher ratios (0.5-0.9) (Table 4-3). Since there are relatively few indoor sources of O₃, and because of reactions of O₃ with indoor surfaces and airborne constituents, indoor O₃ concentrations are often substantially lower than outdoor concentrations (Section 4.3.2). The lack of indoor sources also suggests that fluctuations in ambient O₃ may be primarily responsible for changes in personal exposure, even under low-ventilation, low-concentration conditions.

Another factor that can influence the pattern of exposure is the tendency for people to avoid O₃ exposure by altering their behavior (e.g., reducing outdoor activity levels or time spent being active outdoors) on high-O₃ days. Activity pattern has a substantial effect on ambient O₃ exposure, with time spent outdoors contributing to increased
exposure (Section 4.4.2). Air quality alerts and public health recommendations induce reductions in time spent outdoors on high-O₃ days among some lifestages and populations, particularly for children, older adults, and people with respiratory problems. Such effects are less pronounced in the general population. Limited evidence from an epidemiologic study conducted in the 1990s in Los Angeles, CA reports increased asthma hospital admissions among children and older adults when O₃ alert days (1-h max O₃ concentration >200 ppb) were excluded from the analysis of daily hospital admissions and O₃ concentrations (presumably thereby eliminating averting behavior based on high O₃ forecasts). The lower rate of admissions observed when alert days were included in the analysis suggests that estimates of health effects based on concentration-response functions that do not account for averting behavior may be biased toward the null.

Variations in O₃ concentrations occur over multiple spatial and temporal scales. Near roadways, O₃ concentrations are reduced due to reaction with NO and other species (Section 4.3.4.2). Over spatial scales of a few kilometers and away from roads, O₃ can be somewhat more homogeneous due to its formation as a secondary pollutant, while over scales of tens of kilometers, additional atmospheric processing can result in higher concentrations downwind of an urban area. Although local-scale variability impacts the magnitude of O₃ concentrations, O₃ formation rates are influenced by factors that vary over larger spatial scales, such as temperature (Section 3.2), suggesting that urban monitors may track one another temporally, but miss small-scale variability. This variation in concentrations changes the pattern of exposure people experience as they move through different microenvironments and affects the magnitude of exposures in different locations within an urban area. The various factors affecting exposure patterns and quantification of exposure result in uncertainty which can contribute to exposure measurement error in epidemiologic studies, which typically use fixed-site monitor data as an indicator of exposure. Low personal-ambient ratios result in attenuation of the magnitude of the exposure-based effect estimate or response function relative to the concentration-based response function, although the statistical association is similar for concentration- and exposure-based effect estimates if the ratio is approximately constant over time. Low personal-ambient correlations are a source of exposure error for epidemiologic studies, tending to obscure the presence of potential thresholds, bias effect estimates toward the null, and widen confidence intervals, and this impact may be more pronounced among populations spending substantial time indoors. The impact of this exposure error may tend more toward widening confidence intervals than biasing effect estimates, since epidemiologic studies evaluating the influence of monitor selection indicate that effect estimates are similar across different spatial averaging scales and monitoring sites. In addition, in examinations of respiratory endpoints in epidemiologic studies, associations were similar in magnitude across analyses using several different concentration metrics to estimate exposure, such as on-site measurements, closest-site measurements, and multi-site averages. These exposure estimation methods likely vary in how well ambient O₃ concentrations represent personal exposures and between-subject variability in exposures. Respiratory effects were observed with ambient O₃ concentrations found to have stronger personal-ambient relationships, including those measured on-site during long periods of
outdoor activity. However, such effects were also found with ambient O\textsubscript{3} measurements expected to have weaker personal-ambient relationships, including those measured at home or school, measured at the closest site, averaged from multiple community sites, and measured at a single site. Overall, there was no clear indication that a particular method of exposure estimation produced stronger findings.

### 2.4 Dosimetry and Mode of Action

Upon inspiration, O\textsubscript{3} uptake in the respiratory tract is affected by a number of factors including respiratory tract morphology, and breathing route, frequency, and volume. Additionally, physicochemical properties of O\textsubscript{3} itself and how it is transported, as well as the physical and chemical properties of the extracellular lining fluid (ELF) and tissue layers in the respiratory tract can influence O\textsubscript{3} uptake. Experimental studies and models have suggested that there are differences between the total absorption of O\textsubscript{3} from the inhaled air and the O\textsubscript{3} dose reaching the respiratory tract tissues. The total O\textsubscript{3} absorption gradually decreases with distal progression into the respiratory tract allowing for a proportionally large concentration of O\textsubscript{3} to be absorbed in the upper respiratory tract; thus, causing the nasal membranes to be a potential target site of O\textsubscript{3}-induced injury. In contrast, the primary site of O\textsubscript{3} delivery to the lung epithelium is believed to be the centriacinar region or the junction of the conducting airways with the gas exchange region. In addition, as a large concentration of the O\textsubscript{3} absorbed by the respiratory tract is absorbed in the upper respiratory tract, the nasal membranes are another potential site of O\textsubscript{3}-induced injury.

Ozone uptake is sensitive to a number of factors including tidal volume, breathing frequency, O\textsubscript{3} concentration, and exposure time. Interindividual variability also accounts for a large amount of the variability in local dose due to differences in pulmonary physiology, anatomy, and biochemistry. An increase in tidal volume and breathing frequency are both associated with increased physical activity. These changes and a switch to oronasal breathing during exercise result in deeper penetration of O\textsubscript{3} into the lower respiratory tract in part due to less oral versus nasal uptake efficiency. For these reasons, increased physical activity acts to move the maximum tissue dose of O\textsubscript{3} distally in the respiratory tract and more into the alveolar region.

The ELF is a complex mixture of lipids, proteins, and antioxidants that serves as the first barrier and target for inhaled O\textsubscript{3} (see Figure 5-7). Distinct products with diverse reactivity (i.e., secondary oxidation products), are mainly formed by reactions of O\textsubscript{3} with soluble ELF components. The thickness of the ELF is an important determinant of the dose of O\textsubscript{3} to the tissues; a thicker ELF generally results in a lower dose of O\textsubscript{3} to the tissues. Additionally, the quenching ability and the concentrations of antioxidants and other ELF components are determinants of the formation of secondary oxidation products. These reactions appear to limit interaction of O\textsubscript{3} with underlying tissues and to reduce penetration of O\textsubscript{3} distally into the respiratory tract.
In addition to contributing to the driving force for O₃ uptake, formation of secondary oxidation products contributes to oxidative stress which can lead to cellular injury and altered cell signaling in the respiratory tract. Secondary oxidation products initiate pathways (see Figure 5-8) that provide the mechanistic basis for short- and long-term health effects described in detail in Chapters 6 and 7. Other key events involved in the mode of action of O₃ in the respiratory tract include the activation of neural reflexes, initiation of inflammation, alterations of epithelial barrier function, sensitization of bronchial smooth muscle, modification of innate and adaptive immunity, and airways remodeling. Another key event, systemic inflammation and vascular oxidative/nitrosative stress, may be critical to the extrapulmonary effects of O₃.

Secondary oxidation products can transmit signals to respiratory tract cells resulting in the activation of neural reflexes. Nociceptive sensory nerves mediate the involuntary truncation of respiration, resulting in decreases in lung function (i.e., FVC, FEV₁, and tidal volume), and pain upon deep inspiration. Studies implicate TRPA1 receptors on bronchial C-fibers in this reflex. Another neural reflex involves vagal sensory nerves, which mediate a mild increase in airways obstruction (i.e., bronchoconstriction) following exposure to O₃ via parasympathetic pathways. Substance P release from bronchial C-fibers and the SP-NK receptor pathway can also contribute to this response.

Secondary oxidation products also initiate the inflammatory cascade following exposure to O₃. Studies have implicated eicosanoids, chemokines and cytokines, vascular endothelial adhesion molecules, and tachykinins in mediating this response. Inflammation is characterized by airways neutrophilia as well as the influx of other inflammatory cell types. Recent studies demonstrate a later phase of inflammation characterized by increased numbers of macrophages, which is mediated by hyaluronan. Inflammation further contributes to O₃-induced oxidative stress.

Alteration of the epithelial barrier function of the respiratory tract also occurs as a result of O₃-induced secondary oxidation product formation. Increased epithelial permeability may lead to enhanced sensitization of bronchial smooth muscle, resulting in airways hyperresponsiveness (AHR). Neurally-mediated sensitization also occurs and is mediated by cholinergic postganglionic pathways and bronchial C-fiber release of substance P. Recent studies implicate hyaluronan and Toll-like receptor 4 (TLR4) signaling in bronchial smooth muscle sensitization, while earlier studies demonstrate roles for eicosanoids, cytokines, and chemokines.

Evidence is accumulating that exposure to O₃ modifies innate and adaptive immunity through effects on macrophages, monocytes, and dendritic cells. Enhanced antigen presentation, adjuvant activity, and altered responses to endotoxin have been demonstrated. TLR4 signaling contributes to some of these responses. Effects on innate and adaptive immunity can result in both short- and longer-term consequences related to the exacerbation and/or induction of asthma and to alterations in host defense.
Airways remodeling has been demonstrated following chronic and/or intermittent exposure to O₃ by mechanisms that are not well understood. However, the TGF-β signaling pathway has recently been implicated in O₃-induced deposition of collagen in the airways wall. These studies were conducted in adult animal models and their relevance to effects in humans is unknown.

Evidence is also accumulating that O₃ exposure results in systemic inflammation and vascular oxidative/nitrosative stress. The release of diffusible mediators from the O₃-exposed lung into the circulation can initiate or propagate inflammatory responses in the vascular or in systemic compartments. This may provide a mechanistic basis for extrapulmonary effects, such as vascular dysfunction.

Both dosimetric and mechanistic factors contribute to the understanding of inter-individual variability in response. Inter-individual variability is influenced by variability in respiratory tract volume and thus surface area, breathing route, certain genetic polymorphisms, pre-existing conditions and disease, nutritional status, lifestages, attenuation, and co-exposures. In particular, very young individuals may be sensitive to developmental effects of O₃ since studies in animal models demonstrated altered development of lung and immune system.

Some of these factors are also influential in understanding species homology and sensitivity. Qualitatively, animal models exhibit a similar pattern of tissue dose distribution for O₃ with the largest tissue dose delivered to the centriacinar region. However, due to anatomical and biochemical respiratory tract differences, the actual O₃ dose delivered differs between humans and animal models. Animal data obtained in resting conditions underestimates the dose to the respiratory tract tissue relative to exercising humans. Further, it should be noted that, with the exception of airways remodeling, the mechanistic pathways discussed above have been demonstrated in both animals and human subjects in response to the inhalation of O₃. Even though interspecies differences limit quantitative comparison between species, the short- and long-term functional responses of laboratory animals to O₃ appear qualitatively homologous to those of the human making them a useful tool in determining mechanistic and cause-effect relationships with O₃ exposure. Furthermore, animal studies add to a better understanding of the full range of potential O₃-mediated effects.

### 2.5 Integration of Ozone Health Effects

This section evaluates the evidence from toxicological, controlled human exposure, and epidemiologic studies (which examined the health effects associated with short- and long-term exposure to O₃,) and summarizes the main conclusions of this assessment regarding the health effects of O₃ and the concentrations at which those effects are observed. The results from the health studies, supported by the synthesis of atmospheric chemistry (see Section 2.2) and exposure assessment (see Section 2.3) studies, contribute to the causal determinations made for the health
outcomes discussed in this assessment (see Preamble to this document for details on the causal framework).

Epidemiologic studies generally present O\textsubscript{3}-related effect estimates for mortality and morbidity health outcomes based on an incremental change in exposure, traditionally equal to the interquartile range in O\textsubscript{3} concentrations or some other arbitrary value (e.g., 10 ppb). Additionally, various averaging times are used in O\textsubscript{3} epidemiologic studies, with the three most common being the maximum 1-hour average within a 24-hour period (1-h max), the maximum 8-hour average within a 24-hour period (8-h max), and 24-hour average (24-h avg). For the purpose of presenting results from studies that use different exposure metrics, EPA consistently applies the same O\textsubscript{3} increments to facilitate comparisons between the results of various studies that may use different indices. These increments were derived using the nationwide distributional data for O\textsubscript{3} monitors in U.S. Metropolitan Statistical Areas. They are representative of a low-to-high change in O\textsubscript{3} concentrations and were approximated on the basis of annual mean to 95th percentile differences (Langstaff, 2003). Therefore, throughout Chapter 6, efforts were made to standardize O\textsubscript{3}-related effect estimates using the increments of 20 ppb for 24-h avg, 30 ppb for 8-h max, and 40 ppb for 1-h max O\textsubscript{3} concentrations, except as noted. In long-term exposure studies, typically, O\textsubscript{3} concentrations are lower and less variable when averaged across longer exposure periods, and differences due to the use of varying averaging times (e.g., 1-h max, 24-h avg) become less apparent. As such, in the long-term exposure chapter (Chapter 7) an increment of 10 ppb was consistently applied across studies, regardless of averaging time, to facilitate comparisons between the results from these studies.

2.5.1 Conclusions from Previous O\textsubscript{3} AQCDs

The 2006 O\textsubscript{3} AQCD concluded that there was clear, consistent evidence of a causal relationship between short-term O\textsubscript{3} exposure and respiratory health effects (U.S. EPA, 2006b). The causal relationship for respiratory health effects was substantiated by the coherence of effects observed across controlled human exposure, epidemiologic, and toxicological studies indicating effects of short-term O\textsubscript{3} exposures on a range of respiratory health endpoints from respiratory tract inflammation to respiratory-related emergency department (ED) visits and hospital admissions.

Across disciplines, short-term O\textsubscript{3} exposures induced or were associated with statistically significant declines in lung function. An equally strong body of evidence from controlled human exposure and toxicological studies demonstrated O\textsubscript{3}-induced inflammatory responses, increased epithelial permeability, and airway hyperresponsiveness (both specific and nonspecific). Toxicological studies provided additional evidence for O\textsubscript{3}-induced impairment of host defenses. Combined, these findings from experimental studies provided support for epidemiologic evidence, in which short-term increases in ambient O\textsubscript{3} concentration were consistently associated
with increases in respiratory symptoms and asthma medication use in children with asthma, respiratory-related hospital admissions, and asthma-related ED visits. Although O₃ was consistently associated with nonaccidental and cardiopulmonary mortality, the contribution of respiratory causes to these findings was uncertain.

Collectively, there is a vast amount of evidence spanning several decades that demonstrated that exposure to O₃ induces a range of respiratory effects. The majority of this evidence was derived from studies investigating short-term exposure (i.e., hours to weeks) to O₃. The combined evidence across disciplines led to the causal relationship between short-term O₃ exposure and respiratory effects reported in the 2006 O₃ AQCD.

Mechanistic evidence for the effects of O₃ on the respiratory system was characterized in the 1996 O₃ AQCD (U.S. EPA, 1996a), which identified O₃-induced changes in a variety of lung lipid species whose numerous biologically active metabolites, in turn, can affect host defenses, lung function, and the immune system. As summarized in Section 2.4 and fully characterized in Chapter 5, key events in the toxicity pathway of O₃ have been identified in humans and animal models. They include formation of secondary oxidation products, activation of neural reflexes, initiation of inflammation, alteration of epithelial barrier function, sensitization of bronchial smooth muscle, modification of innate/adaptive immunity, airways remodeling, and systemic inflammation and oxidative/nitrosative stress.

### 2.5.2 Summary of Causal Determinations

Recent studies support or build upon the strong body of evidence presented in the 1996 and 2006 O₃ AQCDs that short-term O₃ exposure is causally associated with respiratory health effects. Recent controlled human exposure studies demonstrate statistically significant group mean decreases in pulmonary function to exposures as low as 60-70 ppb O₃ in young, healthy adults, and are supported by the strong, cumulative evidence from epidemiologic studies. Equally strong evidence demonstrated associations of ambient O₃ with respiratory hospital admissions and ED visits across the U.S., Europe, and Canada. Most effect estimates ranged from a 1.6 to 5.4% increase in daily respiratory-related ED visits or hospital admissions in all-year analyses for unit increases¹ in ambient O₃ concentrations. Several multicity studies and a multicontinent study reported associations between short-term increases in ambient O₃ concentrations and increases in respiratory mortality. This evidence is supported by a large body of individual-level epidemiologic panel studies that demonstrate associations of ambient O₃ with respiratory symptoms in children with asthma. Further support is provided by recent studies that found O₃-associated increases in indicators of airway inflammation and oxidative stress in children with asthma. Across respiratory endpoints, evidence indicates antioxidant capacity may modify the risk of respiratory morbidity associated with O₃ exposure. The potentially elevated risk of populations with diminished antioxidant capacity and the reduced

¹ Effect estimates were standardized to a 40-, 30-, and 20-ppb unit increase for 1-h max, 8-h max, and 24-h avg O₃.
risk of populations with enhanced antioxidant capacity identified in epidemiologic studies is strongly supported by similar findings from controlled human exposure studies and by evidence that characterizes O₃-induced decreases in intracellular antioxidant levels as a mode of action for downstream effects. By demonstrating O₃-induced airway hyperresponsiveness, decreased pulmonary function, allergic responses, lung injury, impaired host defense, and airway inflammation, toxicological studies have characterized O₃ modes of action and provided biological plausibility for epidemiologic associations of ambient O₃ concentrations with lung function and respiratory symptoms, hospital admissions, ED visits, and mortality. Together, the evidence integrated across controlled human exposure, epidemiologic, and toxicological studies and across the spectrum of respiratory health endpoints continues to demonstrate that there is a causal relationship between short-term O₃ exposure and respiratory health effects.

The epidemiologic evidence for a relationship between long-term O₃ exposure and respiratory health effects (including respiratory symptoms, new-onset asthma, and respiratory mortality) is contributed by recent studies that evaluate the associations between long-term exposure to O₃ and respiratory effects that demonstrate interactions between exercise or different genetic variants and both new-onset asthma in children and increased respiratory symptom effects in individuals with asthma. While the evidence is limited, a U.S. multicommunity prospective cohort demonstrates that asthma risk is affected by interactions among genetic variability, environmental O₃ exposure, and behavior. The evidence relating new-onset asthma to long-term O₃ exposure is supported by toxicological studies of asthma in monkeys. This nonhuman primate evidence of O₃-induced changes supports the biologic plausibility of long-term exposure to O₃ contributing to the effects of asthma in children. Early life O₃ exposure can alter airway development and lead to the development of asthma. Other recent epidemiologic studies provide coherent evidence for long-term O₃ exposure and respiratory effects such as first asthma hospitalization, respiratory symptoms in asthmatics, and respiratory mortality. Generally, the epidemiologic and toxicological evidence provides a compelling case that supports the hypothesis that a relationship exists between long-term exposure to ambient O₃ and measures of respiratory health effects and mortality. The evidence for short-term exposure to O₃ and effects on respiratory endpoints provides coherence and biological plausibility for the effects of long-term exposure to O₃. Building upon that evidence, the more recent epidemiologic evidence, combined with toxicological studies in rodents and nonhuman primates, provides biologically plausible evidence that there is likely to be a causal relationship between long-term exposure to O₃ and respiratory health effects.

In past O₃ AQCDs the effects of short-term exposure to O₃ on the cardiovascular system could not be thoroughly evaluated due to the paucity of information available. However, studies investigating O₃-induced cardiovascular events have advanced in the last two decades. Animal toxicological studies, although limited in number, demonstrate O₃-induced cardiovascular effects; specifically enhanced ischemia/reperfusion (I/R) injury, disrupted NO-induced vascular reactivity, decreased cardiac function, and increased heart rate variability (HRV). These effects...
are consistent with cardiovascular system effects observed after long-term O₃ exposure, such as increased vascular disease. These effects may, in part, correspond to the alteration of the autonomic nervous system or to the development and maintenance of systemic oxidative stress and a proinflammatory environment that can result from pulmonary inflammation. Controlled human exposure studies provide some coherence with the evidence from animal toxicological studies, by demonstrating increases and decreases in HRV following relatively low (120 ppb during rest) and high (300 ppb with exercise) O₃ exposures, respectively. Controlled human exposure studies also support the animal toxicology studies by demonstrating O₃-induced effects on blood biomarkers of systemic inflammation and oxidative stress as well as changes in biomarkers suggestive of a pro-thrombogenic response to O₃. The experimental evidence provides initial biological plausibility for the consistently positive associations observed across multiple epidemiologic studies of short-term O₃ exposure and cardiovascular mortality. However, epidemiologic studies generally do not observe associations between short-term exposure to O₃ and cardiovascular morbidity; studies of cardiovascular-related hospital admissions and ED visits and other various cardiovascular effects did not find consistent evidence of a relationship with O₃ exposure. The lack of coherence between the results from studies that examined associations between short-term O₃ exposure and cardiovascular morbidity and subsequently cardiovascular mortality complicate the interpretation of the overall evidence for O₃-induced cardiovascular effects. Overall, animal toxicological studies demonstrate O₃-induced cardiovascular effects, and support the strong body of evidence indicating O₃-induced cardiovascular mortality. Animal toxicological and controlled human exposure studies provide evidence for biologically plausible mechanisms underlying these O₃-induced cardiovascular effects. However, a lack of coherence with epidemiologic studies of cardiovascular morbidity remains an important uncertainty. Taken together, the overall body of evidence across disciplines indicates that there is likely to be a causal relationship between short-term exposures to O₃ and cardiovascular effects.

The 2006 O₃ AQCD concluded that the overall body of evidence was highly suggestive that short-term exposure to O₃ directly or indirectly contributes to nonaccidental and cardiopulmonary-related mortality, but additional research was needed to more fully establish underlying mechanisms by which such effects occur. The evaluation of recent multicity studies and a multicontinent study that examined the association between short-term increases in ambient O₃ concentration and mortality found evidence that supports the conclusions of the 2006 O₃ AQCD. These recent studies reported consistent positive associations between short-term increases in ambient O₃ concentration and total (nonaccidental) mortality, with associations being stronger during the warm season, as well as provided additional support for associations between O₃ concentrations and cardiovascular mortality being similar or larger in magnitude compared to respiratory mortality. Additionally, these new studies examined previously identified areas of uncertainty in the O₃-mortality relationship, and provide additional evidence supporting an association between short-term O₃ exposure and mortality. Taken together, the body of evidence indicates
that there is likely to be a causal relationship between short-term O₃ exposures and total mortality.

The 2006 O₃ AQCD concluded that an insufficient amount of evidence existed to suggest a causal relationship between long-term O₃ exposure and mortality (U.S. EPA, 2006b). A synthesis of the recent and earlier evidence reveals that the strongest evidence for an association between long-term exposure to ambient O₃ concentrations and mortality is derived from associations for respiratory mortality that remained robust after adjusting for PM₂.₅ concentrations. There is inconsistent evidence for an association between long-term exposure to ambient O₃ and cardiopulmonary mortality, with several analyses from the American Cancer Society (ACS) cohort reporting some positive associations, while other studies reported no association. There is generally limited evidence for an association with long-term exposure to ambient O₃ and total mortality. The findings for respiratory mortality are consistent and coherent with the evidence from epidemiologic, controlled human exposure, and animal toxicological studies for the effects of short- and long-term exposure to O₃ on respiratory effects. Respiratory mortality is a relatively small portion of total mortality [about 7.6% of all deaths in 2010 were due to respiratory causes (Murphy et al., 2012)], thus it is not surprising that the respiratory mortality signal may be difficult to detect in studies of cardiopulmonary or total mortality. Based on the recent evidence for respiratory mortality along with limited evidence for total and cardiopulmonary mortality, the evidence is suggestive of a causal relationship between long-term O₃ exposures and total mortality.

In past O₃ AQCDs the effects of long-term exposure to O₃ on the cardiovascular system could not be thoroughly evaluated due to the paucity of information available. However, studies investigating O₃-induced cardiovascular events have advanced in the last two decades. Animal toxicological studies provide evidence for long-term O₃ exposure leading to cardiovascular morbidity, including increased vascular disease. There is limited, inconsistent evidence for cardiovascular morbidity in epidemiologic studies examining long-term exposure to O₃. Overall, animal toxicological studies provide some evidence for O₃-induced cardiovascular effects, but the effects observed were not consistently supported by controlled human exposure studies or epidemiologic studies. Thus, the overall body of evidence across disciplines is suggestive of a causal relationship between long-term exposures to O₃ and cardiovascular effects.

In the 2006 O₃ AQCD, there were a number of health effects for which an insufficient amount of evidence existed to adequately characterize the relationships with exposure to O₃. However, recent evidence suggests that O₃ may impart health effects through exposure durations and biological mechanisms not previously considered. For example, recent toxicological studies add to earlier evidence that short- and long-term exposures to O₃ can produce a range of effects on the central nervous system and behavior. Additionally, an epidemiologic study demonstrated that long-term exposure to O₃ affects memory in humans as well. Together the evidence from studies of short- and long-term exposure to O₃ is suggestive of a causal relationship between O₃ exposure and central nervous system effects. There is also limited though positive toxicological evidence for O₃-induced
developmental effects. Limited epidemiologic evidence exists for an association of 
O₃ concentration with decreased sperm concentration and associations with reduced 
birth weight and restricted fetal growth. Overall, the evidence is suggestive of a 
causal relationship between long-term exposures to O₃ and reproductive and 
developmental effects.

These causal determinations are summarized in Table 2-1, along with the conclusions 
from the previous NAAQS review. Special emphasis and additional details are 
provided in Table 2-1 for respiratory health outcomes, for which there is the 
strongest body of evidence.

| Table 2-1 | Summary of evidence from epidemiologic, controlled human 
| Health Outcome | exposure, and animal toxicological studies on the health effects 
<p>| associated with short- and long-term exposure to O₃. |</p>
<table>
<thead>
<tr>
<th>Health Outcome</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
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<tbody>
<tr>
<td>Short-Term Exposure to O₃</td>
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</table>
| Respiratory effects | The overall evidence supports a 
causal relationship between acute 
ambient O₃ exposures and increased 
respiratory morbidity outcomes. | Evidence integrated across controlled 
human exposure, epidemiologic, and 
toxicological studies and across the 
spectrum of respiratory health endpoints 
continues to demonstrate that there is a 
causal relationship between short-term 
O₃ exposure and respiratory health 
effects. |
| Lung function | Results from controlled human 
exposure studies and animal 
toxicological studies provide clear 
evidence of causality for the 
associations observed between acute 
(≤ 24 h) O₃ exposure and relatively 
small, but statistically significant 
decrements in lung function observed in 
numerous recent epidemiologic 
studies. Declines in lung function are 
particularly noted in children, 
asthmatics, and adults who work or 
exercise outdoors. | Recent controlled human exposure 
 studies demonstrate group mean 
decreases in FEV₁ in the range of 2 to 
3% with 6.6 hour exposures to as low as 
60 ppb O₃. The collective body of 
epidemiologic evidence demonstrates 
associations between short-term ambient 
O₃ exposure and decrements in lung 
function, particularly in children with 
asthma, children, and adults who work or 
exercise outdoors. |
| Airway hyperresponsiveness | Evidence from human clinical and 
animal toxicological studies clearly 
indicate that acute exposure to O₃ 
can induce airway hyperreactivity, 
thus likely placing atopic asthmatics 
at greater risk for more prolonged 
bouts of breathing difficulties due to 
airway constriction in response to 
various airborne allergens or other 
triggering stimuli. | A limited number of studies have 
observed airway hyperresponsiveness in 
rodents and guinea pigs after exposure to 
less than 300 ppb O₃. As previously 
reported in the 2006 O₃ AQCD, increased 
airway responsiveness has been 
demonstrated at 80 ppb in young, healthy 
adults, and at 50 ppb in certain strains of 
rats. |
<table>
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<tr>
<th>Health Outcome</th>
<th>Conclusions from 2006 O₃ AQCD</th>
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<tr>
<td>Pulmonary inflammation, injury and oxidative stress</td>
<td>The extensive human clinical and animal toxicological evidence, together with the limited available epidemiologic evidence, is clearly indicative of a causal role for O₃ in inflammatory responses in the airways.</td>
<td>Epidemiologic studies provided new evidence for associations of ambient O₃ with mediators of airway inflammation and oxidative stress and indicate that higher antioxidant levels may reduce pulmonary inflammation associated with O₃ exposure. Generally, these studies had mean 8-h max O₃ concentrations less than 73 ppb. Recent controlled human exposure studies show O₃-induced inflammatory responses at 60 ppb, the lowest concentration evaluated.</td>
</tr>
<tr>
<td>Respiratory symptoms and medication use</td>
<td>Young healthy adult subjects exposed in clinical studies to O₃ concentrations ≥ 80 ppb for 6 to 8 h during moderate exercise exhibit symptoms of cough and pain on deep inspiration. The epidemiologic evidence shows significant associations between acute exposure to ambient O₃ and increases in a wide variety of respiratory symptoms (e.g., cough, wheeze, production of phlegm, and shortness of breath) and medication use in asthmatic children.</td>
<td>The collective body of epidemiologic evidence demonstrates positive associations between short-term exposure to ambient O₃ and respiratory symptoms (e.g., cough, wheeze, and shortness of breath) in children with asthma. Generally, these studies had mean 8-h max O₃ concentrations less than 69 ppb.</td>
</tr>
<tr>
<td>Lung host defenses</td>
<td>Toxicological studies provided extensive evidence that acute O₃ exposures as low as 80 to 500 ppb can cause increases in susceptibility to infectious diseases due to modulation of lung host defenses. A single controlled human exposure study found decrements in the ability of alveolar macrophages to phagocytize microorganisms upon exposure to 80 to 100 ppb O₃.</td>
<td>Recent controlled human exposure studies demonstrate the increased expression of cell surface markers and alterations in sputum leukocyte markers related to innate adaptive immunity with short-term O₃ exposures of 80-400 ppb. Recent studies demonstrating altered immune responses and natural killer cell function build on prior evidence that O₃ can affect multiple aspects of innate and acquired immunity with short-term O₃ exposures as low as 80 ppb.</td>
</tr>
<tr>
<td>Allergic and asthma related responses</td>
<td>Previous toxicological evidence indicated that O₃ exposure skews immune responses toward an allergic phenotype, and enhances the development and severity of asthma-related responses such as AHR.</td>
<td>Recent controlled human exposure studies demonstrate enhanced allergic cytokine production in atopic individuals and asthmatics, increased IgE receptors in atopic asthmatics, and enhanced markers of innate immunity and antigen presentation in health subjects or atopic asthmatics with short-term exposure to 80-400 ppb O₃, all of which may enhance allergy and/or asthma. Further evidence for O₃-induced allergic skewing is provided by a few recent studies in rodents using exposure concentrations as low as 200 ppb.</td>
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<tr>
<td>Health Outcome</td>
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<tr>
<td>Respiratory Hospital admissions, ED visits, and physician visits</td>
<td>Aggregate population time-series studies observed that ambient O₃ concentrations are positively and robustly associated with respiratory-related hospitalizations and asthma ED visits during the warm season.</td>
<td>Consistent, positive associations of ambient O₃ with respiratory hospital admissions and ED visits in the U.S., Europe, and Canada with supporting evidence from single city studies. Generally, these studies had mean 8-h max O₃ concentrations less than 60 ppb.</td>
</tr>
<tr>
<td>Respiratory Mortality</td>
<td>Aggregate population time-series studies specifically examining mortality from respiratory causes were limited in number and showed inconsistent associations between acute exposure to ambient O₃ exposure and respiratory mortality.</td>
<td>Recent multicity time-series studies and a multicontinent study consistently demonstrated associations between ambient O₃ and respiratory-related mortality visits across the U.S., Europe, and Canada with supporting evidence from single city studies. Generally, these studies had mean 8-h max O₃ concentrations less than 63 ppb.</td>
</tr>
<tr>
<td>Cardiovascular effects</td>
<td>The limited evidence is highly suggestive that O₃ directly and/or indirectly contributes to cardiovascular-related morbidity, but much remains to be done to more fully substantiate the association.</td>
<td>The overall body of evidence across disciplines indicates that there is likely to be a causal relationship for short-term exposures to O₃ and cardiovascular effects.</td>
</tr>
<tr>
<td>Central nervous system effects</td>
<td>Toxicological studies report that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short- and long-term memory, sleep patterns, and histological signs of neurodegeneration.</td>
<td>Together the evidence from studies of short-term exposure to O₃ is suggestive of a causal relationship between O₃ exposure and CNS effects.</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>The evidence is highly suggestive that O₃ directly or indirectly contributes to non-accidental and cardiopulmonary-related mortality.</td>
<td>Taken together, the body of evidence indicates that there is likely to be a causal relationship between short-term exposures to O₃ and total mortality.</td>
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### Long-term Exposure to O₃

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<th>Health Outcome</th>
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<tr>
<td>Respiratory effects</td>
<td>The current evidence is suggestive but inconclusive for respiratory health effects from long-term O₃ exposure.</td>
<td>Recent epidemiologic evidence, combined with toxicological studies in rodents and non-human primates, provides biologically plausible evidence that there is likely to be a causal relationship between long-term exposure to O₃ and respiratory health effects.</td>
</tr>
<tr>
<td>New onset asthma</td>
<td>No studies examining this outcome were evaluated in the 2006 O₃ AQCD.</td>
<td>Evidence that different genetic variants (HMOX, GST, ARG), in combination with O₃ exposure, are related to new onset asthma. These associations were observed when subjects living in areas where the mean annual 8-h max O₃ concentration was 55.2 ppb, compared to those who lived where it was 38.4 ppb.</td>
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<tr>
<td>Asthma hospital admissions</td>
<td>No studies examining this outcome were evaluated in the 2006 O₃ AQCD.</td>
<td>Chronic O₃ exposure was related to first childhood asthma hospital admissions in a positive concentration-response relationship. Generally, these studies had mean annual 8-h max O₃ concentrations less than 41 ppb.</td>
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<tr>
<td>Pulmonary structure and function</td>
<td>Epidemiologic studies observed that reduced lung function growth in children was associated with seasonal exposure to O₃; however, cohort studies of annual or multiyear O₃ exposure observed little clear evidence for impacts of longer-term, relatively low-level O₃ exposure on lung function development in children. Animal toxicological studies reported chronic O₃-induced structural alterations, some of which were irreversible, in several regions of the respiratory tract including the centriacinar region. Morphologic evidence from studies using exposure regimens that mimic seasonal exposure patterns report increased lung injury compared to conventional chronic stable exposures.</td>
<td>Evidence for pulmonary function effects is inconclusive, with some new epidemiologic studies observing positive associations (mean annual 8-h max O₃ concentrations less than 65 ppb). Information from toxicological studies indicates that long-term exposure during development among infant monkeys (500 ppb) and adult rodents (&gt;120 ppb) can result in irreversible morphological changes in the lung, which in turn can influence pulmonary function.</td>
</tr>
<tr>
<td>Pulmonary inflammation, injury and oxidative stress</td>
<td>Extensive human clinical and animal toxicological evidence, together with limited epidemiologic evidence available, suggests a causal role for O₃ in inflammatory responses in the airways.</td>
<td>Several epidemiologic studies (mean 8-h max O₃ concentrations less than 69 ppb) and toxicology studies (as low as 500 ppb) add to observations of O₃-induced inflammation and injury.</td>
</tr>
<tr>
<td>Lung host defenses</td>
<td>Toxicological studies provided evidence that chronic O₃ exposure as low as 100 ppb can cause increases in susceptibility to infectious diseases due to modulation of lung host defenses, but do not cause greater effects on infectivity than short exposures.</td>
<td>Consistent with decrements in host defenses observed in rodents exposed to 100 ppb O₃, recent evidence demonstrates a decreased ability to respond to pathogenic signals in infant monkeys exposed to 500 ppb O₃.</td>
</tr>
<tr>
<td>Allergic responses</td>
<td>Limited epidemiologic evidence supported an association between ambient O₃ and allergic symptoms. Little if any information was available from toxicological studies.</td>
<td>Evidence relates positive outcomes of allergic response and O₃ exposure but with variable strength for the effect estimates; exposure to O₃ may increase total IgE in adult asthmatics. Allergic indicators in monkeys were increased by exposure to O₃ concentrations of 500 ppb.</td>
</tr>
<tr>
<td>Respiratory mortality</td>
<td>Studies of cardio-pulmonary mortality were insufficient to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans.</td>
<td>A single study demonstrated that exposure to O₃ (long-term mean O₃ less than 104 ppb) elevated the risk of death from respiratory causes and this effect was robust to the inclusion of PM₂.₅.</td>
</tr>
</tbody>
</table>
### Health Outcome Conclusions from 2006 O₃ AQCD Conclusions from this ISA

<table>
<thead>
<tr>
<th>Health Outcome</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cardiovascular Effects</td>
<td>No studies examining this outcome were evaluated in the 2006 O₃ AQCD.</td>
<td>The overall body of evidence across disciplines is suggestive of a causal relationship for long-term exposures to O₃ and cardiovascular effects.</td>
</tr>
<tr>
<td>Reproductive and developmental effects</td>
<td>Limited evidence for a relationship between air pollution and birth-related health outcomes, including mortality, premature births, low birth weights, and birth defects, with little evidence being found for O₃ effects.</td>
<td>Overall, the evidence is suggestive of a causal relationship between long-term exposures to O₃ and reproductive and developmental effects.</td>
</tr>
<tr>
<td>Central nervous system effects</td>
<td>Toxicological studies reported that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short and long term memory, sleep patterns, and histological signs of neurodegeneration. Evidence regarding chronic exposure and neurobehavioral effects was not available.</td>
<td>Together the evidence from studies of long-term exposure to O₃ is suggestive of a causal relationship between O₃ exposure and CNS effects.</td>
</tr>
<tr>
<td>Cancer</td>
<td>Little evidence for a relationship between chronic O₃ exposure and increased risk of lung cancer.</td>
<td>Overall, the evidence is inadequate to determine if a causal relationship exists between ambient O₃ exposures and cancer.</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>There is little evidence to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans.</td>
<td>Collectively, the evidence is suggestive of a causal relationship between long-term O₃ exposures and total mortality.</td>
</tr>
</tbody>
</table>

### 2.5.3 Integrated Synthesis of Evidence for Health Effects

This section integrates the evidence for respiratory and cardiovascular effects (including mortality) across scientific disciplines and both short- and long-term exposure periods. Here, the complete body of evidence from both previous and the current NAAQS reviews is synthesized for the broad range of respiratory and cardiovascular effects associated with exposure to O₃.

#### 2.5.3.1 Respiratory Effects

Building on evidence evaluated in previous O₃ AQCDs, recent evidence confirms and extends that O₃ is associated with a broad range of respiratory effects, including altered development of the respiratory tract. Recent animal toxicological studies of long-term exposure to O₃ occurring throughout various lifestages in monkeys, beginning with prenatal and early life exposures, provide novel evidence for effects on the development of the respiratory system, including ultrastructural changes in bronchiole development, increased offspring airway hyper-reactivity (Section 7.4.8).
as well as effects on the developing immune system. The strongest evidence for O₃-induced effects on the developing lung comes from a series of experiments using infant rhesus monkeys episodically exposed to 500 ppb O₃ for approximately 5 months, starting at one month of age. Functional changes in the conducting airways of infant rhesus monkeys exposed to either O₃ alone or O₃ + antigen were accompanied by a number of cellular and morphological changes. In addition to these functional and cellular changes, substantial structural changes in the respiratory tract were observed. Importantly, the O₃-induced structural pathway changes persisted after recovery in filtered air for six months after cessation of the O₃ exposures. Exposure to O₃ has also been associated with similar types of alterations in pulmonary structure, including airways remodeling and pulmonary injury and increased permeability, in all adult laboratory animal species studied, from rats to monkeys (U.S. EPA, 1996a).

In addition to effects on the development and structure of the respiratory tract, there is extensive evidence for the effects of short-term exposure to O₃ on pulmonary inflammation and oxidative stress. Previous evidence from controlled human exposure studies indicated that O₃ causes an inflammatory response in the lungs (U.S. EPA, 1996a). This inflammatory response to O₃ was detected after a single 1-h exposure with exercise to O₃ concentrations of 300 ppb; the increased levels of some inflammatory cells and mediators persisted for at least 18 hours. Toxicological studies provided additional evidence for increases in permeability and inflammation in rabbits at levels as low as 100 ppb O₃. Evidence summarized in the 2006 O₃ AQCD demonstrated that inflammatory responses were observed subsequent to 6.6 hours O₃ exposure to the lowest tested level of 80 ppb in healthy human adults, while toxicological studies provided extensive evidence that short-term (1-3 hours) O₃ exposure in the range of 100-500 ppb could cause lung inflammatory responses. The limited epidemiologic evidence reviewed in the 2006 O₃ AQCD demonstrated an association between short-term increases in ambient O₃ concentration and airways inflammation in children (1-h max O₃ of approximately 100 ppb). Recent studies in animals and in vitro models described inflammatory and injury responses mediated by Toll-like receptors (e.g., TLR4, TLR2), receptors for TNF or IL-1, multiple signaling pathways (e.g., p38, JNK, NFκB, MAPK/AP-1), and oxidative stress (Section 6.2.3.3). Recent epidemiologic studies provide additional supporting evidence by demonstrating associations of ambient O₃ with mediators of airways inflammation and oxidative stress.

The normal inflammatory response in lung tissue is part of host defense that aids in removing microorganisms or particles that have reached the distal airways and alveolar surface. The 1996 O₃ AQCD concluded that short-term exposure to elevated concentrations of O₃ resulted in alterations in these host defense mechanisms in the respiratory system. Specifically, toxicological studies of short-term exposures as low as 100 ppb O₃ for 2 hours were shown to decrease the ability of alveolar macrophages to ingest particles, and short-term exposures as low as 80 ppb for 3 hours prevented mice from resisting infection with streptococcal bacteria and resulted in infection-related mortality. Similarly, alveolar macrophages removed from the lungs of human subjects after 6.6 hours of exposure to 80 and 100 ppb O₃...
had decreased ability to ingest microorganisms, indicating some impairment of host defense capability. These altered host defense mechanisms can lead to increased risk of respiratory infections, which can often predispose individuals to developing asthma when occurring in early life. Despite the strong toxicological evidence, in the limited body of epidemiologic evidence, ambient O₃ concentrations have not been consistently associated with hospital admissions or ED visits for respiratory infection, pneumonia, or influenza (Section 6.2.7.2 and Section 6.2.7.3).

The most commonly observed and strongest evidence for respiratory effects associated with short-term exposure to O₃ is transient decrements in pulmonary function. Controlled human exposure studies reviewed in previous assessments demonstrated O₃-induced decrements in pulmonary function, characterized by alterations in lung volumes and flow and airway resistance and responsiveness for multihour exposures (up to 8 hours) to O₃ concentrations as low as 80 ppb (U.S. EPA, 1996a). A series of mobile laboratory studies of lung function and respiratory symptoms reported pulmonary function decrements at mean ambient O₃ concentrations of 140 ppb in exercising healthy adolescents and increased respiratory symptoms and pulmonary function decrements at 150 ppb in heavily exercising athletes and at 170 ppb in lightly exercising healthy and asthmatic subjects. Epidemiologic and animal toxicological evidence is coherent with the results of the controlled human exposure studies, both indicating decrements in lung function upon O₃ exposure. A combined statistical analysis of epidemiologic studies in children at summer camp with particularly strong exposure assessment demonstrated decrements in FEV₁ of 0.50 mL/ppb with an increase in previous hour O₃ concentration. For preadolescent children exposed to 120 ppb ambient O₃, this estimated volume decrease corresponded to an average decrement of 2.4-3.0% in FEV₁. Key studies of lung function (FEV₁) measured before and after well-defined outdoor exercise events in adults yielded concentration-response slopes of 0.40 and 1.35 mL/ppb ambient O₃ after exposure for up to 1 hour. Animal toxicological studies reported similar respiratory effects in rats at exposures as low as 200 ppb O₃ for 3 hours. The 2006 O₃ AQCD characterized the controlled human exposure and animal toxicological studies as providing clear evidence of causality for the associations observed between short-term (≤ 24 hours) increases in O₃ concentration and relatively small, but statistically significant declines in lung function observed in numerous recent epidemiologic studies. In epidemiologic studies, declines in lung function were particularly noted in children with and without asthma, and adults who work or exercise outdoors.

Recent controlled human exposure studies examined lower concentration O₃ exposures (40-80 ppb) and demonstrated that FEV₁, respiratory symptoms, and inflammatory responses were affected by O₃ exposures of 6.6 hours as low as 60 to 70 ppb (Section 6.2.1.1 and Section 6.2.3.1). These studies demonstrated average O₃-induced decreases in FEV₁ in the range of 2.8 to 3.6% with O₃ exposures to 60 ppb for 6.6 hours. Further, in the controlled human exposure studies evaluating effects of 60 ppb O₃, on average, 10% of the exposed individuals experienced >10% FEV₁ decrements following 6.6 hours of exposure. Considerable intersubject variability has also been reported in studies at higher exposure concentrations.
(≥ 70 ppb) with some subjects experiencing considerably greater decrements than average. Recent epidemiologic studies provide greater insight into individual- and population-level factors that can increase for the risk of O₃-associated respiratory morbidity. In addition to lung function decrements consistently reported in healthy children at summer camp, O₃-associated decreases in lung function were consistently observed in epidemiologic studies that included potentially at-risk populations (e.g., individuals with asthma with concurrent respiratory infection, older adults with AHR or elevated body mass index, or groups with diminished antioxidant capacity).

Exposure to O₃ can also result in respiratory symptoms (e.g., coughing, wheezing, shortness of breath). The 1996 O₃ AQCD identified an association between respiratory symptoms and increasing ambient O₃, particularly among children with asthma. In the 2006 O₃ AQCD, symptoms of cough and pain on deep inspiration were well documented in young healthy adult subjects after exposure of ≥ 80 ppb O₃ for 6-8 hours during moderate exercise. Limited data suggested an increase in respiratory symptoms down to 60 ppb. More recently, these effects have been observed at 70 ppb in healthy adults. Controlled human exposure studies of healthy adults, have also reported an increased incidence of cough with O₃ exposures as low as 120 ppb and 1-3 hours in duration with very heavy exercise. The controlled human exposure studies also demonstrated lesser respiratory symptom responses in children and older adults relative to young healthy adults. Cumulative epidemiologic evidence adds to the findings from controlled human exposure studies for healthy adults by demonstrating the effects of ambient O₃ exposure on respiratory symptoms in children with asthma. Increases in ambient O₃ concentration were associated with a wide variety of respiratory symptoms (e.g., cough, wheeze, and shortness of breath) in children with asthma. Epidemiologic studies also indicated that short-term increases in O₃ concentration are likely associated with increased asthma medication use in children with asthma. Additionally, epidemiologic studies provide evidence for an association between long-term exposure to O₃ and respiratory symptoms (Section 7.2).

Ozone exposure has been shown to result in both specific and non-specific airway hyperresponsiveness (AHR). Increased AHR is an important consequence of exposure to O₃ because its presence represents a change in airway smooth muscle reactivity and implies that the airways are predisposed to narrowing on inhalation of a variety of stimuli (e.g., specific allergens, SO₂, cold air). Specifically, short-term (2 or 3 hours) exposure to 250 or 400 ppb O₃ was found to cause increases in AHR in response to allergen challenges among allergic asthmatic subjects who characteristically already had somewhat increased AHR at baseline. Increased non-specific AHR has been demonstrated in healthy young adults down to 80 ppb O₃ following 6.6 hours of exposure during moderate exercise. While AHR has not been widely examined in epidemiologic studies, findings for O₃-induced increases in AHR in controlled human exposure (Section 6.2.2.1) and toxicological (Section 6.2.2.2) studies provide biological plausibility for associations observed between increases in ambient O₃ concentration and increases in respiratory symptoms in subjects with asthma.
In addition to asthma exacerbations, recent epidemiologic evidence has indicated that long-term ambient O₃ concentrations can contribute to new onset asthma (Section 7.2.1, Table 7-2). The new epidemiologic evidence base consists of studies using a variety of designs and analysis methods evaluating the relationship between long-term annual measures of exposure to ambient O₃ and measures of respiratory morbidity. Studies from two California cohorts have provided evidence for different variants in genes related to oxidative or nitrosative stress (e.g., HMOX, GSTs, ARG) that, depending on community long-term O₃ concentrations, are related to new onset asthma. These cohorts provide evidence that extends beyond the association of short-term exposure to O₃ and asthma exacerbations to suggest that long-term exposure to O₃ may play a role in the development of the disease and contribute to incident cases of asthma.

The frequency of ED visits and hospital admissions due to respiratory symptoms, asthma exacerbations and other respiratory diseases is associated with short- and long-term exposure to ambient O₃ concentrations. Summertime daily hospital admissions for respiratory causes in various locations of eastern North America were consistently associated with ambient concentrations of O₃ in studies reviewed in the 1996 O₃ AQCD. This association remained even with examination of only concentrations below 120 ppb O₃. The 2006 O₃ AQCD concluded that aggregate population time-series studies demonstrate a positive and robust association between ambient O₃ concentrations and respiratory-related hospitalizations and asthma ED visits during the warm season. Recent epidemiologic time-series studies that include additional multicity studies and a multicontinent study further demonstrate that short-term exposures to ambient O₃ concentrations are consistently associated with increases in respiratory hospital admissions and ED visits specifically during the warm/summer months across a range of O₃ concentrations (Section 6.2.7). There is also recent evidence for an association between respiratory hospital admissions and long-term exposure to O₃ (Section 7.2.2).

Finally, O₃ exposure can contribute to death from respiratory causes. Recent evidence from several multicity studies and a multicontinent study demonstrate consistent positive associations between short-term exposure to ambient O₃ concentrations and increases in respiratory mortality (Section 6.2.8). Similarly, a study of long-term exposure to ambient O₃ concentrations also demonstrated an association between O₃ and increases in respiratory mortality (Section 7.7.1). Evidence from these recent mortality studies is consistent and coherent with the evidence from epidemiologic, controlled human exposure, and animal toxicological studies for the effects of short- and long-term exposure to O₃ on respiratory effects. Additionally, the evidence for respiratory morbidity after short- and long-term exposure provides biological plausibility for mortality due to respiratory disease.
2.5.3.2 Cardiovascular Effects

There is an emerging body of animal toxicological evidence demonstrating that short-term exposure to $O_3$ can lead to autonomic nervous system alterations (in heart rate and/or heart rate variability) and suggesting that proinflammatory signals may mediate cardiovascular effects. Interactions of $O_3$ with respiratory tract components result in secondary oxidation products and inflammatory mediators that have the potential to penetrate the epithelial barrier and to initiate toxic effects systemically. In addition, animal toxicological studies of long-term exposure to $O_3$ provide evidence of enhanced atherosclerosis and I/R injury, corresponding with development of a systemic oxidative, proinflammatory environment.

Evidence from controlled human exposure studies also demonstrates cardiovascular effects in response to short-term $O_3$ exposure and provides some coherence with evidence from animal toxicology studies. Controlled human exposure studies support the animal toxicological studies by demonstrating $O_3$-induced effects on blood biomarkers of systemic inflammation and oxidative stress, preliminary evidence for $O_3$-induced modulation of the autonomic nervous system, as well as changes in biomarkers that can indicate a prothrombogenic response to $O_3$. However, epidemiologic studies evaluating cardiovascular morbidity and short- and long-term exposure to $O_3$ do not provide consistent evidence for an association. This is evident by the multiple studies that examined the association between (1) short- and long-term $O_3$ concentrations and cardiovascular-related hospital admissions and ED visits, and (2) cardiovascular disease-related biomarkers, and reported inconsistent results.

When examining mortality due to cardiovascular disease, epidemiologic studies consistently observe positive associations with short-term exposure to $O_3$. Additionally, there is some evidence for an association between long-term exposure to $O_3$ and mortality. However, the association between long-term ambient $O_3$ concentrations and cardiovascular mortality can be confounded by other pollutants as evident by a study of cardiovascular mortality that reported no association after adjustment for PM$_{2.5}$ concentrations.

Overall, animal toxicological studies demonstrate $O_3$-induced cardiovascular effects, and support the strong body of evidence indicating $O_3$-induced cardiovascular mortality. Animal toxicological and controlled human exposure studies provide evidence for biologically plausible mechanisms underlying these $O_3$-induced cardiovascular effects. However, a lack of coherence with epidemiologic studies of cardiovascular morbidity remains an important uncertainty.

2.5.4 Policy Relevant Considerations

This ISA summarizes and integrates the available scientific evidence that best informs consideration of the policy-relevant questions that frame this assessment, presented in the Integrated Review Plan (U.S. EPA, 2011d). This includes considering whether the available body of scientific evidence supports or calls into
question the scientific conclusions reached in the last review regarding health effects related to exposure to O₃, with particular emphasis on exposures and health risks among populations potentially at increased risk. Additional policy relevant considerations include how the scientific information, when available, informs decisions regarding the basic elements of the NAAQS: indicator, averaging time, level, and form.

2.5.4.1 Populations Potentially at Increased Risk

Studies were reviewed to identify populations that are at increased risk for O₃-related health effects. These studies have investigated factors that can cause populations to be at increased risk for O₃-related health effects by conducting stratified epidemiologic analyses; by examining individuals with an underlying health condition, genetic polymorphism, or categorized by age, race, or sex in controlled human exposure studies; or by developing animal models that mimic the pathophysiological conditions associated with a health effect. These studies have identified a multitude of factors that could potentially contribute to whether a population is at increased risk for O₃-related health effects.

The populations identified in Chapter 8 that were examined for their potential for increased risk of O₃-related health effects are listed in Table 8-6 and are classified as providing adequate, suggestive, inadequate, or no evidence of being an at-risk factor. The factors that have adequate evidence to be classified as an at-risk factor for O₃-related health effects are individuals with asthma, younger and older age groups, individuals with reduced intake of certain nutrients (i.e., vitamins C and E), and outdoor workers, based on consistency in findings across studies and evidence of coherence in results from different scientific disciplines. Asthma as a factor affecting risk was supported by controlled human exposure and toxicological studies, as well as some evidence from epidemiologic studies. Generally, studies comparing age groups also reported greater associations for respiratory hospital admissions and ED visits among children than for adults. Biological plausibility for this increased risk is supported by toxicological and controlled human exposure studies. Also, children have higher exposure and dose due to increased time spent outdoors and ventilation rate, and children’s respiratory systems are also still undergoing lung growth. Most studies comparing age groups reported greater effects of short-term O₃ exposure on mortality among older adults, although studies of other health outcomes had inconsistent findings regarding whether older adults were at increased risk. Multiple epidemiologic, controlled human exposure, and toxicological studies reported that diets lower in vitamins E and C are associated with increased risk of O₃-related health effects. Previous studies have shown that increased exposure to O₃ due to outdoor work leads to increased risk of O₃-related health effects and it is clear that outdoor workers have higher exposures, and greater internal doses, of O₃, which may lead to increased risk of O₃-related health effects.
Other potential factors [genetic variants (such as those in GSTM1, HMOX-1, NQO1, and TNF-α), obesity, sex, and SES] provided some suggestive evidence of increased risk, but further investigation is needed. Similarly, many factors had inadequate evidence to determine if they increased the risk of O3-related health effects, including influenza/infection, COPD, CVD, diabetes, hyperthyroidism, smoking, race/ethnicity, and air conditioning use.

### 2.5.4.2 Exposure Metrics in Epidemiologic Studies

Some epidemiologic studies have conducted analyses between O3 concentration and health effects (i.e., mortality, respiratory or cardiovascular) using various exposure metrics (i.e., 1-h max, 8-h max, and 24-h avg). No studies of long-term exposure (i.e., months to years) to O3 have compared the use of different exposure metrics on risk estimation.

Among time-series studies, the limited evidence suggests comparable risk estimates across exposure metrics with some evidence for smaller O3 risk estimates when using a 24-hour average exposure metric. Several panel studies examined whether associations of lung function and respiratory symptoms varied depending on the O3 exposure metric used. Although differences in effect estimates across exposure metrics were found within some studies, collectively, there was no indication that the consistency or magnitude of the observed association was stronger for a particular O3 exposure metric. Several studies examining lung function demonstrated that this was true among populations with and without increased outdoor exposures. It is important to note in these studies, the degree of exposure measurement error associated with use of central site ambient O3 concentrations may vary among O3 averaging times, depending on time spent outdoors. Among studies that examined associations of multiple respiratory symptoms in children with multiple O3 exposure metrics, most did not find higher odds ratios for any particular exposure metric. Overall, the evidence from time-series and panel epidemiologic studies does not indicate that one exposure metric is more consistently or strongly associated with mortality or respiratory-related health effects.

### 2.5.4.3 Lag Structure in Epidemiologic Studies

Epidemiologic studies have attempted to identify the time-frame in which exposure to O3 can impart a health effect. The time period between O3 exposure and health effects can potentially be influenced by a multitude of factors, such as age or existence of pre-existing diseases. Different lag times have been evaluated for specific health outcomes.

The epidemiologic evidence evaluated in the 2006 O3 AQCD indicated that one of the remaining uncertainties in characterizing the O3-mortality relationship was identifying the appropriate lag structure (e.g., single-day lags versus distributed lag
An examination of lag times used in the epidemiologic studies evaluated in this assessment can provide further insight on the characterization of the relationship between \( \text{O}_3 \) exposure and morbidity and mortality outcomes from epidemiologic studies.

The majority of epidemiologic studies that focused on the association between short-term \( \text{O}_3 \) exposure and mortality (i.e., all-cause, respiratory and cardiovascular) examined the average of multiday lags with some studies examining single-day lags. Across a range of multiday lags (i.e., average of 0-1 to 0-6 days), the studies evaluated consistently demonstrate that the \( \text{O}_3 \) effects on mortality occur within a few days of exposure (Figure 6-27).

Epidemiologic studies of lung function, respiratory symptoms, and biological markers of airway inflammation and oxidative stress examined associations with single-day ambient \( \text{O}_3 \) concentrations (using various averaging times) lagged from 0 to 7 days as well as concentrations averaged over 2 to 19 days. Lags of 0 and 1 day ambient \( \text{O}_3 \) concentrations were associated with decreases in lung function and increases in respiratory symptoms, airway inflammation, and oxidative stress. Additionally, several studies found that multiday averages of \( \text{O}_3 \) concentration were associated with these endpoints, indicating that not only single day, but exposures accumulated over several days led to a respiratory health effect. In studies of respiratory hospital admissions and ED visits, investigators either examined the lag structure of associations by including both single-day and the average of multiday lags, or selecting lags a priori. The collective evidence indicates a rather immediate response within the first few days of \( \text{O}_3 \) exposure (i.e., for lags days averaged at 0-1, 0-2, and 0-3 days) for hospital admissions and ED visits for all respiratory outcomes, asthma, and chronic obstructive pulmonary disease in all-year and seasonal analyses.

### 2.5.4.4 Ozone Concentration-Response Relationship

An important consideration in characterizing the \( \text{O}_3 \)-morbidity and mortality association is whether the concentration-response (C-R) relationship is linear across the full concentration range that is encountered or if there are concentration ranges where there are departures from linearity (i.e., nonlinearity). In this ISA studies have been identified that attempt to characterize the shape of the \( \text{O}_3 \) C-R curve along with possible \( \text{O}_3 \) “thresholds” (i.e., \( \text{O}_3 \) concentrations which must be exceeded in order to elicit an observable health response). The epidemiologic studies that examined the shape of the C-R curve and the potential presence of a threshold have indicated a generally linear C-R function with no indication of a threshold in analyses that have examined 8-h max and 24-h avg \( \text{O}_3 \) concentrations. However, there is less certainty in the shape of the C-R curve at the lower end of the distribution of \( \text{O}_3 \) concentrations (below NA background concentrations, 29-40 ppb) due to the low density of data in this range.

Controlled human exposure studies have provided strong and quantifiable C-R data on the human health effects of \( \text{O}_3 \). The magnitude of respiratory effects in these
studies is generally a function of O₃ exposure, i.e., the product of concentration (C), minute ventilation (V̇E), and exposure duration. Several studies provide evidence for a smooth C-R curve in young healthy adults exposed during moderate exercise for 6.6 hours to O₃ concentrations between 40 and 120 ppb (Figure 6-1). It is difficult to characterize the C-R relationship at and below 40 ppb due to uncertainty associated with the sparse data at these lower concentrations.

Although relatively few epidemiologic studies have examined the O₃-health effects C-R relationship, the C-R relationship has been examined across multiple health endpoints and exposure durations. Some studies of populations engaged in outdoor activity found that associations between O₃ and lung function decrements persisted at lower O₃ concentrations with some studies showing larger negative associations in analyses limited to lower O₃ concentrations (e.g., 60-80 ppb; Table 6-6) and shorter exposure durations (i.e., in the range of 30 minutes to less than 8 hours; Table 6-6). A study examining the C-R relationship between short-term O₃ exposure and pediatric asthma ED visits found no evidence of a threshold with a linear relationship evident down to 8-h max O₃ concentrations as low as 30 ppb (Figure 6-17). In an additional study, authors used a smooth function while also accounting for the potential confounding effects of PM₂.₅, to examine whether the shape of the C-R curve for short-term exposure to O₃ and asthma hospital admissions is linear. When comparing the curve to a linear fit, the authors found that the linear fit is a reasonable approximation of the C-R relationship between O₃ and asthma hospital admissions in the mid-range of the data though it can be seen that there is greater uncertainty at the lower end of the distribution of ambient O₃ concentrations, generally below 20 ppb (Figure 6-16) due to sparse data at these lower concentrations.

Several recent studies applied a variety of statistical approaches to examine the shape of the O₃-mortality C-R relationship and existence of a threshold (Section 6.6.2.4). These studies suggest that the shape of the O₃-mortality C-R curve is linear across the range of O₃ concentrations though uncertainty in the relationship increases at the lower end of the distribution (Figure 6-36). Generally, the epidemiologic studies that examined the O₃-mortality C-R relationship do not provide evidence for the existence of a threshold within the range of 24-h average (24-h avg) O₃ concentrations most commonly observed in the U.S. during the O₃ season (i.e., above 20 ppb). However, the evaluation of the C-R relationship for short-term exposure to O₃ and mortality is difficult due to the evidence from multicity studies indicating heterogeneity in O₃-mortality associations across regions of the United States. In addition, there are numerous issues that can influence the shape of the O₃-mortality C-R relationship that need to be taken into consideration including: multiday effects (distributed lags), and potential adaptation and mortality displacement (i.e., hastening of death by a short period). Additionally, given the effect modifiers identified in mortality analyses that are also expected to vary regionally (e.g., temperature, air conditioning prevalence), a national or combined analysis may not be appropriate to identify whether a threshold exists in the O₃-mortality C-R relationship.
In addition, the C-R relationship of long-term exposure to O₃ and birth outcomes has been evaluated. Evidence from the southern California Children’s Health Study identified a C-R relationship of birth weight with 24-h avg O₃ concentrations averaged over the entire pregnancy that was clearest above the 30 ppb level (Figure 7-4).

Generally, both short- and long-term exposure studies indicate a linear, no threshold C-R relationship when examining the association between O₃ exposure and multiple health effects across the range of 8-h max and 24-h avg O₃ concentrations most commonly observed in the U.S. during the O₃ season (i.e., greater than 20 ppb). However, evidence from studies of respiratory health effects and mortality indicates less certainty in the shape of the C-R curve at the lower end of the distribution of O₃ data, which corresponds to 8-h max and 24-h avg O₃ concentrations generally below 20 ppb.

2.5.4.5 Regional Heterogeneity in Risk Estimates

Multicity epidemiologic studies that have examined the relationship between short-term O₃ exposures and mortality have provided evidence of city-to-city and regional heterogeneity in O₃-mortality risk estimates. A possible explanation for this heterogeneity may be differences in community characteristics (individual- or community-level) across cities that could modify the O₃ effect (e.g., activity patterns, housing type and age distribution, prevalence and use of air conditioning). Another possible explanation for the observed heterogeneity could be effect modification by concentrations of other air pollutants or interactions with temperature or other meteorological factors that vary regionally in the U.S.

An examination of community characteristics measured at the individual level that may contribute to the observed heterogeneity in O₃-mortality risk estimates indicates increased risk in older adults (i.e., ≥ 65 years of age), women, African American individuals, individuals with pre-existing diseases/conditions (e.g., diabetes, atrial fibrillation), and lower SES. Furthermore, studies have examined community characteristics measured at the community level and found that higher O₃-mortality risk estimates were associated with higher: percent unemployment, fraction of the population Black/African-American, percent of the population that take public transportation to work; and with lower: temperatures and percent of households with central air conditioning. There is also evidence of greater effects in cities with lower mean O₃ concentrations. Additionally, there is evidence of increased risk of O₃-related mortality as percentage unemployed increases and a reduction in O₃-related mortality as mean temperature increased (i.e., a surrogate for air conditioning rate) in the United States. The lack of a consistent reduction in O₃-risk estimates in cities with a higher percentage of central air conditioning across health outcomes complicates the interpretation of the potential modifying effects of air conditioning use.
Overall, the epidemiologic studies that have examined the city-to-city and regional heterogeneity observed in multicity studies have identified a variety of factors that may modify the O₃-mortality or respiratory hospital admission relationship. Some studies have also examined the correlation with other air pollutants or the potential interactive effects between O₃ and temperature to explain city-to-city heterogeneity in O₃-mortality risk estimates. This includes evidence that O₃-mortality risk estimates in the U.S. varied by mean SO₂ concentrations, the ratio between mean NO₂/PM₁₀ concentrations, and temperatures. However, studies have not consistently identified specific community characteristics that explain the observed heterogeneity.

2.6 Integration of Effects on Vegetation and Ecosystems

Chapter 9 presents the most policy-relevant information related to this review of the NAAQS for the welfare effects of O₃ on vegetation and ecosystems. This section integrates the key findings from the disciplines evaluated in this assessment of the O₃ scientific literature, which includes plant physiology, whole plant biology, ecosystems, and exposure-response.

Overall, exposure to O₃ is causally related or likely to be causally related to effects observed on vegetation and ecosystems. These effects are observed across the entire continuum of biological organization; from the cellular and subcellular level to the whole plant level, and up to ecosystem-level processes. Furthermore, there is evidence that the effects observed across this continuum are related to one another; effects of O₃ at lower levels of organization, such as the leaf of an individual plant, can result in effects at higher levels. Ozone enters leaves through stomata, and can alter stomatal conductance and disrupt CO₂ fixation (Section 9.3). These effects can change rates of leaf gas exchange, growth and reproduction at the individual plant level and result in changes in ecosystems, such as productivity, C storage, water cycling, nutrient cycling, and community composition (Section 9.4). Figure 2-2 is a simplified illustrative diagram of the major endpoints that O₃ may affect in vegetation and ecosystems.

The framework for causal determinations (see Preamble) has been applied to the body of scientific evidence to examine effects attributed to O₃ exposure (Table 2-2). The summary below provides brief integrated summaries of the evidence that supports the causal determinations. The detailed discussion of the underlying evidence used to formulate each causal determination can be found in Chapter 9. This summary ends with a short discussion of policy relevant considerations.

2.6.1 Visible Foliar Injury

Visible foliar injury resulting from exposure to O₃ has been well characterized and documented over several decades of research on many tree, shrub, herbaceous, and crop species (U.S. EPA, 2006b, 1996b, 1984, 1978a) (Section 9.4.2). Ozone-induced
Visible foliar injury symptoms on certain bioindicator plant species are considered diagnostic as they have been verified experimentally in exposure-response studies, using exposure methodologies such as continuous stirred tank reactors (CSTRs), open-top chambers (OTCs), and free-air fumigation. Experimental evidence has clearly established a consistent association of visible injury with O3 exposure, with greater exposure often resulting in greater and more prevalent injury. Since publication of the 2006 O3 AQCD, the results of several multiple-year field surveys of O3-induced visible foliar injury at National Wildlife Refuges in Maine, Michigan, New Jersey, and South Carolina have been published. New sensitive species showing visible foliar injury continue to be identified from field surveys and verified in controlled exposure studies.

**Effects of Ozone Exposure**

- **Leaf metabolism & physiology**
  - Antioxidant metabolism up-regulated
  - Decreased photosynthesis
  - Decreased stomatal conductance or sluggish stomatal response

- **Leaves & canopy**
  - Visible leaf injury
  - Altered leaf senescence
  - Altered leaf chemical composition

- **Plant growth (Fig 9-8)**
  - Decreased biomass accumulation
  - Altered reproduction
  - Altered carbon allocation
  - Altered crop quality

- **Belowground processes (Fig 9-8)**
  - Altered litter production & decomposition
  - Altered soil carbon & nutrient cycling
  - Altered soil fauna & microbial communities

- **Ecosystem services**
  - Decreased productivity
  - Decreased C sequestration
  - Altered water cycling (Fig 9-7)
  - Altered community composition (i.e., plant, insect & microbe)

*Figure 2-2* An illustrative diagram of the major endpoints that O3 may affect in plants and ecosystems.
Table 2-2  Summary of O₃ causal determinations for vegetation and ecosystem effects.

<table>
<thead>
<tr>
<th>Vegetation and Ecosystem Effects</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Visible Foliar Injury</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause impaired aesthetic quality of many native plants and trees by increasing foliar injury.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Effects on Vegetation</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Reduced Vegetation</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause decreased growth and biomass accumulation in annual, perennial and woody plants, including agronomic crops, annuals, shrubs, grasses, and trees.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Growth</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Reduced Productivity in Terrestrial Ecosystems</td>
<td>There is evidence that O₃ is an important stressor of ecosystems and that the effects of O₃ on individual plants and processes are scaled up through the ecosystem, affecting net primary productivity.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Carbon (C) Sequestration</td>
<td>Limited studies from the 2006 O₃ AQCD.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>in Terrestrial Ecosystems</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Reduced Yield and Quality of Agricultural Crops</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause decreased yield and/or nutritive quality in a large number of agronomic and forage crops.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Ecosystem Water Cycling</td>
<td>Ecosystem water quantity may be affected by O₃ exposure at the landscape level.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Below-ground Biogeochemical Cycles</td>
<td>Ozone-sensitive species have well known responses to O₃ exposure, including altered C allocation to below-ground tissues; and also altered rates of leaf and root production, turnover, and decomposition. These shifts can affect overall C loss and nitrogen (N) loss from the ecosystem in terms of respired C, and leached aqueous dissolved organic and inorganic C and N.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Community Composition</td>
<td>Ozone may be affecting above- and below-ground community composition through impacts on both growth and reproduction. Significant changes in plant community composition resulting directly from O₃ exposure have been demonstrated.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
</tbody>
</table>

The use of biological indicators in field surveys to detect phytotoxic levels of O₃ is a longstanding and effective methodology. The USDA Forest Service through the Forest Health Monitoring (FHM) Program (1990-2001) and currently the Forest Inventory and Analysis (FIA) Program has been collecting data regarding the incidence and severity of visible foliar injury on a variety of O₃ sensitive plant species throughout the United States. The network has provided evidence that O₃ concentrations were high enough to induce visible symptoms on sensitive vegetation.
From repeated observations and measurements made over a number of years, specific geographical patterns of visible $O_3$ injury symptoms can be identified. In addition, a study assessed the risk of $O_3$-induced visible foliar injury on bioindicator plants in 244 national parks in support of the National Park Service’s Vital Signs Monitoring Network. The results of the study demonstrated that the estimated risk of visible foliar injury was high in 65 parks (27%), moderate in 46 parks (19%), and low in 131 parks (54%). Some of the well-known parks with a high risk of $O_3$-induced visible foliar injury include Gettysburg, Valley Forge, Delaware Water Gap, Cape Cod, Fire Island, Antietam, Harpers Ferry, Manassas, Wolf Trap Farm Park, Mammoth Cave, Shiloh, Sleeping Bear Dunes, Great Smoky Mountains, Joshua Tree, Sequoia and Kings Canyon, and Yosemite. Overall, evidence is sufficient to conclude that there is a causal relationship between ambient $O_3$ exposure and the occurrence of $O_3$-induced visible foliar injury on sensitive vegetation across the U.S.

### 2.6.2 Growth, Productivity, Carbon Storage and Agriculture

Ambient $O_3$ concentrations have long been known to cause decreases in photosynthetic rates and plant growth. The $O_3$-induced damages at the plant scale may translate to damages at the stand, then ecosystem scales, and cause changes in productivity and C storage. The effects of $O_3$ exposure on photosynthesis, growth, biomass allocation, ecosystem production, and ecosystem C sequestration were reviewed for the natural ecosystems, and crop productivity and crop quality were reviewed for the agricultural ecosystems.

#### 2.6.2.1 Natural Ecosystems

The previous $O_3$ AQCDs concluded that there is strong and consistent evidence that ambient concentrations of $O_3$ decrease plant photosynthesis and growth in numerous plant species across the United States. Studies published since the last review continue to support that conclusion (Section 9.4.3.1). Recent studies, based on the Aspen free-air carbon-dioxide/ozone enrichment (FACE) experiment, found that $O_3$ caused reductions in total biomass relative to the control in aspen, paper birch, and sugar maple communities during the first seven years of stand development. Overall, the studies at the Aspen FACE experiment were consistent with the open-top chamber (OTC) studies that were the foundation of previous $O_3$ NAAQS reviews. These results strengthen the understanding of $O_3$ effects on forests and demonstrate the relevance of the knowledge gained from trees grown in OTC studies.

A set of meta-analyses assessed the effects of $O_3$ on plant photosynthesis and growth across different species and fumigation methods (such as OTC and FACE). Those studies reported that current $O_3$ concentrations in the northern hemisphere are decreasing photosynthesis (~11%) across tree species, and the decreases in photosynthesis are consistent with cumulative uptake of $O_3$ into the leaf. The current ambient $O_3$ concentrations (~40 ppb averaged across all hours of exposure)
decreased annual total biomass growth of forest species by an average of 7%, with potentially greater decreases (11-17%) with elevated O$_3$ exposures (Section 9.4.3.1). The meta-analyses further confirmed that reduction of plant photosynthesis and growth under O$_3$ exposure are coherent across numerous species and various experimental techniques.

Studies during recent decades have also demonstrated O$_3$ alters biomass allocation and plant reproduction (Section 9.4.3). Recent meta-analyses have generally indicated that O$_3$ reduced C allocated to roots. Several recent studies published since the 2006 O$_3$ AQCD further demonstrate that O$_3$ altered reproductive processes, such as timing of flowering, number of flowers, fruits and seeds, in herbaceous and woody plant species. However, a knowledge gap still exists pertaining to the exact mechanism of the responses of reproductive processes to O$_3$ exposure (Section 9.4.3.3).

Studies at the leaf and plant scales show that O$_3$ decreased photosynthesis and plant growth, providing coherence and biological plausibility for the reported decreases in ecosystem productivity. During the previous NAAQS reviews, there were very few studies that investigated the effect of O$_3$ exposure on ecosystem productivity and C sequestration. Recent studies from long-term FACE experiments and ecosystem models provided evidence of the association of O$_3$ exposure and reduced productivity at the ecosystem scale. Elevated O$_3$ reduced stand biomass at Aspen FACE after 7 years of O$_3$ exposure, and annual volume growth at the Kranzberg Forest in Germany. Results across different ecosystem models were consistent with the FACE experimental evidence, which showed that O$_3$ reduced ecosystem productivity (Section 9.4.3.4). In addition to primary productivity, other indicators such as net ecosystem productivity (NEP), net ecosystem CO$_2$ exchange (NEE) and C sequestration were often assessed by model studies. Model simulations consistently found that O$_3$ exposure caused negative impacts on these indicators (Section 9.4.3.4, Table 9-3), but the severity of these impacts was influenced by multiple interactions of biological and environmental factors. The suppression of ecosystem C sinks results in more CO$_2$ accumulation in the atmosphere. A recent study suggested that the indirect radiative forcing caused by O$_3$ exposure through lowering the ecosystem C sink could have an even greater impact on global warming than the direct radiative forcing of O$_3$.

Although O$_3$ generally causes negative effects on ecosystem productivity, the magnitude of the response varies among plant communities (Section 9.4.3.4). For example, O$_3$ had little impact on white fir, but greatly reduced growth of ponderosa pine in southern California. Ozone decreased net primary production (NPP) of most forest types in the Mid-Atlantic region, but had small impacts on spruce-fir forest. Ozone could also affect regional C budgets through interacting with multiple factors, such as N deposition, elevated CO$_2$ and land use history. Model simulations suggested that O$_3$ partially offset the growth stimulation caused by elevated CO$_2$ and N deposition in both Northeast- and Mid-Atlantic-region forest ecosystems of the U.S.
Overall, evidence is sufficient to conclude that there is a causal relationship between ambient O₃ exposure and reduced native plant growth and productivity, and a likely causal relationship between O₃ exposure and reduced carbon sequestration in terrestrial ecosystems.

2.6.2.2 Agricultural Crops

The detrimental effect of O₃ on crop production has been recognized since the 1960s and a large body of research has subsequently stemmed from those initial findings. Previous O₃ AQCDs have extensively reviewed this body of literature. Current O₃ concentrations across the U.S. are high enough to cause yield loss for a variety of agricultural crops including, but not limited to, soybean, wheat, potato, watermelon, beans, turnip, onion, lettuce, and tomato (Section 9.4.4.1). Continued increases in O₃ concentration may further decrease yield in these sensitive crops. Despite the well-documented yield losses due to increasing O₃ concentration, there is still a knowledge gap pertaining to the exact mechanism of O₃-induced yield loss. Research has linked increasing O₃ concentration to decreased photosynthetic rates and accelerated senescence, which are related to yield.

In addition, recent research has highlighted the effects of O₃ on crop quality. Increasing O₃ concentration decreases nutritive quality of grasses, decreases macro- and micro-nutrient concentrations in fruits and vegetable crops, and decreases cotton fiber quality. These areas of research require further investigation to determine the mechanism and dose-responses (Section 9.4.4.2).

During the previous NAAQS reviews, there were very few studies that estimated O₃ impacts on crop yields at large geographical scales (i.e., regional, national or global). Recent modeling studies found that O₃ generally reduced crop yield, but the impacts varied across regions and crop species (Section 9.4.4.1). For example, the largest O₃-induced crop yield losses occurred in high-production areas exposed to high O₃ concentrations, such as the Midwest and the Mississippi Valley regions of the United States. Among crop species, the estimated yield loss for wheat and soybean were higher than rice and maize. Satellite and ground-based O₃ measurements have been used to assess yield loss caused by O₃ over the continuous tri-state area of Illinois, Iowa, and Wisconsin. The results showed that O₃ concentrations reduced soybean yield, which correlates well with the previous results from FACE- and OTC-type experiments (Section 9.4.4.1).

Evidence is sufficient to conclude that there is a causal relationship between O₃ exposure and reduced yield and quality of agricultural crops.

2.6.3 Water Cycling

Ozone can affect water use in plants and ecosystems through several mechanisms including damage to stomatal functioning and loss of leaf area. Section 9.3.6
reviewed possible mechanisms for O₃ exposure effects on stomatal functioning. Regardless of the mechanism, O₃ exposure has been shown to alter stomatal performance, which may affect plant and stand transpiration and therefore possibly affecting hydrological cycling.

Although the evidence was from a limited number of field and modeling studies, these findings showed an association of O₃ exposure and the alteration of water use and cycling in vegetation and ecosystems (Section 9.4.5). There is not a clear consensus on the nature of leaf-level stomatal conductance response to O₃ exposure. When measured at steady-state high light conditions, leaf-level stomatal conductance is often found to be reduced when exposed to O₃. However, measurements of stomatal conductance under dynamic light and vapor pressure deficit conditions indicate sluggish responses under elevated O₃ exposure which could potentially lead to increased water loss from vegetation. In situations where stomata fail to close under low light or water stressed conditions water loss may be greater over time. In other situations it is possible that sluggish stomata may fail to completely open in response to environmental stimuli and result in decreased water loss. Field studies suggested that peak hourly O₃ exposure increased the rate of water loss from several tree species, and led to a reduction in the late-season modeled stream flow in three forested watersheds in eastern Tennessee. Sluggish stomatal responses during O₃ exposure was suggested as a possible mechanism for increased water loss during peak O₃ exposure. Currently, the O₃-induced reduction in stomatal aperture is the biological assumption for most process-based models. Therefore, results of those models normally found that O₃ reduced water loss. For example, one study found that O₃ damage and N limitation together reduced evapotranspiration and increased runoff.

Although the direction of the response differed among studies, the evidence is sufficient to conclude that there is likely to be a causal relationship between O₃ exposure and the alteration of ecosystem water cycling.

2.6.4 Below-ground Processes

Below-ground processes are tightly linked with aboveground processes. The responses of aboveground process to O₃ exposure, such as reduced photosynthetic rates, increased metabolic cost, and reduced root C allocation, have provided biologically plausible mechanisms for the alteration of below-ground processes. Since the 2006 O₃ AQCD, more evidence has shown that although the responses are often species specific, O₃ altered the quality and quantity of C input to soil, microbial community composition, and C and nutrient cycling.

Results from Aspen FACE and other experimental studies consistently found that O₃ reduced litter production and altered C chemistry, such as soluble sugars, soluble phenolics, condensed tannins, lignin, and macro/micro nutrient concentration in litter (Section 9.4.6.1). Under elevated O₃, the changes in substrate quality and quantity could alter microbial metabolism, and therefore soil C and nutrient cycling. Several
studies indicated that O₃ generally suppressed soil enzyme activities (Section 9.4.6.2). However, the impact of O₃ on litter decomposition was inconsistent and varied among species, sites, and exposure length. Similarly, O₃ had inconsistent impacts on dynamics of micro and macro nutrients (Section 9.4.6.4).

Studies from the Aspen FACE experiment suggested that the response of below-ground C cycle to O₃ exposure, such as litter decomposition, soil respiration, and soil C content, changed over time. For example, in the early part of the experiment (1998-2003), O₃ had no impact on soil respiration but reduced the formation rates of total soil C under elevated CO₂. However, after 10 to 11 years of exposure, O₃ was found to increase soil respiration but have no substantial impact on soil C formation under elevated CO₂ (Section 9.4.6.3).

The evidence is sufficient to infer that there is a causal relationship between O₃ exposure and the alteration of below-ground biogeochemical cycles.

### 2.6.5 Community Composition

In the 2006 O₃ AQCD, the impact of O₃ exposure on species competition and community composition was assessed. Ozone was found to be one of the dominant factors causing a decline in ponderosa and Jeffrey pine in the San Bernardino Mountains in southern California. Ozone exposure also tended to shift the grass-legume mixtures in favor of grass species. Since the 2006 O₃ AQCD, more evidence has shown that O₃ exposure changed the competitive interactions and led to loss of O₃ sensitive species or genotypes. Studies found that the severity of O₃ damage on growth, reproduction and foliar injury varied among species (Section 9.4.3), which provided the biological plausibility for the alteration of community composition. Additionally, research since the last review has shown that O₃ can alter community composition and diversity of soil microbial communities.

The decline of conifer forests under O₃ exposure was continually observed in several regions. Ozone damage was believed to be an important causal factor in the dramatic decline of sacred fir in the valley of Mexico, as well as cembran pine in southern France and the Carpathian Mountains, although several factors, such as drought, insect outbreak and forest management, may also contribute to or even be the dominant factors causing the mortality of the conifer trees. Results from the Aspen FACE site indicated that O₃ could alter community composition of broadleaf forests as well. At the Aspen FACE site, O₃ reduced growth and increased mortality of a sensitive aspen clone, while the O₃ tolerant clone emerged as the dominant clone in the pure aspen community. In the mixed aspen-birch and aspen-maple communities, O₃ reduced the competitive capacity of aspen compared to birch and maple (Section 9.4.7.1).

The tendency for O₃-exposure to shift the biomass of grass-legume mixtures in favor of grass species was reported in the 2006 O₃ AQCD and has been generally confirmed by recent studies. However, in a high elevation mature/species-rich grass-
legume pasture, O₃ fumigation showed no substantial impact on community composition (Section 9.4.7.2).

Ozone exposure not only altered community composition of plant species, but also microorganisms. The shift in community composition of bacteria and fungi has been observed in both natural and agricultural ecosystems, although no general patterns could be identified (Section 9.4.7.3).

The evidence is sufficient to conclude that there is likely to be a causal relationship between O₃ exposure and the alteration of community composition of some ecosystems.

2.6.6 Policy Relevant Considerations

2.6.6.1 Air Quality Indices

Exposure indices are metrics that quantify exposure as it relates to measured plant response (e.g., reduced growth). They are summary measures of monitored ambient O₃ concentrations over time intended to provide a consistent metric for reviewing and comparing exposure-response effects obtained from various studies. No recent information is available since 2006 that alters the basic conclusions put forth in the 2006 and 1996 O₃ AQCDs. These AQCDs focused on the research used to develop various exposure indices to help quantify effects on growth and yield in crops, perennials, and trees (primarily seedlings). The performance of indices was compared through regression analyses of earlier studies designed to support the estimation of predictive O₃ exposure-response models for growth and/or yield of crops and tree (seedling) species.

Another approach for improving risk assessment of vegetation response to ambient O₃ is based on determining the O₃ concentration from the atmosphere that enters the leaf (i.e., flux or deposition). Interest has been increasing in recent years, particularly in Europe, in using mathematically tractable flux models for O₃ assessments at the regional, national, and European scale. While some efforts have been made in the U.S. to calculate O₃ flux into leaves and canopies, little information has been published relating these fluxes to effects on vegetation. There is also concern that not all O₃ stomatal uptake results in a yield reduction, which depends to some degree on the amount of internal detoxification occurring with each particular species. Species having high detoxification capacity may show little relationship between O₃ stomatal uptake and plant response. The lack of data in the U.S. and the lack of understanding of detoxification processes have made this technique less viable for vulnerability and risk assessments in the U.S.
The main conclusions from the 1996 and 2006 O₃ AQCDs regarding indices based on ambient exposure remain valid. These key conclusions can be restated as follows:

- ozone effects in plants are cumulative;
- higher O₃ concentrations appear to be more important than lower concentrations in eliciting a response;
- plant sensitivity to O₃ varies with time of day and plant development stage;
- quantifying exposure with indices that cumulate hourly O₃ concentrations and preferentially weight the higher concentrations improves the explanatory power of exposure/response models for growth and yield, over using indices based on mean and peak exposure values.

Various weighting functions have been used, including threshold-weighted (e.g., SUM06) and continuous sigmoid-weighted (e.g., W126) functions. Based on statistical goodness-of-fit tests, these cumulative, concentration-weighted indices could not be differentiated from one another using data from previous exposure studies. Additional statistical forms for O₃ exposure indices are summarized in Section 9.5 of this ISA. The majority of studies published since the 2006 O₃ AQCD do not change earlier conclusions, including the importance of peak concentrations, and the duration and occurrence of O₃ exposures in altering plant growth and yield.

Given the current state of knowledge and the best available data, exposure indices that cumulate and differentially weight the higher hourly average concentrations and also include the mid-level values continue to offer the most scientifically defensible approach for use in developing response functions and comparing studies, as well as for defining future indices for vegetation protection.

2.6.6.2 Exposure-Response

None of the information on effects of O₃ on vegetation published since the 2006 O₃ AQCD has modified the assessment of quantitative exposure-response relationships that was presented in that document (U.S. EPA, 2006b). This assessment updates the 2006 exposure-response models by computing them using the W126 metric, cumulated over 90 days. Almost all of the experimental research on the effects of O₃ on growth or yield of plants published since 2006 used only two levels of exposure. In addition, hourly O₃ concentration data that would allow calculations of exposure using the W126 metric are generally unavailable. However, two long-term experiments, one with a crop species (soybean), one with a tree species (aspen), have produced data that are used in Section 9.6 to validate the exposure-response models presented in the 2006 O₃ AQCD, and the methodology used to derive them. EPA compared predictions from the models presented in the 2006 O₃ AQCD, updated to use the 90 day 12hr W126 metric, with more recent observations for yield of soybean and biomass growth of trembling aspen. The models were parameterized using data from the National Crop Loss Assessment Network (NCLAN) and EPA’s National Health and Environmental Effects Research Laboratory – Western Ecology Division.
(NHEERL-WED) projects, which were conducted in OTCs. The more recent observations were from experiments using FACE technology, which is intended to provide conditions closer to natural environments than OTC. Observations from these new experiments were exceptionally close to predictions from the models. The accuracy of model predictions for two widely different plant species, grown under very different conditions, provides support for the validity of the models for crops and trees developed using the same methodology and data for other species. However, variability observed among species in the NCLAN and NHEERL-WED projects indicates that the range of sensitivity between and among species is likely quite wide.

Results from several meta-analyses have provided approximate values for responses of yield of soybean, wheat, rice and other crops under broad categories of exposure, relative to charcoal-filtered air. Additional reports have summarized yield data for six crop species under various broad comparative exposure categories, and reviewed 263 studies that reported effects on tree biomass. However, these analyses have proved difficult to compare with exposure-response models, especially given that exposure was not expressed using a common metric (i.e., W126).

## 2.7 The Role of Tropospheric O₃ in Climate Change and UV-B Shielding Effects

Atmospheric O₃ plays an important role in the Earth’s energy budget by interacting with incoming solar radiation and outgoing infrared radiation. Tropospheric O₃ makes up only a small portion of the total column of O₃, but it has important incremental effects on the overall radiation budget. Chapter 10 assesses the specific role of tropospheric O₃ in the earth’s radiation budget and how perturbations in tropospheric O₃ might affect (1) climate through its role as a greenhouse gas, and (2) health, ecology and welfare through its role in shielding the earth’s surface from solar ultraviolet (UV) radiation.

### 2.7.1 Tropospheric Ozone as a Greenhouse Gas

Ozone is an important greenhouse gas, and increases in its abundance in the troposphere may contribute to climate change according to the 2007 climate assessment by the Intergovernmental Panel on Climate Change (IPCC). Models calculate that the global burden of tropospheric O₃ has doubled since the pre-industrial era, while observations indicate that in some regions O₃ may have increased by factors as great as 4 or 5. These increases are tied to the rise in emissions of O₃ precursors from human activity, mainly fossil fuel consumption and agricultural processes.

Figure 2-3 shows the main steps involved in the influence of tropospheric O₃ on climate. Emissions of O₃ precursors including CO, VOCs, CH₄, and NOₓ lead to
production of tropospheric O₃. A change in the abundance of tropospheric O₃ perturbs the radiative balance of the atmosphere, an effect quantified by the radiative forcing metric. The earth-atmosphere-ocean system responds to the forcing with a climate response, typically expressed as a change in surface temperature. Finally, the climate response causes downstream climate-related health and ecosystem impacts, such as redistribution of diseases or ecosystem characteristics due to temperature changes. Feedbacks from both the climate response and downstream impacts can, in turn, affect the abundance of tropospheric O₃ and O₃ precursors through multiple feedback mechanisms as indicated in Figure 2-3. Direct feedbacks are discussed in Section 10.3.2.4 and Section 10.3.3.4, while downstream climate impacts and their feedbacks are extremely complex and outside the scope of this assessment.

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**Figure 2-3**  Schematic illustrating the effects of tropospheric O₃ on climate; including the relationship between precursor emissions, tropospheric O₃ abundance, radiative forcing, climate response, and climate impacts.
Radiative forcing by a greenhouse gas or aerosol is a metric used to quantify the change in balance between radiation coming into and going out of the atmosphere caused by the presence of that substance. Tropospheric O\textsubscript{3} is a major greenhouse gas and radiative forcing agent; evidence from satellite data shows a sharp dip in the outgoing infrared radiation in the 9.6 \textmu m O\textsubscript{3} absorption band. Models calculate that the global average concentration of tropospheric O\textsubscript{3} has doubled since the pre-industrial era, while observations indicate that in some regions O\textsubscript{3} may have increased by factors as great as 4 or 5. These increases are tied to the rise in emissions of O\textsubscript{3} precursors from human activity, mainly fossil fuel consumption and agricultural processes. Overall, the evidence supports a causal relationship between changes in tropospheric O\textsubscript{3} concentrations and radiative forcing.

The impact of the tropospheric O\textsubscript{3} change since pre-industrial times on climate has been estimated to be about 25-40\% of the anthropogenic CO\textsubscript{2} impact and about 75\% of the anthropogenic CH\textsubscript{4} impact according to the IPCC, ranking it third in importance after CO\textsubscript{2} and CH\textsubscript{4} according to the Intergovernmental Panel on Climate Change (IPCC) (see Section 10.3). There are large uncertainties in the magnitude of the radiative forcing estimate attributed to tropospheric O\textsubscript{3}, making the impact of tropospheric O\textsubscript{3} on climate more uncertain than the effect of the longer-lived greenhouse gases. Furthermore, radiative forcing does not take into account the climate feedbacks that could amplify or dampen the actual surface temperature response. Quantifying the change in surface temperature requires a complex climate simulation in which all important feedbacks and interactions are accounted for. The modeled surface temperature response to a given radiative forcing is highly uncertain and can vary greatly among models and from region to region within the same model. Even with these uncertainties, global climate models indicate that tropospheric O\textsubscript{3} has contributed to observed changes in global mean and regional surface temperatures. As a result of such evidence presented in climate modeling studies, there is likely to be a causal relationship between changes in tropospheric O\textsubscript{3} concentrations and effects on climate.

2.7.2 Tropospheric Ozone and UV-B Shielding Effects

UV radiation emitted from the Sun contains sufficient energy when it reaches the Earth to break (photolyze) chemical bonds in molecules, thereby leading to damaging effects on living organisms and materials. Atmospheric O\textsubscript{3} plays a crucial role in reducing exposure to solar UV radiation at the Earth’s surface. Ozone in the stratosphere is responsible for the majority of this shielding effect, as approximately 90\% of total atmospheric O\textsubscript{3} is located there over mid-latitudes. Ozone in the troposphere provides supplemental shielding of radiation in the wavelength band from 280-315 nm, referred to as UV-B radiation. UV-B radiation has important effects on human health and ecosystems, and is associated with materials damage.

Human health effects associated with solar UV-B radiation exposure include erythema, skin cancer, ocular damage, and immune system suppression. A potential
human health benefit of increased UV-B exposure involves the UV-induced production of vitamin D which may help reduce the risk of metabolic bone disease, type I diabetes, mellitus, and rheumatoid arthritis, and may provide beneficial immunomodulatory effects on multiple sclerosis, insulin-dependent diabetes mellitus, and rheumatoid arthritis. Ecosystem and materials damage effects associated with solar UV-B radiation exposure include terrestrial and aquatic ecosystem impacts, alteration of biogeochemical cycles, and degradation of man-made materials.

EPA has found no published studies that adequately examine the incremental health or welfare effects (adverse or beneficial) attributable specifically to changes in UV-B exposure resulting from perturbations in tropospheric O₃ concentrations. While the effects are expected to be small, they cannot yet be critically assessed within reasonable uncertainty. Overall, the evidence is inadequate to determine if a causal relationship exists between changes in tropospheric O₃ concentrations and effects on health and welfare related to UV-B shielding.

### 2.8 Summary of Causal Determinations for Health Effects and Welfare Effects

This chapter has provided an overview of the underlying evidence used in making the causal determinations for the health and welfare effects of O₃. This review builds upon the conclusions of the previous AQCDs for O₃.

The evaluation of the epidemiologic, toxicological, and controlled human exposure studies published since the completion of the 2006 O₃ AQCD have provided additional evidence for O₃-related health outcomes. Table 2-3 provides an overview of the causal determinations for all of the health outcomes evaluated. Causal determinations for O₃ and welfare effects are included in Table 2-4, while causal determinations for climate change and UV-B shielding effects are in Table 2-5. Detailed discussions of the scientific evidence and rationale for these causal determinations are provided in subsequent chapters of this ISA.
Table 2-3  Summary of O₃ causal determinations by exposure duration and health outcome.

<table>
<thead>
<tr>
<th>Health Outcome</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Short-Term Exposure to O₃</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Respiratory effects</td>
<td>The overall evidence supports a causal relationship between acute ambient O₃ exposures and increased respiratory morbidity outcomes.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Cardiovascular effects</td>
<td>The limited evidence is highly suggestive that O₃ directly and/or indirectly contributes to cardiovascular-related morbidity, but much remains to be done to more fully substantiate the association.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Central nervous system effects</td>
<td>Toxicological studies report that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short and long term memory, sleep patterns, and histological signs of neurodegeneration.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>The evidence is highly suggestive that O₃ directly or indirectly contributes to non-accidental and cardiopulmonary-related mortality.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td><strong>Long-term Exposure to O₃</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Respiratory effects</td>
<td>The current evidence is suggestive but inconclusive for respiratory health effects from long-term O₃ exposure.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Cardiovascular Effects</td>
<td>No studies from previous review</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Reproductive and developmental effects</td>
<td>Limited evidence for a relationship between air pollution and birth-related health outcomes, including mortality, premature births, low birth weights, and birth defects, with little evidence being found for O₃ effects.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Central nervous system effects</td>
<td>Evidence regarding chronic exposure and neurobehavioral effects was not available.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
<tr>
<td>Cancer</td>
<td>Little evidence for a relationship between chronic O₃ exposure and increased risk of lung cancer.</td>
<td>Inadequate to infer a Causal Relationship</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>There is little evidence to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans.</td>
<td>Suggestive of a Causal Relationship</td>
</tr>
</tbody>
</table>
Table 2-4  Summary of \( \text{O}_3 \) causal determination for welfare effects.

<table>
<thead>
<tr>
<th>Vegetation and Ecosystem Effects</th>
<th>Conclusions from 2006 ( \text{O}_3 ) AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Visible Foliar Injury Effects on Vegetation</td>
<td>Data published since the 1996 ( \text{O}_3 ) AQCD strengthen previous conclusions that there is strong evidence that current ambient ( \text{O}_3 ) concentrations cause impaired aesthetic quality of many native plants and trees by increasing foliar injury.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Vegetation Growth</td>
<td>Data published since the 1996 ( \text{O}_3 ) AQCD strengthen previous conclusions that there is strong evidence that current ambient ( \text{O}_3 ) concentrations cause decreased growth and biomass accumulation in annual, perennial and woody plants, including agronomic crops, annuals, shrubs, grasses, and trees.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Productivity in Terrestrial Ecosystems</td>
<td>There is evidence that ( \text{O}_3 ) is an important stressor of ecosystems and that the effects of ( \text{O}_3 ) on individual plants and processes are scaled up through the ecosystem, affecting net primary productivity.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Carbon (C) Sequestration in Terrestrial Ecosystems</td>
<td>Limited studies from previous review</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Reduced Yield and Quality of Agricultural Crops</td>
<td>Data published since the 1996 ( \text{O}_3 ) AQCD strengthen previous conclusions that there is strong evidence that current ambient ( \text{O}_3 ) concentrations cause decreased yield and/or nutritive quality in a large number of agronomic and forage crops.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Ecosystem Water Cycling</td>
<td>Ecosystem water quantity may be affected by ( \text{O}_3 ) exposure at the landscape level.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Below-ground Biogeochemical Cycles</td>
<td>Ozone-sensitive species have well known responses to ( \text{O}_3 ) exposure, including altered C allocation to below-ground tissues, and altered rates of leaf and root production, turnover, and decomposition. These shifts can affect overall C and N loss from the ecosystem in terms of respired C, and leached aqueous dissolved organic and inorganic C and N.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Community Composition</td>
<td>Ozone may be affecting above- and below-ground community composition through impacts on both growth and reproduction. Significant changes in plant community composition resulting directly from ( \text{O}_3 ) exposure have been demonstrated.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
</tbody>
</table>
Table 2-5  Summary of O₃ causal determination for climate and UV-B shielding effects.

<table>
<thead>
<tr>
<th>Effects</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Radiative Forcing</td>
<td>Climate forcing by O₃ at the regional scale may be its most important impact on climate.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Climate Change</td>
<td>While more certain estimates of the overall importance of global-scale forcing due to tropospheric O₃ await further advances in monitoring and chemical transport modeling, the overall body of scientific evidence suggests that high concentrations of O₃ on the regional scale could have a discernible influence on climate, leading to surface temperature and hydrological cycle changes.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Health and Welfare Effects</td>
<td>UV-B has not been studied in sufficient detail to allow for a credible health benefits assessment. In conclusion, the effect of changes in surface-level O₃ concentrations on UV-induced health outcomes cannot yet be critically assessed within reasonable uncertainty.</td>
<td>Inadequate to Determine if a Causal Relationship Exists</td>
</tr>
</tbody>
</table>
References


Zhang, L; Jacob, DJ; Downey, NV; Wood, DA; Blewitt, D; Carouge, CC; Van donkelaar, A; Jones, DBA; Murray, LT; Wang, Y. (2011). Improved estimate of the policy-relevant background ozone in the United States using the GEOS-Chem global model with 1/2 2/3 horizontal resolution over North America. Atmos Environ 45: 6769-6776. [http://dx.doi.org/10.1016/j.atmosenv.2011.07.054](http://dx.doi.org/10.1016/j.atmosenv.2011.07.054)
3 ATMOSPHERIC CHEMISTRY AND AMBIENT CONCENTRATIONS

3.1 Introduction

In the stratosphere, ozone (O$_3$) serves the beneficial role of absorbing the Sun’s harmful ultraviolet radiation and preventing the majority of this radiation from reaching the Earth’s surface. In the troposphere, however, O$_3$ and other photochemical oxidants are air pollutants that can exert harmful effects on humans, animals, and vegetation. This chapter discusses the atmospheric chemistry associated with tropospheric O$_3$ and other related photochemical oxidants and provides a detailed description of their surface-level concentrations. The focus of this chapter is on O$_3$ since it is the NAAQS indicator for all photochemical oxidants. To the extent possible, other photochemical oxidants are discussed, but limited information is currently available. Although O$_3$ is involved in reactions in indoor air, the focus in this chapter will be on chemistry occurring in outdoor, ambient air.

The material in this chapter is organized as follows. Section 3.2 outlines the physical and chemical processes involved in O$_3$ formation and removal. Section 3.3 describes the latest methods used to model global O$_3$ concentrations, and Section 3.4 describes the application of these methods for estimating background concentrations of O$_3$ that are useful for risk and policy assessments informing decisions about the NAAQS. Section 3.5 includes a comprehensive description of available O$_3$ monitoring techniques and monitoring networks, while Section 3.6 presents information on the spatial and temporal variability of O$_3$ concentrations across the U.S. and their associations with other pollutants using available monitoring data. Section 3.7 summarizes the main conclusions from Chapter 3. Finally, Section 3.8 provides supplemental material on atmospheric model simulations of background O$_3$ concentrations (referenced in Section 3.4) and Section 3.9 contains supplemental material on observed ambient O$_3$ concentrations (referenced in Section 3.6).

3.2 Physical and Chemical Processes

Ozone in the troposphere is a secondary pollutant formed by photochemical reactions of precursor gases and is not directly emitted from specific sources. Ozone and other oxidants, such as peroxyacetyl nitrate (PAN) and H$_2$O$_2$ form in polluted areas by atmospheric reactions involving two main classes of precursor pollutants: VOCs and NO$_x$. $^1$ Carbon monoxide (CO) is also important for O$_3$ formation in polluted areas and in the remote troposphere. The formation of O$_3$, other oxidants and oxidation products from these precursors is a complex, nonlinear function of many factors.

$^1$ The term VOCs refers to all organic gas-phase compounds in the atmosphere, both biogenic and anthropogenic in origin. This definition excludes CO and CO$_2$, NO$_x$, also referred to as nitrogen oxides, is equal to the sum of NO and NO$_2$. 

3-1
including (1) the intensity and spectral distribution of sunlight; (2) atmospheric mixing; (3) concentrations of precursors in the ambient air and the rates of chemical reactions of these precursors; and (4) processing on cloud and aerosol particles.

Ozone is present not only in polluted urban atmospheres, but throughout the troposphere, even in remote areas of the globe. The same basic processes involving sunlight-driven reactions of NOX, VOCs, and CO contribute to O3 formation throughout the troposphere. These processes also lead to the formation of other photochemical products, such as PAN, HNO₃, and H₂SO₄, and to other compounds, such as HCHO and other carbonyl compounds, and to secondary components of particulate matter.

A schematic overview of the major photochemical cycles influencing O3 in the troposphere and the stratosphere is given in Figure 3-1. Included in the figure are reactions involving radicals derived from man-made chemicals and that are responsible for depleting stratospheric O3. Most (approximately 90%) of the total O3 column in the earth’s atmosphere resides in the stratosphere, and it is responsible for absorbing harmful solar ultraviolet radiation, the harmful effects of which are discussed in Chapter 10. This solar ultraviolet radiation also initiates the photochemical reactions that are responsible for producing O3 in the troposphere. The processes responsible for producing summertime O3 episodes are fairly well understood, and were covered in detail in the 2006 O3 AQCD (U.S. EPA, 2006b). This section focuses on topics that form the basis for discussions in later chapters, and for which there is substantial new information since the previous O3 review.
Major episodes of high O₃ concentrations in the eastern U.S. and in Europe are associated with slow moving high pressure systems. High pressure systems during the warmer seasons are associated with the sinking of air, resulting in warm, generally cloudless skies, with light winds. The sinking of air results in the development of stable conditions near the surface which inhibit or reduce the vertical mixing of O₃ precursors. Photochemical activity involving these precursors is enhanced because of higher temperatures and the availability of sunlight during the warmer seasons. In the eastern U.S., concentrations of O₃ and other secondary pollutants are determined by meteorological and chemical processes extending typically over areas of several hundred thousand square kilometers (Civerolo et al., 2003; Rao et al., 2003). Ozone episodes are thus best regarded as regional in nature. The conditions conducive to formation of high O₃ can persist for several days. These conditions have been described in greater detail in the 1996 and 2006 O₃ AQCDs (U.S. EPA, 2006b, 1996a). The transport of pollutants downwind of major urban centers is characterized by the development of urban plumes. Mountain barriers limit mixing (as in Los Angeles and Mexico City) and result in a higher frequency and duration of days with high O₃ concentrations. However, orographic lifting over the San Gabriel Mountains results in O₃ transport from Los Angeles to areas hundreds of kilometers downwind (e.g., in Colorado and Utah) (Langford et al., 2009). Ozone concentrations in southern urban areas (such as Houston, TX and Atlanta, GA) tend
to decrease with increasing wind speed. In northern U.S. cities (such as Chicago, IL; New York, NY; Boston, MA; and Portland, ME), the average O₃ concentrations over the metropolitan areas increase with wind speed, indicating that transport of O₃ and its precursors from upwind areas is important (Schichtel and Husar, 2001; Husar and Renard, 1998).

Aircraft observations indicate that there can be substantial differences in mixing ratios of key species between the surface and the overlying atmosphere (Berkowitz and Shaw, 1997; Fehsenfeld et al., 1996). In particular, mixing ratios of O₃ can (depending on time and location) be higher in the lower free troposphere (aloft) than in the planetary boundary layer (PBL) during multiday O₃ episodes (Taubman et al., 2006; Taubman et al., 2004). Convective processes and turbulence transport O₃ and other pollutants both upward and downward throughout the planetary boundary layer and the free troposphere. During the day, convection (driven by heating of the earth’s surface) results in a deeper PBL, with vertically well mixed O₃ and precursors. As solar heating of the surface decreases going into night, the daytime boundary layer collapses leaving behind O₃ and its precursors in a residual layer above a shallow nighttime boundary layer. Pollutants in the residual layer have now become essentially part of the free troposphere, as described in Annex AX2.3.2 of the 2006 O₃ AQCD (U.S. EPA, 2006b). Winds in the free troposphere tend to be stronger than those closer to the surface and so are capable of transporting pollutants over long distances. Thus, O₃ and its precursors can be transported vertically by convection into the upper part of the mixed layer on one day, then transported overnight as a layer of elevated mixing ratios, and then entrained into a growing convective boundary layer downwind and brought back down to the surface.

High O₃ concentrations showing large diurnal variations at the surface in southern New England were associated with the presence of such layers (Berkowitz et al., 1998). Winds several hundred meters above the ground can bring pollutants from the west, even though surface winds are from the southwest during periods of high O₃ in the eastern U.S. (Blumenthal et al., 1997). These considerations suggest that in many areas of the U.S., O₃ and its precursors can be transported over hundreds if not thousands of kilometers.

Nocturnal low level jets (LLJs) are an efficient means for transporting pollutants that have been entrained into the residual boundary layer over hundreds of kilometers. LLJs are most prevalent in the central U.S. extending northward from eastern Texas, and along the Atlantic states extending southwest to northeast. LLJs have also been observed off the coast of California. Turbulence induced by wind shear associated with LLJs brings pollutants to the surface and results in secondary O₃ maxima during the night and early morning in many locations (Corsmeier et al., 1997). Comparison of observations at low elevation surface sites with those at nearby high elevation sites at night can be used to discern the effects of LLJs. For example, Fischer (2004) found occasions when O₃ at the base of Mt. Washington during the night was much higher than typically observed, and closer to those observed at the summit of Mt. Washington. They suggested that mechanically driven turbulence due to wind shear caused O₃ from aloft to penetrate the stable nocturnal inversion thus causing O₃ to
increase near the base of Mt. Washington. The high wind speeds causing this mechanically driven turbulence could have resulted from the development of a LLJ. Stratospheric intrusions and intercontinental transport of O₃ are also important and are covered in Section 3.4 in relation to background concentrations.

### 3.2.1 Sources of Precursors Involved in O₃ Formation

Emissions of O₃ precursor compounds (NOₓ, VOCs, and CO) can be divided into natural and anthropogenic source categories. Natural sources can be further divided into biogenic from vegetation, microbes, and animals, and abiotic from biomass combustion, lightning, and geogenic sources. However, the distinction between natural and anthropogenic sources is often difficult to make in practice, as human activities directly or indirectly affect emissions from what would have been considered natural sources during the pre-industrial era. Thus, emissions from plants and animals used in agriculture have been referred to as anthropogenic or biogenic in different applications. Wildfire emissions can be considered natural, except that forest management practices can lead to buildup of fuels on the forest floor, thereby altering the frequency and severity of forest fires.

Estimates of emissions for NOₓ, VOCs, and CO from the 2005 National Emissions Inventory (NEI) (U.S. EPA, 2008a) are shown in Figure 3-2 to provide a general indication of the relative importance of the different sources in the U.S. as a whole. The magnitudes of the sources are strongly location and time dependent and so should not be used to apportion sources of exposure. Shown in Figure 3-2 are Tier 1 categories. The miscellaneous category can be quite large compared to total emissions, especially for CO and VOCs. The miscellaneous category includes agriculture and forestry, wildfires, prescribed burns, and a much more modest contribution from structural fires.
Note: NO\textsubscript{x} (top), VOCs (middle), and CO (bottom) in the U.S. in million metric tons (MT) per year. Source: U.S. EPA (2008a).

Figure 3-2 Estimated anthropogenic emissions of O\textsubscript{3} precursors for 2005.
Anthropogenic NO\textsubscript{X} emissions are associated with combustion processes. Most emissions are in the form of NO, which is formed at high combustion temperatures from atmospheric nitrogen (N\textsubscript{2}) and oxygen (O\textsubscript{2}) and from fuel nitrogen (N). According to the 2005 NEI, the largest sources of NO\textsubscript{X} are on- and off-road (such as construction equipment, agricultural equipment, railroad trains, ships, and aircraft) mobile sources and electric power generation plants. Emissions of NO\textsubscript{X} therefore are highest in areas having a high density of power plants and in urban regions having high traffic density. Dallmann and Harley (2010) compared NO\textsubscript{X} emissions estimates from the 2005 NEI mobile sector data with an alternative method based on fuel consumption and found reasonable agreement in total U.S. anthropogenic emissions between the two techniques (to within about 5%). However, emissions from on-road diesel engines in the fuel based inventory constituted 46% of total mobile source NO\textsubscript{X} compared to 35% in the EPA inventory. As a result, emissions from on-road diesel engines in the fuel based approach are even larger than electric power generation as estimated in the 2005 NEI, and on-road diesel engines might represent the largest single NO\textsubscript{X} source category. Differences between the two techniques are largely accounted for by differences in emissions from on-road gasoline engines. Uncertainties in the fuel consumption inventory ranged from 3% for on-road gasoline engines to 20% for marine sources, and in the EPA inventory uncertainties ranged from 16% for locomotives to 30% for off-road diesel engines. It should be noted that the on-road diesel engine emissions estimate by Dallmann and Harley (2010) is still within the uncertainty of the EPA estimate (22%). Because of rapid changes to heavy duty diesel NO\textsubscript{X} controls, emissions are likely to also rapidly change.

Satellite-based techniques have been used to obtain tropospheric concentrations of O\textsubscript{3} precursors (e.g., NO\textsubscript{2}, VOCs, and CO). Such satellite-based measurements provide a large-scale picture of spatial and temporal distribution of NO\textsubscript{2}, VOCs, and CO that can be used to evaluate emissions inventories produced using the bottom-up approach and to produce top-down emissions inventories of these species. Although there are uncertainties associated with satellite-based measurements, several studies have shown the utility of top-down constraints on the emissions of O\textsubscript{3} precursors (McDonald-Buller et al., 2011 and references therein). Following mobile sources, power plants are considered the second largest anthropogenic source of NO\textsubscript{X}. Over the past decade, satellite measurements have shown appreciable reductions in NO\textsubscript{X} power plant emissions across the U.S. as a result of emission abatement strategies (Stavrakou et al., 2008; Kim et al., 2006). For instance, Kim et al. (2006) observed a 34% reduction in NO\textsubscript{X} emission over the Ohio River Valley from 1999-2006 due to such strategies. Based on these results, less than 25% of anthropogenic NO\textsubscript{X} emissions were expected to originate from power plants in this region. Uncertainty in NO\textsubscript{X} satellite measurements are impacted by several factors, such as cloud and aerosol properties, surface albedo, stratospheric NO\textsubscript{X} concentration, and solar zenith angle. Boersma et al. (2004) estimated an overall uncertainty between 35-60% for satellite-retrieved NO\textsubscript{X} measurements in urban, polluted regions. Although trends in satellite-retrieved NO\textsubscript{X} power plant emissions reported by Kim et al. (2006) are uncertain to some extent, similar reductions were reported by region-wide power plant measurements (e.g., Continuous Emission Monitoring System observations, CEMS).
Major natural sources of NO$_X$ in the U.S. include lightning, soils, and wildfires. Uncertainties in natural NO$_X$ emissions are much larger than for anthropogenic NO$_X$ emissions. Fang et al. (2010) estimated lightning generated NO$_X$ of ~0.6 MT for July 2004. This value is ~40% of the anthropogenic emissions for the same period, but the authors estimated that ~98% is formed in the free troposphere and so contributions to the surface NO$_X$ burden are low because most of this NO$_X$ is oxidized to nitrate containing species during downward transport into the planetary boundary layer. The remaining 2% is formed within the planetary boundary layer. Both nitrifying and denitrifying organisms in the soil can produce NO$_X$, mainly in the form of NO. Emission rates depend mainly on fertilization amount and soil temperature and moisture. Nationwide, about 60% of the total NO$_X$ emitted by soils is estimated to occur in the central corn belt of the United States. Spatial and temporal variability in soil NO$_X$ emissions leads to considerable uncertainty in emissions estimates. However, these emissions are relatively low, only ~0.97 MT/year, or about 6% of anthropogenic NO$_X$ emissions. However, these emissions occur mainly during summer when O$_3$ is of most concern and occur across the entire country including areas where anthropogenic emissions are low.

Hundreds of VOCs, containing mainly 2 to ~12 carbon (C) atoms, are emitted by evaporation and combustion processes from a large number of anthropogenic sources. The two largest anthropogenic source categories in the U.S. EPA’s emissions inventories are industrial processes and transportation. Emissions of VOCs from highway vehicles account for roughly two-thirds of the transportation-related emissions. The accuracy of VOC emission estimates is difficult to determine, both for stationary and mobile sources. Evaporative emissions, which depend on temperature and other environmental factors, compound the difficulties of assigning accurate emission factors. In assigning VOC emission estimates to the mobile source category, models are used that incorporate numerous input parameters (e.g., type of fuel used, type of emission controls, and age of vehicle), each of which has some degree of uncertainty.

On the U.S. and global scales, emissions of VOCs from vegetation are much larger than those from anthropogenic sources. Emissions of VOCs from anthropogenic sources in the 2005 NEI were ~17 MT/year (wildfires constitute ~1/6 of that total and were included in the 2005 NEI under the anthropogenic category, but see Section 3.4 for how wildfires are treated for background O$_3$ considerations), but were 29 MT/year from biogenic sources. Uncertainties in both biogenic and anthropogenic VOC emission inventories prevent determination of the relative contributions of these two categories, at least in many areas. Vegetation emits substantial quantities of VOCs, such as terpenoid compounds (isoprene, 2-methyl-3-buten-2-ol, monoterpenes), compounds in the hexanal family, alkenes, aldehydes, organic acids, alcohols, ketones, and alkanes. The major chemicals emitted by plants are isoprene (40%), other terpenoid and sesqui-terpenoid compounds (25%) and the remainder consists of assorted oxygenated compounds and hydrocarbons according to the 2005 NEI. Most biogenic emissions occur during the summer because of their dependence on temperature and incident sunlight. Biogenic emissions are also higher in southern states than in northern states for these reasons and because of species variations.
The uncertainty in natural emissions is about 50% for isoprene under midday summer conditions and could be as much as a factor of ten higher for some compounds (Guenther et al., 2000). In EPA’s regional modeling efforts, biogenic emissions of VOCs are estimated using the Biogenic Emissions Inventory System (BEIS) model (U.S. EPA, 2010b) with data from the Biogenic Emissions Landuse Database (BELD) and annual meteorological data. However, other emissions models are used such as Model of Emissions of Gases and Aerosols from Nature (MEGAN) (Guenther et al., 2006), especially in global modeling efforts.

Satellite measurements of HCHO, produced by the oxidation of isoprene and other VOCs, have also been used to estimate biogenic VOC emissions attributed to isoprene (Millet et al., 2008; Millet et al., 2006). Millet et al. (2008) demonstrated that both satellite-based and model techniques capture the spatial variability of biogenic isoprene emissions in the U.S. reasonably well (satellite [versus MEGAN] isoprene estimates, $R^2 = 0.48$ or 0.68 depending on vegetation database used). However, MEGAN tends to overestimate emissions compared to satellite-based measurements. The uncertainty in satellite derived isoprene emissions is roughly 40%, based on combined uncertainty in satellite retrieval and isoprene yield from isoprene oxidation (Millet et al., 2006), which is similar to the error associated with model-based techniques (~50%) (e.g., Millet et al., 2006; Guenther et al., 2000).

Anthropogenic CO is emitted primarily by incomplete combustion of carbon-containing fuels. In general, any increase in fuel oxygen content, burn temperature, or mixing time in the combustion zone will tend to decrease production of CO relative to CO$_2$. However, it should be noted that controls mute the response of CO formation to fuel-oxygen. CO emissions from large fossil-fueled power plants are typically very low since the boilers at these plants are tuned for highly efficient combustion with the lowest possible fuel consumption. Additionally, the CO-to-CO$_2$ ratio in these emissions is shifted toward CO$_2$ by allowing time for the furnace flue gases to mix with air and be oxidized by OH to CO$_2$ in the hot gas stream before the OH concentrations drop as the flue gases cool. Nationally, on-road mobile sources constituted about half of total CO emissions in the 2005 NEI. When emissions from off-highway (non-road) vehicles are included, it can be seen from Figure 3-2 that all mobile sources accounted for about three-quarters of total anthropogenic CO emissions in the U.S.

Analyses by Harley et al. (2005) and Parrish (2006) are consistent with the suggestion in Pollack et al. (2004) that the EPA MOBILE6 vehicle emissions model (U.S. EPA, 2010d) overestimates vehicle CO emissions by a factor of ~2. Field measurements by Bishop and Stedman (2008) were in accord with the Parrish (2006) findings that the measured trends of CO and NO$_X$ concentrations from mobile sources in the U.S. indicated that modeled CO emission estimates were substantially too high. Hudman et al. (2008) found that the NEI overestimated anthropogenic CO emissions by 60% for the eastern U.S. during the period July 1-August 15, 2004 based on comparison of aircraft observations of CO from the International Consortium for Atmospheric Research on Transport and Transformation (ICARTT) campaign (Fehsenfeld et al., 2006) and results from a tropospheric chemistry model.
(GEOS-Chem). Improvements in emissions technologies not correctly represented in MOBILE emission models have been suggested as one cause for this discrepancy. For example, Pokharel et al. (2003, 2002) demonstrated substantial decrements in the CO fraction of tailpipe exhaust in several U.S. cities and Burgard et al. (2006) documented improvements in emission from heavy-duty on-road diesel engines. Some of the largest errors in the MOBILE models are addressed in the successor model, MOVES (U.S. EPA, 2011f).

Estimates of biogenic CO emissions in the 2005 NEI are made in a manner similar to that for VOCs. National biogenic emissions, excluding fires, were estimated to contribute ~7% and wildfires added another ~16% to the national CO emissions total. Photodecomposition of organic matter in oceans, rivers, lakes, and other surface waters, and from soil surfaces also releases CO (Goldstein and Galbally, 2007). However, soils can act as a CO source or a sink depending on soil moisture, UV flux reaching the soil surface, and soil temperature (Conrad and Seiler, 1985). Soil uptake of CO is driven by anaerobic bacteria (Inman et al., 1971). Emissions of CO from soils appear to occur by abiogenic processes, such as thermodecomposition or photodecomposition of organic matter. In general, warm and moist conditions found in most soils favor CO uptake, whereas hot and dry conditions found in deserts and some savannas favor the release of CO (King, 1999).

### 3.2.2 Gas Phase Reactions Leading to O₃ Formation

Photochemical processes involved in O₃ formation have been extensively reviewed in a number of books (Jacobson, 2002; Jacob, 1999; Seinfeld and Pandis, 1998; Finlayson-Pitts and Pitts, 1986) and in the 1996 and 2006 O₃ AQCDs (U.S. EPA, 2006b, 1996a). The photochemical formation of O₃ in the troposphere proceeds through the oxidation of NO to nitrogen dioxide (NO₂) by organic-peroxy (RO₂) or hydro-peroxy (HO₂) radicals. The peroxy radicals oxidizing NO to NO₂ are formed during the oxidation of VOCs as presented in Annex AX2.2.2 of the 2006 O₃ AQCD (U.S. EPA, 2006b). The photolysis of NO₂ yields NO and a ground-state oxygen atom, O(3P), which then reacts with molecular oxygen to form O₃.

VOCs important for the photochemical formation of O₃ include alkanes, alkenes, aromatic hydrocarbons, carbonyl compounds (e.g., aldehydes and ketones), alcohols, organic peroxides, and halogenated organic compounds (e.g., alkyl halides). This array of compounds encompasses a wide range of chemical properties and lifetimes: isoprene has an atmospheric lifetime of approximately an hour, whereas methane has an atmospheric lifetime of about a decade.

In urban areas, compounds representing all classes of VOCs and CO are important for O₃ formation. In non-urban vegetated areas, biogenic VOCs emitted from vegetation tend to be the most important. In the remote troposphere, methane (CH₄) and CO are the main carbon-containing precursors to O₃ formation. The oxidation of VOCs is initiated mainly by reaction with hydroxyl (OH) radicals. The primary source of OH radicals in the atmosphere is the reaction of electronically excited
oxygen atoms, \( O(1D) \), with water vapor. \( O(1D) \) is produced by the photolysis of \( O_3 \) in the Hartley bands. In polluted areas, the photolysis of aldehydes (e.g., HCHO), HONO and H\(_2\)O\(_2\) can also be appreciable sources of OH, or HO\(_2\) radicals that can rapidly be converted to OH (Eisele et al., 1997). Ozone can oxidize alkenes, as can NO\(_3\) radicals. NO\(_3\) radicals are most effective at night when they are most abundant. In coastal environments and other selected environments, atomic Cl and Br radicals can also initiate the oxidation of VOCs as discussed in Annex AX2.2.3 of the 2006 \( O_3 \) AQCD (U.S. EPA, 2006b). It was also emphasized in Annex AX2.2.9 of the 2006 \( O_3 \) AQCD (U.S. EPA, 2006b) that the reactions of oxygenated VOCs are important components of \( O_3 \) formation. They may be present in ambient air not only as the result of the atmospheric oxidation of hydrocarbons but also by direct emissions. For example, motor vehicles (including compressed natural gas vehicles) and some industrial processes emit formaldehyde (Rappenglück et al., 2009) and vegetation emits methanol.

There are a large number of oxidized N-containing compounds in the atmosphere including NO, NO\(_2\), NO\(_3\), HNO\(_2\), HNO\(_3\), N\(_2\)O\(_5\), HNO\(_4\), PAN (and its homologues), and other organic nitrates (such as alkyl nitrates, isoprene nitrates), and particulate nitrate. Collectively these species are referred to as NO\(_Y\). Oxidized nitrogen compounds are emitted to the atmosphere mainly as NO, which rapidly interconverts with NO\(_2\) and so NO and NO\(_2\) are often “lumped” together into their own group or family, which is referred to as NO\(_X\). All the other species mentioned above in the definition of NO\(_Y\) are products of NO\(_X\) reactions are referred to as NO\(_Z\), such that NO\(_Y\) = NO\(_X\) + NO\(_Z\). The major reactions involving interconversions of oxidized N species were covered in Annex AX2.2.4 of the 2006 \( O_3 \) AQCD (U.S. EPA, 2006b). Mollner et al. (2010) identified pernitrous acid (HOONO), an unstable isomer of nitric acid, as a product of the major gas phase reaction forming HNO\(_3\). However, since pernitrous acid is unstable, it is not a substantial reservoir for NO\(_X\). This finding stresses the importance of identifying products in addition to measuring the rate of disappearance of reactants in kinetic studies.

The photochemical cycles by which the oxidation of hydrocarbons leads to \( O_3 \) production are best understood by considering the oxidation of methane, structurally the simplest VOC. The CH\(_4\) oxidation cycle serves as a model for the chemistry of the relatively clean or unpolluted troposphere (although this is a simplification because vegetation releases large quantities of complex VOCs, such as isoprene, into the atmosphere). In the polluted atmosphere, the underlying chemical principles are the same, as discussed in Annex AX2.2.5 of the 2006 \( O_3 \) AQCD (U.S. EPA, 2006b). The conversion of NO to NO\(_2\) occurring with the oxidation of VOCs is accompanied by the production of \( O_3 \) and the efficient regeneration of the OH radical, which in turn can react with other VOCs as shown in Figure 3-1.

The oxidation of alkanes and alkenes in the atmosphere has been treated in depth in the 1996 \( O_3 \) AQCD (U.S. EPA, 1996a) and was updated in Annexes AX2.2.6 and AX2.2.7 of the 2006 \( O_3 \) AQCD (U.S. EPA, 2006b). In contrast to simple hydrocarbons containing one or two C atoms, detailed kinetic information about the gas phase oxidation pathways of many anthropogenic hydrocarbons (e.g., aromatic
compounds such as benzene and toluene), biogenic hydrocarbons (e.g., isoprene, the monoterpenes), and their intermediate oxidation products (e.g., peroxides, nitrates, carbonyls and epoxides) is lacking. This information is crucial even for compounds formed in low yields, such as isoprene epoxides, as they are major precursors to secondary organic aerosol formation (see, e.g., Surratt et al., 2010). Reaction with OH radicals represents the major loss process for alkanes. Reaction with chlorine (Cl) atoms is an additional sink for alkanes. Stable products of alkane photo-oxidation are known to include a wide range of compounds and concentrations including carbonyl compounds, alkyl nitrates, and d-hydroxycarbonyls. Major uncertainties in the atmospheric chemistry of the alkanes concern the chemistry of alkyl nitrate formation; these uncertainties affect the amount of NO-to-NO₂ conversion occurring and, hence, the amounts of O₃ formed during photochemical degradation of the alkanes.

The reaction of OH radicals with aldehydes produced during the oxidation of alkanes forms acyl (R’CO) radicals, and acyl peroxy radicals (R’C(O)−O₂) are formed by the further addition of O₂. As an example, the oxidation of ethane (C₂H₅-H) yields acetaldehyde (CH₃-CHO). The reaction of CH₃-CHO with OH radicals yields acetyl radicals (CH₃-CO). The acetyl radicals will then participate with O₂ in a termolecular recombination reaction to form acetyl peroxy radicals, which can then react with NO to form CH₃ + CO₂ or they can react with NO₂ to form PAN. PAN acts as a temporary reservoir for NO₂. Upon the thermal decomposition of PAN, either locally or elsewhere, NO₂ is released to participate in the O₃ formation process again.

Alkenes react in ambient air with OH, NO₃, and Cl radicals and with O₃. All of these reactions are important atmospheric transformation processes, and all proceed by initial addition to the carbon double bonds. Major products of alkene photo-oxidation include carbonyl compounds. Hydroxynitrates and nitratocarbonyls, and decomposition products from the energy-rich biradicals formed in alkene-O₃ reactions are also produced. Major uncertainties in the atmospheric chemistry of the alkenes concern the products and mechanisms of their reactions with O₃, especially the yields of radicals that participate in O₃ formation. Examples of oxidation mechanisms of complex alkanes and alkenes can be found in comprehensive texts such as Seinfeld and Pandis (1998).

Although the photochemistry of isoprene is crucial for understanding O₃ formation, there are major uncertainties in its oxidation pathways that still need to be addressed. Apart from the effects of the oxidation of isoprene on production of radicals and O₃ formation, isoprene nitrates (RONO₂) appear to play an important role as NOₓ reservoirs over the eastern U.S. (e.g., Perring et al., 2009). Their decomposition leads to the recycling of NOₓ, which can participate in the O₃ formation process. Laboratory and field-based approaches support yields for RONO₂ formation from isoprene oxidation ranging from 4% to 12% (see summaries in, Lockwood et al., 2010; Perring et al., 2009; Horowitz et al., 2007; von Kuhlmann et al., 2004). The rate at which RONO₂ reacts to recycle NOₓ is poorly understood (Archibald et al., 2010; Paulot et al., 2009) with ranges from 0 to 100% in global chemical
transport models. This range affects the sign of the \(O_3\) response to changes in biogenic VOC emissions as well as the sensitivity of \(O_3\) to changes in NO\(_X\) emissions (Archibald et al., 2011; Ito et al., 2009; Weaver et al., 2009; Horowitz et al., 2007; Fiore et al., 2005). In models that assume zero RONO\(_2\) recycling (Zhang et al., 2011; Wu et al., 2007; Fiore et al., 2003) \(O_3\) production is suppressed relative to a model that recycles NO\(_X\) from RONO\(_2\) (Kang et al., 2003). A related issue concerns the lack of regeneration of OH + HO\(_2\) radicals especially in low NO\(_X\) (<~1 ppb) environments. The isomerization of the isoprene peroxy radicals that are formed after initial OH attack and subsequent reactions could help resolve this problem (Peeters and Müller, 2010; Peeters et al., 2009) and result in increases in OH concentrations from 20 to 40% over the southeastern U.S. (Archibald et al., 2011). However, the effectiveness of this pathway is uncertain and depends on the fraction of isoprene-peroxy radicals reacting by isomerization. Crounse et al. (2011) estimated that only 8-11% of the isoprene-peroxy radicals isomerizes to reform HO\(_2\) radicals. Hofzumahaus et al. (2009) also found under predictions of OH in the Pearl River Delta and they also note that the sequence of reactions beginning with OH attack on VOCs introduces enormous complexity which is far from being fully understood.

The oxidation of aromatic hydrocarbons constitutes an important component of the chemistry of \(O_3\) formation in urban atmospheres as discussed in Annex AX2.2.8 of the 2006 \(O_3\) AQCD (U.S. EPA, 2006b). Virtually all of the important aromatic hydrocarbon precursors emitted in urban atmospheres are lost through reaction with the hydroxyl radical. Loss rates for these compounds vary from slow (e.g., benzene) to moderate (e.g., toluene), to very rapid (e.g., xylene and trimethylbenzene isomers). However, the mechanism for the oxidation of aromatic hydrocarbons following reaction with OH is poorly understood, as is evident from the poor mass balance of the reaction products. The mechanism for the oxidation of toluene has been studied most thoroughly, and there is general agreement on the initial steps in the mechanism. However, at present there is no promising approach for resolving the remaining issues concerning the later steps. The oxidation of aromatic hydrocarbons also leads to particle formation that could remove gas-phase constituents that participate in \(O_3\) formation.

Adequate analytical techniques needed to identify and quantify key intermediate species are not available for many compounds. In addition, methods to synthesize many of the suspected intermediate compounds are not available so that laboratory studies of their reaction kinetics cannot be performed. Similar considerations apply to the oxidation of biogenic hydrocarbons besides isoprene. These considerations are important because oxidants (other than \(O_3\)) that are formed from the chemistry described above could exert effects on human health and perhaps also on vegetation (Doyle et al., 2007; Doyle et al., 2004; Sexton et al., 2004). Gas phase oxidants include PAN, H\(_2\)O\(_2\), CH\(_3\)OOH, and other organic hydroperoxides.

Ozone is lost through a number of gas phase reactions and deposition to surfaces. The reaction of \(O_3\) with NO to produce NO\(_2\), for example in urban centers near roads, mainly results in the recycling of \(O_3\) downwind via the recombination of
O$_3$(P) with O$_2$ to re-form O$_3$. By itself, this reaction does not lead to a net loss of O$_3$ unless the NO$_2$ is converted to stable end products such as HNO$_3$. Ozone reacts with unsaturated hydrocarbons and with OH and HO$_2$ radicals.

Perhaps the most recent field study aimed at obtaining a better understanding of atmospheric chemical processes was the Second Texas Air Quality Field Study (TexAQS-II) conducted in Houston in August and September 2006 (Olaguer et al., 2009). The TexAQS-II Radical and Aerosol Measurement Project (TRAMP) found evidence for the importance of short-lived radical sources such as HCHO and HONO in increasing O$_3$ productivity. During TRAMP, daytime HCHO pulses as large as 32 ppb were observed and attributed to industrial activities upwind in the Houston Ship Channel (HSC) and HCHO peaks as large as 52 ppb were detected by in situ surface monitors in the HSC. Primary HCHO produced in flares from local refineries and petrochemical facilities could increase peak O$_3$ by ~30 ppb (Webster et al., 2007). Other findings from TexAQS-II included substantial concentrations of HONO during the day, with peak concentrations approaching 1 ppb at local noon. These concentrations are well in excess of current air quality model predictions using gas phase mechanisms alone (Sarwar et al., 2008) and multiphase processes are needed to account for these observations. Olaguer et al. (2009) also noted that using measured HONO brings modeled O$_3$ concentrations into much better agreement with observations and could result in the production of an additional 10 ppb O$_3$. Large nocturnal vertical gradients indicating a surface or near-surface source of HONO, and large concentrations of night-time radicals (~30 ppt HO$_2$) were also found during TRAMP.

### 3.2.3 Multiphase Processes

In addition to the gas phase, chemical reactions also occur on the surfaces of or within cloud droplets and airborne particles. Their collective surface area is huge, implying that collisions with gas phase species occur on very short time scales. In addition to hydrometeors (e.g., cloud and fog droplets and snow and ice crystals) there are also potential reactions involving atmospheric particles of varying composition (e.g., wet [deliquesced] inorganic particles, mineral dust, carbon chain agglomerates and organic carbon particles). Multiphase reactions are involved in the formation of a number of species such as particulate nitrate, and gas phase HONO that can act to both increase and reduce the rate of O$_3$ formation in the polluted troposphere. Data collected in Houston as part of TexAQS-II summarized by Olaguer et al. (2009) indicate that concentrations of HONO are much higher than can be explained by gas phase chemistry and by tailpipe emissions. Photolysis of HONO formed in multiphase reactions in addition to the other sources can help to reduce the model underestimate of simulated O$_3$ in Houston.

Multiphase processes have been associated with the release of gaseous halogen compounds from marine aerosol, mainly in marine and coastal environments. However, Thornton et al. (2010) found production rates of gaseous nitryl chloride.
near Boulder, Colorado, from reaction of N₂O₅ with particulate Cl⁻, similar to those found in coastal and marine environments. CINO₂ readily photolyzes, to yield Cl. They also found that substantial quantities of N₂O₅ are recycled through CINO₂ back into NOₓ instead of forming HNO₃ (a stable reservoir for reactive nitrogen compounds). The oxidation of hydrocarbons by Cl radicals released from the marine aerosol could lead to the rapid formation of peroxo radicals and higher rates of O₃ production. It should be noted that in addition to production from marine aerosol, reactive halogen species are also produced by the oxidation of halogenated organic compounds (e.g., CH₃Cl, CH₃Br, and CH₃I). The atmospheric chemistry of halogens is complex because Cl⁻, Br⁻, and I-containing species can react among themselves and with hydrocarbons and other species and could also be important for O₃ destruction, as has been noted for the lower stratosphere (McElroy et al., 1986; Yung et al., 1980). For example, the reactions of Br⁻ and Cl-containing radicals deplete O₃ in selected environments such as the Arctic during the spring (Barrie et al., 1988), the tropical marine boundary layer (Dickerson et al., 1999), and inland salt flats and salt lakes (Stutz et al., 2002). Mahajan et al. (2010) found that I and Br species acting together resulted in O₃ depletion that was much larger than would have been expected if they acted individually and did not interact with each other; see Annex AX2.2.10.3 of the 2006 O₃ AQCD (U.S. EPA, 2006b).

Multiphase processes have also been associated with the uptake of reactive gas phase species affecting global budgets of O₃ and nitrogen oxides among others. The uptake of N₂O₅ on aerosols or cloud droplets leads to the loss of O₃ and NOₓ and the production of aqueous phase nitric acid, aerosol nitrate, and gaseous halogen nitrites. In addition to loss of HO₂, the uptake of HO₂ radicals on aerosol surfaces potentially reduces O₃ concentrations and increases formation of sulfate (if H₂O₂ is formed after uptake). Macintyre and Evans (2011) developed a parameterization for uptake of HO₂ based on laboratory studies, which were about a factor of seven lower than previously estimated. However they note that some of the earlier studies reporting higher values might have been influenced by transition metal ions (e.g., Cu(II), Fe(II)), which are highly spatially variable and could be important catalysts in areas with high concentrations of these ions. Although the global change in O₃ was small (≈-0.3%) much larger regional changes were found (e.g., up to -27% at the surface over China).

Uptake coefficients for these species vary widely among laboratory studies. Macintyre and Evans (2010) showed that the sensitivity of key tropospheric species such as O₃ varies from very small to significant over the range of uptake coefficients for N₂O₅ obtained in laboratory studies. For example, global O₃ loss ranges from 0 to over 10%, with large regional variability over the range of N₂O₅ uptake coefficients reported. In this regard, it should be stressed that knowledge of multiphase processes is still evolving and there are still many questions that remain to be answered. However, it is becoming clear that multiphase processes are important for O₃ chemistry.

Reactions of O₃ with monoterpenes have been shown to produce oxidants in the aerosol phase, principally as components of ultrafine particles. Docherty et al. (2005)
found evidence for the substantial production of organic hydroperoxides in secondary organic aerosol (SOA) resulting from the reaction of monoterpenes with \( \text{O}_3 \). Analysis of the SOA formed in their environmental chamber indicated that the SOA consisted mainly of organic hydroperoxides. In particular, they obtained yields of 47% and 85% of organic peroxides from the oxidation of \( \alpha \)- and \( \beta \)-pinene. The hydroperoxides then react with aldehydes in particles to form peroxynediacetals, which can either rearrange to form other compounds such as alcohols and acids or revert back to the hydroperoxides. The aldehydes are also produced in large measure during the ozonolysis of the monoterpenes. Monoterpenes also react with OH radicals resulting in the production of more lower-molecular-weight products than in the reaction with monoterpenes and \( \text{O}_3 \). Bonn et al. (2004) estimated that hydroperoxides lead to 63% of global SOA formation from the oxidation of terpenes. The oxidation of anthropogenic aromatic hydrocarbons by OH radicals could also produce organic hydroperoxides in SOA (Johnson et al., 2004). Recent measurements show that the abundance of oxidized SOA exceeds that of more reduced hydrocarbon like organic aerosol in Pittsburgh (Zhang et al., 2005) and in about 30 other cities across the Northern Hemisphere (Zhang et al., 2007b). Based on aircraft and ship-based sampling of organic aerosols over coastal waters downwind of northeastern U.S. cities, de Gouw et al. (2008) reported that 40-70% of measured organic mass was water soluble and estimated that approximately 37% of SOA is attributable to aromatic precursors, using PM yields estimated for NO\textsubscript{X}-limited conditions. Uncertainties still exist as to the pathways by which the oxidation of isoprene leads to the formation of SOA. Nozière et al. (2011) found that a substantial fraction of 2-methyltetrols are primary in origin, although these species have been widely viewed solely as products of the atmospheric oxidation of isoprene. This finding points to lingering uncertainty in reaction pathways in the oxidation of isoprene and in estimates of the yield of SOA from isoprene oxidation.

Reactions of \( \text{O}_3 \) on the surfaces of particles, in particular those with humic acid like composition, are instrumental in the processing of SOA and the release of low molecular weight products such as HCHO (D'Anna et al., 2009). However, direct reactions of \( \text{O}_3 \) and atmospheric particles appear to be too slow to represent a major \( \text{O}_3 \) sink in the troposphere (D'Anna et al., 2009).

### 3.2.3.1 Indoor Air

Except when activities such as photocopying or welding are occurring, the major source of \( \text{O}_3 \) to indoor air is through infiltration of outdoor air. Reactions involving ambient \( \text{O}_3 \) with NO either from exhaled breath or from gas-fired appliances, surfaces of furnishings and terpenoid compounds from cleaning products, air fresheners and wood products also occur in indoor air as was discussed in the 2006 \( \text{O}_3 \) AQCD (U.S. EPA, 2006b). The previous \( \text{O}_3 \) review also noted that the ozonolysis of terpenoid compounds could be a substantial source of secondary organic aerosol in the ultrafine size fraction. Chen et al. (2011) examined the formation of secondary
organic aerosol from the reaction of \( \text{O}_3 \) that has infiltrated indoors with terpenoid components of commonly used air fresheners. They focused on the formation and decay of particle bound reactive oxygen species (ROS) and on their chemical properties. They found that the ROS content of samples can be decomposed into fractions that differ in terms of reactivity and volatility; however, the overall ROS content of samples decays and over 90% is lost within a day at room temperature. This result also suggests loss of ROS during sampling periods longer than a couple of hours.

### 3.2.4 Temperature and Chemical Precursor Relationships

As might be expected based on the temperature dependence of many reactions involved in the production and destruction of \( \text{O}_3 \) and the temperature dependence of emissions processes such as evaporation of hydrocarbon precursors and the emissions of biogenically important precursors such as isoprene, ambient concentrations of \( \text{O}_3 \) also show temperature dependence. Bloomer et al. (2009) determined the sensitivity of \( \text{O}_3 \) to temperature at rural sites in the eastern United States. They found that \( \text{O}_3 \) increased on average at rural Clean Air Status and Trends Network (CASTNET) sites by \( \sim 3.2 \text{ ppbv/}^\circ \text{C} \) before 2002; and after 2002 by \( \sim 2.2 \text{ ppbv/}^\circ \text{C} \). This change in sensitivity was largely the result of reductions in \( \text{NO}_x \) emissions from power plants. These results are in accord with model predictions by Wu et al. (2008b) showing that the sensitivity of \( \text{O}_3 \) to temperature decreases with decreases in precursor emissions. Rasmussen et al. (2012) recently extended the work of Bloomer et al. (2009) to quantify seasonal changes in the sensitivity of \( \text{O}_3 \) to temperature as well as regional variability (3-6 ppb/\(^\circ\)C over the Northeast and mid-Atlantic; 3-4 ppb/\(^\circ\)C over the Great Lakes region) and to evaluate the capability of a chemistry-climate model to capture \( \text{O}_3 \) sensitivity to temperature. However, the associations of \( \text{O}_3 \) with temperature are not as clear in the West as they might be in the East. For example, sites downwind of Phoenix, AZ showed basically no sensitivity of \( \text{O}_3 \) to temperature (\( R^2 = 0.02 \)) (U.S. EPA, 2006b). However, Wise and Comrie (2005) did find that meteorological parameters (mixing height and temperature) typically accounts for 40 to 70% of the variability in \( \text{O}_3 \) in the five southwestern cities (including Phoenix) they examined. It is likely that differences in the nature of sites chosen (urban versus rural) accounted for this difference and are at least partially responsible for the difference in results. Jaffe et al. (2008) regressed \( \text{O}_3 \) on temperature at Yellowstone and Rocky Mountain NP and found weak associations (\( R^2 = 0.09 \) and 0.16). They found that associations with area burned by wildfires are much stronger. Other sources as discussed in Section 3.4 might also be more important in the West than in the East.

The warmer months of the year are generally regarded as being the most conducive to higher \( \text{O}_3 \) concentrations. However, Schnell et al. (2009) reported observations of high \( \text{O}_3 \) concentrations (maximum 1-h avg of 140 ppb; maximum 8-h avg of 120 ppb) in the Jonah-Pinedale gas fields in Wyoming during winter at temperatures of \(-17^\circ \text{C}\). Potential factors contributing to these anomalously high concentrations
include a highly reflective snow surface, emissions of NO\textsubscript{X}, hydrocarbons and short-lived radical reservoirs (e.g., HONO and HCHO) and a very shallow, stable boundary layer trapping these emissions close to the surface (Schnell et al., 2009). Multiphase processes might also be involved in the production of these short-lived reservoirs. At a temperature of -17\textdegree C, the production of hydroxyl radicals (by the photolysis of O\textsubscript{3} yielding O\textsuperscript{1}D followed by the reaction, O(\textsuperscript{1}D) + H\textsubscript{2}O, needed to initiate hydrocarbon oxidation) is severely limited, suggesting that another source of radicals is needed. Radicals can be produced by the photolysis of molecules such as HONO and HCHO which photolyze in optically thin regions of the solar spectrum. A similar issue, in part due to the under-prediction of radicals, has arisen in the Houston airshed where chemistry-transport models (CTMs; discussed further in Section 3.3) under-predict O\textsubscript{3} (Olaguer et al., 2009). Carter and Seinfeld (2012) modeled several of the events using the SAPRC-07 chemical mechanism and found that the release of HONO from the snow surface aids in the formation of O\textsubscript{3}. The chemical mechanism they used—including the temperature dependence of rate coefficients—was developed for application at higher temperatures. They also note that temperature changes will also affect the distribution of products and radicals formed when individual VOCs react, but the current version of the mechanism represents these by lumped overall processes in which the product and radical distributions are treated as if they are temperature independent. It is not clear how this treatment of radical production might affect their results.

Rather than varying directly with emissions of its precursors, O\textsubscript{3} changes in a nonlinear fashion with the concentrations of its precursors. At the low NO\textsubscript{X} concentrations found in remote continental areas to rural and suburban areas downwind of urban centers (low-NO\textsubscript{X} regime), the net production of O\textsubscript{3} typically increases with increasing NO\textsubscript{X}. In the low-NO\textsubscript{X} regime, the overall effect of the oxidation of VOCs is to generate (or at least not consume) free radicals, and O\textsubscript{3} production varies directly with NO\textsubscript{X}. In the high-NO\textsubscript{X} regime, NO\textsubscript{2} reacts with OH radicals to form HNO\textsubscript{3} (e.g., Hameed et al., 1979). These OH radicals would otherwise oxidize VOCs to produce peroxy radicals, which in turn would oxidize NO to NO\textsubscript{2}. In this regime, O\textsubscript{3} production is limited by the availability of radicals (Tonnesen and Jeffries, 1994) and O\textsubscript{3} shows only a weak dependence on NO\textsubscript{X} concentrations. The production of radicals is in turn limited by the availability of solar UV radiation capable of photolyzing O\textsubscript{3} (in the Hartley bands) and aldehydes and/or by the abundance of VOCs whose oxidation produce more radicals than they consume. At the even higher NO\textsubscript{X} concentrations found in downtown metropolitan areas, especially near busy streets and roads, and in power plant plumes, there is scavenging (sometimes referred to as titration) of O\textsubscript{3} by reaction with NO to form NO\textsubscript{2} leading to depletion of O\textsubscript{3}. However, as urban plumes are transported and diluted, this NO\textsubscript{2} can lead to photochemical production of O\textsubscript{3} downwind of the source areas.

The production of radicals can also be limited by the availability of solar UV radiation capable of photolyzing O\textsubscript{3} (in the Hartley bands) and aldehydes. When solar radiation is blocked by clouds or reduced during winter, VOC and NO\textsubscript{X} may
both be available, but O$_3$ production is limited by the availability of solar radiation, and this has been defined as the “light-limited” regime (Hess et al., 1992).

There are a number of ways to refer to the chemistry of O$_3$ production in these different chemical regimes. Sometimes the terms VOC-limited and NO$_X$-limited are used. However, there are difficulties with this usage because (1) VOC measurements are not as abundant as they are for nitrogen oxides; (2) rate coefficients for reaction of individual VOCs with radicals (e.g., OH, Cl) vary over an extremely wide range; and (3) consideration is not given to CH$_4$ or CO and other reactions that can produce radicals without involving VOCs (e.g., photolysis of HONO). Many of these difficulties are overcome by the terms NO$_X$-limited and radical-limited (Tonnesen and Dennis, 2000a, b). This usage recognizes that OH radicals are needed to react with VOCs to form O$_3$ and that either low NO or high NO$_2$ can limit the production of OH radicals. This usage also implicitly considers the importance of processes such as the availability of solar radiation and photolysis in generating radicals. The terms NO$_X$-limited and NO$_X$-saturated (Jaegle et al., 2001) have also been used to describe these two regimes. However, the terminology used in original articles will also be used here. In addition, in the remote marine troposphere, NO$_X$ concentrations can be ~20 ppt or less. Under these very low NO$_X$ conditions, which are not likely to be found in the continental U.S., but can characterize inflowing air, HO$_2$ and CH$_3$O$_2$ radicals react with each other and HO$_2$ radicals undergo self-reaction (to form H$_2$O$_2$), OH radicals efficiently convert NO$_2$ to HNO$_3$, and OH and HO$_2$ react with O$_3$, leading to net destruction of O$_3$ and inefficient OH radical regeneration. In addition, halogen-containing radicals also react with O$_3$ acting to keep its concentrations very low. This is in contrast to the situation in areas of the U.S. outside of urban cores, where HO$_2$ and CH$_3$O$_2$ radicals react with NO to convert NO to NO$_2$, regenerate the OH radical, and, through the photolysis of NO$_2$, produce O$_3$ as noted in Annex AX2.2.5 of the 2006 O$_3$ AQCD (U.S. EPA, 2006b).

There are no definitive rules governing the concentrations of NO$_X$ at which the transition from NO$_X$-limited to NO$_X$-saturated conditions occurs. The transition between these two regimes is highly spatially and temporally dependent and depends also on the nature and abundance of the hydrocarbons that are present. In a NO$_X$-limited (or NO$_X$-sensitive) regime, O$_3$ formation is not completely insensitive to radical production or the flux of solar UV photons, just that O$_3$ formation is more sensitive to NO$_X$. For example, global tropospheric O$_3$ is sensitive to the concentration of CH$_4$ even though the troposphere is predominantly NO$_X$-limited. Likewise, in a NO$_X$-saturated regime there can still be some peroxyl-peroxy radical interactions depending on NO$_X$ concentrations.

These considerations introduce a high degree of uncertainty into attempts to relate changes in O$_3$ concentrations to emissions of precursors. The chemistry of OH radicals, which are responsible for initiating the oxidation of hydrocarbons, shows behavior similar to that for O$_3$ with respect to NO$_X$ concentrations (Poppe et al., 1993; Zimmermann and Poppe, 1993; Hameed et al., 1979).

Trainer et al. (1993) and Olszyna et al. (1994) have shown that O$_3$ and NO$_X$ are highly correlated in rural areas in the eastern United States. Trainer et al. (1993) also
showed that O₃ concentrations correlate even better with NOₓ than with NOₓ, as may be expected because NOₓ represents the amount of NOₓ that has been oxidized, forming O₃ in the process. NOₓ is equal to the difference between measured total reactive nitrogen (NOₓ) and NOₓ and represents the summed products of the oxidation of NOₓ. NOₓ is composed mainly of HNO₃, PAN and other organic nitrates, particulate nitrate, and HNO₄. Trainer et al. (1993) also suggested that the slope of the regression line between O₃ and NOₓ can be used to estimate the rate of O₃ production per NOₓ oxidized (also known as the O₃ production efficiency [OPE]). Ryerson et al. (2001); Ryerson et al. (1998) used measured correlations between O₃ and NOₓ to identify different rates of O₃ production in plumes from large point sources. A number of studies in the planetary boundary layer over the continental U.S. have found that the OPE ranges typically from 1 to nearly 10. However, it may be higher in the upper troposphere and in certain areas, such as the Houston-Galveston area in Texas. Observations indicate that the OPE depends mainly on the abundance of NOₓ and also on availability of solar UV radiation, VOCs and O₃ itself.

Various techniques have been proposed to use ambient NOₓ and VOC measurements, i.e., as indicators to derive information about the dependence of O₃ production on their concentrations. For example, it has been suggested that O₃ formation in individual urban areas could be understood in terms of measurements of ambient NOₓ and VOC concentrations during the early morning (NRC, 1991). In this approach, the ratio of summed (unweighted) VOC to NOₓ is used to determine whether conditions were NOₓ-limited or VOC-limited. This approach is inadequate because it does not consider many factors that are important for O₃ production such as the impact of biogenic VOCs (which are typically not present in urban centers during early morning); important differences in the ability of individual VOCs to generate radicals (rather than just total VOC) and other differences in O₃ forming potential for individual VOCs (Carter, 1995); and changes in the VOC to NOₓ ratio due to photochemical reactions and deposition as air moves downwind from urban areas (Milford et al., 1994).

Photochemical production of O₃ generally occurs simultaneously with the production of various other species such as HNO₃, organic nitrates, and other oxidants such as hydrogen peroxide. The relative rate of production of O₃ and other species varies depending on photochemical conditions, and can be used to provide information about O₃-precursor sensitivity. Sillman (1995) and Sillman and He (2002) identified several secondary reaction products that show different correlation patterns for NOₓ-limited and NOₓ-saturated conditions. The most important correlations are for O₃ versus NOₓ, O₃ versus NOₓ, O₃ versus HNO₃, and H₂O₂ versus HNO₃. The correlations between O₃ and NOₓ, and O₃ and NOₓ are especially important because measurements of NOₓ and NOₓ are more widely available than for VOCs. Measured O₃ versus NOₓ (Figure 3-3) shows distinctly different patterns in different locations. In rural areas and in urban areas such as Nashville, TN, O₃ is highly correlated with NOₓ. By contrast, in Los Angeles, CA, O₃ is not as highly correlated with NOₓ, and the rate of increase of O₃ with NOₓ is lower and the O₃ concentrations for a given NOₓ value are generally lower. The different O₃ versus
NO$_2$ relations in Nashville, TN and Los Angeles, CA reflects the difference between NO$_X$-limited conditions in Nashville versus an approach to NO$_X$-saturated conditions in Los Angeles.

Note: (NOY–NOX) during the afternoon at rural sites in the eastern U.S. (the grey circles) and in urban areas and urban plumes associated with Nashville, TN (the gray dashes); Paris, France (the black diamonds); and Los Angeles, CA (the Xs).

Source: Adapted with permission of American Geophysical Union, (Sillman and He, 2002; Sillman et al., 1998; Trainer et al., 1993).

**Figure 3-3 Measured concentrations of O$_3$ and NO$_2$.**

The difference between NO$_X$-limited and NO$_X$-saturated regimes is also reflected in measurements of H$_2$O$_2$. H$_2$O$_2$ production is highly sensitive to the abundance of radicals and is thus favored in the NO$_X$-limited regime. Measurements in the rural eastern U.S. (Jacob et al., 1995), Nashville, TN (Sillman et al., 1998), and Los Angeles, CA (Sakugawa and Kaplan, 1989), show large differences in H$_2$O$_2$ concentrations between likely NO$_X$-limited and NO$_X$-saturated locations. Tonnesen and Dennis (2000a) reviewed the use of many indicator ratios and Tonnesen and Dennis (2000b) proposed HCHO/NO$_2$ as an indicator to distinguish between NO$_X$-limited and radical-limited regimes.

The applications of indicator species mentioned above are mainly limited to individual urban areas because these are the areas where it is often not clear which regime is prevalent. Satellites provide a platform for greatly extending the range of applicability of the indicator technique and also have the resolution necessary to examine urban to rural differences. Martin et al. (2004) and Duncan et al. (2010) used satellite data from Ozone Monitoring Instrument (OMI) for HCHO to NO$_2$ column ratios to diagnose NO$_X$-limited and radical-limited (NO$_X$-saturated) regimes. HCHO can be used as an indicator of VOCs as it is a common, short-lived, oxidation
product of many VOCs that is a source of HOX \( (\text{Sillman, 1995}) \). In adopting the satellite approach, CTMs are used to estimate the fractional abundance of the indicator species in the planetary boundary layer. Duncan et al. \( (2010) \) found that O\(_3\) formation over most of the U.S. became more sensitive to NO\(_X\) over most of the U.S. from 2005 to 2007 largely because of decreases in NO\(_X\) emissions. They also found that surface temperature is correlated with the ratio of HCHO to NO\(_2\) especially in cities in the Southeast where emissions of isoprene (a major source of HCHO) are high due to high temperatures in summer.

### 3.3 Atmospheric Modeling

CTMs have been widely used to compute the interactions among atmospheric pollutants and their transformation products, and the transport and deposition of pollutants. They have also been widely used to improve basic understanding of atmospheric chemical processes and to develop control strategies. The spatial scales over which pollutant fields are calculated range from intra-urban to regional to global. Generally, these models are applied to problems on different spatial scales but efforts are underway to link across spatial scales for dealing with global scale environmental issues that affect population health within cities. Many features are common to all of these models and hence they share many of the same problems. On the other hand, there are appreciable differences in approaches to parameterizing physical and chemical processes that must be addressed in applying these models across spatial scales.

CTMs solve a set of coupled, non-linear partial differential equations, or continuity equations, for relevant chemical species. Jacobson \( (2005) \) described the governing partial differential equations, and the methods that are used to solve them. Because of limitations imposed by the complexity and spatial-temporal scales of relevant physical and chemical processes, the CTMs must include parameterizations of these processes, which include atmospheric transport; the transfer of solar radiation through the atmosphere; chemical reactions; and removal to the surface by turbulent motions and precipitation. Development of parameterizations for use in CTMs requires data for three dimensional wind fields, temperatures, humidity, cloudiness, and solar radiation; emissions data for primary (i.e., directly emitted from sources) species such as NO\(_X\), SO\(_2\), NH\(_3\), VOCs, and primary PM; and chemical reactions.

The domains of CTMs extend from a few hundred kilometers on a side to the entire globe. Most major regional (i.e., sub-continental) scale air-related modeling efforts at EPA rely on the Community Multi-scale Air Quality (CMAQ) modeling system \( (\text{Byun and Schere, 2006}; \text{Byun and Ching, 1999}) \). CMAQ’s horizontal domain typically extends over North America with efforts underway to extend it over the entire Northern Hemisphere. Note that CTMs can be ‘nested’ within each other as shown in Figure 3-4 which shows domains for CMAQ (Version 4.6.1); additional details on the model configuration and application are found elsewhere \( (\text{U.S. EPA, 2009e}) \). The figure shows the outer domain (36 km horizontal grid spacing) and two
12 km spatial resolution (east and west) sub-domains. The upper boundary for CMAQ is typically set at about 100 hPa, or at about 16 km altitude on average, although in some recent applications the upper boundary has been set at 50 hPa. These domains and grid spacings are quite common and can also be found in a number of other models.

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**Figure 3-4** Sample Community Multi-scale Air Quality (CMAQ) modeling domains.

The main components of a CTM such as EPA’s CMAQ are summarized in Figure 3-5. The capabilities of a number of CTMs designed to study local- and regional-scale air pollution problems were summarized by Russell and Dennis (2000) and in the 2006 O₃ AQCD (U.S. EPA, 2006b). Historically, CMAQ has been driven most often by the MM5 mesoscale meteorological model (Seaman, 2000), though it could be driven by other meteorological models including the Weather Research and Forecasting (WRF) model and the Regional Atmospheric Modeling System (RAMS) (ATMET, 2011).
Figure 3-5  **Main components of a comprehensive atmospheric chemistry modeling system, such as the U.S. EPA’s Community Multi-scale Air Quality (CMAQ) modeling system.**

Simulations of pollution episodes over regional domains have been performed with a horizontal resolution down to 1 km; see the application and general survey results reported in Ching et al. (2006). However, simulations at such high resolution require better parameterizations of meteorological processes such as boundary layer fluxes, deep convection, and clouds (Seaman, 2000). Finer spatial resolution is necessary to resolve features such as urban heat island circulation; sea, bay, and land breezes; mountain and valley breezes; and the nocturnal low-level jet; all of which can affect pollutant concentrations. Other major air quality systems used for regional scale applications include the Comprehensive Air Quality Model with extensions (CAMx) (ENVIRON, 2005) and the Weather Research and Forecast model with Chemistry (WRF/Chem) (NOAA, 2010).

CMAQ and other grid-based or Eulerian air quality models subdivide the modeling domain into a three-dimensional array of grid cells. The most common approach to setting up the horizontal domain is to nest a finer grid within a larger domain of coarser resolution. The use of finer horizontal resolution in CTMs will necessitate finer-scale inventories of land use and better knowledge of the exact paths of roads, locations of factories, and, in general, better methods for locating sources and estimating their emissions. The vertical resolution of CTMs is variable and usually configured to have more layers in the PBL and fewer in the free troposphere.
The meteorological fields are produced either by other numerical prediction models such as those used for weather forecasting (e.g., MM5, WRF), and/or by assimilation of satellite data. The flow of information shown in Figure 3-5 has most often been unidirectional in the sense that information flows into the CTM (large box) from outside; feedbacks on the meteorological fields and on boundary conditions (i.e., out of the box) have not been included. However, CTMs now have the capability to consider these feedbacks as well; see, for example, Binkowski et al. (2007) and WRF/Chem (NOAA, 2010).

Because of the large number of chemical species and reactions that are involved in the oxidation of realistic mixtures of anthropogenic and biogenic hydrocarbons, condensed mechanisms must be used in atmospheric models. These mechanisms can be tested by comparison with smog chamber data. However, the existing chemical mechanisms often neglect many important processes such as the formation and subsequent reactions of long-lived carbonyl compounds, the incorporation of the most recent information about intermediate compounds, and heterogeneous reactions involving cloud droplets and aerosol particles. To the extent that information is available, models like CMAQ and CAMx do include state-of-the-science parameterization for some of these processes such as heterogeneous N₂O₅ chemistry.

The initial conditions, or starting concentration fields of all species computed by a model, and the boundary conditions, or concentrations of species along the horizontal and upper boundaries of the model domain throughout the simulation, must be specified at the beginning of the simulation. Both initial and boundary conditions can be estimated from models or data or, more generally, model plus data hybrids. Because data for vertical profiles of most species of interest are very sparse, results of model simulations over larger, usually global, domains are often used.

Chemical kinetics mechanisms representing the important reactions occurring in the atmosphere are used in CTMs to estimate the rates of chemical formation and destruction of each pollutant simulated as a function of time. The Master Chemical Mechanism (MCM) (Univ of Leeds, 2010) is a comprehensive reaction database providing as near an explicit treatment of chemical reactions in the troposphere as is possible. The MCM currently includes over 12,600 reactions and 4,500 species. However, mechanisms that are this comprehensive are still computationally too demanding to be incorporated into CTMs for regulatory use. Simpler treatments of tropospheric chemistry have been assembled by combining chemical species into mechanisms that group together compounds with similar chemistry. It should be noted that because of different approaches to the lumping of organic compounds into surrogate groups for computational efficiency, chemical mechanisms can produce different results under similar conditions. Jimenez et al. (2003) briefly described the features of the seven main chemical mechanisms in use and compared concentrations of several key species predicted by these mechanisms in a box-model simulation over 24 hours. Several of these mechanisms have been incorporated into CMAQ including extensions of the Carbon Bond (CB) mechanism (Luecken et al., 2008), SAPRC (Luecken et al., 2008), and the Regional Atmospheric Chemistry Mechanism, version 2 (RACM2) (Fuentes et al., 2007). The CB mechanism is currently undergoing
extension (CB06) to include, among other things, longer lived species to better simulate chemistry in the remote and upper troposphere. These mechanisms were developed primarily for homogeneous gas phase reactions and treat multiphase chemical reactions in a very cursory manner, if at all. As a consequence of neglecting multiphase chemical reactions, models such as CMAQ could have difficulties capturing the regional nature of O\(_3\) episodes, in part because of uncertainty in the chemical pathways converting NO\(_X\) to HNO\(_3\) and recycling of NO\(_X\) (Godowitch et al., 2008; Hains et al., 2008). Much of this uncertainty also involves multiphase processes as described in Section 3.2.3.

CMAQ and other CTMs incorporate processes and interactions of aerosol-phase chemistry (Zhang and Wexler, 2008; Gaydos et al., 2007; Binkowski and Roselle, 2003). There have also been several attempts to study the feedbacks of chemistry on atmospheric dynamics using meteorological models like MM5 and WRF (Liu et al., 2001; Park et al., 2001; Grell et al., 2000; Lu et al., 1997). This coupling is necessary to accurately simulate feedbacks from PM (Park et al., 2001; Lu et al., 1997) over areas such as Los Angeles or the Mid-Atlantic region. Photolysis rates in CMAQ can now be calculated interactively with model produced O\(_3\), NO\(_2\), and aerosol fields (Binkowski et al., 2007).

Spatial and temporal characterizations of anthropogenic and biogenic precursor emissions can be specified as inputs to a CTM or these emissions can be calculated in-line in CMAQ. Emissions inventories have been compiled on grids of varying resolution for many hydrocarbons, aldehydes, ketones, CO, NH\(_3\), and NO\(_X\). Preprocessing of emissions data for CMAQ is done by the Spare-Matrix Operator Kernel Emissions (SMOKE) system (UNC, 2011). For many species, information on temporal variability of emissions is lacking, so long-term annual averages are used in short-term, episodic simulations. Annual emissions estimates can be modified by the model to produce emissions more characteristic of the time of day and season. Appreciable errors in emissions can occur if inappropriate time dependence is applied.

Each of the model components described above has associated uncertainties; and the relative importance of these uncertainties varies with the modeling application. Large errors in photochemical modeling arise from the meteorological, chemical and emissions inputs to the model (Russell and Dennis, 2000). While the effects of poorly specified boundary conditions propagate through the model’s domain, the effects of these errors remain undetermined. Because many meteorological processes occur on spatial scales smaller than the model’s vertical or horizontal grid spacing and thus are not calculated explicitly, parameterizations of these processes must be used. These parameterizations introduce additional uncertainty.

The performance of CTMs must be evaluated by comparison with field data as part of a cycle of model evaluations and subsequent improvements (NRC, 2007). However, they are too computationally demanding to have the full range of their sensitivities examined using Monte Carlo techniques (NRC, 2007). Models of this complexity are evaluated by comparison with field observations for O\(_3\) and other species. Evaluations of the performance of CMAQ are given in Arnold et al. (2003),
Discrepancies between model predictions and observations can be used to point out
gaps in current understanding of atmospheric chemistry and to spur improvements in
parameterizations of atmospheric chemical and physical processes. Model evaluation
does not merely involve a straightforward comparison between model predictions
and the concentration field of the pollutant of interest. Such comparisons may not be
meaningful because it is difficult to determine if agreement between model
predictions and observations truly represents an accurate treatment of physical and
chemical processes in the CTM or the effects of compensating errors in complex
model routines (in other words, it is important to know if the right answer is obtained
for the right reasons). Each of the model components (emissions inventories,
chemical mechanism, and meteorological driver) should be evaluated individually as
has been done to large extent in some major field studies such as TexAQS I and II
and CalNex. Comparisons of correlations between measured and modeled VOCs and
NOX are useful for evaluating results from CTMs and can provide information about
the chemical state of the atmosphere. A CTM that accurately computes both VOC
and NOX along with the spatial and temporal relations among the critical secondary
species associated with O3 has a higher probability of representing O3-precursor
relations correctly than one that does not.

The above evaluation techniques are sometimes referred to as “static” in the sense
that individual model variables are compared to observations. It is also crucial to
understand the dynamic response to changes in inputs and to compare the model
responses to those that are observed. These tests might involve changes in some
natural forcing or in emissions from an anthropogenic source. As an example,
techniques such as the direct decoupled method (DDM) could be used in CTMs to
determine their first order sensitivity to emissions changes (Zhang, 2005; Dunker et
al., 2002). However, the observational basis for comparing a model’s response is
largely unavailable for many problems of interest, in large part because
meteorological conditions are also changing while the emissions are changing. As a
result, methods such as DDM are used mainly to assess the potential effectiveness of
emissions controls (Arunachalam, 2009). Because the chemistry of O3 formation is
non-linear near strong sources, and higher order terms taking into account the
curvature of the response surface of O3 with respect to changes in sources are needed
when using DDM. These additional terms are incorporated in the higher order
decoupled method, or HDDM, (see the U.S. EPA technical memorandum on
applications of HDDM, particularly with respect to adjusting O3 concentrations in
response to emissions reductions [or “roll back”] at

### 3.3.1 Global Scale CTMs

With recognition of the global nature of many air pollution problems, global scale
CTMs have been applied to regional scale pollution problems (NRC, 2009). Global-
scale CTMs are used to address issues associated with global change, to characterize
long-range transport of air pollutants, and to provide boundary conditions for the regional-scale models. The upper boundaries of global scale CTMs extend anywhere from the tropopause (~8 km at the poles to ~16 km in the tropics) to the mesopause at ~80 km, in order to obtain more realistic boundary conditions for problems involving stratospheric dynamics and chemistry. The global-scale CTMs consider the same processes shown in Figure 3-5 for the regional scale models. In addition, many of the same issues that have arisen for the regional models have also arisen for the global scale models (Emmerson and Evans, 2009). For example, after adjusting lightning NOx to better match observed constraints in the MOZART-4 model, simulated HNO3 was too low and PAN too high in the mid-troposphere, though observations were captured in the upper troposphere, over the U.S. during summer 2004 in the MOZART-4 model (Fang et al., 2010). In contrast, summer 2004 simulations with improved lightning NOx in GEOS-Chem indicate that PAN is too low but HNO3 is overestimated throughout the mid- and upper troposphere (Hudman et al., 2007). Predictions of HNO3 were too high and PAN too low over the U.S. during summer in the MOZART model (Fang et al., 2010). Similar findings were obtained in a box model of upper tropospheric chemistry (Henderson et al., 2011), indicating a need for improved constraints on processes controlling NOx distributions in the free troposphere.

The GEOS-Chem model is a community-owned, global scale CTM that has been widely used to study issues associated with the hemispheric transport of pollution and global change (Harvard University, 2010a). Comparisons of the capabilities of GEOS-Chem and several other models to simulate intra-hemispheric transport of pollutants are given in a number of articles (Fiore et al., 2009; Reidmiller et al., 2009). Reidmiller et al. (2009) compared the ensemble average of 18 global models to spatially and monthly averaged observations of O3 at CASTNET sites in the U.S. (see Figure 3-6). These results show that the multi-model ensemble agrees much better with observations than do most of the individual models. The GEOS-Chem model was run for two grid spacings (4º × 4.5º and 2º × 2.5º) over the U.S. with very similar results that lie close to the ensemble average. In general, the model ensemble average and the two GEOS-Chem simulations are much closer to observations in the Intermountain West Region than in the Southeast Region during summer, when most major O3 episodes occur (Note, though, that more current versions of GEOS-Chem are now in use.) However, there are also sizable over-predictions by many models in both regions during summer.
In their review, McDonald-Buller et al. (2011) noted that global scale chemical transport models exhibit biases in monthly mean of the daily maximum 8-h avg (MDA8) $O_3$ in some regions of the U.S., including the Gulf Coast, regions affected by fires, and regions with complex topography, which have implications for model estimates of background $O_3$; and they also have difficulty representing the fine structures of $O_3$ events at sub-grid scales at relatively remote monitoring sites that include contributions to $O_3$ from background sources.

Global models are not alone in overestimating $O_3$ in the Southeast. Godowitch et al. (2008), Gilliland et al. (2008) and Nolte et al. (2008) found positive $O_3$ biases in regional models over the eastern U.S., as well, which they largely attributed to uncertainties in temperature, relative humidity and planetary boundary layer height. Agreement between monthly average values is expected to be better than with daily values because of a number of factors including the increasing uncertainty of emissions at finer time resolution. Kasibhatla and Chameides (2000) found that the
accuracy of simulations improved as the averaging time of both the simulation and the observations increased.

Simulations of the effects of long-range transport at particular locations must be able to link multiple horizontal resolutions from the global to the local scale. Because of computational limitations, global simulations are not made at the same horizontal resolutions found in the regional scale models, i.e., down to 1-4 km² horizontal resolution. They are typically conducted with a horizontal grid spacing of 1°-2° of latitude and longitude (or roughly 100-200 km at mid-latitudes). Some models such as GEOS-Chem have the capability to include nested models at a resolution of 0.5° × 0.667° (Wang et al., 2009a) and efforts are underway to achieve even higher spatial resolution. Another approach is to nest regional models within GEOS-Chem. Caution must be exercised with nesting different models because of differences in chemical mechanisms and numerical schemes, and in boundary conditions between the outer and inner models. As an example of these issues, surface O₃ concentrations that are too high have been observed in models in which CMAQ was nested inside of GEOS-Chem. The high O₃ results in large measure from stratospheric O₃ intruding into the CMAQ domain [see (Lam and Fu, 2010) for one way to address this issue]. Large vertical O₃ gradients in the upper troposphere must be preserved to accurately represent downward transport of stratospheric O₃. This complicates efforts to link global and regional models with different vertical grid spacing. Efforts are also underway to extend the domain of CMAQ over the entire Northern Hemisphere. In this approach, the same numerical schemes are used for transporting species and the same chemistry is used throughout all spatial scales. Finer resolution in models of any scale can only improve scientific understanding to the extent that the governing processes are accurately described. Consequently, there is a crucial need for observations at the appropriate scales to evaluate the scientific understanding represented by the models.

3.4 Background O₃ Concentrations

Background concentrations of O₃ have been given various definitions in the literature over time. An understanding of the sources and contributions of background O₃ to O₃ concentrations in the U.S. is potentially useful in reviewing the O₃ NAAQS, especially related to days at the upper end of the distribution of O₃ concentrations. In the context of a review of the NAAQS, it is useful to define background O₃ concentrations in a way that distinguishes between concentrations that result from precursor emissions that are relatively less controllable from those that are relatively more controllable through U.S. policies. In previous NAAQS reviews, a specific definition of background concentrations was used and referred to as policy relevant background (PRB). In those previous reviews, PRB concentrations were defined by EPA as those concentrations that would occur in the U.S. in the absence of anthropogenic emissions in continental North America (CNA), defined here as the U.S., Canada, and Mexico. There is no chemical difference between background O₃ and O₃ attributable to CNA anthropogenic sources. However, to inform policy
considerations regarding the current or potential alternative standards, it is useful to understand how total O₃ concentrations can be attributed to different sources.

For this document, EPA has considered background O₃ concentrations more broadly by considering three different definitions of background. The first is natural background which includes contributions resulting from emissions from natural sources (e.g., stratospheric intrusion, wildfires, biogenic methane, and more short-lived VOC emissions) throughout the globe simulated in the absence of all anthropogenic emissions. The second is North American background (NA background) which includes contributions from natural background throughout the globe and emissions of anthropogenic pollutants contributing to global concentrations of O₃ (e.g., anthropogenic methane) from countries outside North America. The third is United States background (U.S. background) which includes contributions from natural background throughout the globe and emissions from anthropogenic pollutants contributing to global concentrations of O₃ from countries outside the United States. U.S. background differs from NA background in that it includes anthropogenic emissions from neighboring Canada and Mexico. These three definitions have been explored in recent literature and are discussed further below.

Sources included in the definitions of NA-background and U.S.-background O₃ are shown schematically in Figure 3-7. Definitions of background and approaches to derive background concentrations were reviewed in the 2006 O₃ AQCD (U.S. EPA, 2006b) and in Reid et al. (2008). Further detail about the processes involved in these sources is given in Section 3.4.1 and Section 3.4.2 and application to models calculating background concentrations is presented in Section 3.4.3.
Note: Background concentrations are O₃ concentrations that would exist in the absence of anthropogenic emissions from the U.S., Canada, and Mexico. United States (U.S.) background is similarly defined, but includes transport from Canada and Mexico in addition to intercontinental transport.

Figure 3-7 Schematic overview of contributions to North American (NA) background concentrations of O₃.

3.4.1 Contributions from Natural Sources

Natural sources contributing to background O₃ include the stratospheric-tropospheric exchange (STE) of O₃ and photochemical reactions involving natural O₃ precursor emissions of VOCs, NOₓ, and CO. Natural sources of O₃ precursors include biogenic emissions, wildfires, and lightning. Biogenic emissions from agricultural activities in CNA (or the U.S.) are not considered in the formation of NA- (or U.S.-)background O₃. Contributions from natural sources are an important component of background concentrations and are discussed in greater detail below.

3.4.1.1 Contributions from the Stratosphere

The basic atmospheric dynamics and thermodynamics of STE were outlined in the 2006 O₃ AQCD (U.S. EPA, 2006b); as noted there, stratospheric air rich in O₃ is transported into the troposphere. Ozone is produced naturally by photochemical reactions in the stratosphere as shown in Figure 3-1. Some of this O₃ is transported downward into the troposphere throughout the year, with maximum contributions at mid-latitudes during late winter and early spring mainly coming from a process known as tropopause folding. These folds occur behind most cold fronts, bringing
stratospheric air with them. The tropopause should not be interpreted as a material surface through which there is no exchange. Rather these folds should be thought of as regions in which mixing of tropospheric and stratospheric air is occurring (Shapiro, 1980). This imported stratospheric air contributes to the natural background of O$_3$ in the troposphere, especially in the free troposphere during winter and spring. Significant intrusions of stratospheric air occur in “ribbons” ~200 to 1000 km in length, 100 to 300 km wide and about 1 to 4 km thick (Wimmers et al., 2003; Hoskins, 1972). Thus, these intrusions are large scale three-dimensional events and should not be thought of as one-dimensional. STE also occurs during other seasons including summer.

Methods for estimating the contribution of stratospheric intrusions rely on the use of tracers of stratospheric origin that can be either dynamical or chemical. Thompson et al. (2007) found that roughly 20-25% of free tropospheric O$_3$ over northeastern North America during July-August 2004 was of stratospheric origin based on an analysis of ozonesonde data. This O$_3$ can be mixed into the PBL where it can either be destroyed or transported to the surface. They relied on the combined use of low relative humidity and high (isentropic) potential vorticity (PV) (>2 PV units) to identify stratospheric contributions. PV has been a widely used tracer for stratospheric air; see the 2006 O$_3$ AQCD (U.S. EPA, 2006b). However, bear in mind that this analysis pertains to a single year only with no indication of interannual variability. Lefohn et al. (2011) used these and additional criteria to assess stratospheric influence on sites in the Intermountain West and in the Northern Tier. Additional criteria include consideration of trajectories originating at altitudes above the 380 K potential temperature surface with a residence time requirement at these heights. Based on these criteria, they identified likely stratospheric influence at the surface sites on a number of days during spring of 2006 to 2008. However, they noted that their analysis of stratospheric intrusions captures only the frequency and vertical penetration of the intrusions but does not provide information about the contribution of the intrusions to the measured O$_3$ concentration. These results are all generally consistent with what was noted in the 2006 O$_3$ AQCD (U.S. EPA, 2006b).

Fischer (2004) analyzed the O$_3$ record during summer at Mount Washington and identified a stratospheric contribution to 5% of events during the summers of 1998 - 2003 when O$_3$ was >65 ppb; the air was dry and trajectories originated from altitudes where PV was elevated (PV >1 PV unit). However, this analysis did not quantify the relative contributions of anthropogenic and stratospheric O$_3$ sources more precisely, because as they note identifying stratospheric influences is complicated by transport over industrialized/urban source regions. Stratospheric O$_3$ was hypothesized to influence the summit during conditions also potentially conducive to photochemical O$_3$ production, which make any relative contribution calculations difficult without additional measurements of anthropogenic and stratospheric tracers.

Although most research has been conducted on tropopause folding as a source of stratosphere to troposphere exchange, this is not the only mechanisms by which stratospheric O$_3$ can be brought to lower altitudes. Tang et al. (2011) estimated that deep convection capable of penetrating the tropopause can increase the overall downward flux of O$_3$ by ~20%. This mechanism operates mainly during summer in
contrast with tropopause folding which is at a maximum from late winter through spring and at lower latitudes. Yang et al. (2010) estimated that roughly 20% of free tropospheric O₃ above coastal California in 2005 and 2006 was stratospheric in origin. Some of this O₃ could also contribute to O₃ at the surface.

It should be noted that there is considerable uncertainty in the magnitude and distribution of this potentially important source of tropospheric O₃. Stratospheric intrusions that reach the surface are much less frequent than intrusions which penetrate only to the middle and upper troposphere. However, O₃ transported to the upper and middle troposphere can still affect surface concentrations through various exchange mechanisms that mix air from the free troposphere with air in the PBL. Several instances of STE producing high concentrations of O₃ around Denver and Boulder, Colorado, were analyzed by Langford et al. (2009) and several likely instances of STE, including one of the cases analyzed by Langford et al. (2009) were also cited in Annex AX2-3 of the 2006 O₃ AQCD (U.S. EPA, 2006b). Clear examples of STE have also been observed in southern Quebec province by Hocking et al. (2007), in accord with previous estimates by Wernli and Bourqui (2002) and James et al. (2003). As also noted in the 2006 O₃ AQCD (U.S. EPA, 2006b), the identification of stratospheric O₃ and the calculation of its contributions to ambient air requires data for other tracers of stratospheric origin. In some cases, stratospheric O₃ intrusions can be identified based on measurements of low relative humidity, high potential vorticity and low ratios of O₃/PM. Strong stratospheric O₃ intrusion events that typically occur during winter or spring have been readily identified using these types of data (Langford et al., 2009). However, it remains challenging to accurately estimate the contributions from smaller direct or indirect (i.e., resulting from shallow intrusions into the mid and upper troposphere that are then mixed downward into the planetary boundary layer) contributions of stratospheric O₃ to ambient air.

### 3.4.1.2 Contributions from Other Natural Sources

Biomass burning consists of wildfires and the intentional burning of vegetation to clear new land for agriculture and for population resettlement; to control the growth of unwanted plants on pasture land; to manage forest resources with prescribed burning; to dispose of agricultural and domestic waste; and as fuel for cooking, heating, and water sterilization. Biomass burning also exhibits strong seasonality and interannual variability (van der Werf et al., 2006), with most biomass burned during the local dry season. This is true for both prescribed burns and wildfires. Globally, most wildfires may be ignited directly as the result of human activities, leaving only 10-30% initiated by lightning (Andreae, 1991). However, because fire management practices suppress natural wildfires, the buildup of fire fuels increases the susceptibility of forests to more severe but less frequent fires. Thus there is considerable uncertainty in attributing the fraction of wildfire emissions to human activities because the emissions from naturally occurring fires that would have been present in the absence of fire suppression practices are not known. Contributions to
NOX, CO, and VOCs from wildfires and prescribed fires are considered as precursors to background O3 formation in this assessment.

Estimating contributions from wildfires is subject to considerable uncertainty. McDonald-Buller et al. (2011) note that “Models generally find little O3 production in wildfire plumes for short aging times (days) because NOX emissions are low and conversion to peroxyacetylnitrate (PAN) is rapid. In contrast, observations show large O3 production from at least some regional wildfires that may significantly elevate O3 at low altitude sites on a monthly basis, and persist over long distances from the burned region.” They also note that fire plumes transported on intercontinental scales can contain very high O3 concentrations. However Singh et al. (2010b) found appreciable increases of O3 in California fire plumes only when they are mixed with urban pollution. Jaffe and Wigder (2012) note that this result could have also been due to suppression of O3 production near the source. Indeed, Singh et al. (2012) note that low O3 found in the California fire plumes of 2008 were associated with low concentrations of NOX and that NOX from urban emissions was needed to produce high O3. Factors such as the stage of combustion (smoldering to flaming), fuel nitrogen content, ambient meteorological conditions, and the availability of solar ultraviolet radiation need to be considered when evaluating the potential of fires for producing O3.

Jaffe et al. (2008) examined the effects of wildfires on O3 in the western United States. They found a strong relation (R² = 0.60) between summer mean O3 measured at various national park and CASTNET sites and area burned in the western United States. They also found generally higher concentrations within surrounding 5° × 5° and 10° × 10° of burned areas. Smaller correlations were found within the surrounding 1°× 1° areas, reflecting near source consumption of O3 and the time necessary for photochemical processing of emissions to form O3. Jaffe et al. (2008) estimate that burning 1 million acres in the western U.S. during summer results in an increase in O3 of 2 ppb across the region; this translates to an average O3 increase across the entire western U.S. of 3.5 and 8.8 ppb during mean and maximum fire years. The unusually warm and dry weather in central Alaska and western Yukon in the summer of 2004 contributed to the burning of 11 million acres there. Subsequent modeling by Pfister et al. (2005) showed that the CO contribution from these fires in July 2004 was 33.1 (± 5.5) MT that summer, roughly comparable to total U.S. anthropogenic CO emissions during the same period.

These results underscore the importance of wildfires as a source of important O3 precursors. In addition to emissions from forest fires in the U.S., emissions from forest fires in other countries can be transported to the U.S., for example from boreal forest fires in Canada (Mathur, 2008), Siberia (Generoso et al., 2007) and tropical forest fires in the Yucatan Peninsula and Central America (Wang et al., 2006). These fires have all resulted in notable increases in O3 concentrations in the U.S.

Estimates of biogenic VOC, NOx, and CO emissions can be made using the BEIS model with data from the BELD and annual meteorological data or MEGAN. VOC emissions from vegetation were described in Section 3.2.
As discussed in Section 3.2.1, NO$_X$ is produced by lightning. Kaynak et al. (2008) found lightning contributes 2 to 3 ppb to surface-level background O$_3$ centered mainly over the southeastern U.S. during summer. Although total column estimates of lightning produced NO$_X$ are large compared to anthropogenic NO$_X$ during summer, lightning produced NO$_X$ does not contribute substantially to the NO$_X$ burden in the continental boundary layer. For example, (Fang et al., 2010) estimated that only 2% of NO$_X$ production by lightning occurs within the boundary layer and most occurs in the free troposphere. In addition, much of the NO$_X$ produced in the free troposphere is converted to more oxidized N species during downward transport. Note that contributions of natural sources to North American background arise from everywhere in the world.

3.4.2 Contributions from Anthropogenic Emissions

In addition to emissions from North America, anthropogenic emissions from Eurasia have contributed to the global burden of O$_3$ in the atmosphere and to the U.S. (NRC, 2009, and references therein). Because the mean tropospheric lifetime of O$_3$ is on the order of a few weeks (Hsu and Prather, 2009), O$_3$ can be transported from continent to continent and around the globe in the Northern Hemisphere. Ozone produced by U.S. emissions can, therefore, be recirculated around northern mid-latitudes back to the United States. High elevation sites are most susceptible to the intercontinental transport of pollution especially during spring. For example, a number of occurrences of O$_3$>60 ppb from mid-April to mid-May of 2006 were observed at Mt. Bachelor Observatory, OR (elevation 2,700 meters) with a maximum O$_3$ concentration of ~85 ppb observed on April 22, 2006. Calculations using GEOS-Chem, a global-scale CTM, indicate that Asia contributed 9 ± 3 ppb to a modeled mean concentration of 53 ± 9 ppb O$_3$ at Mt. Bachelor during the same period compared to measured concentrations of 54 ± 10 ppb (Zhang et al., 2008). Zhang et al. (2008) also calculated a contribution of 5 to 7 ppb to surface O$_3$ over the western U.S. during that period from Asian anthropogenic emissions. They also estimated an increase in NO$_X$ emissions of ~44% from Asia from 2001 to 2006 resulting in an increase of 1-2 ppb in O$_3$ over North America.

Cooper et al. (2010) analyzed all available O$_3$ measurements in the free troposphere above western North America at altitudes of 3-8 km (above sea level) during April and May of 1995 to 2008 (i.e., times when intercontinental transport is most prominent). They derived a trend of +0.63 ± 0.34 ppb/year in median O$_3$ concentrations with indication of a similar rate of increase since 1984. Back trajectories that were likely to have been strongly and recently influenced by North American emissions were filtered out, resulting in a trend of +0.71 ± 0.45 ppb/year. Considering only trajectories with an Asian origin resulted in a trend of +0.80 ± 0.34 ppb/year. These results suggest that local North American emissions were not responsible for the measured O$_3$ increases. This O$_3$ could have been produced from natural and anthropogenic precursors in Asia and Europe with some contribution from North American emissions that have circled the globe. Cooper et
al. (2010) also found that it is unlikely that the trends in tropospheric O₃ are associated with trends in stratospheric intrusions. Note, however, that these results relate to O₃ trends above ground level and not to surface O₃. Model results (Zhang et al., 2008) show that surface O₃ contributions from Asia are much smaller than those derived in the free troposphere because of dilution and chemical destruction during downward transport to the surface. These processes tend to reduce the strength of associations between free tropospheric and surface O₃ especially if air from other sources is sampled by the surface monitoring sites.

Trinidad Head, CA is one sampling location at which measurements might be expected to reflect in large measure NA background O₃ contributions, at times during the spring (Oltmans et al., 2008; Goldstein et al., 2004). The monitoring station at Trinidad Head is on an elevated peninsula extending out from the mainland of northern California, and so might be expected at times to intercept air flowing in from the Pacific Ocean with little or no influence from sources on the mainland. Figure 3-8 shows the time series of MDA8 O₃ concentrations measured at Trinidad Head from April 18, 2002 through December 31, 2009. The data show pronounced seasonal variability with spring maxima and summer minima. Springtime concentrations typically range from 40 to 50 ppb with a number of occurrences >50 ppb. The two highest daily maxima were 60 and 62 ppb. The data also show much lower concentrations during summer, with concentrations typically ranging between 20 and 30 ppb. Oltmans et al. (2008) examined the time series of O₃ and back trajectories reaching Trinidad Head. They found that springtime maxima (April-May) were largely associated with back trajectories passing over the Pacific Ocean and most likely entraining emissions from Asia, with minimal interference from local sources. However, Parrish et al. (2009) noted that only considering trajectories coming from a given direction is not sufficient for ruling out local continental influences, as sea breeze circulations are complex phenomena involving vertical mixing and entrainment of long-shore components. They found that using a wind speed threshold in addition to a criterion for wind direction allowed determination of background trajectories not subject to local influence. This was confirmed by measurements of chemical tracers of local influence such as CO₂, MTBE and radon. By applying the two criteria for wind speed and direction, they found that Trinidad Head met these criteria only 43% of the time during spring. Goldstein et al. (2004) used CO₂ as an indicator of exchange with the local continental environment and found that O₃ concentrations were higher by about 2-3 ppb when filtered against local influence indicating higher O₃ in air arriving from over the Pacific Ocean. At other times of the year, Trinidad Head is less strongly affected by air passing over Asia and the northern Pacific Ocean; and many trajectories have long residence times over the semi-tropical and tropical Pacific Ocean where O₃ concentrations are much lower than they are at mid-latitudes. The use of the Trinidad Head data to derive contributions from background sources requires the use of screening procedures adopted by Parrish et al. (2009) and the application of photochemical models to determine the extent either of titration of O₃ by fresh NOₓ emissions and the extent of local production of O₃ from these emissions. Although O₃ concentrations at Trinidad Head might at times be representative of Pacific air arriving over the
U.S., they cannot be viewed as NA background over the continental U.S. because of deposition to the surface and chemical loss over the continental United States. As noted above, anthropogenic emissions from North America also contribute to hemispheric background and must be filtered out from observations at coastal sites such as Trinidad Head even when it is thought that air sampled came directly from over the Pacific Ocean and was not influenced by local pollutant emissions.

Figure 3-8  Time series of MDA8 O$_3$ concentrations (ppm) measured at Trinidad Head, CA, from April 18, 2002 through December 31, 2009.

Parrish et al. (2009) also examined data obtained at other marine boundary layer sites on the Pacific Coast. These include Olympic NP, Redwood NP, Point Arena, and Point Reyes. Using data from these sites, they derived trends in O$_3$ of +0.46 ppb/year (with a 95% confidence interval of 0.13 ppb/year) during spring and +0.34 ppb/year (0.09 ppb/year) for the annual mean O$_3$ increase in air arriving from over the Pacific during the past two decades. Although O$_3$ data are available from the Channel Islands, Parrish et al. (2009) noted that these data are not suitable for determining background influence because of the likelihood of circulating polluted air from the South Coast Basin.
The 2010 Intercontinental Chemical Transport Experiment Ozone Network Study (IONS-2010) and Research at the Nexus of Air Quality and Climate Change (CalNex) study conducted in May through June of 2010 had discerning the contributions of Asian emissions to air quality in California as a major focus. Cooper et al. (2011) examined O₃ profiles measured above four coastal sites in California, including Trinidad Head. Based on trajectory analyses coupled with comparison with the O₃ profiles, they suggested that Asian pollution, stratospheric intrusions, and international shipping made substantial contributions to lower tropospheric O₃ (typically 0 to 3 km above sea level, meant as a rough approximation of planetary boundary layer height) measured at inland California sites. These contributions tended to increase on a relative basis in going from south to north. In particular, no contribution from local pollution was needed to explain lower tropospheric O₃ in the northern Central Valley; and the contribution of local pollution to lower tropospheric O₃ in the LA basin ranged from 32 to 63% (depending on layer depth; either 0 to 1.5 km or 0 to 3 km). It should be noted that the extent of photochemical production and loss occurring in the descending air masses between the coastal and inland sites remains to be determined. Cooper et al. (2011) also note that very little of the O₃ observed above California reaches the eastern United States. However, this does not necessarily mean that the pathways by which Asian O₃ could reach the eastern U.S. were fully captured in this analysis.

Lin et al. (2012) used the AM3 model (~50 × 50 km resolution globally) and satellite data to characterize the influence of Asian emissions and stratospheric intrusions on O₃ concentrations in southern California and Arizona during CalNex (May-June 2010). The model simulates sharp O₃ gradients in the upper troposphere and the interweaving and mixing of stratospheric air and Asian plumes. Similar phenomena were also found during field campaigns conducted in the North Atlantic as noted in Annex AX2.3.1 of the 2006 O₃ AQCD (U.S. EPA, 2006b) and introduces uncertainty into attempts to attribute O₃ to these sources, based solely on observations, because this mixing will affect relationships between CO (mainly a marker for polluted air that is commonly used to separate air influenced by anthropogenic pollution from stratospheric air) and O₃ (a pollutant and a stratospheric component). Lin et al. (2012) found that Asian emissions contributed from 20-30% to O₃ in the mid troposphere over the California coast and remnants of stratospheric intrusions contributed from 50 to 60% to O₃ in discrete layers in the same altitude range. This O₃ then has the potential to mix downward into the planetary boundary layer. Lin et al. (2012) also found evidence of Asian contributions of up to 8 -15 ppb in surface air during strong transport events in southern California. These contributions are in addition to contributions from dominant local pollution sources. Their results suggest that the influence of background sources on high surface O₃ concentrations is not always confined to high elevation sites. However, it is not clear to what extent the contributions inferred by Cooper et al. (2011) and Lin et al. (2012) are likely to be found in other years or how long they would extend into summer.
3.4.3 Estimating Background Concentrations

Historically, two approaches to estimating NA background concentrations (previously referred to as PRB) have been considered in past O\textsubscript{3} assessments. In the 1996 and earlier O\textsubscript{3} AQCDs, measurements from remote monitoring sites were used. In the 2006 O\textsubscript{3} AQCD, the use of CTMs was adopted, because as noted in Section 3.9 of the 2006 O\textsubscript{3} AQCD (U.S. EPA, 2006b), estimates of background concentrations cannot be obtained directly by examining measurements of O\textsubscript{3} obtained at relatively remote monitoring sites in the U.S. because of the long-range transport from anthropogenic source regions within North America. The 2006 O\textsubscript{3} AQCD (U.S. EPA, 2006b) also noted that it is impossible to determine sources of O\textsubscript{3} without ancillary data that could be used as tracers of sources or to calculate photochemical production and loss rates. As further noted by Reid et al. (2008), the use of monitoring data for estimating background concentrations is essentially limited to the edges of the domain of interest. The current definition of NA background implies that only CTMs (see Section 3.3 for description and associated uncertainties) can be used to estimate the range of background concentrations. An advantage to using models is that the entire range of O\textsubscript{3} concentrations measured in different environments can be used to evaluate model performance. In this regard, data from the relatively small number of monitoring sites at which large background contributions are expected are best used to evaluate model predictions.

Estimates of NA background concentrations in the 2006 O\textsubscript{3} AQCD (U.S. EPA, 2006b) were based on output from the GEOS-Chem (v4.3.3) model (Fiore et al., 2003) with 2° × 2.5° horizontal resolution. The GEOS-Chem model estimates indicated that NA background O\textsubscript{3} concentrations in eastern U.S. surface air were 25 ± 10 ppb (or generally 15-35 ppb) from June through August, based on conditions for 2001. Values reported by Fiore et al. (2003) represent averages from 1 p.m. to 5 p.m.; all subsequent values given for background concentrations refer to MDA8 O\textsubscript{3} concentrations. Background concentrations decline from spring to summer. Background O\textsubscript{3} concentrations may be higher, especially at high altitude sites during the spring, due to enhanced contributions from (1) pollution sources outside North America; and (2) stratospheric O\textsubscript{3} exchange. At the time, only the GEOS-Chem model (Harvard University, 2010b) was documented in the literature for calculating background O\textsubscript{3} concentrations (Fiore et al., 2003). The simulated monthly mean concentrations in different quadrants of the U.S. were typically within 5 ppbv of observations at CASTNET sites, with no discernible bias, except in the Southeast in summer when the model was 8-12 ppbv too high. This bias was attributed to excessive background O\textsubscript{3} transported in from the Gulf of Mexico and the tropical Atlantic Ocean in the model (Fiore et al., 2003).

Although many of the features of the day-to-day variability in O\textsubscript{3} at relatively remote monitoring sites in the U.S. were simulated reasonably well by GEOS-Chem (Fiore et al., 2003), uncertainties in the calculation of the temporal variability of O\textsubscript{3} originating from different sources on shorter time scales must be recognized. The uncertainties stem in part from an underestimate in the seasonal variability in the
STE of O₃ (Fusco and Logan, 2003), the geographical variability of this exchange, and the variability in the exchange between the free troposphere and the PBL in the model. In addition, the relatively coarse spatial resolution in that version of GEOS-Chem (2° × 2.5°) limited the ability to provide separate estimates for cities located close to each other, and so only regional estimates were provided for the 2006 O₃ AQCD (U.S. EPA, 2006b) based on the results of Fiore et al. (2003).

Wang et al. (2009a) recomputed NA background concentrations for 2001 using GEOS-Chem (v7-01-01) at higher spatial resolution (1° × 1°) over North America and not only for afternoon hours but for the daily maximum 8-h avg O₃ concentration. The resulting background concentrations, 26.3 ± 8.3 ppb for summer, are consistent with those of 26 ± 7 ppb for summer reported by Fiore et al. (2003), suggesting horizontal resolution was not a substantial factor limiting the accuracy of the earlier results. In addition to computing NA background concentrations, Wang et al. (2009a) also computed U.S. background concentrations of 29.6 ± 8.3 ppb with higher concentrations in the Northeast (up to 15 ppb higher) and the Southwest (up to 13 ppb higher) for summer means.

3.4.3.1 Updated GEOS-Chem Model Estimates of Background Concentrations

Zhang et al. (2011) computed NA background, U.S. background and natural background O₃ concentrations using GEOS-Chem (v8-02-03) at an even finer grid spacing of 0.5° × 0.667° over North America for 2006 through 2008. For March through August 2006, mean NA background O₃ concentrations of 29 ± 8 ppb at low elevation (<1,500 meters) and 40 ± 8 ppb at high elevation (>1,500 meters) were predicted. Spring and summer mean O₃ concentrations calculated for the base case (i.e., including all natural and anthropogenic sources worldwide), U.S. background, and NA background in surface air for spring and summer 2006 calculated by Zhang et al. (2011) are shown in the upper, middle and lower panels of Figure 3-9.
Note: Values in parentheses refer to continental U.S. means and are shown as black squares in the color bar for summer and white squares for spring.

Source: Adapted from Zhang et al. (2011).

**Figure 3-9** Mean MDA8 $O_3$ concentrations in surface air for spring and summer 2006 calculated by GEOS-Chem for the base case (Base), U.S. background (USB), and NA background (NAB).
As noted above, Zhang et al. (2011) found increases in Asian emissions only accounted for an average increase of between 1 to 2 ppb in background O₃ across the U.S. even though Asian emissions have increased by about 44% from 2001 to 2006. As can be seen from Figure 3-9, U.S. background and NA background concentrations are very similar throughout most of the United States. Zhang et al. (2011) also found that NA background concentrations are ~4 ppb higher, on average, in the 0.5° × 0.667° version than in the coarser 2° × 2.5° version. This difference was partially due to higher resolution (~1 to 2 ppb) and the remainder to the combination of changes in lightning and Asian emission estimates as well as higher model resolution.

As can be seen from the middle and lower panels in Figure 3-9, U.S. background and NA background concentrations tend to be higher in the West, particularly in the Intermountain West and in the Southwest compared to the East in both spring and summer. U.S. background and NA background concentrations tend to be highest in the Southwest during summer in the GEOS-Chem model, driven by lightning NOₓ.

Intercontinental transport and stratospheric intrusions are major contributors to the high elevation, Intermountain West during spring with wildfires becoming more important sources during summer. The base case O₃ concentrations (upper panels) show two broad maxima with highest concentrations extending throughout the Southwest, Intermountain West, and the East; in both spring and summer. These maxima extend over many thousands of kilometers demonstrating that O₃ is a regional pollutant. Low-level outflow from the Northeast out over the Atlantic Ocean and from the Southeast out over the Gulf of Mexico is also apparent.

Note that background concentrations tend to increase with increasing base model (and measured) concentrations at higher elevation sites, particularly during spring. However, it is not only background O₃ that increases with elevation. Convection efficiently lifts precursor emissions and O₃ from the polluted boundary layer in North America as well as in Europe and Asia to the mid and upper troposphere. Significant production of O₃ from precursor emissions and from lightning can occur aloft, providing a diffuse source in the mid and upper troposphere because the O₃ production efficiency with respect to NOₓ is much higher aloft than at low elevations. Higher wind speeds aloft, coupled with a lifetime of O₃ that increases with height in the troposphere, allow pollutants to be transported much farther and to mix over wider areas than they would at lower elevations. Thus, through these mechanisms, O₃ aloft at a given location can be higher than at low elevations and caution should be observed in attempting to ascribe increases in O₃ with elevation to particular sources.

Although the results of Zhang et al. (2011) are broadly consistent with results from earlier coarser resolution versions of GEOS-Chem used by Fiore et al. (2003) and Wang et al. (2009a), there are some apparent differences. Concentrations of O₃ for both the base case and the NA background case in Zhang et al. (2011) are higher in the Intermountain West than in earlier versions. In addition, background concentrations in many eastern areas tend to be higher on days when predicted total
O$_3$ is > 60 ppb or at least do not decrease with increasing total O$_3$ Zhang et al. (2011).

Figure 3-10 shows seasonal mean estimates of contributions to O$_3$ from Canadian and Mexican emissions calculated by Zhang et al. (2011) as the difference between U.S. background and NA background values and then averaged over spring and summer following the procedure in Wang et al. (2009a). U.S. background concentrations are on average 3 ppb higher than NA background concentrations during spring and summer across the United States. Highest values in Figure 3-10 (in the U.S.) are found over the Northern Tier of New York State (19.1 ppb higher than NA background) in summer. High values are also found in other areas bordering Canada and Mexico. Although the contributions from Canada and Mexico were obtained by differencing, it should be remembered that relations between O$_3$ and precursors are subject to non-linear effects that are strongest near concentrated sources of precursors, as noted in Section 3.2.4. Therefore, the values shown in the figure are only estimates of contributions to total O$_3$ coming from Canada and Mexico.

Figure 3-10 Spring and summer mean Canadian and Mexican (CM) contributions to MDA8 O$_3$ determined as the difference between the U.S. background and NA background.

Figure 3-11 shows MDA8 O$_3$ concentrations for spring (March-May) and summer (June-August) 2006 simulated by GEOS-Chem (versus measured) by the ensemble of CASTNET sites in the Intermountain West, Northeast, Great Lakes, and Southeast (Zhang et al., 2011). Shown is the 1:1 line and NA background (blue) and natural background (green) model statistics as box plots (minimum, 25th, 50th, 75th
percentile, and maximum) for 10-ppbv bins of observed $O_3$ concentrations. These plots show that NA background constitute a larger fraction of modeled base case $O_3$ at the upper end of the concentration distribution for the Intermountain West than for other regions of the country.

Figure 3-11  MDA8 $O_3$ concentrations for spring (March-May) and summer (June-August) 2006 simulated by GEOS-Chem vs. measured by the ensemble of CASTNET sites in the Intermountain West, Northeast, Great Lakes, and Southeast.
Comparisons between GEOS-Chem and measurements of the mean MDA8 O\textsubscript{3} between March and August at individual CASTNET sites across the country are shown as supplemental material in Section 3.8, Figure 3-58 through Figure 3-64. In general, the GEOS-Chem predictions tend to show better agreement with observations at the high-altitude sites than at the low-altitude sites. Overall agreement between model results for the base case and measurements is within a few parts per billion for spring-summer means in the Northeast (see Figure 3-58 in Section 3.8) and the Southeast (see Figure 3-59 in Section 3.8), except in and around Florida where the base case over-predicts O\textsubscript{3} by 10 ppb on average. In the Upper Midwest (Figure 3-60 in Section 3.8), the Intermountain West (Figure 3-61 and Figure 3-62 in Section 3.8), and the West (Figure 3-63 in Section 3.8) including most sites in California (Figure 3-64 in Section 3.8), the model predictions are within 5 ppb of measurements. The model under-predicts O\textsubscript{3} by 10 ppb at the Yosemite site (Figure 3-64 in Section 3.8). These results suggest that the model is capable of calculating March to August mean MDA8 O\textsubscript{3} to within ~5 ppb at most (26 out of 28) sites chosen.

Comparison between results in Wang et al. (2009a) for 2001 with data obtained in the Virgin Islands indicate that GEOS-Chem over-predicts summer mean MDA8 O\textsubscript{3} concentrations there by 10 ppb (28 versus 18 ppb). Ozone concentrations at the Virgin Islands NP site appear not to have been affected by U.S. emissions, based on the close agreement between the base case and the NA background case. Wind roses calculated for the Virgin Islands site indicate that wind patterns affecting this site are predominantly easterly/southeasterly in spring and summer. The over-predictions at the Virgin Islands site imply that modeled O\textsubscript{3} over the tropical Atlantic Ocean is too high. As a result, inflow of O\textsubscript{3} over Florida and into the Gulf of Mexico is also likely to be too high as winds are predominantly easterly at these low latitudes. Similar considerations apply to the results of Zhang et al. (2011). Possible explanations include deficits in model chemistry (for example, reactions involving halogens are not included) and/or subsidence that is too strong over tropical oceans in the model. No clear explanation can be provided on why the model under-predicts mean O\textsubscript{3} at Yosemite (elevation 1,680 meters) by ~10 ppb (see Figure 3-64 in Section 3.8). However, March to August mean MDA8 O\textsubscript{3} concentrations are simulated to within a few parts per billion at an even higher elevation site in California (Converse Station, elevation 1,837 meters) and at the low elevation sites.

Figure 3-65 in Section 3.8 shows a comparison of GEOS-Chem output with measurements at Mt. Bachelor, OR and Trinidad Head, CA from March-August, 2006 from Zhang et al. (2011). For the Mt. Bachelor model runs, model estimates are given for both a coarse (2° × 2.5°) and fine (0.5° × 0.667°) resolution model. In general, mean concentrations are simulated reasonably well at both coarse and finer grid resolution versions of the model with mean values 2 ppb higher in the finer resolution model. Although the finer resolution version provides some additional day to day variability and can capture the timing of peaks, it still does not adequately resolve peak concentrations as can be seen for an event in the second half of April.
Figure 3-66 in Section 3.8 shows a comparison of vertical profiles (mean ± 1σ) calculated by GEOS-Chem with ozonesondes launched at Trinidad Head, CA and Boulder, Colorado. As can be seen from the figure, variability in both model and measurements increases with altitude, but variability in the model results is much smaller at high altitudes than seen in the observations. This may be due in part to the inability of grid-point models to capture the fine-scale, layered structure often seen in O_3 in the mid and upper troposphere (Rastigejev et al., 2010; Newell et al., 1999) and to inadequacies in parameterizations of relevant chemistry and dynamics. Figure 3-67 and Figure 3-68 in Section 3.8 show a comparison of vertical profiles simulated by AM3 at 50 × 50 km global resolution (Lin et al., 2012) with ozonesondes launched at several locations in California during May-June 2010. Note that in contrast to comparing measured mean monthly O_3 profiles to monthly mean profiles calculated by GEOS-CHEM (see, for example, Figure 3-66 in Section 3.8), AM3 is sampled for comparison to individual measurements of O_3 profiles. This model has likely had the most success in simulating vertical O_3 gradients in the upper troposphere and in capturing layered structures in the mid and upper troposphere.

The natural background for O_3 averages 18 ± 6 ppbv at the low elevation sites and 27 ± 6 ppbv at the high elevation sites in the GEOS-Chem model Zhang et al. (2011). In regions where non-linear effects are small, far from concentrated sources of O_3 precursors, the difference between NA background and natural background O_3 concentrations provides an estimate of contributions from intercontinental pollution including anthropogenic methane (given by the difference between values in 2006 and the pre-industrial era, or 1,760 ppb and 700 ppb). The difference between the two backgrounds averages 9 ppbv at the low elevation sites and 13 ppbv at sites in the Intermountain West. Based on the Zhang et al. (2011) model runs, anthropogenic methane emissions are estimated to contribute ~4-5 ppb to global annual mean O_3 surface concentrations. North American emissions of methane are uncertain, but are a small fraction of total anthropogenic input. This suggests that slightly less than half of the difference between North American background and natural background is due to the increase of methane since the beginning of the industrial era and the other half is due to anthropogenic emissions of shorter lived VOCs and NOX. However, the relative importance of methane for O_3 production is expected to increase in the future. Indeed, variations in methane concentrations account for approximately 75% of the wide spread (~5 ppb) in tropospheric O_3 projections between Representative Concentration Pathway (RCP) scenarios for the next century (Wild et al., 2012); see Section 10.3.6.1 in Chapter 10 for more on the RCP scenarios.

Figure 3-12 shows frequency distributions for observations at low-altitude and high-altitude CASTNET sites along with GEOS-Chem frequency distributions for the base case, NA background, and natural background. Most notable is the shift to higher concentrations and the narrowing of the concentration distributions for all three simulations and the observations in going from low to high altitudes. However, maximum concentrations show little if any dependence on altitude, except for the natural background (which tends to be slightly higher at high altitude sites).
Figure 3-12  Frequency distributions of MDA8 O₃ concentrations in March-August 2006 for the ensemble of low-altitude (<1,500 meters) and high-altitude CASTNET sites (>1,500 meters) in the U.S.

3.4.3.2 Using Other Models to Estimate Background Concentrations

Another approach to modeling background concentrations involves using a regional CTM such as CMAQ or CAMx with boundary conditions taken from a global scale CTM such as GEOS-Chem (see Section 3.3 for discussion of this approach). Mueller and Mallard (2011a), while not calculating NA background values exactly as defined here, calculated contributions from natural sources and inflow from the boundaries to O₃ for 2002 using MM5 and CMAQ for the outermost domain (36 km resolution) shown in Figure 3-4 with boundary conditions from GEOS-Chem. The overall bias based on comparison with AQS monitors for the base case is about 3 ppb; the annual mean fractional bias and mean fractional error were 7% and 21% for the ozone season across the United States. Note that Figure 2 in their paper is mislabeled, as it
should refer to the case with total emissions - not to natural emissions in North America only (Mueller and Mallard, 2011b). However, boundary conditions are fixed according to monthly averages based on an earlier version of GEOS-Chem and do not reflect shorter term variability or trends in Northern Hemispheric emissions of pollution. In addition, fluxes of O₃ from the stratosphere are not included explicitly. Note that their natural background includes North American natural background emissions only and influence from boundary conditions and thus is not a global natural background. Calculated values including natural emissions from North America and from fluxes through the boundaries are somewhat larger than given in Zhang et al. (2011), in large measure because of much larger contributions from wildfires and lightning. Wildfire contributions reach values of ~140 ppb in Redwoods National Park, CA and higher elsewhere in the U.S. and in Quebec in the simulations by Mueller and Mallard (2011a). Lightning contributions (ranging up to ~30 ppb) are substantially larger than estimated by Kaynak et al. (2008) (see Section 3.4.1.2). The reasons for much larger contributions from wildfires and lightning found by Mueller and Mallard (2011a) are not clear and need to be investigated further.

Emery et al. (2012) used CAMx with boundary conditions taken from the coarse resolution version of GEOS-Chem (2 × 2.5° or ~200 km resolution) to derive NA background concentrations of O₃. The nested CAMx simulations were run at a horizontal resolution of 12 km separately for the eastern and western United States. The following paragraphs compare results from the Emery et al. (2012) nested CAMx simulations at 12 km resolution, with those obtained by Zhang et al. (2011) using GEOS-Chem simulations at 0.5° × 0.667° (~50 km) resolution. This is in contrast to the comparison reported in Emery et al. (2012) in which results from CAMx at 12 km resolution were compared to results from the 2° × 2.5° (~200 km) resolution version of the GEOS-Chem model over the United States.

Figure 3-13 shows seasonal mean MDA8 O₃ concentrations calculated by Emery et al. (2012) using CAMx for 2006 for the base case and for NA background. Figure 3-14 shows a comparison of monthly average O₃ concentrations calculated by GEOS-Chem (Zhang et al., 2011) with those calculated by CAMx (Emery et al., 2012). Comparison of the base case for GEOS-Chem with that for CAMx in Figure 3-14 indicates broad agreement in spatial patterns.
Figure 3-13  Mean MDA8 O₃ concentrations in surface air during spring and summer 2006 (top) calculated by GEOS-Chem/CAMx for the base case (Base, top) and NA background (NAB, bottom).
Figure 3-14  
Monthly average MDA8 O₃ concentrations observed (Obs) and predicted for the base case and NA background (NAB) by GEOS-Chem (GC) and GEOS-Chem/CAMx (CX) at CASTNET sites above 1,500 meters elevation (upper panel) and CASTNET sites below 1,500 meters elevation (lower panel).
Supplemental figures (Figure 3-69 through Figure 3-74) in Section 3.8 show box plots comparing MDA8 O₃ concentrations calculated by GEOS-Chem at 0.5° × 0.667° resolution and CAMx for March-August 2006 at the combined set of CASTNET sites used by both groups for model evaluation. Note that the individual model results and the observations are unpaired in time. At CASTNET sites in the Northern Rockies, both models tend to under-predict maximum O₃ concentrations, but they are generally higher in CAMx than in GEOS-Chem (typically by 5-10 ppb). The distribution of MDA8 values from GEOS-Chem is consistent with measured distributions (i.e., cannot be rejected using Mann-Whitney rank sum test, p-value <0.01) at 18 of 39 sites in spring and 21 of 39 sites in summer. The distribution of MDA8 values from CAMx is consistent with measured distributions at 13 of 39 sites in spring and 18 of 39 sites in summer. When spring and summer are pooled, both simulations are consistent with measured distributions at 16 out of 39 sites (but not the same 16 sites). There are examples in which either model over- or under-simulates maximum concentrations. However, over-predictions are made more often by CAMx. At high elevations in the Intermountain West (see Figure 3-72 in Section 3.8), both models tend to under-predict maxima, but their interquartile range agrees much better with observations. As McDonald-Buller et al. (2011) noted, complex topography in some regions of the U.S. could influence surface O₃ through fine-scale, orographically induced flow regimes. In addition, numerical diffusion broadly affects the ability of models to capture observed maxima, particularly at mountain sites. McDonald-Buller et al. (2011) also note there are regions in the U.S. where global models show consistent biases. For example, models are generally unable to simulate the low O₃ concentrations observed at Gulf Coast sites in summer during onshore flow from the Gulf of Mexico, which could reflect marine boundary layer chemistry and/or stratification that is not properly represented in the model. Both models over-predict O₃ at two sites in Florida (Sumatra and Indian River Lagoon). However, further inland, CAMx tended to over-predict O₃ at the Coffeeville, MS; Sand Mountain, AL; and Georgia Station, GA sites whereas GEOS-Chem did not. The same is true for higher elevation sites (Great Smoky Mountain, NC-TN; Shenandoah, VA). In the Northeast, there is a general tendency for both models to over-predict the measured distributions with somewhat higher maximum concentrations in CAMx compared to GEOS-Chem and observations (see Figure 3-69 to Figure 3-74 in Section 3.8).

The most readily discernible differences in model formulation are in the model grid spacing and the treatment of wildfires. The finer resolution in CAMx allows for topography to be better-resolved producing higher maximum O₃ concentrations in the Intermountain West. For wildfires, treatment differences include emission composition, emission time averaging, and associated chemistry. Wildfires produce more O₃ in CAMx simulations than in GEOS-Chem simulations, and Emery et al. (2012) attribute these enhancements to shorter emission time averaging. The CAMx emissions average fire emissions at hourly resolution based on the SmartFire algorithm, whereas GEOS-Chem uses monthly averages from GFED2. Each model representation also uses different emission compositions. The emissions used by Emery et al. (2012) include a larger number of VOCs and additional categories of VOCs than used by Zhang et al. (2011). Following emission, Emery et al. (2012)
note that photochemical aging of wildfire emissions depends on the chemical mechanism. Neither chemical mechanism was designed specifically for these type of events. GEOS-Chem has traditionally focused on the chemistry of the non-urban troposphere and does not represent secondary products of fast reacting VOCs as does CB05. A lack of reactivity of secondary products would cause a dampening of fire contributions to O$_3$. CB05 has traditionally focused on urban chemistry and does not explicitly includes ketones (Henderson et al., 2011), which are among the top ten VOCs emitted from fires (Andreae and Merlet, 2001). The O$_3$ increases seen in Emery et al. (2012) and Mueller and Mallard (2011a), however, are subject to uncertainties in the representation of physics in the wildfire plumes. The improvements in characterizing emissions would lead to smoke plumes that attenuate light, thereby reducing photolysis and photoreactivity (e.g., Real et al., 2007). The wildfires would also alter temperature and convective activity that influences plume rise and the height of the planetary boundary layer. Emery et al. (2012) note the need for more research to improve simulation of O$_3$ from fires. Using a sensitivity analysis of CAMx, the authors showed that removing wildfires in the West (California, Oregon, and Idaho) resulted in reductions of NA background O$_3$ of 10 to 50 ppb, with smaller reductions elsewhere. Further, Emery et al. (2012) note that their calculated O$_3$ increases in the vicinity of wildfires is consistent with that of Mueller and Mallard (2011a).

Emery et al. (2012) captured the timing of a possible stratospheric intrusion at Gothic, Colorado, on April 19-20, 2006, and predicted an MDA8 O$_3$ value of ~73 ppb using CAMx on April 20, compared to a measured observation value of 87 ppb. GEOS-Chem (at 0.5° × 0.667°) predicted ~65 ppb for this event. The higher spatial resolution in CAMx likely contributed to the improvement in model performance, but this may not be the only factor. AM3, another global scale CTM (Lin et al., 2012) at ~2° × 2.5° resolution predicted ~75 ppb for that event; suggesting that differences in dynamical cores between WRF and AM3, different treatments of the stratospheric O$_3$ source, and perhaps the spatial extent of the intrusion’s effect on surface O$_3$ should be considered in addition to model resolution. Typically, these strong intrusions have spatial extent of thousands of kilometers along their axis and hundreds of kilometers transversely to this axis, as noted in the 2006 O$_3$ AQCD (U.S. EPA, 2006b). Note that all three models (CAMx, GEOS-Chem, and AM3) underpredicted the magnitude of this event. These results indicate a need for process-oriented evaluation and targeted measurements that yield insight into both chemical and dynamical processes. The $R^2$ for comparison of AM3 with observations of MDA8 O$_3$ from March-August 2006 was 0.33, with lower $R^2$ for GEOS-Chem and CAMx. All three models predicted very similar means for March to August 2006: 54.9 ppb (simulated by the fine resolution version of GEOS-Chem), 55.0 ppb (simulated by CAMx), and 58.6 ppb (simulated by AM3), compared to 55.9 ppb for observed concentrations. See Figure 3-75 in Section 3.8; however, note that Figure 3-75 does not show the CAMx mean of 55.0 ppb, which was reported in Emery et al. (2012).

The results from either model have also been compared to more urban oriented sites in the AQS network. As noted earlier, comparisons between model results and
observations become problematic near concentrated sources of O₃ precursors (NOₓ and VOCs) in urban cores. Emery et al. (2012) note that in coarse resolution models rural biogenic and urban precursor emissions are mixed immediately leading to higher production efficiency for O₃. Finer resolution models are better able to separate these two source categories and to resolve features of urban chemistry such as titration of O₃ by NOₓ emitted by traffic and subsequent processing of NOₓ emissions during transport downwind. CAMx at 12 × 12 km resolution is better able to capture these features than GEOS-Chem at 50 × 50 km resolution. Both models tend to over-predict O₃ at the low O₃ concentrations in areas where O₃ scavenging by NOₓ is evident. In these situations, NA background O₃ concentrations are often higher than in the respective models for the base case. At high O₃ concentrations downwind of source areas, both models predict NA background O₃ concentrations that are much lower than observed or base case O₃. The latter results are in accord with results shown in Figure 3-11 for rural CASTNET sites at low elevations, which show lower ratios between NA background O₃ and either observations or base case O₃ at high O₃ than at low O₃ concentrations.

Figure 3-15 shows the annual 4th-highest MDA8 O₃ predicted by GEOS-Chem (at 0.5° × 0.667° resolution) for the base case (upper panel), and corresponding U.S. background (middle panel) and NA background (lower panel) MDA8 O₃ on the same days for 2006. Figure 3-16 shows corresponding values predicted by CAMx for the base case (upper panel) and NA background (lower panel) MDA8 O₃ on the same days for 2006. As can be seen from Figure 3-15 and Figure 3-16, on those days when models predicted their annual 4th-highest MDA8 O₃, the corresponding NA background concentrations are 36 ± 9 ppb in the eastern United States. Base case concentrations are much higher indicating that regional pollution is mainly responsible for the models 4th-highest concentrations. In the western U.S. on the other hand, NA background concentrations are generally higher and make up a larger fraction of the calculated 4th-highest MDA8 O₃ in both models, but for different reasons. GEOS-Chem predicts highest values in the Southern Rockies because of over-production of NOₓ by lightning. CAMx predicts highest values in ID, OR and WA from wildfires. The CAMx run includes day specific values for area burned, but GEOS-Chem uses monthly averages. (A more recent version of GEOS-Chem also incorporates day specific estimates for area burned.) Remaining areas of relatively high background levels (>60 ppb) are due mainly to some combination of stratospheric intrusions and Eurasian emissions. There are a few examples that can be used to give a rough idea of the magnitudes of episodically high background contributions. A comparison of the annual 4th-highest MDA8 O₃ concentration simulated by CAMx including wildfires and omitting them indicates that wildfires contributed ~30 to 40 ppb in Idaho, Montana, and Washington with a potentially larger contribution in the upper northwestern corner of California. Estimated contributions from strong stratospheric intrusions to surface O₃ in AM3 could range up to ~55 ppb in the western United States. It should be borne in mind in using these figures, that they are model derived and hence could be model specific. Issues related to calculating O₃ formation in wild fire plumes by CAMx were mentioned above. The method for calculating the stratospheric contributions in AM3 is based on the amount of O₃ carried downward through a given surface thus raising the possibility
that O$_3$ could be generated in the upper troposphere, transported into the stratosphere and then transported downward again. This could indeed lead to an overestimate of the baseline stratospheric attribution in the model. However, Lin et al. (2012) focus on the "delta during events" rather than the baseline, because during specific intrusion events this recycling of O$_3$ between troposphere and stratosphere is less likely to be a problem. Therefore, there is greater confidence in the anomaly from stratospheric O$_3$ rather than the absolute amount. Tagging O$_3$ according to where it was produced will remove this ambiguity in future model runs.

All models undergo continuous updating of inputs, parameterizations of physical and chemical processes, and improvements in model resolution. Inputs that might be considered most relevant include emissions inventories, chemical reactions, and meteorological fields. This leads to uncertainty in model predictions in part because there is typically a lag between updated information for the above inputs—as outlined in Section 3.2 for chemical processes and emissions and in Section 3.3 for model construction—and their implementation in CTMs including GEOS-Chem or the other models described above. Quantitative estimates of uncertainties from meteorological and emission inputs and chemical mechanisms are problematic because simulations designed to quantify uncertainties from these sources have not been performed for these model runs. At best, these uncertainties can be estimated by comparison with observations while recognizing that compensating errors likely exist.
Figure 3-15  Annual 4th-highest MDA8 O₃ predicted by GEOS-Chem (0.5° × 0.667°) for the base case (Base) with corresponding U.S. background (USB) and NA background (NAB) MDA8 O₃ for the same days in 2006.
Figure 3-16  Annual 4th-highest MDA8 O$_3$ predicted by CAMx for the base case (Base) and corresponding NA background (NAB) MDA8 O$_3$ for the same days in 2006.
Since NA background is a construct that cannot be directly measured, the range of background O₃ concentrations must be estimated using CTMs. Results from the Zhang et al. (2011) GEOS-Chem and Emery et al. (2012) CAMx model estimates were chosen for further analysis because these models have produced the latest estimates for background O₃ concentrations documented in the open literature. The main results from these two modeling efforts can be described as follows:

- Both models show background concentrations vary spatially and temporally;
- Simulated mean background concentrations are highest in the Intermountain West (i.e., at high altitude) in spring and lowest in the Northeast during summer;
- Background concentrations tend to increase with total (i.e., base case) O₃ concentrations at high elevation; but that tendency is not as clear at low elevations.

The most pronounced differences between the Zhang et al. (2011) GEOS-Chem and the Emery et al. (2012) CAMx models—when compared with observations—are at the upper end of the concentration distribution. At high elevations, differences are likely to be the result of under-predictions of background contributions which are driven mainly by episodic events such as stratospheric intrusions and wildfires. In general, CAMx predicts higher values at the upper end of the concentration distribution than does GEOS-Chem. At low elevations (<1,500 meters)—located mainly in the East—the reasons for under-predictions at the upper end of the concentration distribution are more complex and likely involve extensive interactions between anthropogenic and natural sources.

Table 3-1 summarizes modeling results for seasonal mean MDA8 O₃ by region simulated by the two models. The regions in Table 3-1 are shown in Figure 3-50. As can be seen from the table, seasonal means predicted by GEOS-Chem are within a few parts per billion of measurements in both spring and summer for all regions shown except for California in the spring. Although CASTNET sites are meant to represent regional background air, they can be heavily influenced by polluted air masses, particularly in California where the under-predictions are largest. Seasonal means are simulated by CAMx to within 2-5 ppb except in California in the spring where they are under-predicted by 8 ppb and at sites in the Northeast and Southeast where they are over-predicted by 8-9 ppb in summer. When compared to observations, the mean R² within each region—except for California in the spring—is higher for CAMx than for GEOS-Chem suggesting better ability to track day-to-day variability by CAMx. It is clear from these results that model resolution (at least for the model resolutions considered here) is not the dominant factor determining agreement of the means between simulations or between simulations and measurements. Differences in model chemistry and physics must also be considered.

Table 3-2 summarizes modeling results for the annual 4th-highest (99th –percentile) MDA8 O₃ for the same seasons and regions used in Table 3-1. As can be seen in the table, the GEOS-Chem and the CAMx models both underestimate mean paired (day-specific) 4th-highest values in California by ~20 ppb. In general, CAMx simulates
paired annual 4th-highest MDA8 O₃ concentrations that are higher and in better agreement with measurements. Shown alongside the model estimates is the number of days the modeled MDA8 O₃ concentrations were within 5 ppb of observed concentrations. The unpaired entries for the models in Table 3-2 show model predicted 4th-highest MDA8 O₃ concentrations that are not calculated on the same day as the 4th-highest values observed at CASTNET sites. It can be seen that simulated regional means of the 4th-highest MDA8 O₃ are in better agreement with measurements when results are unpaired by date. In other words, the models do not necessarily predict the annual 4th-highest MDA8 O₃ concentrations on the same day as they are observed.

These results underscore the uncertainties inherent in any model’s attempts to simulate day specific 4th-highest O₃ concentrations. As noted earlier, uncertainties in calculating day specific O₃ concentrations are especially challenging because of the lack of day specific data for emissions of many species. While progress is being made in obtaining day specific data for lightning strikes and area burned in wildfires, the emission factors for precursors from these episodic sources such as lightning and wildfires are still uncertain. In addition to uncertainty in emissions, uncertainties in the treatment of transport and chemical mechanisms in the models must also be considered.
Table 3-1  Comparison of Zhang et al. (2011) and Emery et al. (2012) results for MDA8 O₃ concentrations (ppbv) with measurements at selected CASTNET sites.

<table>
<thead>
<tr>
<th>Region</th>
<th>Observation</th>
<th>Model</th>
<th>Mean MDA8 O₃ concentration (ppbv)ᵃ</th>
<th>Mean R²³⁻</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>Spring</td>
<td>Summer</td>
</tr>
<tr>
<td>California (5 sites)</td>
<td>CASTNET ⁶⁻</td>
<td>GEOS-Chem ⁴⁻</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>Base ⁵⁻</td>
<td>52 ± 11</td>
<td>66 ± 18</td>
</tr>
<tr>
<td></td>
<td></td>
<td>U.S. bkg ⁶⁻</td>
<td>38 ± 7</td>
<td>37 ± 9</td>
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<tr>
<td></td>
<td></td>
<td>NA bkg ⁷⁻</td>
<td>37 ± 6</td>
<td>35 ± 9</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMx ⁸⁻</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>Base ⁹⁻</td>
<td>50 ± 10</td>
<td>66 ± 13</td>
</tr>
<tr>
<td></td>
<td></td>
<td>NA bkg ¹⁰⁻</td>
<td>39 ± 6</td>
<td>42 ± 6</td>
</tr>
</tbody>
</table>

| West (14 sites) | CASTNET ⁶⁻ | GEOS-Chem ⁴⁻ |         |         |         |         |
|                 |             | Base ⁵⁻        | 53 ± 7  | 55 ± 11 | 0.30   | 0.12   |
|                 |             | U.S. bkg ⁶⁻    | 42 ± 6  | 40 ± 9  |         |         |
|                 |             | NA bkg ⁷⁻      | 41 ± 6  | 38 ± 9  |         |         |
|                 |             | CAMx ⁸⁻        |         |         |         |         |
|                 |             | Base ⁹⁻        | 49 ± 8  | 57 ± 10 | 0.39   | 0.33   |
|                 |             | NA bkg ¹⁰⁻     | 40 ± 7  | 41 ± 8  |         |         |

| North Central (6 sites) | CASTNET ⁶⁻ | GEOS-Chem ⁴⁻ |         |         |         |         |
|                        |             | Base ⁵⁻        | 47 ± 8  | 51 ± 14 | 0.52   | 0.44   |
|                        |             | U.S. bkg ⁶⁻    | 33 ± 6  | 27 ± 7  |         |         |
|                        |             | NA bkg ⁷⁻      | 30 ± 7  | 24 ± 7  |         |         |
|                        |             | CAMx ⁸⁻        |         |         |         |         |
|                        |             | Base ⁹⁻        | 45 ± 11 | 54 ± 13 | 0.63   | 0.48   |
|                        |             | NA bkg ¹⁰⁻     | 30 ± 6  | 31 ± 5  |         |         |

| Northeast (5 sites) | CASTNET ⁶⁻ | GEOS-Chem ⁴⁻ |         |         |         |         |
|                     |             | Base ⁵⁻        | 45 ± 7  | 45 ± 13 | 0.44   | 0.47   |
|                     |             | U.S. bkg ⁶⁻    | 33 ± 7  | 24 ± 7  |         |         |
|                     |             | NA bkg ⁷⁻      | 29 ± 6  | 18 ± 6  |         |         |
|                     |             | CAMx ⁸⁻        |         |         |         |         |
|                     |             | Base ⁹⁻        | 46 ± 11 | 53 ± 14 | 0.53   | 0.54   |
|                     |             | NA bkg ¹⁰⁻     | 30 ± 5  | 27 ± 6  |         |         |

| Southeast (9 sites) | CASTNET ⁶⁻ | GEOS-Chem ⁴⁻ |         |         |         |         |
|                    |             | Base ⁵⁻        | 51 ± 7  | 54 ± 9  | 0.42   | 0.21   |
|                    |             | U.S. bkg ⁶⁻    | 32 ± 7  | 29 ± 10 |         |         |
|                    |             | NA bkg ⁷⁻      | 29 ± 7  | 28 ± 9  |         |         |
|                    |             | CAMx ⁸⁻        |         |         |         |         |
|                    |             | Base ⁹⁻        | 54 ± 9  | 61 ± 12 | 0.56   | 0.45   |
|                    |             | NA bkg ¹⁰⁻     | 33 ± 6  | 30 ± 6  |         |         |

ᵃ Seasonal (spring, summer) mean MDA8 O₃ concentration ± standard deviation;
ᵇ Mean R² of all individual model-measurement pairs at individual CASTNET sites;
ᶜ Observed concentrations at CASTNET sites;
ᵈ Modeled concentrations at CASTNET sites;
ᵉ Modeled U.S. background concentrations at CASTNET sites,
⁰ Modeled NA background concentrations at CASTNET sites;
Source: Data from Zhang et al. (2011) for GEOS-Chem and Emery et al. (2012) for CAMx.
<table>
<thead>
<tr>
<th>Region</th>
<th>Observation</th>
<th>Model</th>
<th>4th-highest MDA8 O₃ concentration (ppbv)ᵃ</th>
<th>Number of Days within 5 ppbᵇ</th>
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</thead>
<tbody>
<tr>
<td>California (5 sites)</td>
<td>CASTNETᶜ</td>
<td>GEOS-Chem (paired)ᵈ</td>
<td>90 ± 13ᶜ</td>
<td>0</td>
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<td></td>
<td></td>
<td>GEOS-Chem (unpaired)ᵉ</td>
<td>85 ± 19</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (paired)⁰</td>
<td>71 ± 9</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (unpaired)ᵐ</td>
<td>85 ± 13</td>
<td></td>
</tr>
<tr>
<td>West (14 sites)</td>
<td>CASTNETᶜ</td>
<td>GEOS-Chem (paired)ᵈ</td>
<td>70 ± 4ᶜ</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>GEOS-Chem (unpaired)ᵉ</td>
<td>62 ± 8</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (paired)⁰</td>
<td>63 ± 8</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (unpaired)ᵐ</td>
<td>71 ± 7</td>
<td></td>
</tr>
<tr>
<td>North Central (6 sites)</td>
<td>CASTNETᶜ</td>
<td>GEOS-Chem (paired)ᵈ</td>
<td>71 ± 5ᶜ</td>
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<td></td>
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<td></td>
<td></td>
<td>CAMX (unpaired)ᵐ</td>
<td>73 ± 8</td>
<td></td>
</tr>
<tr>
<td>Northeast (5 sites)</td>
<td>CASTNETᶜ</td>
<td>GEOS-Chem (paired)ᵈ</td>
<td>71 ± 4ᶜ</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>GEOS-Chem (unpaired)ᵉ</td>
<td>61 ± 6</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (paired)⁰</td>
<td>72 ± 7</td>
<td>3</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (unpaired)ᵐ</td>
<td>75 ± 3</td>
<td></td>
</tr>
<tr>
<td>Southeast (9 sites)</td>
<td>CASTNETᶜ</td>
<td>GEOS-Chem (paired)ᵈ</td>
<td>76 ± 8ᶜ</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>GEOS-Chem (unpaired)ᵉ</td>
<td>61 ± 6</td>
<td>2</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (paired)⁰</td>
<td>71 ± 11</td>
<td>5</td>
</tr>
<tr>
<td></td>
<td></td>
<td>CAMX (unpaired)ᵐ</td>
<td>79 ± 9</td>
<td></td>
</tr>
</tbody>
</table>

ᵃAnnual 4th-highest (99th-percentile) MDA8 O₃ concentration regional means (ppb) ± standard deviation;
ᵇNumber of days the model predicted MDA8 O₃ concentrations were within 5 ppb of the observed 4th-highest concentrations;
ᶜObserved concentrations at CASTNET sites;
ᵈModeled concentrations at CASTNET sites on days when the 4th-highest MDA8 O₃ concentration was observed (paired by date);
ᵉModel predicted annual 4th-highest MDA8 O₃ concentration at CASTNET sites (unpaired by date).

Source: Data from Zhang et al. (2011) for GEOS-Chem and Emery et al. (2012) for CAMx.
Comparison of GEOS-Chem results for natural and NA background indicate that methane is also a major contributor to NA background $O_3$, accounting for slightly less than half of the increase in background since the pre-industrial era and whose relative contribution is projected to grow in the future. U.S. background concentrations are on average 2.6 ppb higher than NA background concentrations during spring and 2.7 ppb during summer across the United States. Highest values for U.S. background (in the U.S.) are found over the Northern Tier of New York State (19.1 ppb higher than local NA background concentrations) in summer. High values are also found in other areas bordering Canada and Mexico.

Analyses of results from GEOS-Chem and CAMx presented here and shown in Table 3-1 and Table 3-2 are in accord with results from Kasibhatla and Chameides (2000) who found that the accuracy of simulations improved as the averaging time of both the simulation and the observations increased (see Section 3.3). Note that any CTM—not just the ones considered here—will have difficulty in predicting day specific quantities. When analyzing results over long time periods (e.g., months), special care should be taken to examine temporal trends in bias because this will improve understanding of the modeling results.

Overall, these results suggest that GEOS-Chem is capable of simulating seasonal or monthly mean MDA8 $O_3$ to within a few parts per billion on a regional basis throughout the U.S., except in California. These results suggest that CAMx is capable of simulating seasonal or monthly mean MDA8 $O_3$ to within a few ppb, though, CAMx also shows relatively large disagreements in California and, in addition, shows relatively large positive bias in seasonal mean MDA8 $O_3$ in the eastern United States. However, differences between the models in the East are likely to narrow with updates to chemistry. Neither model is capable of simulating 4th-highest MDA8 $O_3$ to within suitable bounds on a day-specific basis at all sites, or even most sites. However, agreement between simulated versus observed 4th-highest MDA8 $O_3$ is improved for either model when the models and the measurements are sampled on different days.

Note that the calculations of background concentrations presented in this section were formulated to answer the question, “what would $O_3$ concentrations be if there were no anthropogenic sources.” This is different from asking, “how much of the $O_3$ measured or simulated in a given area is due to background contributions.” Because of potentially strong non-linearities (i.e., the fate, or lifetime, of the background $O_3$ transported into the urban area will depend on the concentration of the background $O_3$ in addition to interactions of background $O_3$ with the local chemical regime) in many urban areas, these estimates by themselves should not be used to answer the second question posed above. The extent of these non-linearities will generally depend on location and time, the strength of concentrated sources and the nature of the chemical regime.
3.5 Monitoring

3.5.1 Routine Monitoring Techniques

The federal reference method (FRM) for O<sub>3</sub> measurement is called the Chemiluminescence Method (CLM) and is based on the detection of chemiluminescence resulting from the reaction of O<sub>3</sub> with ethylene gas. The UV absorption photometric analyzers were approved as federal equivalent methods (FEMs) in 1977 and gained rapid acceptance for NAAQS compliance purposes due to ease of operation, relatively low cost, and reliability. The UV absorption method is based on the principle that O<sub>3</sub> molecules absorb UV radiation at a wavelength of 254 nm from a mercury lamp. The concentration of O<sub>3</sub> is computed from Beer’s law using the radiation absorbed across a fixed path length, the absorption coefficient, and the measured pressure and temperature in the detection cell. UV absorption photometry is the predominant method for assessing compliance with the NAAQS for O<sub>3</sub>. Almost all of the state and local air monitoring stations (SLAMS) that reported data to EPA AQS from 2005 to 2009 used UV absorption photometer FEMs. No CLM monitors, approved as FRMs or FEMs, reported O<sub>3</sub> data to AQS from 2005 to 2009 and only one monitor reported data using a long-path or open path Differential Optical Absorption Spectrometer (DOAS) FEM during this period.

The rationale, history, and calibration of O<sub>3</sub> measurements were summarized in the 1996 and 2006 O<sub>3</sub> AQCDs (U.S. EPA, 2006b, 1996a) and focused on the state of ambient O<sub>3</sub> measurements at that time as well as evaluation of interferences and new developments. This discussion will continue with the current state of O<sub>3</sub> measurements, interferences, and new developments for the period 2005 to 2010.

UV O<sub>3</sub> monitors use mercury lamps as the source of UV radiation and employ an O<sub>3</sub> scrubber (typically manganese dioxide) to generate an O<sub>3</sub>-free air flow to serve as a reference channel for O<sub>3</sub> measurements. There are known interferences with UV O<sub>3</sub> monitors. The 2006 O<sub>3</sub> AQCD (U.S. EPA, 2006b) reported on the investigation of the effects of water vapor, aromatic compounds, ambient particles, mercury vapor and alternative materials in the instrument’s O<sub>3</sub> scrubber. The overall conclusions from the review of the scientific literature covered in the 2006 O<sub>3</sub> AQCD (U.S. EPA, 2006b) are briefly summarized below.

Kleindienst et al. (1993) found water vapor to have no measurable impact and aromatic compounds to have a minor impact (as much as 3% higher than the FRM extrapolated to ambient conditions) on UV absorption measurements. UV O<sub>3</sub> monitor response evaluated by chamber testing using cigarette smoke, reported an elimination of the O<sub>3</sub> monitor response to the smoke when a particle filter was used that filtered out particles less than 0.2 μm in diameter (Arshinov et al., 2002). One study (Leston et al., 2005) in Mexico City compared a UV O<sub>3</sub> FEM to a CLM FRM. The UV FEM reported consistently higher O<sub>3</sub> than the CLM FRM. They suggested that O<sub>3</sub> measured in ambient air could be too high by 20 to 40 ppb under specific conditions due to positive interference by a number of organic compounds, mainly...
those produced during the oxidation of aromatic hydrocarbons and some primary compounds such as styrene and naphthalene. However, the concentrations of these compounds were many times higher in both of these environments than are typically found at ambient air monitoring sites in the United States. Although Hg is also potentially a strong interfering agent, because the Hg resonance line is used in this technique, its concentration would also have to be many times higher than is typically found in ambient air, e.g., as might be found in power plant plumes. Thus, it seems unlikely that such interferences would amount to more than one or two ppb (within the design specifications of the FEM), except under conditions conducive to producing high concentrations of the substances they identified as causing interference. Leston et al. (2005) also presented smog chamber data which demonstrated that heated metal and heated silver wool scrubbers perform better in the presence of aromatic hydrocarbon irradiations than manganese dioxide scrubbers when compared to the FRM. They also suggested the use of humidified calibration gas and alternative scrubber materials to improve UV O₃ measurements. Some O₃ monitor manufacturers now offer heated silver wool scrubbers as an alternative to manganese dioxide. Another possible solution to the O₃ scrubber problem may be the use of a gas phase scrubber such as NO. A commercial version of this has recently been introduced by 2B Technologies as an option on their model 202 FEM; however, it has not been field tested or approved for use as an FEM.

Review of the recent literature is summarized below. Study of UV monitors by Williams et al. (2006) concluded that well maintained monitors showed no substantial interferences when operated in locations with high concentrations of potentially interfering VOCs including Nashville, Houston, and the Gulf of Maine. Monitors were tested in urban and suburban environments, as well as on board a ship in both polluted and clean marine air. Comparisons of UV measurements to a non-FRM/FEM NO based CLM demonstrated agreement to within 1%. At the Houston location, they did observe a brief period on one day for about 30 minutes where the UV measurements exceeded the CLM by about 8 ppb (max). This was attributed to probable instrument malfunction.

Wilson and Birks (2006) investigated water vapor interference in O₃ measurements by four different UV monitors. In extreme cases where a rapid step change in relative humidity between 0 and 90% was presented, large transitory responses (tens to hundreds of ppb) were found for all monitors tested. Rapid changes in relative humidity such as this would not be expected during typical ambient O₃ measurements and could only be expected during measurement of vertical profiles from balloon or aircraft. The magnitude of the interference and the direction (positive or negative) was dependent on the manufacturer and model. Wilson and Birks (2006) also hypothesized that water vapor interference is caused by physical interactions of water vapor on the detection cell. The O₃ scrubber was also thought to act as a reservoir for water vapor and either added or removed water vapor from the air stream, subsequently affecting the detector signal and producing either a positive or negative response. They demonstrated that the use of a Nafion permeation membrane just before the O₃ detection cell to remove water vapor eliminated this interference.
Dunlea et al. (2006) evaluated multiple UV O₃ monitors with two different O₃ scrubber types (manganese dioxide and heated metal wool) in Mexico City. Large spikes in O₃ concentrations were observed while measuring diesel exhaust where large increases in particle number density were observed. The interference due to small particles passing through the Teflon filter and scattering/absorbing light in the detection cell were estimated to cause at most a 3% increase in measurements in typical ambient air environments. This estimate pertains to measurements in the immediate vicinity of fresh diesel emissions and most monitor siting guidelines would not place the monitor close to such sources, so actual interferences are expected to be much less than 3%. Dunlea et al. (2006) also observed no evidence for either a positive or negative interference or dependence due to variations in aromatics during their field study.

Li et al. (2006c) verified early reports of gas phase mercury interference with the UV O₃ measurement. They found that 300 ng/m³ of mercury produced an instrument response of about 35 ppb O₃. Background concentrations of mercury are around 1-2 ng/m³ and expected to produce an O₃ response that would be <1 ppb.

Spicer et al. (2010) examined potential UV O₃ monitor interferences by water vapor, mercury, aromatic compounds, and reaction products from smog chamber simulations. Laboratory tests showed little effect of changing humidity on conventional FEM UV O₃ monitors with manganese dioxide or heated metal wool scrubbers in the absence of other interferences. Mercury vapor testing produced an O₃ response by the UV monitors that was <1 ppb O₃ per 1 ppt (about 8 ng/m³) mercury vapor. Interference by aromatic compounds at low (3% RH) and high (80% RH) humidity showed some positive responses that varied by UV monitor and ranged from 0 to 2.2 ppb apparent O₃ response, per ppb of aromatic compound tested. The authors acknowledged that the aromatic compounds most likely to interfere are rarely measured in the atmosphere and therefore, make it difficult to assess the impact of these compounds during ambient air monitoring. Comparison of UV and CLM responses to photochemical reaction products in smog chamber simulations at 74% to 85% RH showed varied responses under low (0.125 ppmv/0.06 ppmv) to high (0.50 ppmv/0.19 ppmv) hydrocarbon/NOₓ conditions. The conventional UV monitors were as much as 2 ppb higher than the CLM under low hydrocarbon/NOₓ conditions and 6 ppb higher under the high hydrocarbon/NOₓ conditions. Two FEM UV monitors were also co-located at six sites in Houston from May to October, 2007 with one UV monitor equipped with Nafion permeation membrane. The average difference between 8-h daily max O₃ concentrations using the UV and the UV with Nafion permeation membrane ranged from -4.0 to 4.1 ppb.
### 3.5.2 Precision and Bias

In order to provide decision makers with an assessment of data quality, EPA’s Quality Assurance (QA) group derives estimates of both precision and bias for O₃ and the other gaseous criteria pollutants from the biweekly single point quality control (QC) checks using calibration gas, performed at each site by the monitoring agency. The single-point QC checks are typically performed at concentrations around 90 ppb. Annual summary reports of precision and bias can be obtained for each monitoring site at [http://www.epa.gov/ttn/amtic/qareport.html](http://www.epa.gov/ttn/amtic/qareport.html). The assessment of precision and bias are based on the percent-difference values, calculated from single-point QC checks. The percent difference is based on the difference between the pollutant concentration indicated by monitoring equipment and the known (actual) concentration of the standard used during the QC check. The monitor precision is estimated from the 90% upper confidence limit of the coefficient of variation (CV) of relative percent difference (RPD) values. The bias is estimated from the 95% upper confidence limit on the mean of the absolute values of percent differences. The data quality goal for O₃ precision and bias at the 90 and 95% upper confidence limits is 7% (40 CFR Part 58, Appendix A). Table 3-3 presents a summary of the number of monitors that meet the precision and bias goal of 7% for 2005 to 2009. Greater than 96% of O₃ monitors met the precision and bias goal between 2005 and 2009. Another way to look at the precision (CV) and bias (percent difference) information using the single-point QC check data from the monitoring network is to present box plots of the monitors’ individual precision and percent-difference data; Figure 3-17 and Figure 3-18 include this information for O₃ monitors operating from 2005 to 2009.

<table>
<thead>
<tr>
<th>Year</th>
<th>Number of Monitors</th>
<th>Monitors with Acceptable Precision (%)</th>
<th>Monitors with Acceptable Bias (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2005</td>
<td>879</td>
<td>96.5</td>
<td>96.7</td>
</tr>
<tr>
<td>2006</td>
<td>881</td>
<td>98.1</td>
<td>97.6</td>
</tr>
<tr>
<td>2007</td>
<td>935</td>
<td>98.1</td>
<td>98.1</td>
</tr>
<tr>
<td>2008</td>
<td>955</td>
<td>97.1</td>
<td>96.7</td>
</tr>
<tr>
<td>2009</td>
<td>958</td>
<td>97.4</td>
<td>97.5</td>
</tr>
</tbody>
</table>
Figure 3-17  Box plots of precision data by year (2005-2009) for all O₃ monitors reporting single-point QC check data to AQS.

Figure 3-18  Box plots of percent-difference data by year (2005-2009) for all O₃ monitors reporting single-point QC check data to AQS.
3.5.2.1 Precision from Co-located UV O₃ Monitors in Missouri

The Missouri Department of Natural Resources (MODNR) maintains a network of co-located UV O₃ analyzers. The MODNR provided co-located data from four monitors: two co-located at the same monitoring site in Kansas City (AQS ID 290370003) and two co-located at the same monitoring site in St. Louis (AQS ID 291831002). Hourly observations for the co-located measurements at these two sites between April and October, 2006-2009 were used to evaluate precision from co-located UV monitors. These data were then compared with the precision obtained by the biweekly single point QC checks for all sites reporting single-point QC check data to AQS between 2005 and 2009; the method normally used for assessing precision. Box plots of the RPD between the primary and co-located hourly O₃ measurements in Missouri are shown in Figure 3-19 and box plots of the RPD between the actual and indicated QC check for all U.S. sites are shown in Figure 3-20. As mentioned above, the average concentration of the single-point QC check is 90 ppb, whereas the average ambient O₃ concentration measured at the two sites in Missouri was 34 ppb. The mean RPD for the co-located monitors in Missouri and the single-point QC check data from all sites were less than 1 percent.

![Figure 3-19](image)

**Figure 3-19** Box plots of RPD data by year for the co-located O₃ monitors at two sites in Missouri from 2006-2009.
Figure 3-20  Box plots of RPD data by year for all of the United States. Ozone sites reporting single-point QC check data to AQS from 2005-2009.

3.5.3  Performance Specifications

The performance specifications for evaluating and approving new FEMs in accordance with 40 CFR Part 53 are provided in Table 3-4. These specifications were developed and originally published in the Federal Register in 1975. Modern, commercially-available instruments can now perform much better than the requirements specified below. For example, the lower detectable limit (LDL) performance specification is 10 ppb and the typical vendor-stated performance for the LDL is now less than 0.60 ppb. The amount of allowable interference equivalent for total interference substances is 60 ppb, and the current NAAQS for $\text{O}_3$ is 75 ppb, with an averaging time of 8 hours. Improvements in new measurement technology have occurred since these performance specifications were originally developed. These specifications should be revised to more accurately reflect the necessary performance requirements for $\text{O}_3$ monitors used to support the current NAAQS.
Table 3-4 Performance specifications for O₃ based in 40 CFR Part 53.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Specification</th>
</tr>
</thead>
<tbody>
<tr>
<td>Range</td>
<td>0 – 0.5 ppm (500 ppb)</td>
</tr>
<tr>
<td>Noise</td>
<td>0.005 ppm (5 ppb)</td>
</tr>
<tr>
<td>LDL – defined as two times the noise</td>
<td>0.01 ppm (10 ppb)</td>
</tr>
<tr>
<td>Interference equivalent</td>
<td></td>
</tr>
<tr>
<td>Each interfering substance</td>
<td>± 0.02 ppm (20 ppb)</td>
</tr>
<tr>
<td>Total interfering substances</td>
<td>0.06 ppm (60 ppb)</td>
</tr>
<tr>
<td>Zero drift</td>
<td></td>
</tr>
<tr>
<td>12 h</td>
<td>± 0.02 ppm (20 ppb)</td>
</tr>
<tr>
<td>24 h</td>
<td>± 0.02 ppm (20 ppb)</td>
</tr>
<tr>
<td>Span Drift, 24 h</td>
<td></td>
</tr>
<tr>
<td>20% of upper range limit</td>
<td>± 20.0%</td>
</tr>
<tr>
<td>80% of upper range limit</td>
<td>± 5.0%</td>
</tr>
<tr>
<td>Lag time</td>
<td>20 min</td>
</tr>
<tr>
<td>Rise time</td>
<td>15 min</td>
</tr>
<tr>
<td>Fall time</td>
<td>15 min</td>
</tr>
<tr>
<td>Precision</td>
<td></td>
</tr>
<tr>
<td>20% of upper range limit</td>
<td>0.01 ppm (10 ppb)</td>
</tr>
<tr>
<td>80% of upper range limit</td>
<td>0.01 ppm (10 ppb)</td>
</tr>
</tbody>
</table>

3.5.4 Monitor Calibration

The calibration of O₃ monitors was summarized in detail in the 1996 O₃ AQCD (U.S. EPA, 1996a). The calibration of O₃ monitors is done using an O₃ generator and UV photometers. UV photometry is the prescribed procedure for the calibration of reference methods to measure O₃ in the atmosphere. Because O₃ is unstable and cannot be stored, the O₃ calibration procedure specifically allows the use of transfer standards for calibrating ambient O₃ monitors. A transfer standard is calibrated against a standard of high authority and traceability and then moved to another location for calibration of O₃ monitors. The EPA and the National Institute of Standards and Technology (NIST) have established a network of standard reference photometers (SRPs) that are used to verify transfer standards. The International Bureau of Weights and Measures (BIPM) maintain one NIST SRP (SRP27) as the World’s O₃ reference standard. NIST maintains two SRPs (SRP0 and SRP2) that are used for comparability to ten other SRPs maintained by the EPA’s Regional QA staff.

SRPs have been compared to other reference standards. Tanimoto et al. (2006) compared NIST SRP35, owned by the National Institute for Environmental Studies
in Japan, to gas phase titration (GPT). The SRP was found to be 2% lower than GPT. GPT is no longer used as a primary or transfer standard in the United States. Viallon et al. (2006) compared SRP27 built at BIPM to four other NIST SRPs maintained by BIPM (SRP28, SRP31, SRP32, and SRP33). A minimum bias of +0.5% was found for all SRP measurement results, due to use of the direct cell length measurement for the optical path length; this bias was accounted for by applying the appropriate correction factor. Study of the bias-corrected SRPs showed systematic biases and measurement uncertainties for the BIPM SRPs. A bias of -0.4% in the instrument O₃ mole fraction measurement was identified and attributed to non-uniformity of the gas temperature in the instrument gas cells, which was compensated by a bias of +0.5% due to an under-evaluation of the UV light path length in the gas cells. The relative uncertainty of the O₃ absorption cross section was 2.1% at 253.65 nm and this was proposed as an internationally accepted consensus value until sufficient experimental data is available to assign a new value.

In November, 2010, the EPA revised the Technical Assistance Document for Transfer Standards for Calibration of Air Monitoring Analyzers for Ozone (U.S. EPA, 2010f) that was first finalized in 1979 (U.S. EPA, 1979b). The revision removed methods no longer in use and updated definitions and procedures where appropriate. In the revised document, the discussion of transfer standards for O₃ applies to the family of standards that are used beyond SRPs or Level 1 standards. To reduce confusion, EPA reduced the number of common terms that were used in the past such as: primary standard, local primary standard, transfer standard, and working standard. Beyond the SRPs, all other standards are considered transfer standards.

3.5.5 Other Monitoring Techniques

3.5.5.1 Portable UV O₃ Monitors

Small, lightweight, and portable UV O₃ monitors with low power consumption are commercially available. These monitors are based on the same principle of UV absorption by O₃ at 254 nm. Monitors of this type are typically used for vertical profiling using balloons, kites, or light aircraft where space and weight are limited. They have also been used for monitoring at remote locations such as National Parks. Burley and Ray (2007) compared portable O₃ monitor measurements to those from a conventional UV monitor in Yosemite National Park. Calibrations of the portable O₃ monitors against a transfer standard resulted in an overall precision of ± 4 ppb and accuracy of ± 6%. Field measurement comparisons between the portable and conventional monitor at Turtleback Dome showed the portable monitor to be 3.4 ppb lower on average, with daytime deviation typically on the order of 0-3 ppb. Agreement between the portable and conventional monitor during daylight hours (9:00 a.m. to 5:00 p.m. PST) resulted in an R² of 0.95, slope of 0.95, and intercept of 0.36 ppb. Substantial deviations were observed in the predawn hours where the
portable monitor was consistently low. These deviations were attributed to the difference in sampling inlet location. The portable monitor was located at 1.3 meters above ground and the conventional monitor was located at 10 meters above ground. Agreement between the portable and conventional monitors for all hours sampled resulted in an $R^2$ of 0.88, slope of 1.06, and intercept of -6.8 ppb. (Greenberg et al., 2009) also compared a portable UV O$_3$ monitor to a conventional UV monitor in Mexico City and obtained good agreement for a 14 day period with an $R^2$ of 0.97, slope of 0.97, and intercept of 6 ppb. One portable O$_3$ monitor was recently approved as an FEM (EQOA-0410-190) on April 27, 2010 (75 FR 22126).

3.5.5.2  NO-based Chemiluminescence Monitors

One commercially available NO-based chemiluminescence monitor has been approved as an FEM (EQOA-0611-199) on October 7, 2011 (75 FR 62402). It may also be designated as a second or replacement FRM since the ethene based FRMs are no longer manufactured. Although this is a relatively new monitor, other NO-based CLM instruments have been custom built for various field studies since the early 1970s. A commercial version that measured both O$_3$ and NO$_X$ was offered in the early 1970s but failed to gain commercial acceptance. Initial testing with SO$_2$, NO$_2$, Cl$_2$, C$_2$H$_2$, C$_2$H$_4$ and C$_3$H$_6$ (Stedman et al., 1972) failed to identify any interferences. In the intervening years, custom built versions have not been found to have any interference; however, they do experience a slight decrease in response with increasing relative humidity (due to quenching of the excited species by the water molecules). The new NO-based CLM solves this problem with the use of a Nafion membrane dryer. A custom built NO-based CLM similar to the FEM was used by Williams et al. (2006) in Houston, TX; Nashville, TN; and aboard ship along the New England coast. It was found to be in good agreement with a standard UV based FEM and with a custom built DOAS.

3.5.5.3  Passive Air Sampling Devices and Sensors

A passive O$_3$ sampling device depends on the diffusion of O$_3$ in air to a collecting or indicating medium. In general, passive samplers are not adequate for compliance monitoring because of the limitations in averaging time (typically one week or more), particularly for O$_3$. However, these devices are valuable for personal human exposure estimates and for obtaining long-term data in rural areas where conventional UV monitors are not practical or feasible to deploy. The 1996 O$_3$ AQCD (U.S. EPA, 1996a) provided a detailed discussion of passive samplers, along with the limitations and uncertainties of the samplers evaluated and published in the literature from 1989 to 1995. The 2006 O$_3$ AQCD (U.S. EPA, 2006b) provided a brief update on available passive samplers developed for use in direct measurements of personal exposure published through 2004. The 2006 O$_3$ AQCD (U.S. EPA, 2006b) also noted the sensitivity of these samplers to wind velocity, badge
placement, and interference by other copollutants that may result in measurement error.

Subsequent evaluations of passive diffusion samplers in Europe showed good correlation when compared to conventional UV O₃ monitors, but a tendency for the diffusion samplers to overestimate the O₃ concentration (Gottardini et al., 2010; Vardoulakis et al., 2009; Buzica et al., 2008). The bias of O₃ diffusion tubes were also found to vary with concentration, season, and exposure duration (Vardoulakis et al., 2009). Development of simple, inexpensive, passive O₃ measurement devices that rely on O₃ detection papers and a variety of sensors with increased time resolution (sampling for hours instead of weeks) and improved sensitivity have been reported (Maruo et al., 2010; Ebeling et al., 2009; Miwa et al., 2009; Ohira et al., 2009; Maruo, 2007; Utembe et al., 2006). Limitations for some of these sensors and detection papers include air flow dependence and relative humidity interference.

### 3.5.5.4 Differential Optical Absorption Spectrometry

Optical remote sensing methods can provide direct, sensitive, and specific measurements of O₃ over a broad area or open path in contrast with conventional single-point UV monitors. The 1996 O₃ AQCD (U.S. EPA, 1996a) provided a brief discussion of DOAS for O₃ measurements and cited references to document the sensitivity (1.5 ppb for a 1-minute averaging time), correlation (r = 0.89), and agreement (on the order of 10%) with UV O₃ monitors (Stevens et al., 1993). The 2006 O₃ AQCD (U.S. EPA, 2006b) provided an update on DOAS where a positive interference due to an unidentified absorber was noted (Reisinger, 2000). More recent study of the accuracy of UV absorbance monitors by Williams et al. (2006) compared UV and DOAS measurements at two urban locations. In order to compare the open path measurements and UV, the data sets were averaged to 30-minute periods and only data when the boundary layer was expected to be well mixed (between 10:00 a.m. and 6:00 p.m. CST) were evaluated. The comparisons showed variations of no more than ± 7% (based on the slope of the linear least squares regression over a concentration range from about 20 to 200 ppb) and good correlation (R² = 0.96 and 0.98). Lee et al. (2008b) evaluated DOAS and UV O₃ measurements in Korea and found the average DOAS concentration to be 8.6% lower than the UV point measurements with a good correlation (R² = 0.94).

DOAS has also been used for the measurement of HNO₂ (or HONO). DOAS was compared to chemical point-measurement methods for HONO. Acker et al. (2006) obtained good results when comparing wet chemical and DOAS during well mixed atmospheric conditions (wet chemical = 0.009 + 0.92 × DOAS; r = 0.7). Kleffmann and Wiesen (2008) noted that interferences with the HONO wet chemical methods can affect results from inter-comparison studies if not addressed. In an earlier study, Kleffmann et al. (2006) demonstrated that when the interferences were addressed, excellent agreement with DOAS can be obtained. Stutz et al. (2010) found good agreement (15% or better) between DOAS and a wet chemical method (Mist
DOAS remains attractive due to its sensitivity, speed of response, and ability to simultaneously measure multiple pollutants; however, further inter-comparisons and interference testing are recommended.

### 3.5.5.5 Satellite Remote Sensing

Satellite observations for O₃ are growing as a resource for many purposes, including model evaluation, assessing emissions reductions, pollutant transport, and air quality management. Satellite remote sensing instruments do not directly measure the composition of the atmosphere. Satellite retrievals are conducted using the solar backscatter or thermal infrared emission spectra and a variety of algorithms. Most satellite measurement systems have been developed for stratospheric measurement of the total O₃ column. Mathematical techniques have been developed and must be applied to derive information from these systems about tropospheric O₃ (Tarasick and Slater, 2008; Ziemke et al., 2006). Direct retrieval of global tropospheric O₃ distributions from solar backscattered UV spectra have been reported from OMI and the Global Ozone Monitoring Experiment (GOME) (Liu et al., 2006). Another satellite measurement system, Tropospheric Emission Spectrometer (TES), produces global-scale vertical concentration profiles of tropospheric O₃ from measurements of thermal infrared emissions. TES has been designed specifically to focus on mapping the global distribution of tropospheric O₃ extending from the surface to about 10-15 km altitude (Beer, 2006). Satellite measurements of tropospheric O₃ generally require monthly averaging to reduce noise, and thus are of little use for observing synoptic-scale variability.

In order to improve the understanding of the quality and reliability of the data, satellite-based observations of total column and tropospheric O₃ have been validated in several studies using a variety of techniques, such as aircraft observations, ozonesondes, CTMs, and ground-based spectroradiometers. Antón et al. (2009) compared satellite data from two different algorithms (OMI-DOAS and OMI-TOMS) with total column O₃ data from ground-based spectroradiometers at five locations. The satellite total column O₃ data underestimated ground-based measurements by less than 3%. Richards et al. (2008) compared TES tropospheric O₃ profiles using airborne differential absorption lidar (DIAL) and found TES to have a 7 ppbv positive bias relative to DIAL throughout the troposphere. Nassar et al. (2008) compared TES O₃ profiles and ozonesonde coincidences and found a positive bias of 3-10 ppbv for TES. Worden et al. (2007a) also compared TES with ozonesondes and found TES O₃ profiles to be biased high in the upper troposphere (average bias of 16.8 ppbv for mid-latitudes and 9.8 ppbv for the tropics) and biased low in the lower troposphere (average bias of -2.6 ppbv for mid-latitudes and -7.4 ppbv for the tropics). Comparisons of TES and OMI with ozonesondes by Zhang et al. (2010b) showed a mean positive bias of 5.3 ppbv (10%) for TES and 2.8 ppbv (5%) for OMI at 500 hPa. In addition, Zhang et al. (2010b) used a CTM (GEOS-
Chem) to determine global differences between TES and OMI. They found differences between TES and OMI were generally ± 10 ppbv except at northern mid-latitudes in summer and over tropical continents. Satellite observations have also been combined (e.g., OMI and TES) to improve estimates of tropospheric O$_3$ (Worden et al., 2007b).

Satellite measurements are also available for O$_3$ precursors such as CO, NO$_2$, and HCHO. These measurements are useful for constraining model estimates of O$_3$ precursor emissions and long-range transport of pollution (Section 3.4). Zhang et al. (2008) used satellite measurements of CO and NO$_2$ along with O$_3$ to constrain estimates of background O$_3$ precursors in Asia and O$_3$ produced during long range transport.

### 3.5.6 Ambient O$_3$ Network Design

#### 3.5.6.1 Monitor Siting Requirements

To monitor compliance with the NAAQS, state and local monitoring agencies operate O$_3$ monitoring sites at various locations depending on the area size (population and geographic characteristics$^1$) and typical peak concentrations (expressed in percentages below, or near the O$_3$ NAAQS). SLAMS make up the ambient air quality monitoring sites that are primarily needed for NAAQS comparisons, but may also serve some other basic monitoring objectives that include: providing air pollution data to the general public in a timely manner; emissions strategy development; and support for air pollution research. SLAMS include National Core (NCore), Photochemical Assessment Monitoring Stations (PAMS), and all other State or locally-operated stations except for the monitors designated as special purpose monitors (SPMs).

The SLAMS minimum monitoring requirements to meet the O$_3$ design criteria are specified in 40 CFR Part 58, Appendix D. Although NCore and PAMS are a subset of SLAMS, the monitoring requirements for those networks are separate and discussed below. The minimum number of O$_3$ monitors required in a Metropolitan Statistical Area (MSA) ranges from zero for areas with a population of at least 50,000 and under 350,000 with no recent history of an O$_3$ design value$^2$ greater than 85 percent of the NAAQS, to four for areas with a population greater than 10 million and an O$_3$ design value greater than 85 percent of the NAAQS. Within an O$_3$ network, at least one site for each MSA, or Combined Statistical Area (CSA) if multiple MSAs are involved, must be designed to record the maximum concentration for that particular metropolitan area. More than one maximum concentration site may

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$^1$ Geographic characteristics such as complexity of terrain, topography, land use, etc.

$^2$ A design value is a statistic that describes the air quality status of a given area relative to the level of the NAAQS. Design values are typically used to classify nonattainment areas, assess progress toward meeting the NAAQS, and develop control strategies. See [http://epa.gov/airtrends/values.html](http://epa.gov/airtrends/values.html) (U.S. EPA, 2010a) for guidance on how these values are defined.
be necessary in some areas. The spatial scales for O₃ sites are neighborhood, urban and regional.

- Neighborhood scale: represents concentrations within some extended area of the city that has relatively uniform land use with dimensions in the 0.5-4.0 km range. The neighborhood and urban scales listed below have the potential to overlap in applications that concern secondary or homogeneously distributed primary air pollutants.
- Urban scale: represents concentrations within an area of city-like dimensions, on the order of 4-50 km. Within a city, the geographic placement of sources may result in there being no single site that can be said to represent air quality on an urban scale.
- Regional scale: usually defines a rural area of reasonably homogeneous geography without large sources, and extends from tens to hundreds of kilometers.

Since O₃ concentrations decrease appreciably in the colder parts of the year in many areas, O₃ is required to be monitored at SLAMS monitoring sites only during the “ozone season.” Table D-3 of 40 CFR Part 58, Appendix D lists the beginning and ending month of the ozone season for each U.S. state or territory. Most operate O₃ monitors only during the ozone season. Those that operate some or all of their O₃ monitors on a year-round basis include Arizona, California, Hawaii, Louisiana, Nevada, New Mexico, Puerto Rico, Texas, American Samoa, Guam and the Virgin Islands.

The total number of SLAMS O₃ sites needed to support the basic monitoring objectives includes more sites than the minimum numbers required in 40 CFR Part 58, Appendix D. In 2010, there were 1250 O₃ monitoring sites reporting values to the EPA AQS database (Figure 3-21). Monitoring site information for EPA’s air quality monitoring networks is available in spreadsheet format (CSV) and keyhole markup language format (KML or KMZ) that is compatible with Google Earth™ and other software applications on the AirExplorer website (U.S. EPA, 2011e). States may operate O₃ monitors in non-urban or rural areas to meet other objectives (e.g., support for research studies of atmospheric chemistry or ecosystem impacts). These monitors are often identified as SPMs and can be operated up to 24 months without being considered in NAAQS compliance determinations. The current monitor and probe siting requirements have an urban focus and do not address the siting for SPMs or monitors in non-urban, rural areas to support ecosystem impacts and the secondary standards.
NCore is a new multipollutant monitoring network implemented to meet multiple monitoring objectives. Those objectives include: timely reporting of data to the public through AirNow (U.S. EPA, 2011a); support for the development of emission reduction strategies; tracking long-term trends of criteria pollutants and precursors; support to ongoing reviews of the NAAQS and NAAQS compliance; model evaluation; support for scientific research studies; and support for ecosystem assessments. Each state is required to operate at least one NCore site. The NCore monitoring network began January 1, 2011 at about 80 stations (about 60 urban and 20 rural sites). NCore has leveraged the use of sites in existing networks; for example, some IMPROVE sites also serve as rural NCore sites. In addition to O₃, other components including CO, NOₓ, NOₓ, SO₂, and basic meteorology are also measured at NCore sites. The spatial scale for urban NCore stations is urban or neighborhood; however, a middle-scale¹ site may be acceptable in cases where the site can represent many such locations throughout a metropolitan area. Rural NCore sites are located at a regional or larger scale, away from any large local emission sources so that they represent ambient concentrations over an extensive area. Ozone monitors at NCore sites are operated year round.

¹ Middle scale defines an area up to several city blocks in size with dimensions ranging from about 100 to 500 meters.
PAMS provides more comprehensive data on O₃ in areas classified as serious, severe, or extreme nonattainment for O₃. In addition to O₃, PAMS provides data for NOₓ, NOₓ, VOCs, carbonyls, and meteorology. The PAMS network design criteria are based on locations relative to O₃ precursor source areas and predominant wind directions associated with high O₃ concentrations. The overall network design is location specific and geared toward enabling characterization of precursor emission sources in the area, O₃ transport, and photochemical processes related to O₃ nonattainment. Minimum monitoring for O₃ and its precursors is required annually during the months of June, July, and August when peak O₃ concentrations are expected. In 2006, the EPA reduced the minimum PAMS monitoring requirements (71 FR 61236). There were a total of 92 PAMS sites reporting values to the AQS data base in 2010.

CASTNET is a regional monitoring network established to assess trends in acidic deposition due to emission reduction regulations. CASTNET also provides concentration measurements of air pollutants involved in acidic deposition, such as sulfate and nitrate, in addition to the measurement of O₃. CASTNET O₃ monitors operate year round and are primarily located in rural areas. In 2010, there were 80 CASTNET sites located in, or near, rural areas. As part of CASTNET, the National Park Service (NPS) operates 23 sites located in national parks and other Class-I areas. Ozone data collected at the 23 NPS sites is compliant with the SLAMS QA requirements in 40 CFR Part 58, Appendix A. Ozone measurements at the remaining CASTNET sites were not collected with the QA requirements for SLAMS outlined in 40 CFR Part 58, Appendix A, and therefore, these O₃ data cannot be used for NAAQS compliance purposes. The SLAMS QA requirements and procedures are currently being implemented at the remaining sites.

The NPS also operates a Portable Ozone Monitoring Systems (POMS) network. The POMS couples the small, low-power O₃ monitor with a data logger, meteorological measurements, and solar power in a self contained system for monitoring in remote locations. Typical uses for the POMS data include research projects, survey monitoring, and assessments of spatial O₃ distribution. The portable O₃ monitor in use by the NPS was recently designated as an equivalent method for O₃ (75 FR 22126). Seventeen NPS POMS monitors were operating in 2010 (NPS, 2011). A map of the rural NCore sites, along with the CASTNET, and the NPS POMS sites are shown in Figure 3-22. As can be seen from Figure 3-21 and Figure 3-22, vast rural areas of the country still exist without any monitor coverage. Monitoring opportunities exist in these areas where relatively few and easily characterized precursor sources dominate and could be used to improve understanding of O₃ formation.
3.5.6.2 Probe/Inlet Siting Requirements

Probe and monitoring path siting criteria for ambient air quality monitoring are contained in 40 CFR Part 58, Appendix E. For O\textsubscript{3}, the probe must be located between 2 and 15 meters above ground level and be at least 1 meter away (both in the horizontal and vertical directions) from any supporting structure, walls, etc. If it is located on the side of a building, it must be located on the windward side, relative to prevailing wind direction during the season of highest potential O\textsubscript{3} concentration. Ozone monitors are placed to determine air quality in larger areas (neighborhood, urban, or regional scales) and therefore, placement of the monitor probe should not be near local, minor sources of NO, O\textsubscript{3}-scavenging hydrocarbons, or O\textsubscript{3} precursors. The probe or inlet must have unrestricted air flow in an arc of at least 180 degrees and be located away from any building or obstacle at a distance of at least twice the height of the obstacle. The arc of unrestricted air flow must include the predominant wind direction for the season of greatest O\textsubscript{3} concentrations. Some exceptions can be made for measurements taken in street canyons or sites where obstruction by buildings or other structures is unavoidable. The scavenging effect of trees on O\textsubscript{3} is greater than other pollutants and the probe/inlet must be located at least 10 meters from the tree drip line to minimize interference with normal air flow. When siting O\textsubscript{3} monitors near roadways, it is important to minimize the destructive interferences.
from sources of NO, since NO reacts readily with O₃. For siting neighborhood and urban scale O₃ monitors, guidance on the minimum distance from the edge of the nearest traffic lane is based on roadway average daily traffic count (40 CFR Part 58, Appendix E, Table E-1). The minimum distance from roadways is 10 meters (average daily traffic count ≤ 1,000) and increases to a maximum distance of 250 meters (average daily traffic count ≥ 110,000).

3.6 Ambient Concentrations

This section investigates spatiotemporal variability in ambient O₃ concentrations and associations between O₃ and copollutants. To set the stage for the rest of the section, common O₃ measurement units, metrics, and averaging times are described and compared in Section 3.6.1. Spatial variability is covered in Section 3.6.2 and is divided into urban-focused variability and rural-focused variability. Urban-focused variability is organized by scale, extending from national-scale down to neighborhood-scale and the near-road environment. Rural-focused variability is organized by region and includes observations of ground-level vertical O₃ gradients where available. Temporal variability is covered in Section 3.6.3 and is organized by time, extending from multiyear trends down to hourly (diel) variability. In many instances, spatial and temporal variability are inseparable (e.g., seasonal dependence to spatial variability), resulting in some overlap between Section 3.6.2 and Section 3.6.3. Finally, Section 3.6.4 covers associations between O₃ and copollutants including CO, SO₂, NO₂, PM₂.₅, and PM₁₀.

As noted in the 2006 O₃ AQCD (U.S. EPA, 2006b), O₃ is the only photochemical oxidant other than nitrogen dioxide (NO₂) that is routinely monitored and for which a comprehensive database exists. Data for other photochemical oxidants (e.g., PAN, H₂O₂, etc.) typically have been obtained only as part of special field studies. Consequently, no data on nationwide patterns of occurrence are available for these other oxidants; nor are extensive data available on the relationships of concentrations and patterns of these oxidants to those of O₃. As a result, this section focuses solely on O₃, the NAAQS indicator for photochemical oxidants. The majority of ambient O₃ data reported in this section were obtained from AQS, EPA’s repository for detailed, hourly data that has been subject to EPA quality control and assurance procedures (the AQS network was described in Section 3.5).

3.6.1 Measurement Units, Metrics, and Averaging Times

Several approaches are commonly used for reporting O₃ data. In atmospheric sciences and epidemiology, O₃ is frequently reported as a concentration, expressed as a volume-to-volume mixing ratio, commonly measured in ppm or ppb. In human exposure, O₃ is frequently reported as a cumulative exposure, expressed as a mixing ratio times time (e.g., ppm-h). In ecology, cumulative exposure indicators are frequently used that extend over longer time periods, such as growing season or year.
This section focuses on ambient concentrations derived primarily from hourly average O₃ measurements and concentrations are reported in ppb wherever possible. Further details on human and ecological exposure metrics can be found in Chapter 4 and Chapter 9, respectively.

As discussed in Section 3.5, most continuous O₃ monitors report hourly average concentrations to AQS with a required precision of 10 ppb and LDL of 10 ppb (see Table 3-4). This data can be used as reported (1-h avg), or further summarized in one of several ways to focus on important aspects of the data while simultaneously reducing the volume of information. Three common daily reporting metrics include: (1) the average of the hourly observations over a 24-hour period (24-h avg); (2) the maximum hourly observation occurring in a 24-hour period (1-h daily max); and (3) the maximum 8-h running average of the hourly observations occurring in a 24-hour period (8-h daily max). Throughout this ISA and the literature, O₃ concentrations are reported using different averaging times as appropriate, making it important to recognize the differences between these metrics.

Nation-wide, year-round 1-h avg O₃ data reported to AQS from 2007-2009 was used to compare these different daily metrics. Correlations between the 24-h avg, 1-h daily max and 8-h daily max metrics were generated on a site-by-site basis. Figure 3-23 contains box plots of the distribution in correlations from all sites. The top comparison in Figure 3-23 is between 8-h daily max and 1-h daily max O₃. Not surprisingly, these two metrics are very highly correlated (median r = 0.97, IQR = 0.96-0.98). There are a couple outlying sites, with correlations between these two metrics as low as 0.63, but 95% of sites have correlations above 0.93.

The middle comparison in Figure 3-23 is between 8-h daily max and 24-h avg O₃. For these metrics, the distribution in correlations is shifted down and broadened out (median r = 0.89, IQR = 0.86-0.92). Finally, the bottom comparison in Figure 3-23 is between 1-h daily max and 24-h avg O₃. Again, for these metrics the distribution in correlations is shifted down and broadened out relative to the other two comparisons (median r = 0.83, IQR = 0.78-0.88). The correlation between the two daily maximum metrics (1-h daily max and 8-h daily max) are quite high for most sites, but correlations between the daily maximum metrics and the daily average metric (24-h avg) are lower. This illustrates the influence of the overnight period on the 24-h avg O₃ concentration. In contrast, the 1-h daily max and 8-h daily max are more indicative of the daytime, higher O₃ periods. The correlation between these metrics, however, can be very site-specific, as is evident from the broad range in correlations in Figure 3-23 for all three comparisons. Therefore, understanding which O₃ metric is being used in a given study is very important since they capture different aspects of O₃ temporal variability.

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For O₃ regulatory monitoring purposes, the 8-h daily max is calculated by first generating all 8-h running averages and storing these averages hourly by the first hour in the 8-hour period. The 8-h daily max is then set equal to the maximum of the 24 individual 8-h averages occurring in a given day.
The median 1-h daily max, 8-h daily max, and 24-h avg O₃ concentrations across all sites included in the 3-year nation-wide data set were 44, 40, and 29 ppb, respectively. Representing the upper end of the distribution, the 99th percentiles of these same metrics across all sites were 94, 80, and 60 ppb, respectively. While the ratio of these metrics will vary by location, typically the 1-h daily max will be the highest value representing peak concentrations and the 24-h avg will be considerably lower representing daily average concentrations incorporating the overnight period. The 8-h daily max typically represents the higher mid-day concentrations and will generally lie somewhere between the other two metrics.

3.6.2 Spatial Variability

3.6.2.1 Urban-Focused Variability

National-Scale Variability

AQS contains a large depository of national O₃ data collected to meet the monitoring objectives described in Section 3.5.6. In many areas, O₃ concentrations decrease appreciably during months with lower temperatures and decreased sunlight. As a
result, year-round O₃ monitoring is only required in certain areas. Table D-3 of 40 CFR Part 58, Appendix D lists the beginning and ending month of the ozone season (defined in Section 3.5.6.1) by geographic area and Figure 3-24 illustrates these time periods on a monitor-by-monitor basis. Monitoring is optional outside the ozone season and many states elect to operate their monitors year-round or for time periods outside what is strictly mandated.


**Figure 3-24** Required O₃ monitoring time periods (ozone season) identified by monitoring site.
Hourly FRM and FEM O₃ data reported to AQS for the period 2007-2009 were used to investigate national-scale spatial variability in O₃ concentrations. Given the variability in O₃ monitoring time periods available in AQS as a result of the regionally-varying ozone seasons, the analyses in this section were based on two distinct data sets:

- a **year-round** data set: data only from monitors reporting year-round;
- a **warm-season** data set: data from all monitors reporting May through September.

The warm-season data set was used to capture the majority of ozone season data while providing a consistent time-frame for comparison across states. All available monitoring data including data from year-round monitors was included in the warm-season data set after removing observations outside the 5-month window. Data were retrieved from AQS on February 25, 2011 for these two data sets, and all validated data was included regardless of flags or regional concurrence¹. A summary of the two O₃ data sets including the applied completeness criteria is provided in Table 3-5. Figure 3-25 and Figure 3-26 show the location of the 457 year-round and 1,064 warm-season monitors meeting the completeness criteria for all three years (2007-2009).

<table>
<thead>
<tr>
<th>Table 3-5</th>
<th>Summary of O₃ data sets originating from AQS.</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Year-Round Data Set</td>
</tr>
<tr>
<td>Months</td>
<td>January - December (12 mo)</td>
</tr>
<tr>
<td>Completeness Criteria</td>
<td>75% of hours in a day</td>
</tr>
<tr>
<td></td>
<td>75% of days in a calendar quarter</td>
</tr>
<tr>
<td></td>
<td>All 4 quarters per year</td>
</tr>
<tr>
<td>Number of monitors meeting completeness criteria</td>
<td>618 containing at least one valid year in 2007-2009</td>
</tr>
<tr>
<td></td>
<td>550 containing at least two valid years in 2007-2009</td>
</tr>
<tr>
<td></td>
<td>457 containing all three valid years in 2007-2009</td>
</tr>
</tbody>
</table>

¹ Concentrations that might have been affected by exceptional events (and contribute to a violation of the NAAQS) can be flagged in the Air Quality System (AQS) by the reporting organization. Exceptional events are defined as unusual or naturally occurring events that can affect air quality but are not reasonably controllable using techniques that tribal, state or local air agencies may implement in order to attain and maintain the National Ambient Air Quality Standards (NAAQS). The corresponding EPA Regional Office is responsible for reviewing the data and evidence of the event, and deciding whether to concur with the flag. Flagged data that has been concurred by the Regional office is typically excluded for regulatory purposes.
Figure 3-25  Location of the 457 O$_3$ monitors meeting the year-round data set completeness criterion for all 3 years between 2007 and 2009.

Figure 3-26  Location of the 1,064 O$_3$ monitors meeting the warm-season data set completeness criteria for all 3 years between 2007 and 2009.
Tabulated statistics generated from the year-round and warm-season data sets are included in Table 3-6 and Table 3-7, respectively. This information was used to compare (1) the year-round and warm-season data sets; (2) the $O_3$ distribution variability across years (2005-2009); and (3) four different averaging times (1-h avg, 24-h avg, 1-h daily max, and 8-h daily max). Summary statistics for 2005 and 2006 were added to these tables in order to gain a broader view of year-to-year variability, but the year-round and warm-season data sets used for analyses in the rest of this section are limited to 2007-2009 as described above and in Table 3-5. The 8-h daily max pooled by site was also included in these tables to show the distribution of the annual and 3-year (2007-2009) site-averages of the 8-h daily max statistic.

The year-round data set includes data from roughly half the number of monitors as the warm-season data set and a larger fraction of the year-round monitors are located in the southern half of the U.S. due to extended monitoring requirements in these areas. Despite these differences, the mean, SD and percentiles of the nation-wide $O_3$ concentrations were quite similar for the year-round data presented in Table 3-6 and the warm-season data presented in Table 3-7. In both data sets, there was very little variability across years in the central statistics; for example, the median 1-h avg concentrations between 2005 and 2009 ranged from 28 to 29 ppb for the year-round data and from 29 to 30 ppb for the warm-season data. The 8-h daily max showed similar uniformity in median across the five years, with concentrations ranging from 39 to 41 ppb for the year-round data and from 40 to 43 for the warm-season data. The upper percentiles (95th and above) showed a general downward trend from 2005 to 2009 in both nation-wide data sets. For example, the 99th percentile of the 8-h daily max observed in the warm-season data dropped from 85 ppb in 2005 to 75 ppb in 2009. Trends in $O_3$ concentrations investigated over a longer time period are included in Section 3.6.3.1.
Table 3-6 Nationwide distributions of O₃ concentrations (ppb) from the year-round data set.

| Time Period | N Monitors | N Obs | Mean | SD | Min | 1 | 5 | 10 | 25 | 50 | 75 | 90 | 95 | 98 | 99 | Max | Site ID<sup>b</sup> |
|-------------|------------|-------|------|----|-----|---|---|---|---|---|---|---|---|---|---|---|---|-----------------|
| **1-h avg** |            |       |      |    |     |   |   |   |   |   |   |   |   |   |   |   |   |                 |
| 2005        | 499        | 4,284,219 | 29  | 18 | 2   | 2 | 2 | 15 | 28 | 41 | 53 | 61 | 71 | 78 | 182 | 78 | 610710005      |
| 2006        | 532        | 4,543,205 | 30  | 18 | 2   | 2 | 2 | 15 | 26 | 42 | 54 | 61 | 71 | 78 | 175 | 72 | 60370016       |
| 2007        | 522        | 4,547,280 | 29  | 18 | 2   | 2 | 2 | 16 | 29 | 41 | 52 | 60 | 68 | 75 | 237 | 72 | 45079021       |
| 2008        | 520        | 4,470,065 | 30  | 17 | 2   | 2 | 2 | 17 | 29 | 41 | 52 | 59 | 67 | 74 | 222 | 72 | 450210002      |
| 2009        | 551        | 4,716,962 | 29  | 16 | 2   | 2 | 2 | 16 | 17 | 29 | 40 | 50 | 56 | 64 | 70 | 188 | 72 | 720770001      |
| 2007-2009   | 599        | 13,734,307 | 29  | 17 | 2   | 2 | 2 | 16 | 17 | 29 | 40 | 51 | 58 | 67 | 73 | 237 | 72 | 45079021       |
| **24-h avg** |            |       |      |    |     |   |   |   |   |   |   |   |   |   |   |   |   |   |                 |
| 2005        | 504        | 183,815  | 29  | 13 | 2   | 4 | 9 | 13 | 20 | 28 | 37 | 46 | 51 | 57 | 61 | 103 | 60179002       |
| 2006        | 536        | 194,884  | 30  | 13 | 2   | 5 | 10| 14 | 21 | 29 | 38 | 47 | 52 | 58 | 62 | 102 | 60170009       |
| 2007        | 531        | 194,873  | 29  | 12 | 2   | 5 | 11| 14 | 20 | 29 | 37 | 45 | 50 | 56 | 60 | 96  | 60065010       |
| 2008        | 528        | 191,875  | 30  | 12 | 2   | 5 | 11| 14 | 21 | 29 | 38 | 46 | 50 | 56 | 61 | 98  | 60170005       |
| 2009        | 556        | 202,142  | 29  | 11 | 2   | 6 | 11| 14 | 21 | 28 | 37 | 44 | 48 | 53 | 57 | 95  | 60170005       |
| 2007-2009   | 611        | 588,890  | 29  | 12 | 2   | 5 | 11| 14 | 21 | 29 | 37 | 45 | 49 | 55 | 60 | 98  | 60170005       |
| **1-h daily max** |        |       |      |    |     |   |   |   |   |   |   |   |   |   |   |   |   |   |                 |
| 2005        | 504        | 183,815  | 48  | 18 | 2   | 11| 21| 26 | 35 | 46 | 58 | 71 | 80 | 91 | 100 | 182 | 60179005       |
| 2006        | 536        | 194,884  | 48  | 18 | 2   | 13| 23| 28 | 36 | 46 | 58 | 71 | 80 | 91 | 100 | 175 | 60370016       |
| 2007        | 531        | 194,873  | 47  | 17 | 2   | 14| 23| 28 | 36 | 45 | 57 | 69 | 77 | 87 | 94  | 237 | 45079021       |
| 2008        | 528        | 191,875  | 47  | 17 | 2   | 14| 23| 27 | 35 | 45 | 56 | 67 | 76 | 87 | 96  | 222 | 45021002       |
| 2009        | 556        | 202,142  | 45  | 15 | 2   | 14| 22| 27 | 35 | 44 | 54 | 64 | 72 | 83 | 91  | 188 | 720770001      |
| 2007-2009   | 611        | 588,890  | 46  | 16 | 2   | 14| 23| 27 | 35 | 45 | 55 | 67 | 75 | 86 | 94  | 237 | 45079021       |
| **8-h daily max (pooled by site)** |        |       |      |    |     |   |   |   |   |   |   |   |   |   |   |   |   |   |                 |
| 2005        | 508        | 183,279  | 42  | 16 | 2   | 7 | 16| 21| 30 | 40 | 52 | 63 | 70 | 78 | 84 | 145 | 60170005       |
| 2006        | 536        | 194,285  | 42  | 16 | 2   | 9 | 18| 23| 31 | 41 | 52 | 63 | 70 | 79 | 85 | 142 | 60170005       |
| 2007        | 528        | 194,266  | 41  | 15 | 2   | 10| 19| 23| 31 | 40 | 51 | 61 | 68 | 75 | 81 | 137 | 60170005       |
| 2008        | 528        | 191,283  | 41  | 15 | 2   | 11| 19| 23| 31 | 40 | 51 | 60 | 66 | 75 | 82 | 172 | 45021002       |
| 2009        | 556        | 201,536  | 40  | 14 | 2   | 11| 18| 23| 30 | 39 | 49 | 57 | 63 | 71 | 77 | 128 | 601710002      |
| 2007-2009   | 608        | 587,085  | 46  | 16 | 2   | 10| 19| 23| 31 | 40 | 50 | 60 | 66 | 74 | 80 | 172 | 45021002       |

<sup>a</sup>Includes all validated data regardless of flags or regional concurrence and therefore may differ from data used for regulatory purposes

<sup>b</sup>AQS Site ID corresponding to the observation in the Max column
Table 3-7  Nationwide distributions of O₃ concentrations (ppb) from the warm-season data set.

<table>
<thead>
<tr>
<th>Period</th>
<th>N Monitors</th>
<th>N Obs</th>
<th>Mean</th>
<th>SD</th>
<th>Min</th>
<th>1</th>
<th>5</th>
<th>10</th>
<th>25</th>
<th>50</th>
<th>75</th>
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*aIncludes all validated data regardless of flags or regional concurrence and therefore may differ from data used for regulatory purposes.

bAQS Site ID corresponding to the observation in the Max column.
Given the strong diurnal pattern in \( \text{O}_3 \) concentrations, the selection of averaging time has a substantial effect on the magnitude of concentration reporting. The nation-wide median 1-h avg, 24-h avg, 1-h daily max, and 8-h daily max concentrations for the year-round data set in 2009 were 29, 28, 44 and 39 ppb, respectively. The median concentrations for the warm-season data set in 2009 were: 29, 29, 45 and 40 ppb, respectively. The 1-h avg and 24-h avg both include the lowest concentrations typically observed in the overnight period which lowers their values relative to the daily maximum statistics.

A strong seasonal pattern in \( \text{O}_3 \) concentrations can also be seen in the year-round data. **Table 3-8** shows the 8-h daily max stratified by season, with the seasons defined as:

- winter: December-February;
- spring: March-May;
- summer: June-August;
- fall: September-November.

In addition, warm-season (May-Sept) and cold-season (Oct-Apr) stratifications of the year-round data set are included in the table for comparison with the four seasonal stratifications. Substantial seasonal variability in the 8-h daily max concentration for the period 2007-2009 was evident with lower concentrations present in fall (median = 36 ppb) and winter (median = 32 ppb) and higher concentrations in spring (median = 47 ppb) and summer (median = 46 ppb). The seasonal differences were even more pronounced in the upper percentiles. For example, the 99th percentile in the 8-h daily max over the 2007-09 time period ranged from 52 ppb in winter to 90 ppb in summer. The distribution in 8-h daily max \( \text{O}_3 \) during the warm-season (as defined above) and during summer were very similar, which is not surprising given their close overlap in months. The distribution during the cold-season (as defined above) is shifted toward higher 8-h daily max \( \text{O}_3 \) concentrations compared with the distribution during winter. This is a result of including the four transition months (Oct, Nov, Mar and Apr) in the cold-season when high \( \text{O}_3 \) concentrations can occur. Further investigation of temporal variability including multiyear trends and diel behavior is included in **Section 3.6.3**.
Table 3-8  Seasonally stratified distributions of 8-h daily max O₃ concentrations (ppb) from the year-round data set (2007-2009).

| Time Period                        | N Monitors | N Obs     | Mean | SD  | Min | 1  | 5  | 10 | 25 | 50 | 75 | 90 | 95 | 98 | 99 | Max | Max Site ID<sup>b</sup> |
|-----------------------------------|------------|-----------|------|-----|-----|----|----|----|----|----|----|----|----|----|----|----|----|------------------------|
| **8-h daily max (2007-2009)<sup>a</sup>** |            |           |      |     |     |    |    |    |    |    |    |    |    |    |    |    |    |                       |
| Year-round                        | 608        | 587,085   | 41   | 15  | 2   | 10 | 19 | 23 | 31 | 40 | 50 | 60 | 66 | 74 | 80 | 90 | 172 | 450210002               |
| **8-h daily max by season (2007-2009)<sup>a</sup>** |            |           |      |     |     |    |    |    |    |    |    |    |    |    |    |    |    |                       |
| Winter (Dec-Feb)                  | 608        | 143,855   | 31   | 10  | 2   | 6  | 14 | 18 | 25 | 32 | 38 | 43 | 46 | 49 | 52 | 172 | 450210002               |
| Spring (Mar-May)                 | 612        | 148,409   | 47   | 12  | 2   | 20 | 28 | 33 | 40 | 47 | 55 | 62 | 67 | 72 | 77 | 118 | 060370016               |
| Summer (June-Aug)                | 613        | 148,280   | 47   | 16  | 2   | 16 | 22 | 26 | 35 | 46 | 57 | 67 | 75 | 84 | 90 | 137 | 060710005               |
| Fall (Sept-Nov)                  | 608        | 146,541   | 37   | 13  | 2   | 10 | 17 | 21 | 28 | 36 | 45 | 54 | 61 | 68 | 75 | 116 | 060370016               |
| Warm-season (May-Sept)           | 616        | 246,233   | 47   | 16  | 2   | 16 | 22 | 27 | 35 | 46 | 57 | 66 | 73 | 81 | 87 | 137 | 060710005               |
| Cold-season (Oct-Apr)            | 608        | 340,852   | 36   | 12  | 2   | 8  | 16 | 21 | 28 | 36 | 44 | 52 | 57 | 63 | 67 | 172 | 450210002               |

<sup>a</sup>Includes all validated data regardless of flags or regional concurrence and therefore may differ from data used for regulatory purposes.

<sup>b</sup>AQS Site ID corresponding to the observation in the Max column.

A national picture of AQS O₃ concentrations was generated from the year-round and warm-season data sets by aggregating the 8-h daily max observations by U.S. county. For this purpose, the 8-h daily max concentrations at each site were averaged over one or more calendar years and then the highest site in each county was selected for that county. Figure 3-27 contains the county-scale 8-h daily max O₃ concentrations from the year-round data set for 2007-2009 (top map) with seasonal stratification (bottom four maps). Figure 3-28 contains the county-scale 8-h daily max O₃ concentrations from the warm-season data set for 2007-2009 (top map) along with individual maps for each calendar year between 2007 and 2009 (bottom three maps). These maps are meant to illustrate the general national-scale distribution in long-term average 8-h daily max O₃ concentrations and are not representative of O₃ concentrations at all locations or times within the counties shown; considerable spatial variability can exist within a county. This is particularly important in the West where counties are larger on average than in the East. These maps are limited by monitor availability, resulting in the majority of U.S. counties not having available data (the white regions in Figure 3-27 and Figure 3-28).
Figure 3-27  Highest monitor (by county) 3-year avg (2007-2009) of the 8-h daily max O₃ concentration based on the year-round data set (the top map) with seasonal stratification (the four bottom maps).
Figure 3-28  Highest monitor (by county) 3-year avg (2007-2009) of the 8-h daily max $O_3$ concentration based on the warm-season data set (the top map) with annual stratification (the three bottom maps).
As shown in the top county-scale map generated from the 2007-2009 year-round data set in Figure 3-27, the highest 3-year avg 8-h daily max O₃ concentrations (≥ 50 ppb) occur in counties in central and southern California, Arizona, Colorado and high elevation counties in Tennessee. The highest year-round average concentration of 61 ppb over this period comes from Site #060719002 located at an elevation of 1,244 meters in Joshua Tree National Monument, San Bernardino County, CA. The lowest 3-year avg 8-h daily max O₃ concentrations (<30 ppb) occur in Pacific Coast counties in northern California and Washington as well as in two northeastern counties in Pennsylvania and Massachusetts. The seasonally-stratified county-scale maps in Figure 3-28 reinforce the strong seasonality in 8-h daily max O₃ concentrations shown in Table 3-8. The highest wintertime concentrations (≥ 40 ppb) occur in the West with the highest 3-year wintertime avg of 46 ppb calculated for Site #080690007 located at an elevation of 2,743 meters near Rocky Mountain National Park, Larimer County, Colorado. In spring and summer, the concentrations increase considerably across all counties, with the highest concentrations (≥ 60 ppb) occurring during the summer in 15 counties in California, 3 counties in Colorado and 1 county each in Nevada and Arizona. Many counties in rural Wyoming, Montana, North Dakota, Maine, and along the Gulf Coast peak in the spring instead of the summer. In the fall, 8-h daily max O₃ concentrations drop back down below their spring and summer concentrations.

The top county-scale map in Figure 3-28 based on the 2007-2009 warm-season data set looks similar to the corresponding map in Figure 3-27 based on the year-round data set. The warm-season map, however, incorporates approximately twice as many monitors across the U.S., providing more spatial coverage. Several counties in Utah, New Mexico, Indiana, Ohio, Maryland, North Carolina, and Georgia in addition to California, Arizona, Colorado and Tennessee identified above have 3-year avg (2007-2009) 8-h daily max O₃ concentrations ≥ 50 ppb based on the warm-season data set. The individual yearly average county-maximum 8-h daily max O₃ concentrations in the lower half of Figure 3-27 show a general decrease in most counties from 2007 to 2009. The number of counties containing a monitor reporting an annual average 8-h daily max O₃ concentration above 50 ppb dropped from 230 counties in 2007 to 30 counties in 2009. This is consistent with the general decrease across these years shown in Table 3-6 and Table 3-7 for the upper percentiles of the 8-h daily max O₃ concentration.

**Urban-Scale Variability**

Statistical analysis of the human health effects of airborne pollutants based on aggregate population time-series data have often relied on ambient concentrations of pollutants measured at one or more central monitoring sites in a given metropolitan area. The validity of relying on central monitoring sites is strongly dependent on the spatial variability in concentrations within a given metropolitan area. To investigate urban-scale variability, 20 focus cities were selected for closer analysis of O₃ concentration variability; these cities are listed in Table 3-9 and were selected based on their importance in O₃ epidemiology studies and on their geographic distribution...
across the United States. In order to provide a well-defined boundary around each city, the combined statistical area (CSA) encompassing each city was used. If the city was not within a CSA, the smaller core-based statistical area (CBSA) was selected. The CSAs/CBSAs are defined by the U.S. Census Bureau (2011) and have been used to establish analysis regions around cities in previous ISAs for particulate matter (PM) (U.S. EPA, 2009d) and carbon monoxide (CO) (U.S. EPA, 2010c).

The distribution of the 8-h daily max O₃ concentrations from 2007-2009 for each of the 20 focus cities is included in Table 3-10. These city-specific distributions were extracted from the warm-season data set and can be compared to the nationwide warm-season 8-h daily max distribution for 2007-2009 in Table 3-7 (and repeated in the first line of Table 3-10 for reference). The median 8-h daily max concentration in these focus cities was 41 ppb, similar to the nationwide median of 42 ppb. Seattle had the lowest median (31 ppb) and Salt Lake City had the highest median (53 ppb) of the 20 cities investigated. The 99th percentile of the 8-h daily max concentration in the focus cities was 84 ppb; similar once again to the nationwide 99th percentile of 80 ppb. Seattle had the lowest 99th percentile (64 ppb) and Los Angeles had the highest 99th percentile (98 ppb) of the 20 cities investigated. In aggregate, the 20 focus cities selected are similar in distribution to the nationwide data set, but there is substantial city-to-city variability in the individual distributions of the 8-h daily max concentrations based on the warm-season data set.

Maps showing the location of central monitoring sites with O₃ monitors reporting to AQS for each of the 20 focus cities are included as supplemental material in Section 3.9.1, Figure 3-76 through Figure 3-95; examples for Atlanta, GA, Boston, MA, and Los Angeles, CA, are shown in Figure 3-29 through Figure 3-31. The sites are delineated in the maps as year-round or warm-season based on their inclusion in the year-round data set and the warm-season data set (the warm-season data set includes May-September data from both the warm-season monitors and the year-round monitors meeting the warm-season data inclusion criteria). The maps also include the CSA/CBSA boundary selected for monitor inclusion, the location of urban areas and water bodies, the major roadway network, as well as the population gravity center based on the entire CSA/CBSA and the individual focus city boundaries. Population gravity center is calculated from the average longitude and latitude values for the input census tract centroids and represents the mean center of the population in a given area. Census tract centroids are weighted by their population during this calculation.

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¹A CBSA represents a county-based region surrounding an urban center of at least 10,000 people determined using 2000 census data and replaces the older Metropolitan Statistical Area (MSA) definition from 1990. The CSA represents an aggregate of adjacent CBSAs tied by specific commuting behaviors. The broader CSA definition was used when selecting monitors for the cities listed above with the exception of Phoenix and San Antonio, which are not contained within a CSA. Therefore, the smaller CBSA definition was used for these metropolitan areas.
<table>
<thead>
<tr>
<th>Focus City</th>
<th>Short Name</th>
<th>CSA/CBSA Name</th>
<th>Year-Round O&lt;sub&gt;3&lt;/sub&gt; Monitoring Sites&lt;sup&gt;b&lt;/sup&gt;</th>
<th>Warm-Season O&lt;sub&gt;3&lt;/sub&gt; Monitoring Sites&lt;sup&gt;c&lt;/sup&gt;</th>
<th>Included in Prior ISAs&lt;sup&gt;d&lt;/sup&gt;</th>
</tr>
</thead>
<tbody>
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<td>Atlanta-Sandy Springs-Gainesville</td>
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<td>11</td>
<td>CO, PM, SO&lt;sub&gt;x&lt;/sub&gt;, NO&lt;sub&gt;x&lt;/sub&gt;</td>
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<td>Baltimore CSA</td>
<td>Washington-Baltimore-northern VA</td>
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<td>Chicago CSA</td>
<td>Chicago-Naperville-Michigan City</td>
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<td>PM, NO&lt;sub&gt;x&lt;/sub&gt;</td>
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<td>Los Angeles-Long Beach-Riverside</td>
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<sup>a</sup>Defined based on 2000 Census data from the U.S. Census Bureau (2011).

<sup>b</sup>The number of sites within each CSA/CBSA with AQS monitors meeting the year-round data set inclusion criteria.

<sup>c</sup>The number of sites within each CSA/CBSA with AQS monitors meeting the warm-season data set inclusion criteria; the warm-season data set includes May - September data from both the warm-season and year-round monitors meeting the warm-season data set inclusion criteria.

<sup>d</sup>Boundaries for the 2010 CO ISA (U.S. EPA, 2010c) and 2009 PM ISA (U.S. EPA, 2009d) focus cities were based on CSA/CBSA definitions; boundaries for the 2008 SO<sub>x</sub> ISA (U.S. EPA, 2008d) and 2008 NO<sub>x</sub> ISA (U.S. EPA, 2008c) focus cities were based on similar metropolitan statistical area (MSA) definitions from the 1990 U.S. Census.
Table 3-10  City-specific distributions of 8-h daily max O₃ concentrations (ppb) from the warm-season data set (2007-2009).

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<th>SD</th>
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<th>98</th>
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<th>Max</th>
<th>Max Site ID&lt;sup&gt;b&lt;/sup&gt;</th>
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<sup>a</sup>Includes all validated data regardless of flags or regional concurrence and therefore may differ from data used for regulatory purposes.

<sup>b</sup>AQS Site ID corresponding to the observation in the Max column.
Figure 3-29  Map of the Atlanta, Georgia, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-30  Map of the Boston, Massachusetts, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-31 Map of the Los Angeles, California, CSA including O3 monitor locations, population gravity centers, urban areas, and major roadways.

The Atlanta, GA, CSA contains 11 warm-season monitors distributed evenly yet sparsely around the city center (Figure 3-29). The population gravity center for the city and the larger CSA are only separated by 4 km, indicating that the majority of the population lives within or evenly distributed around the city limits. Atlanta is landlocked with a radial network of interstate highways leading to the city center. The Boston, MA, CSA contains 3-year-round and 18 warm-season monitors spread evenly throughout the CSA. Boston is a harbor city with the Atlantic Ocean to the east, resulting in the city-based population gravity center being located 17 km east of the CSA-based population gravity center. The Los Angeles, CA, CSA contains the largest number of monitors of the 20 CSA/CBSAs investigated with 47 year-round and 3 warm-season monitors. These monitors are primarily concentrated in the Los Angeles urban area with relatively few monitors extending out to the northern and eastern reaches of the CSA. These unmonitored areas are very sparsely populated, resulting in only 15 km separating the city-based and the CSA-based population gravity centers despite the vast area of the Los Angeles CSA.

Other CSAs/CBSAs (see Section 3.9.1) with monitors concentrated within the focus city limits include Birmingham, AL, Chicago, IL, Denver, CO, Houston, TX, Phoenix AZ, San Antonio, TX, and Salt Lake City, UT. The remaining CSAs/CBSAs have monitors distributed more evenly throughout the CSA/CBSA area. Baltimore, MD, is contained within the same CSA as Washington D.C. and suburbs, resulting in
a 50-km separation (the largest of the focus cities investigated) between the city-based population gravity center for Baltimore and the CSA-based population gravity center for the Washington-Baltimore-Northern Virginia CSA.

Box plots depicting the distribution of 2007-2009 warm-season 8-h daily max O₃ data from each individual monitor in the 20 focus cities are included as supplemental material in Section 3.9.2, Figure 3-96 through Figure 3-115; examples for Atlanta, GA, Boston, MA, and Los Angeles, CA, are shown in Figure 3-32 through Figure 3-34. The Atlanta CSA has little spatial variability in 8-h daily max O₃ concentrations with median concentrations ranging from 47 ppb at Sites I and J located far from the city center to 54 ppb at Site A located closest to the city center. The variation in warm-season 8-h daily max concentrations are also relatively similar across monitors with IQRs ranging from 17 ppb at Site J to 23 ppb at Site B. The Boston CSA has more spatial variability in 8-h daily max O₃ concentrations than the Atlanta CSA with median concentrations ranging from 33 ppb at Site A nearest to the city center to 46 ppb at Site L located 84 km west of the city center. For monitors located within and just adjacent to the Boston city limits (Sites A-D), the O₃ concentrations can vary over relatively short distances owing to differing degrees of NOₓ titration and influence from the local topography. Like the Atlanta CSA, the variation in warm-season 8-h daily max concentrations are relatively similar across monitors within the Boston CSA with IQRs ranging from 15 ppb at Site U to 21 ppb at Site K. The Los Angeles CSA exhibits the most variability in O₃ concentrations between monitors of all the CSAs/CBSAs investigated. The median 8-h daily max O₃ concentration in the Los Angeles CSA ranged from 20 ppb at Site AM in the south-central extreme of the CSA to 80 ppb at Site AE near Crestline, CA in the San Bernardino National Forest just north of San Bernardino, CA. These two sites are at approximately the same longitude and are separated by only 85 km, but the Crestline site is downwind of the Los Angeles basin, resulting in substantially higher O₃ concentrations. Site AM also contains data for only 2009, which could explain some of the deviation when comparing this site with others in the Los Angeles CSA. Sites AM and AE also had the lowest (8 ppb) and highest (28 ppb) IQR, respectively. The remaining focus cities exhibited spatial variability ranging from uniform as in the Atlanta CSA to non-uniform as observed in the Los Angeles CSA (see supplemental figures in Section 3.9.2).
Figure 3-32 Site information, statistics and box plots for 8-h daily max O₃ from AQS monitors meeting the warm-season data set inclusion criteria within the Atlanta CSA.

Figure 3-33 Site information, statistics and box plots for 8-h daily max O₃ from AQS monitors meeting the warm-season data set inclusion criteria within the Boston CSA.

3-100
Figure 3-34 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Los Angeles CSA.
Pair-wise monitor comparisons were used to further evaluate spatial variability between monitors within the 20 focus cities. In the particular case of ground-level O₃, central-site monitoring has been justified as a regional measure of exposure mainly on the grounds that correlations between concentrations at neighboring sites measured over time are usually high. In areas with multiple monitoring sites, averages over the monitors have often been used to characterize population exposures. However, substantial differences in concentrations between monitors can exist even though concentrations measured at the monitoring sites are highly correlated, thus leading to the potential for exposure misclassification error. Therefore, both the Pearson correlation coefficient and the coefficient of divergence (COD) were calculated for each monitor pair within the CSA/CBSAs using the 8-h daily max O₃ data. The correlation provides an indication of temporal linear dependence across sites while the COD provides an indication of the variability in absolute concentrations across sites. The COD is defined as follows:

\[
COD_{jk} = \sqrt{\frac{1}{p} \sum_{t=1}^{p} \left( \frac{X_{ij} - X_{ik}}{X_{ij} + X_{ik}} \right)^2}
\]

Equation 3-1

where \(X_{ij}\) and \(X_{ik}\) represent observed concentrations averaged over some measurement averaging period \(i\) (hourly, daily, etc.) at sites \(j\) and \(k\), and \(p\) is the number of paired observations. A COD of 0 indicates there are no differences between concentrations at paired sites (spatial homogeneity), while a COD approaching 1 indicates extreme spatial heterogeneity. These methods for analysis of spatial variability follow those used in previous ISAs for CO, PM, SOₓ and NOₓ as well as those used in Pinto et al. (2004) for PM₂.₅.

Histograms and contour matrices of the Pearson correlation coefficient between 8-h daily max O₃ concentrations from each monitor pair are included as supplemental material in Section 3.9.3, Figure 3-116 through Figure 3-135; examples for Atlanta, Boston and Los Angeles are shown in Figure 3-35 through Figure 3-37. Likewise, histograms, contour matrices, and scatter plots of the coefficient of divergence (COD) between 8-h daily max O₃ concentrations from each monitor pair are included as supplemental material in Section 3.9.3, Figure 3-136 through Figure 3-155; examples for Atlanta, Boston and Los Angeles are shown in Figure 3-38 through Figure 3-40. These figures also contain scatter plots of correlation and COD as a function of straight-line distance between monitor pairs.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of the correlations.

**Figure 3-35**  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max \( \text{O}_3 \) in the Atlanta CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of the correlations.

**Figure 3-36** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Boston CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of the correlations.

**Figure 3-37** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Los Angeles CSA.
Figure 3-38  Pair-wise monitor coefficient of divergence (COD) expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Atlanta CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of the CODs.

**Figure 3-39** Pair-wise monitor coefficient of divergence (COD) expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Boston CSA.
Figure 3-40  Pair-wise monitor coefficient of divergence (COD) expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Los Angeles CSA.
The monitor pairs within the Atlanta CSA (Figure 3-35) were generally well correlated with correlations between 8-h daily max O₃ concentrations ranging from 0.61 to 0.96. The correlations shown in the scatter plot were highest for close monitor pairs and dropped off with distance in a near-linear form. At a monitor separation distance of 50 km or less, the correlations ranged from 0.79 to 0.96. The monitor pairs within the Boston CSA (Figure 3-36) were also generally well correlated with correlations ranging from 0.49 to 0.96. Again, the correlations shown in the scatter plot were highest for close monitor pairs, but there was slightly more scatter in correlation as a function of distance in the Boston CSA compared with the Atlanta CSA. At a monitor separation distance of 50 km or less, the correlations ranged from 0.81 to 0.96. The monitor pairs within the Los Angeles CSA (Figure 3-37) showed a much broader range in correlations, extending from -0.06 to 0.97. At a monitor separation distance of 50 km or less, the correlations shown in the scatter plot ranged from 0.21 to 0.97. The negative and near-zero correlations were between monitors with a relatively large separation distance (>150 km), but even some of the closer monitor pairs were not very highly correlated. For example, Site AL located at Emma Wood State Beach in Ventura and Site AK situated in an agricultural valley surrounded by mountains 20 km inland (see map in Figure 3-41) had a correlation coefficient of only 0.21 over the 2007-2009 warm-season time period. This was slightly lower than the correlation between Site AL and Site AX on the Arizona border, 441 km away (R = 0.28). San Francisco and Seattle (Figure 3-133 and Figure 3-134 in Section 3.9.3) also showed a broad range in pairwise correlations, likely resulting from their similar geography where background air coming in from the Pacific Ocean rapidly mixes with urban pollutants such as NOₓ and VOCs from coastal cities and is transported downwind into diversified terrain to create spatially and temporally varying O₃ concentrations.
Note: Site AL is near shore, 3 meters above sea level, while Site AK is in an agricultural valley surrounded by mountains, 262 meters above sea level.

**Figure 3-41** Terrain map showing the location of two nearby AQS O₃ monitoring sites (red dots) along the western edge of the Los Angeles CSA.

The COD between 8-h daily max O₃ measured at paired monitors in all CSAs/CBSAs (Figure 3-136 through Figure 3-155 in Section 3.9.3) were generally low, with values similar to those shown in Figure 3-38 and Figure 3-39 for Atlanta and Boston. This suggests a generally uniform distribution in the 8-h daily max O₃ concentration across monitors within these cities and is consistent with the uniformity observed in the box plots (e.g., Figure 3-32, Figure 3-33, and Figure 3-96 through Figure 3-115 in Section 3.9.2). Los Angeles (Figure 3-34) and San Francisco (Figure 3-153 in Section 3.9.3), however, had several monitor pairs with COD >0.30 indicating greater spatial heterogeneity. This is consistent with the variability observed in the box plots for these two CSAs (Figure 3-34 and Figure 3-113 in Section 3.9.2). In particular, Site AM in the Los Angeles CSA had consistently lower concentrations (median = 20 ppb, IQR = 17-25 ppb) relative to other sites in the CSA.
(Figure 3-31), resulting in high CODs with other monitors as shown in Figure 3-40. The O₃ monitor at Site AM is located on the Pechanga Tribal Government Building in Temecula, CA, and began collecting data on June 9, 2008. It is located in a suburban setting and is classified as a general background monitor. Another close by site (site ID = 060731201) located in the Pala Reservation, 9.5 km south of this one (just outside the boundary of the Los Angeles CSA) reported similarly low 2009 8-h daily max O₃ concentrations (median = 28 ppb, IQR = 23-32 ppb) between May-June, 2009 (the only warm-season months with available data from this site between 2007 and 2009).

Note: Site characteristics range from Site A near downtown at 6 meters above sea level to Site D in a forested area on Blue Hill at 192 meters above sea level.

Figure 3-42 Terrain map showing the location of four AQS O₃ monitoring sites (red dots) located in or near the city limits in the center of the Boston CSA.
There are instances where sites in an urban area may exhibit substantial differences in median concentrations, but still be moderately well correlated in time. For example, Sites A and D in Boston (see terrain map in Figure 3-42) have an 11 ppb difference in median 8-h daily max O₃ concentration (COD = 0.16), but a high correlation (R = 0.90). In this example, Site A is located in the Boston city limits at an elevation of 6 meters while Site D is located 13 km to the south in a forested area on Blue Hill, the highest point in Norfolk County (elevation = 192 meters). The difference in median O₃ concentration at these two sites can be attributed to differing degrees of NOₓ titration between the neighborhood scale site (Site A) and the regional scale site (Site D) and to the influence of local topography.

Comparison of monitoring data within the selected focus cities has demonstrated considerable variability between cities in the behavior of the O₃ concentration fields. Median O₃ concentrations vary more within certain urban areas than others. Likewise, pair-wise monitor statistics (R and COD) are dependent on the urban area under investigation. These conclusions are consistent with those drawn in the 2006 O₃ AQCD (U.S. EPA, 2006b) where a subset of these focus cities were investigated using similar statistics. As a result, caution should be observed in using data from a sparse network of ambient O₃ monitors to approximate community-scale exposures.

**Neighborhood-Scale Variability and the Near-Road Environment**

Ozone is a secondary pollutant formed in the atmosphere from precursor emissions and therefore is generally more regionally homogeneous than primary pollutants emitted from stationary or mobile point sources. However, O₃ titration from primary NO emissions does result in substantial localized O₃ gradients. This is evident in the near-road environment where fresh NO emissions from motor vehicles titrate O₃ present in the urban background air, resulting in an O₃ gradient down-wind from the roadway. Ozone titration occurring in street canyons where NO emissions are continuously being generated is more efficient because of inhibited transport away from the source of NO.

Several studies have reported O₃ concentrations that increase with increasing distance from the roadway, both upwind and downwind of the road. Beckerman et al. (2008) measured O₃ profiles in the vicinity of heavily traveled roadways with Annual Average Daily Traffic (AADT) >340,000 vehicles in Toronto, Canada. Ozone was observed to increase with increasing distance from the roadway, both upwind and downwind of the road. This is consistent with scavenging of O₃ in the near-road environment by reaction with NO to form NO₂. Upwind of the road, concentrations were >75% of the maximum observed value at >100 meters from the road; downwind, concentrations were approximately 60% of the maximum within 200-400 meters of the road. The O₃ concentration adjacent to the road on the upwind side was approximately 40% of the maximum value observed at the site. Concentrations measured with Ogawa passive samplers over a 1-week period ranged from 7.3-19.4 ppb with the mean at the two sites ranging from 13.0-14.7 ppb. In a study of patrol cars during trooper work shifts, Riediker et al. (2003) made
simultaneous 9-h O₃ measurements inside patrol cars, at the roadside, and at a centrally-located ambient monitoring site. The roadside concentrations were approximately 81% of the ambient values (mean of 22.8 ppb versus 28.3 ppb). Wind direction relative to the roadway was not reported.

Johnson (1995) measured O₃, NO, and CO concentrations at upwind and downwind locations near a variety of roadways in Cincinnati, OH. The effects of O₃ scavenging by NO were apparent in the O₃ reduction in the interval between 9 meters upwind and 82 meters downwind of the road. A similar effect was observed by Rodes and Holland (1981) during an earlier study in which outdoor O₃ concentrations were monitored downwind of a freeway in Los Angeles, CA. In this study, O₃ concentrations measured near the roadway were approximately 20% of the concentrations measured simultaneously at more distant locations judged to be unaffected by the roadway. Minimal separation distances of the samplers from the roadway to eliminate measurable influence were estimated to be approximately 400-500 meters for NO, NO₂, and O₃. Similar results have been observed outside the U.S., for example in the city of Daegu, Korea, where the yearly roadside concentrations of CO and SO₂ showed a well-defined decreasing trend with distance from the roadway, whereas concentrations of NO₂ and O₃ exhibited the reverse trend (Jo and Park, 2005). During the peak O₃ month of May, O₃ concentrations in a residential neighborhood were approximately 40% higher than concentrations at roadside monitors located 1 meter from the edge of multiple-lane freeways.

3.6.2.2 Rural-Focused Variability and Ground-level Vertical Gradients

AQS O₃ data for monitors located at several rural monitoring sites (e.g., national parks, national forests, state parks, etc.) were used to investigate rural-focused O₃ concentration variability in contrast with the urban-focused variability discussed in Section 3.6.2.1. These rural monitoring sites tend to be less directly affected by anthropogenic pollution sources than urban sites. However, they can be regularly affected by transport of O₃ or O₃ precursors from upwind urban areas, or by local anthropogenic sources within the rural areas such as emissions from motor vehicles, power generation, biomass combustion, or oil and gas operations. As a result, monitoring data from these rural locations are not unaffected by anthropogenic emissions.

Six rural focus areas were selected for their geographic distribution across the U.S. as well as their unique topography and relevance to the ecological assessment in Chapter 9. Table 3-11 lists the rural focus areas and provides some cursory site information along with the number of available AQS monitors reporting year-round and only during the warm-season. Accompanying box plots depicting the distribution of 2007-2009 warm-season 8-h daily max O₃ data from each individual monitor in the six rural focus areas are included in Figure 3-43. This analysis was restricted to AQS monitors meeting the same data completeness criteria outlined in Table 3-5 for
a direct comparison with the 20 urban focus areas investigated in Section 3.6.2.1. Given the population-center emphasis of the AQS network, limited monitoring sites (between one and five) were available for each rural focus area. Expanded analyses of O$_3$ concentrations measured using the more rural-focused CASTNET monitoring network are included in Chapter 9.

### Table 3-11  Rural focus areas.

<table>
<thead>
<tr>
<th>Focus Area</th>
<th>Short Name</th>
<th>Year-Round O$_3$ Monitoring Sites$^a$</th>
<th>Warm-Season O$_3$ Monitoring Sites$^b$</th>
<th>Monitor Elevation (meters)</th>
<th>Site Descriptions</th>
</tr>
</thead>
<tbody>
<tr>
<td>Adirondack State Park, NY</td>
<td>ADSP</td>
<td>1</td>
<td>0</td>
<td>1,483</td>
<td>One site on the summit of Whiteface Mountain in the Adirondack Mountains</td>
</tr>
<tr>
<td>Mount Mitchell State Park, NC</td>
<td>MMSP</td>
<td>0</td>
<td>1</td>
<td>1,982</td>
<td>One site near the summit of Mount Mitchell (highest point in the eastern U.S.), in the Appalachian Mountains</td>
</tr>
<tr>
<td>Great Smoky Mountain National Park, NC-TN</td>
<td>SMNP</td>
<td>2</td>
<td>3</td>
<td>564 to 2,021</td>
<td>Five different locations within Great Smoky Mountain National Park in the Appalachian Mountains</td>
</tr>
<tr>
<td>Rocky Mountain National Park, CO</td>
<td>RMNP</td>
<td>1</td>
<td>0</td>
<td>2,743</td>
<td>One site in a valley at the foot of Longs Peak in the Rocky Mountains</td>
</tr>
<tr>
<td>San Bernardino National Forest, CA</td>
<td>SBNF$^c$</td>
<td>1</td>
<td>0</td>
<td>1,384</td>
<td>One site in Lake Gregory Regional Park (near Crestline, CA) in the San Bernardino Mountains</td>
</tr>
<tr>
<td>Sequoia National Park, CA</td>
<td>SENP</td>
<td>2</td>
<td>0</td>
<td>560 to 1,890</td>
<td>Two contrasting sites at different elevations within Sequoia NP in the Sierra Nevada Mountains</td>
</tr>
</tbody>
</table>

$^a$Number of AQS monitors meeting the year-round data set inclusion criteria; the year-round data set is limited to these monitors.

$^b$Number of AQS monitors meeting the warm-season data set inclusion criteria; the warm-season data set includes May-September data from both the warm-season and year-round monitors.

$^c$Same AQS site as Site AE in the Los Angeles CSA shown in Figure 3-31.
**Rural Focus Areas**

<table>
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<tr>
<th>Site ID</th>
<th>Years</th>
<th>N</th>
<th>Mean</th>
<th>SD</th>
<th>Median</th>
<th>IQR</th>
<th>Area</th>
<th>Site</th>
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<td>14</td>
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<td>68</td>
<td>15</td>
<td>69</td>
<td>19</td>
<td>B</td>
<td></td>
</tr>
</tbody>
</table>

Note: Includes: Adirondack State Park, NY (ADSP); Mount Mitchell State Park, NC (MMSP); Great Smoky Mountain National Park, NC-TN (SMNP); Rocky Mountain National Park, Colorado (RMNP); San Bernardino National Forest, CA (SBNF); and Sequoia National Park, CA (SENP).

**Figure 3-43** Rural focus area site information, statistics and box plots for 8-h daily max O₃ from AQS monitors meeting the warm-season data set inclusion criteria within the rural focus areas.

**Eastern Rural Focus Areas**

In the East, the distribution in warm-season 8-h daily max O₃ concentrations from the Adirondack State Park (ADSP) site on Whiteface Mountain in Upstate NY (median = 49 ppb) ([Figure 3-43](#)) was among the lowest of the rural focus monitors investigated, but was still higher than concentration distributions measured in the Boston CSA (medians ranging from 33 to 46 ppb) ([Figure 3-33](#)) located 320 km to the southeast. The ADSP AQS site was included in an analysis for the 2006 O₃ AQCD ([U.S. EPA, 2006b](#)) and had the lowest year-round median hourly O₃ concentration of the rural forested sites investigated (including Yellowstone NP, the Great Smoky Mountains NP, and Shenandoah NP). For the Appalachian Mountain monitors in Mount Mitchell State Park, NC (MMSP) and Great Smoky Mountain National Park, NC-TN (SMNP), there was a fair amount of variability in concentration distribution. Within SMNP, the median warm-season 8-h daily max O₃ concentration ranged from 47 ppb at the lowest elevation site (elevation = 564 meters; site ID = 470090102) to 60 ppb at the highest elevation site (elevation = 2,021 meters; site ID = 471500102); these sites are shown on the terrain map in [Figure 3-44](#). The warm-season median 8-h daily max O₃ concentration gradient between these two sites located 26.2 km apart in SMNP was 0.9 ppb per 100 meters elevation gain.

Data from the five sites within SMNP allowed for further investigation of spatial variability within the park; [Figure 3-45](#) contains histograms, contour plots and scatter
plots as a function of distance for the pair-wise correlation and COD (defined in Equation 3-1) for SMNP. The correlations between the five sites ranged from 0.78 to 0.92 and the CODs ranged from 0.04 to 0.16. The plots of correlation and COD as a function of distance between SMNP monitor pairs in Figure 3-45 show a large degree of spatial variability between monitors over relatively short distances. A host of factors may contribute to these variations, including proximity to local O₃ precursor emissions, variations in boundary-layer influences, meteorology and stratospheric intrusion as a function of elevation, and differences in wind patterns and transport behavior due to local topography.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histograms includes the number of monitor pairs per bin and the contour matrix includes the numeric values of the correlations and CODs.

**Figure 3-45**  Pair-wise monitor correlations (left) and coefficients of divergence (CODs) (right) expressed as a histogram (top), contour matrix (middle) and scatter plot vs. distance between monitors (bottom) for 8-h daily max O₃ in Great Smoky Mountain National Park, NC-TN (SMNP).

**Western Rural Focus Areas**

The Rocky Mountain National Park (RMNP) site in Colorado at 2,743 meters in elevation had a warm-season 8-h daily max O₃ concentration distribution (median = 56 ppb, IQR = 11 ppb) (Figure 3-43) that is comparable to the distributions at sites in the Denver CSA located 75 km southeast at elevations around 1,600 meters (medians ranging from 41 to 59 ppb, IQRs ranging from 10 to 16 ppb; see Figure 3-102 in Section 3.9.1). In nearby Boulder County, Colorado, a 1-year time-series (Sept 2007-Aug 2008) of ambient surface-level O₃ measurements was collected by Brodin et al. (2010) along an elevation gradient ranging from 1,608 meters to 3,528 meters. The 7 sites used in this study are shown in Figure 3-46 along with the RMNP site and the 15 Denver CSA sites. In fall, winter, and spring, they observed a clear monotonic increase in O₃ concentration with elevation, with a rate of increase in the mean O₃ concentration of 1.5 ppb per 100 meters elevation gain during winter. In summer, the O₃ gradient was similar in magnitude over the
seven-site transect (1.3 ppb per 100 meters), but much less monotonic; the majority of the vertical gradient occurred between the lowest two sites (4.5 ppb per 100 meters) and between the highest two sites (5.5 ppb per 100 meters), with the middle five sites all having approximately equal median O₃ concentrations. Ozone concentrations at the lowest site in Boulder were influenced by NO titration as evidenced by traffic-related diel cycles in O₃ concentrations, but the remaining six sites were located at elevation in more rural/remote settings and illustrate a positive surface-level O₃ elevation gradient similar to that seen in SMNP and typical of areas under less direct influence of boundary layer pollution.

Note: Elevations range from approximately 1,600 meters above sea level in Denver and Boulder, Colorado, to 3,528 meters above sea level at the highest mountainous site. Blue circles indicate monitoring sites used in the Brodin et al. (2010) study.

Figure 3-46 Terrain map showing the location of the AQS O₃ monitoring site in Rocky Mountain National Park, Colorado (black/green star) and the Denver, Colorado, CSA (red dots) along with O₃ monitoring sites used in the Brodin et al. (2010) study (blue circles).
The three sites in California—one in San Bernardino National Forest (SBNF) and two in Sequoia National Park (SENP)—had the highest distribution of 8-h daily max O₃ concentrations of the selected rural focus area monitors included in Figure 3-43. The SBNF site had a warm-season 8-h daily max O₃ concentration mean of 80 ppb and a maximum of 137 ppb measured on July 1, 2007. This site is located in Crestline, CA, 90 km down-wind of Los Angeles in the San Bernardino Mountains. This site was included in the Los Angeles CSA shown in Figure 3-31 (Site AE) and had the highest median 8-h daily max O₃ concentration of any AQS site in the Los Angeles CSA during this time period (Figure 3-34). This site was also included in an analysis performed for the 2006 O₃ AQCD (U.S. EPA, 2006b) where similarly high O₃ concentrations were observed using 2004 year-round hourly observations.

Note: The lower site (site ID = 061070009) is 560 meters above sea level and the higher site (site ID = 061070006) is 1,890 meters above sea level.

Figure 3-47  Terrain map showing the location of two AQS O₃ monitoring sites (black/green stars) in Sequoia National Park, CA.
The two sites in SENP are located 9.7 km apart at contrasting elevations as is illustrated in the terrain map in Figure 3-47. The correlation in 8-h daily max O₃ between these two sites was 0.86 and the COD was 0.09, which are within the range in correlations and CODs for SMNP (Figure 3-45). The distribution of 8-h daily max O₃ concentrations at the lower elevation site (elevation = 560 meters; site ID = 061070009) is shifted slightly higher with a median of 76 ppb compared to the higher elevation site (elevation = 1,890 meters; site ID = 061070006) with a median of 69 ppb. The lower elevation site is located at the entrance to the park and is at a low enough elevation to be influenced by boundary layer pollution coming upwind from Fresno and the San Joaquin Valley. The higher elevation site is in the free troposphere above the planetary boundary layer and is less influenced by such pollution. This gives rise to a negative average surface-level elevation gradient of -0.5 ppb per 100 meters elevation gain in SENP, illustrating the location-specific complexities inherent to high-altitude surface-level O₃ concentrations.

Since O₃ produced from emissions in urban areas is transported to more rural downwind locations, elevated O₃ concentrations can occur at considerable distances from urban centers. In addition, major sources of O₃ precursors such as highways, power plants, biomass combustion, and oil and gas operations are commonly found in rural areas, adding to the O₃ in these areas. Due to lower chemical scavenging in non-urban areas, O₃ tends to persist longer in rural than in urban areas which tends to lead to higher cumulative exposures in rural areas influenced by anthropogenic precursor emissions. The persistently high O₃ concentrations observed at many of these rural sites investigated here indicate that cumulative exposures for humans and vegetation in rural areas can be substantial and often higher than cumulative exposures in urban areas.

### 3.6.3 Temporal Variability

#### 3.6.3.1 Multiyear Trends

As reported in the 2010 National Air Quality Status and Trends report (U.S. EPA, 2010e), nation-wide surface-level O₃ concentrations in the U.S. have declined gradually over the last decade. Figure 3-48 shows the downward trend in the annual 4th-highest 8-h daily max O₃ concentration from 870 surface-level monitors across the United States. Figure 3-49 shows a similar trend in the annual 2nd highest 1-h daily max O₃ concentration from 875 surface-level monitors. The median annual 4th-highest 8-h daily max dropped from 88 ppb in 1998 to 71 ppb in 2010. Likewise, the median annual 2nd-highest 1-h daily max dropped from 109 ppb in 1998 to 86 ppb in 2010. The large decreases in 2003 and 2004 in both figures coincides with NOₓ emissions reductions resulting from implementation of the NOₓ State Implementation Plan (SIP) Call rule, which began in 2003 and was fully
implemented in 2004. This rule was designed to reduce NO\textsubscript{X} emissions from power plants and other large combustion sources in the eastern United States. Reductions in mobile NO\textsubscript{X} emissions nationwide from the implementation of recent vehicle and fuel standards could also be adding to the gradual decline in nationwide surface-level O\textsubscript{3} concentrations (Dallmann and Harley, 2010).

**Figure 3-48**: National 8-h daily max O\textsubscript{3} trend and distribution across 870 U.S. O\textsubscript{3} monitors, 1998-2010 (annual 4th-highest 8-h daily max O\textsubscript{3} concentrations in ppm).
The distributional percentiles (10th, 25th, 75th, and 90th) displayed in Figure 3-48 and Figure 3-49 reveal a gradual tightening of the O₃ concentration distribution observed across monitors. For the annual 4th-highest 8-h daily max O₃ concentration, the IQR decreased from 17 ppb in 1998 to 9 ppb in 2010. Likewise, for the annual 2nd highest 1-h daily max O₃ concentration, the IQR decreased from 23 ppb in 1998 to 16 ppb in 2010. A similar tightening was observed for the wider percentiles (90th-10th) for both averaging times.

Weather can have a strong influence on the O₃ trends shown in Figure 3-48 and Figure 3-49. The number of hot, dry days can substantially alter the number of high O₃ days in any given year, even if O₃ forming emissions do not change. To better evaluate the progress and effectiveness of emissions reduction programs, EPA uses a statistical model to estimate the influence of atypical weather on O₃ formation (U.S. EPA, 2010e). After adjusting for the influence of weather, the downward trend in surface-level national 8-h daily max O₃ concentrations between 2001 and 2008 increased slightly from an 8% reduction prior to adjustment for weather to an 11% reduction after adjustment for weather (U.S. EPA, 2010e).

A regional breakdown of the trend in O₃ concentrations for the 8-hour and 1-hour metrics is included in Figure 3-50 and Figure 3-51, respectively. In general, the trends are region-specific with a substantial amount of year-to-year variability. The reduction in NOₓ and O₃ during the 2003-2004 timeframe is particularly evident in the North Central and Northeastern U.S. where the NOₓ SIP Call was focused.
The western region (including Alaska and Hawaii but excluding California) started out with the lowest annual O₃ concentration in 1998 and exhibits the least amount of reduction when compared to 2010 concentrations (11% reduction in the average annual 4th-highest 8-h daily max and 13% reduction in the average annual 2nd-highest 1-h daily max). In contrast, California—which has some of the highest concentrations of the identified regions—shows a larger downward trend in O₃ concentrations over the same time period (19% reduction in the average annual 4th-highest 8-h daily max and 22% reduction in the average annual 2nd-highest 1-h daily max).

**Figure 3-50**  
Trend in mean 8-h daily max O₃ by region, 1998-2010 (mean of the annual 4th-highest 8-h daily max O₃ concentrations in ppm).
Figure 3-51  Trend in mean 1-h daily max O$_3$ by region, 1998-2010 (mean of the annual 2nd-highest 1-h daily max O$_3$ concentrations in ppm).

Narrowing the focus to changes in O$_3$ concentrations at the individual monitor level, Figure 3-52 displays the 8-h O$_3$ design value (4th-highest 8-h daily max O$_3$ concentration occurring within a three-year period) for all available monitors for the 2008-2010 period (Figure 3-52A) as well as the change in this design value between the 2001-2003 period and the 2008-2010 period (Figure 3-52B). Figure 3-53 displays analogous information for a 1-h O$_3$ design value (4th-highest 1-h daily max O$_3$ concentration occurring within a three-year period). As can be seen in both figures, the majority of monitors recorded a decrease in design values when comparing the 2001-2003 period to the 2008-2010 period. Specifically, 699 of 853 sites (82%) included in Figure 3-52B for the 8-h design value and 747 of 869 sites (86%) included in Figure 3-53B for the 1-h design value reported a decrease of at least 6 ppb in the respective design values. The highest density of monitors reporting decreases occurs in the Northeast. Only 8 sites (1%) reported an increase of more than 5 ppb in the 8-h design value and only 16 sites (2%) reported an increase of more than 5 ppb in the 1-h design value. These sites reporting an increase between the 2001-2003 and the 2008-2010 periods were located primarily in the West. More in depth trends analyses have been performed to support the O$_3$ Risk and Exposure Analysis and are available through EPA’s Technology Transfer Network website for the O$_3$ review (Wells et al., 2012).
Figure 3-52  Individual monitor 8-h daily max O₃ design values displayed: (A) for the 2008-2010 period, and (B) as the change since the 2001-2003 period.
Figure 3-53  Individual monitor 1-h daily max O₃ design values displayed: (A) for the 2008-2010 period, and (B) as the change since the 2001-2003 period.
Similar findings were reported for regional trends in the 4th-highest 8-h daily max O3 concentration between 2001 and 2008 in the 2010 National Air Quality Status and Trends report (U.S. EPA, 2010e). Individual sites that showed the greatest reduction in O3 between 2001 and 2008 were in or near the following metropolitan areas: Anderson, IN; Chambersburg, PA; Chicago, IL; Cleveland, OH; Houston, TX; Michigan City, IN; Milwaukee, WI; New York, NY; Racine, WI; Watertown, NY; and parts of Los Angeles, CA. Individual sites that showed an increase in O3 over this time period and had measured concentrations above the O3 standard1 during the 2006-2008 time period were located in or near the following metropolitan areas: Atlanta, GA; Baton Rouge, LA; Birmingham, AL; Denver, Colorado; El Centro, CA; San Diego, CA; Seattle, WA; and parts of Los Angeles, CA.

Pegues et al. (2012) investigated changes in 3-year average 8-h daily max O3 design values between 2003 and 2009 and found reductions at the majority of sites across the U.S.; consistent with the findings in this section and in the 2010 National Air Quality Status and Trends report (U.S. EPA, 2010e). Furthermore, they compared trends in O3 design values between areas that were or were not classified as nonattainment of the 84 ppb O3 standard in the 2004 designations. Monitors designated nonattainment achieved O3 design value reductions of 13.3 ppb on average while monitors designated in attainment achieved reductions of 7.0 ppb on average.

Looking further back in time, Leibensperger et al. (2008) included an analysis of June-August 8-h daily max O3 trends from 1980-2006 using AQS data from over 2000 sites in the contiguous United States. They created an index for “pollution days” representing days when the 8-h daily max O3 concentration was greater than 84 ppb. The observed trend in summertime O3 pollution days over this 27 year period decreased at an average rate of -0.84 days/year. The authors used several methods to deconstruct this trend into a component coming from reductions in O3 precursor emissions (-1.50 days/year) and a component coming from climate change (+0.63 days/year). The climate change impact is a result of decreases in frequency of mid-latitude cyclones which serve to ventilate surface air over the United States. Leibensperger et al. (2008) conclude that the reduction in frequency of mid-latitude cyclones over the 1980-2006 time period has offset almost half of the air quality gains in the Northeastern U.S. that should have been achieved from reductions of anthropogenic emissions alone over that period. This conclusion is based on the assumption of a linear additive relationship between O3 precursor emission changes and cyclone frequency variations on the rate of change of high O3 days, and does not account for nonlinearities inherent in O3 chemistry as discussed in Section 3.2.

A more recent analysis by Turner et al. (2012) addressed this issue by utilizing the GFDL CM3 global coupled chemistry climate model to assess relationships between summertime cyclones and O3 pollution episodes. They also found a robust decline in cyclone frequency in modeled scenarios incorporating climate change, but in their models, less than 10% of the variability in high O3 days in the Northeastern U.S. was explained by cyclone activity. As a result, Turner et al. (2012) caution against over-

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1 The 2008 O3 NAAQS include primary and secondary standards of 0.075 ppm (8-h daily max).
interpreting the strong association observed by Leibensperger et al. (2008) between mid-latitude cyclone frequency and the occurrence of high \( \text{O}_3 \) days.

Averaging time can have an impact on perceived trends in surface-level \( \text{O}_3 \) concentrations. Lefohn et al. (2008) investigated the impact of using different exposure indices on trends in surface-level \( \text{O}_3 \) concentrations in the U.S. by comparing the annual 2nd-highest 1-h avg concentration, the annual 4th-highest daily maximum 8-h avg concentration, and the seasonally corrected 24-h \( \text{W126} \) cumulative exposure index. Between 1980 and 2005, most of the urban and rural sites across the U.S. included in this study showed decreasing or zero trend for all three of these metrics. However, the magnitude of this trend varied greatly by exposure index. The largest downward trend in the 1-h and 8-h metrics listed above were observed in Southern California (>2%/yr downward trend) but the \( \text{W126} \) cumulative exposure metric showed large (>2%/yr) downward trends in many locations across the U.S. including Southern California, the Midwest and Northeast. By contrasting the 1980 – 2005 trends with more recent 1990 – 2005 trends, Lefohn et al. (2008) reported that a large number of sites (44%, 35% and 25% of sites for the 1-h, 8-h and \( \text{W126} \) metrics, respectively) shifted from a negative trend to no trend. These shifts in trends were attributed to slow changes in mid-level concentrations (i.e., 60-90 ppb) following a more rapid change in peak concentrations in the early years. A similar conclusion was drawn from nationwide \( \text{O}_3 \) data between 1980 – 2008 (Lefohn et al., 2010b), suggesting a shift in the \( \text{O}_3 \) distribution over this time period.

In contrast to the mostly urban observations included in the Pegues et al. (2012) study above, several studies focusing on rural western monitors have reported positive trends in \( \text{O}_3 \) concentrations over the last few decades. Jaffe and Ray (2007) investigated daytime (10 a.m. – 6 p.m. local time) \( \text{O}_3 \) concentrations at rural sites in the northern and western U.S. between 1987-2004. They found significantly positive trends in seven of the eleven sites selected ranging from 0.19 ppb/yr in Gothic, Colorado to 0.51 ppb/yr in Rocky Mountain NP, Colorado (mean trend of 0.26 ppb/yr at these seven sites). No significant trend was observed for the two sites in Alaska and one site each in Wyoming and Montana. Seasonal analyses were conducted on the sites having the longest records in Rocky Mountain NP, Yellowstone, NP and Lassen NP and positive trends were found for all seasons at all sites. As noted in the 2006 \( \text{O}_3 \) AQCD (U.S. EPA, 2006b), caution should be exercised in using trends calculated at national parks to infer contributions from distant sources either inside or outside of North America because of the influence of regional pollution (see Section 3.4 for a discussion of background \( \text{O}_3 \) concentrations and international transport).

Trends in baseline \( \text{O}_3 \) concentrations, defined as \( \text{O}_3 \) concentrations at a given site in the absence of strong local influences, were estimated by region and season in the U.S. in Chan and Vet (2010). The temperature-adjusted decadal (1997-2006) trends in estimated baseline \( \text{O}_3 \) varied substantially by region and season. In the Pacific coastal regions, the trends increased in all seasons except fall, but none of the trends were statistically significant. In the eastern U.S., negative trends were observed in all
seasons with the exception of (1) insignificant positive trends in northeast Maine in summer, fall and winter; (2) significant positive trends in the Midwest in winter; and 3) significant positive trends at one site in Vermont in the summer. The density of sites in the central and western U.S. were much lower than the coastal and eastern areas, but in general all sites showed trends that tended to be negative in the spring and fall but positive in the summer and winter.

Positive trends in marine boundary layer O₃ concentrations at several sites on the Pacific Coast have been reported by other sources in the literature. Parrish et al. (2009) used observations from multiple coastal sites in California and Washington and reported a positive annual mean trend of 0.34 ± 0.09 ppb/yr between the mid-1980s and 2007 (exact dates varied by site depending on available data). A seasonal stratification of the data at these sites showed the largest positive trend in the spring (0.46 ± 0.13 ppb/yr) with a smaller and non-significant positive trend during fall (0.12 ± 0.14 ppb/yr). These results agree with positive trends in springtime O₃ mixing ratios reported in an earlier study (Jaffe et al., 2003). Positive trends in O₃ measurements in the free troposphere above western North America at altitudes of 3-8 km (above sea level) during April and May of 1995 to 2008 were reported by Cooper et al. (2010) and discussed in Section 3.4.2 as they relate to intercontinental transport. Comparable trends were observed in the median as well as 5th, 33rd, 67th, and 95th percentiles of observations. Note, however, that these results relate to O₃ trends above ground level and not to surface O₃.

Extending back to the 19th Century, Volz and Kley (1988) report a series of historic O₃ measurements from Europe. Comparing these with more contemporary measurements, Parrish et al. (2009) report a 2 to 3 fold increase in boundary layer O₃ mixing ratios over the last 130 years with no indication of stabilization in recent years. Other long-term observations of global trends in the burden of tropospheric O₃ as they relate to climate change are discussed in Chapter 10, Section 10.3.3.1.

### 3.6.3.2 Hourly Variations

Ozone concentrations frequently possess a strong degree of diel variability resulting from daily patterns in temperature, sunlight, and precursor emissions. Other factors, such as the relative importance of transport versus local photochemical production and loss rates, the timing for entrainment of air from the nocturnal residual boundary layer, and the diurnal variability in mixing layer height also play a role in daily O₃ patterns. The 2006 O₃ AQCD (U.S. EPA, 2006b) looked at composite urban diel variations from April to October 2000 to 2004 and found 1-h maxima to occur in mid-afternoon and 1-h minima to occur in early morning. On a national basis, however, there was a high degree of spread in these times and caution was raised in extrapolating results from one city to another in determining the time of day for O₃ maxima and minima.

Urban diel variability in O₃ concentrations was investigated for the 20 focus cities listed in Table 3-9 using 1-h avg O₃ data from AQS. The year-round data set
described in Table 3-5 was used to compare diel patterns during cold months (October - April) and warm months (May - September) between 2007 and 2009. The warm-season data set, also described in Table 3-5, was used to compare weekday and weekend diel patterns. Figure 3-156 through Figure 3-160, in the supplemental material in Section 3.9.4, show these patterns for each of the 20 cities; examples for Atlanta, Boston, and Los Angeles are shown in Figure 3-54.

Figure 3-54  Diel patterns in 1-h avg O₃ for Atlanta, Boston and Los Angeles between 2007 and 2009.
In general, all the urban areas showed 1-h daily max concentrations occurring typically in the early afternoon. In all cities, these afternoon peaks were more pronounced in the warm months than in the cold months. However, a small peak was still present during the cold months. During warm months, the difference between the median daily extrema varied considerably by city. For example, in Los Angeles, the median 1-h daily min (10 ppb) at ~5:00 a.m. was 50 ppb less than the median 1-h daily max (60 ppb) at ~2:00 p.m. By contrast, in Boston, the median 1-h daily min (13 ppb) occurred at the same time, but was only 25 ppb less than the median 1-h daily max (38 ppb). Cities with large daily swings (>40 ppb) in median 1-h O₃ concentrations included Atlanta, Birmingham, Los Angeles, Phoenix, Pittsburgh, and Salt Lake City (Figure 3-156 through Figure 3-160 in Section 3.9.4). Cities with small daily swings (<25 ppb) in median 1-h O₃ concentrations included Boston, Minneapolis, San Francisco and Seattle (Figure 3-156 through Figure 3-160 in Section 3.9.4). These results are very similar to those found in the 2006 O₃ AQCD (U.S. EPA, 2006b) where many of these same urban areas were investigated. This supports the conclusions drawn in the previous O₃ review that diel patterns in O₃ have remained stable over the last 20 years, with times of occurrence of the daily maxima varying by no more than an hour from year to year.

Using the warm-season data, there was little difference in the median diel profiles for weekdays compared with weekends across all urban areas. This result stresses the complexity of O₃ formation and the importance of meteorology, entrainment, biogenic precursor emissions, and transport in addition to anthropogenic precursor emissions. There was, however, a subtle deviation between weekdays and weekends in the lower percentiles (1st and 5th) of the distribution. The lower end of the distribution tended to be lower on weekdays relative to weekends. This is consistent with analyses in the 2006 O₃ AQCD (U.S. EPA, 2006b) and is a result of lower traffic volumes on weekends relative to weekdays, leading to less NO emissions and O₃ titration on the weekends.

Seasonal and site-to-site variations in diel patterns within a subset of the urban focus areas presented here were investigated in the 2006 O₃ AQCD (U.S. EPA, 2006b). In northern cities, there was substantial seasonal variability in the diel patterns with higher extreme values in the O₃ distribution during the warm season than during the cold season. In southern cities, the seasonal differences in extreme O₃ concentrations were much smaller, and some of the highest O₃ concentrations in the Houston CSA were found outside of summer. The general pattern that emerged from investigating site-to-site variability within the urban areas was that peaks in 1-h avg O₃ concentrations are higher and tend to occur later in the day at downwind sites relative to sites located in the urban core. Differences between sites were not only related to the distance between them, but also depend on the presence or absence of nearby O₃ sources or sinks.

Rural diel variability in O₃ concentrations was investigated for the six rural focus areas listed in Table 3-11 using 1-h avg O₃ data from AQS. As with the urban analysis, the year-round data set described in Table 3-5 was used to compare diel patterns during cold months (October - April) and warm months (May - September).
between 2007 and 2009. The warm-season data set, also described in Table 3-5, was used to compare weekday and weekend diel patterns. Figure 3-55 shows the diel patterns for each of the rural areas investigated.

There was considerable variability in the diel patterns observed in the six rural focus areas. The selected mountainous eastern sites in ADSP, MMSP, and SMNP exhibited a generally flat profile with little hourly variability in the median concentration and the upper percentiles. In SMNP, there was some diel variability in the lower percentiles, with higher values during the daylight hours in the warm season data. This behavior was not present in the data coming from the two year-round monitors located at lower elevation sites (Sites C and Site D; see map in Figure 3-44), however, possibly resulting from differing impacts from local sources within SMNP. For the western rural areas, there was a clear diel pattern to the hourly O₃ data with a peak in concentration in the afternoon similar to those seen in the urban areas in Figure 3-54 and Figure 3-156 through Figure 3-160 in Section 3.9.4. This was especially obvious at the SBNF site which sits 90 km east of Los Angeles in the San Bernardino Mountains at an elevation of 1,384 meters. This site was located here to monitor O₃ transported downwind from major urban areas in the South Coast Air Basin. It had the highest 2007-2009 median 8-h daily max O₃ concentration of any AQS site in the Los Angeles CSA (see Figure 3-34), and is clearly impacted by the upwind urban plume which has sufficient time and sunlight to form O₃ from precursor emissions and concentrate the O₃ in the shallow boundary layer present at this elevation.

As with the urban analysis, there was little difference observed in the weekday and weekend diel profiles using the warm-season data, even down at the lower percentiles in the distribution. This is consistent with the regional nature of tropospheric O₃. Using the year-round data, there was an upward shift in the distribution going from the cold months to the warm months, and in some instances the general shape of the distribution changed considerably as was seen in several urban sites.
Note: Uses the year-round data set for the cold month/warm month comparison (left half) and the warm-season data set for the weekday/weekend comparison (right half). Mt. Mitchell SP, NC had no year-round monitors available for the cold month/warm month comparison.

Figure 3-55  Diel patterns in 1-h avg O₃ for six rural focus areas between 2007 and 2009.
3.6.4 Associations with Copollutants

Correlations between O₃ and other criteria pollutants are discussed in this section. Since O₃ is a secondary pollutant formed in the atmosphere from precursor emissions, its correlation with primary pollutants such as CO and NOₓ can vary substantially by location. Furthermore, O₃ formation is strongly influenced by meteorology, entrainment, and transport of both O₃ and O₃ precursors, resulting in a broad range in correlations with other pollutants which can vary substantially with season. This section focuses on correlations between O₃ and other criteria pollutants measured at the mostly urban AQS sites: a more detailed discussion of O₃ and O₃-precursor relationships is included in Section 3.2.4. To investigate correlations with copollutants, 8-h daily max O₃ from the year-round and warm-season data sets (Table 3-6 and Table 3-7) were compared with co-located 24-h avg CO, SO₂, NO₂, PM₂.₅ and PM₁₀ obtained from AQS for 2007-2009. Figure 3-56 and Figure 3-57 contain copollutant box plots of the correlation between co-located monitors for the year-round data set and the warm-season data set, respectively.

The year-round 8-h daily max O₃ data (Figure 3-56) had a very wide range in correlations with all the 24-h avg copollutants. A clearer pattern emerged when the data were stratified by season (bottom four plots in Figure 3-56) with mostly negative correlations in the winter and mostly positive correlations in the summer for all copollutants. In summer, the IQR in correlations is positive for all copollutants. However, the median seasonal correlations are still modest at best with the highest positive correlation at 0.52 for PM₂.₅ in the summer and the highest negative correlation at -0.38 for PM₂.₅ in the winter. Spring and fall lie in between with spring having a slightly narrower distribution than fall for all copollutants. The warm-season 8-h daily max O₃ data (Figure 3-57) shows a very similar distribution to the summer stratification of the year-round data due to their overlap in time periods (May-Sept and June-Aug, respectively).
Note: Year round (Top figure), and with seasonal stratification (Bottom four figures). Shown are the median (red line), mean (green star), inner-quartile range (box), 5th and 95th percentiles (whiskers) and extremes (black circles).

**Figure 3-56** Distribution of Pearson correlation coefficients for comparison of 8-h daily max $O_3$ from the year-round data set with co-located 24-h avg CO, SO$_2$, NO$_2$, PM$_{10}$ and PM$_{2.5}$ from AQS, 2007-2009.
Note: Shown are the median (red line), mean (green star), inner-quartile range (box), 5th and 95th percentiles (whiskers), and extremes (black circles).

**Figure 3-57** Distribution of Pearson correlation coefficients for comparison of 8-h daily max O$_3$ from the warm-season (May-Sept) data set with co-located 24-h avg CO, SO$_2$, NO$_2$, PM$_{10}$ and PM$_{2.5}$ from AQS, 2007-2009.

The seasonal fluctuations in correlations present in **Figure 3-56** result in part from the mixture of primary and secondary sources for the copollutants. For example, O$_3$ is a secondary pollutant whereas PM$_{2.5}$ has both primary and secondary origins and these two pollutants show the largest summertime/wintertime swing in correlation distributions. This situation arises because the secondary component to PM$_{2.5}$ is larger during the summer and is formed in conditions conducive to secondary O$_3$ formation. This results in positive correlations between O$_3$ and PM$_{2.5}$ during the summer. During the winter, photochemical production of O$_3$ is much smaller than during summer and O$_3$ comes mainly from aloft, i.e., the free troposphere (see Section 3.4.1.1 for further details). In addition, concentrations of PM$_{2.5}$ are much lower aloft. On relatively clean days, this can lead to high concentrations of O$_3$ and lower concentrations of primary pollutants such as PM$_{2.5}$ or NO. On relatively dirty days with elevated NO and PM$_{2.5}$, the intruding O$_3$ is readily titrated by NO in the boundary layer. These processes result in negative correlations between O$_3$ and PM$_{2.5}$ during the winter.
3.7 Chapter Summary

This section contains a summary of the major topics included in this chapter on the atmospheric chemistry and ambient concentrations of tropospheric O₃ and other related photochemical oxidants. This chapter has built upon information previously reported in the 2006 O₃ AQCD (U.S. EPA, 2006b) and includes updated material on: (1) physical and chemical processes of O₃ formation and removal; (2) atmospheric modeling; (3) background O₃ concentrations; (4) monitoring techniques and networks; and (5) ambient concentrations.

3.7.1 Physical and Chemical Processes

Ozone in the troposphere is a secondary pollutant; it is formed by photochemical reactions of precursor gases and is not directly emitted from specific sources. Ozone precursor gases originate from both anthropogenic and natural source categories. Ozone attributed to anthropogenic sources is formed in the atmosphere by photochemical reactions involving sunlight and precursor pollutants including VOCs, NOₓ, and CO. Ozone attributed to natural sources is formed through similar photochemical reactions involving natural emissions of precursor pollutants from vegetation, microbes, animals, biomass burning, lightning, and geogenic sources. The distinction between natural and anthropogenic sources of O₃ precursors is often difficult to make in practice, as human activities affect directly or indirectly emissions from what would have been considered natural sources during the pre-industrial era. The formation of O₃, other oxidants, and oxidation products from these precursors is a complex, nonlinear function of many factors including: (1) the intensity and spectral distribution of sunlight reaching the lower troposphere; (2) atmospheric mixing; (3) concentrations of precursors in the ambient air and the rates of chemical reactions of these precursors; and (4) processing on cloud and aerosol particles.

Ozone is present not only in polluted urban atmospheres but throughout the troposphere, even in remote areas of the globe. The same basic processes involving sunlight-driven reactions of NOₓ, VOCs, and CO contribute to O₃ formation throughout the troposphere. These processes also lead to the formation of other photochemical products, such as PAN, nitric acid, and sulfuric acid, and to other compounds, such as formaldehyde and other carbonyl compounds. In urban areas, NOₓ, VOCs, and CO are all important for O₃ formation. In non-urban vegetated areas, biogenic VOCs emitted from vegetation tend to be the most important precursor to O₃ formation. In the remote troposphere, methane (structurally the simplest VOC) and CO are the main carbon-containing precursors to O₃ formation. Ozone is subsequently removed from the troposphere through a number of gas phase reactions and deposition to surfaces.

Convective processes and turbulence transport O₃ and other pollutants both upward and downward throughout the planetary boundary layer and the free troposphere.
In many areas of the U.S., $O_3$ and its precursors can be transported over long distances, aided by vertical mixing. The transport of pollutants downwind of major urban centers is characterized by the development of urban plumes. Meteorological conditions, small-scale circulation patterns, localized chemistry, and mountain barriers can influence mixing on a smaller scale, resulting in frequent heterogeneous $O_3$ concentrations across individual urban areas.

### 3.7.2 Atmospheric Modeling

CTMs have been widely used to compute the interactions among atmospheric pollutants and their transformation products, and the transport and deposition of pollutants. They have also been widely used to improve basic understanding of atmospheric chemical processes and to develop control strategies. The domains of CTMs extend from a few hundred kilometers on a side to the entire globe.

Most major regional (i.e., sub-continental) scale air-related modeling efforts at EPA rely on the CMAQ modeling system. The horizontal domain for CMAQ typically extends over North America with efforts underway to extend it over the entire Northern Hemisphere. The upper boundary for CMAQ is typically set at 100 hPa, which is located on average at an altitude of ~16 km. CMAQ is most often driven by the MM5 mesoscale meteorological model, though it may be driven by other meteorological models including the WRF model and the RAMS. Other major air quality systems used for regional scale applications include CAMx and WRF/Chem.

Fine scale resolution is necessary to resolve features which can affect pollutant concentrations such as urban heat island circulation; sea breezes; mountain and valley breezes; and the nocturnal low-level jet. Horizontal domains are typically modeled by nesting a finer grid model within a larger domain model of coarser resolution. Caution must be exercised in using nested models because certain parameterizations like those for convection might be valid on a relatively coarse grid scale but may not be valid on finer scales and because incompatibilities can occur at the model boundaries. The use of finer resolution in CTMs will require advanced parameterizations of meteorological processes such as boundary layer fluxes, deep convection, and clouds, and necessitate finer-scale inventories of land use, source locations, and emission inventories.

Because of the large number of chemical species and reactions that are involved in the oxidation of realistic mixtures of anthropogenic and biogenic hydrocarbons, condensed mechanisms must be used to simplify atmospheric models. These mechanisms can be tested by comparison with smog chamber data. However, the existing chemical mechanisms often neglect many important processes such as the formation and subsequent reactions of long-lived carbonyl compounds, the incorporation of the most recent information on reactions of halogenated species, and heterogeneous reactions involving cloud droplets and aerosol particles. As a result, models such as CMAQ have had difficulties with capturing the regional nature of $O_3$. 

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episodes, in part because of uncertainty in the chemical pathways converting NO\textsubscript{X} to isoprene nitrates and recycling of NO\textsubscript{X}.

Errors in photochemical modeling arise from meteorological, chemical, and emissions inputs to the model. Algorithms must be used for simulating meteorological processes that occur on spatial scales smaller than the model’s grid spacing and for calculating the dependence of emissions on meteorology and time. Large uncertainties exist in the mechanism for oxidizing compounds of importance for atmospheric chemistry such as isoprene. Appreciable errors in emissions can occur if inappropriate assumptions are used in these parameterizations.

The performance of CTMs must be evaluated by comparison with field data as part of a cycle of model evaluations and subsequent improvements. Discrepancies between model predictions and observations can be used to point out gaps in current understanding of atmospheric chemistry and to spur improvements in parameterizations of atmospheric chemical and physical processes.

3.7.3 Background Concentrations

Because the mean tropospheric lifetime of O\textsubscript{3} is on the order of a few weeks, O\textsubscript{3} can be transported from continent to continent. The degree of influence from intercontinental transport varies greatly by location and time. For instance, high elevation sites are most susceptible to the intercontinental transport of pollution, particularly during spring. However, because the atmospheric chemistry of O\textsubscript{3} is quite complex and can be highly non-linear in environments close to sources of precursors, isolating the influence of intercontinental transport of O\textsubscript{3} and O\textsubscript{3} precursors on urban air quality is particularly problematic.

A number of recent studies indicate that natural sources such as wildfires and stratospheric intrusions and the intercontinental transport of pollution can significantly affect O\textsubscript{3} air quality in the United States. Two major modeling/field studies that focused on discerning the contributions of Asian emissions to air quality in California were the IONS-2010 and the CalNex studies conducted in May through June of 2010. Modeling and observational components of these studies found evidence for substantive contributions from stratospheric intrusions and Eurasian pollution to boundary layer O\textsubscript{3}. In particular, one modeling study found evidence of Asian contributions of 8-15 ppb in surface air during strong transport events in southern California. These contributions are in addition to contributions from dominant local pollution sources. Their results suggest that the influence of background sources on high O\textsubscript{3} concentrations at the surface is not always confined to high elevation sites. It is not clear to what extent the contributions inferred by these studies are likely to be found in other years, during other seasons, or in other locations. To gain a broader perspective and to isolate the influence of natural or transported O\textsubscript{3}, estimates from CTMs must be used. This is because observations within the U.S. (even at relatively remote monitoring sites) are impacted by transport from anthropogenic source regions within the U.S. borders.
In the context of a review of the NAAQS, it is useful to define background O\textsubscript{3} concentrations in a way that distinguishes between concentrations that result from precursor emissions that are relatively less controllable from those that are relatively more controllable through U.S. policies. For this assessment, three definitions of background O\textsubscript{3} concentrations are considered, including (1) NA background (simulated O\textsubscript{3} concentrations that would exist in the absence of anthropogenic emissions from the U.S., Canada and Mexico), (2) U.S. background (simulated O\textsubscript{3} concentrations that would exist in the absence of anthropogenic emissions from the U.S.), and (3) natural background (simulated O\textsubscript{3} concentrations in the absence of all anthropogenic emissions globally). Each definition of background O\textsubscript{3} includes contributions resulting from emissions from natural sources (e.g., stratospheric intrusion, wildfires, biogenic methane and more short-lived VOC emissions) throughout the globe. There is no chemical difference between background O\textsubscript{3} and O\textsubscript{3} attributable to U.S. or North American anthropogenic sources. However, to inform policy considerations regarding the current or potential alternative standards, it is useful to understand how total O\textsubscript{3} concentrations can be attributed to different sources.

Since background O\textsubscript{3} concentrations as defined above are a construct that cannot be directly measured, the range of background O\textsubscript{3} concentrations is estimated using CTMs. For the current assessment, recently published results from Zhang et al. (2011) using the GEOS-Chem model at 0.5° × 0.667° (~50 km × 50 km) horizontal resolution and Emery et al. (2012) using a GEOS-Chem/CAMx model (hereafter referred to as CAMx) at finer horizontal resolution (12 km × 12 km) were used. Results from these models represent the latest estimates for background O\textsubscript{3} concentrations documented in the peer-reviewed literature.

The main results from these modeling efforts can be summarized as follows. Simulated regional and seasonal means of base-case O\textsubscript{3} using both models generally agree to within a few ppb with observations for most of the United States. However, neither model is currently capable of simulating day specific base-case O\textsubscript{3} concentrations within reasonable bounds. Both models show background concentrations vary spatially and temporally. NA background concentrations are generally higher in spring than in summer across the United States. Simulated mean NA background concentrations are highest in the Intermountain West (i.e., at high altitude) in spring and in the Southwest in summer. Lowest estimates of NA background occur in the East in the spring and the Northeast in summer. NA background concentrations tend to increase with total (i.e., base case) O\textsubscript{3} concentrations at high elevation; but that tendency is not as clear at low elevations. Comparison of NA background and natural background indicate that methane is a major contributor to NA background O\textsubscript{3}, accounting for slightly less than half of the increase in background since the pre-industrial era; and whose relative contribution is projected to grow in the future. U.S. background concentrations are on average 2.6 ppb higher than NA background concentrations during spring and 2.7 ppb during summer across the U.S. with highest increases above NA background over the Northern Tier of New York State (19.1 ppb higher than NA background) in summer. High values for U.S. background are also found in other areas bordering Canada and
Mexico. Contributions to background O\textsubscript{3} can be episodic or non-episodic; high background concentrations are driven primarily by the episodic events such as stratospheric intrusions and wildfires. The most pronounced differences between these model results and observations are at the upper end of the concentration distribution, particularly at high elevations. In general, these model simulations provide a consistent representation of average background concentrations over seasons and broad spatial areas, but are not able to capture background concentrations at finer spatial (i.e., urban) and temporal (i.e., specific day) scales.

Note that the calculations of background concentrations presented in this chapter were formulated to answer the question, “what would O\textsubscript{3} concentrations be if there were no anthropogenic sources.” This is different from asking, “how much of the O\textsubscript{3} measured or simulated in a given area is due to background contributions.” Because of potentially strong non-linearities—particularly in many urban areas—these estimates by themselves should not be used to answer the second question posed above. The extent of these non-linearities will generally depend on location and time, the strength of concentrated sources, and the nature of the chemical regime. Further work is needed on how these estimates of background concentrations can be used to help determine the contributions of background sources of O\textsubscript{3} to urban concentrations.

### 3.7.4 Monitoring

The FRM for O\textsubscript{3} measurement is the CLM and is based on the detection of chemiluminescence resulting from the reaction of O\textsubscript{3} with ethylene gas. Almost all of the SLAMS that reported data to AQS from 2005 to 2009 used UV absorption photometer FEMs and greater than 96% of O\textsubscript{3} monitors met precision and bias goals during this period.

State and local monitoring agencies operate O\textsubscript{3} monitors at various locations depending on the area size and typical peak concentrations (expressed in percentages below, or near the O\textsubscript{3} NAAQS). SLAMS make up the ambient air quality monitoring sites that are primarily needed for NAAQS comparisons and include PAMS, NCore, and all other State or locally-operated stations except for the monitors designated as SPMs.

In 2010, there were 1250 SLAMS O\textsubscript{3} monitors reporting values to the EPA AQS database. Since O\textsubscript{3} levels decrease appreciably in the colder parts of the year in many areas, O\textsubscript{3} is required to be monitored at SLAMS monitoring sites only during the “ozone season” which varies by state. PAMS provides more comprehensive data on O\textsubscript{3} in areas classified as serious, severe, or extreme nonattainment for O\textsubscript{3}. There were a total of 119 PAMS reporting values to the EPA AQS database in 2009. NCore is a new multipollutant monitoring network currently being implemented to meet multiple monitoring objectives. Each state is required to operate at least one NCore site and the network will consist of about 60 urban and 20 rural sites nationwide.
CASTNET is a regional monitoring network established to assess trends in acidic deposition and also provides concentration measurements of O₃. CASTNET O₃ monitors operate year round and are primarily located in rural areas. At the beginning of 2010, there were 80 CASTNET sites located in, or near, rural areas. The NPS also operates a POMS network. The POMS couples the small, low-power O₃ monitor with a data logger, meteorological measurements, and solar power in a self contained system for monitoring in remote locations. Twenty NPS POMS reported O₃ data to AQS in 2010. A map of the current and proposed rural NCore sites, along with the CASTNET, and the NPS POMS sites was shown in Figure 3-22.

Satellite observations for O₃ are growing as a resource for many purposes, including model evaluation, assessing emissions reductions, pollutant transport, and air quality management. Satellite retrievals are conducted using the solar backscatter or thermal infrared emission spectra and a variety of algorithms. Most satellite measurement systems have been developed for measurement of the total O₃ column. Mathematical techniques have been developed and must be applied to derive information from these systems about tropospheric O₃. Satellite observations of O₃ precursors such as CO, NO₂, and HCHO are also available and useful for constraining model predictions of emissions of precursors and formation of O₃ during intercontinental transport.

3.7.5 Ambient Concentrations

Ozone is the only photochemical oxidant other than NO₂ that is routinely monitored and for which a comprehensive database exists. Other photochemical oxidants are typically only measured during special field studies. Therefore, the concentration analyses contained in this chapter have been limited to widely available O₃ data obtained directly from AQS for the period from 2007 to 2009.

The median 24-h avg, 8-h daily max, and 1-h daily max O₃ concentrations across all U.S. sites reporting data to AQS between 2007 and 2009 were 29, 40, and 44 ppb, respectively. Representing the upper end of the distribution, the 99th percentiles of these same metrics across all sites were 60, 80, and 94 ppb, respectively.

To investigate urban-scale O₃ variability, 20 focus cities were selected for closer analysis; these cities were selected based on their importance in O₃ epidemiologic studies and on their geographic distribution across the United States. Several of these cities had relatively little spatial variability in 8-h daily max O₃ concentrations (e.g., inter-monitor correlations ranging from 0.61 to 0.96 in Atlanta) while other cities exhibited considerably more variability in O₃ concentrations (e.g., inter-monitor correlations ranging from -0.06 to 0.97 for Los Angeles). The negative and near-zero correlations in Los Angeles were between monitors with a relatively large separation distance (>150 km), but even some of the closer monitor pairs were not very highly correlated. Similar to the correlation, the coefficient of divergence was found to be highly dependent on the urban area under investigation. As a result,
caution should be observed in using data from a sparse network of ambient O₃ monitors to approximate community-scale exposures.

To investigate rural-focused O₃ variability using AQS data, all monitors located within six rural monitoring areas were examined. These rural monitoring sites are impacted by transport of O₃ or O₃ precursors from upwind urban areas, and by local anthropogenic emissions within the rural areas such as emissions from motor vehicles, power generation, biomass combustion, or oil and gas operations. As a result, monitoring data from these rural locations are not unaffected by anthropogenic emissions. The rural area investigated with the largest number of available AQS monitors was Great Smoky Mountain National Park in NC and TN where the median warm-season 8-h daily max O₃ concentration ranged from 47 ppb at the lowest elevation site (elevation = 564 meters; site ID = 470090102) to 60 ppb at the highest elevation site (elevation = 2,021 meters; site ID = 471550102), with correlations between the 5 sites ranging from 0.78 to 0.92 and CODs ranging from 0.04 to 0.16. A host of factors may contribute to variations observed at these rural sites, including proximity to local O₃ precursor emissions, variations in boundary-layer influences, meteorology and stratospheric intrusion as a function of elevation, and differences in wind patterns and transport behavior due to local topography.

Since O₃ produced from emissions in urban areas is transported to more rural downwind locations, elevated O₃ concentrations can occur at considerable distances from urban centers. In addition, major sources of O₃ precursors such as highways, power plants, biomass combustion, and oil and gas operations are commonly found in rural areas, adding to the O₃ in these areas. Due to lower chemical scavenging in non-urban areas, O₃ tends to persist longer in rural than in urban areas which tends to lead to higher cumulative exposures in rural areas influenced by anthropogenic precursor emissions. The persistently high O₃ concentrations observed at many of these rural sites investigated here indicate that cumulative exposures for humans and vegetation in rural areas can be substantial and often higher than cumulative exposures to O₃ in urban areas.

Nation-wide surface-level O₃ concentrations in the U.S. have declined gradually over the last decade. A noticeable decrease in O₃ concentrations between 2003 and 2004, particularly in the eastern U.S., coincided with NOₓ emissions reductions resulting from implementation of the NOₓ SIP Call rule, which began in 2003 and was fully implemented in 2004. This rule was designed to reduce NOₓ emissions from power plants and other large combustion sources in the eastern United States. Downward trends in O₃ concentrations in the western U.S. have not been as substantial and several individual monitors have reported increases in O₃ concentrations when 2001-2003 design values are compared with 2008-2010 design values. Over a longer time scale, several observational studies investigating O₃ concentrations in the marine layer off the Pacific Coast of the U.S. have reported a steady rise in O₃ concentrations over the last few decades. And global scale observations have indicated a general rise in O₃ by a factor of 2 or more since pre-industrial times, as discussed in Chapter 10, Section 10.3.3.1.
Urban O₃ concentrations show a strong degree of diel variability resulting from daily patterns in temperature, sunlight, and precursor emissions. Other factors, such as the relative importance of transport versus local photochemical production and loss rates, the timing for entrainment of air from the nocturnal residual boundary layer, and the diurnal variability in mixing layer height also play a role in daily O₃ patterns. Urban diel variations investigated in this assessment show no substantial change in patterns since the 2006 O₃ AQCD (U.S. EPA, 2006b). The 1-h max concentrations tend to occur in mid-afternoon and 1-h min concentrations tend to occur in early morning, with more pronounced peaks in the warm months relative to the cold months. There is city-to-city variability in these times, however, and caution is raised in extrapolating results from one city to another in determining the time of day for O₃ maxima and minima.

Rural O₃ concentrations show a varying degree of diel variability depending on their location relative to larger urban areas. Three rural areas investigated in the east showed relatively little diel variability, reflecting the regional nature of O₃ in the east. In contrast, three rural areas investigated in the west did display diel variability resulting from their proximity to fresh urban emissions. These six areas investigated were selected as illustrative examples and do not represent all rural areas in the U.S.

Since O₃ is a secondary pollutant formed in the atmosphere from precursor emissions, its correlation with primary pollutants such as CO and NOₓ can vary substantially by location. Furthermore, O₃ formation is strongly influenced by meteorology, entrainment, and transport of both O₃ and O₃ precursors, resulting in a broad range in correlations with other pollutants which can vary substantially with season. In the copollutant analyses shown in Figure 3-56, the year-round 8-h daily max O₃ data exhibited a very wide range in correlations with all the criteria pollutants. A clearer pattern emerged when the data are stratified by season with mostly negative correlations in the winter and mostly positive correlations in the summer for all copollutants. The median seasonal correlations are modest at best with the highest positive correlation at 0.52 for PM₂.₅ in the summer and the highest negative correlation at -0.38 for PM₂.₅ in the winter. Therefore, statistical analyses that may be sensitive to correlations between copollutants need to take seasonality into consideration, particularly when O₃ is being investigated.

### 3.8 Supplemental Information on O₃ Model Predictions

This section contains supplemental comparisons between GEOS-Chem simulations of MDA8 O₃ concentrations with observations for 2006 from Zhang et al. (2011) and Emery et al. (2012). Further details on these simulations can be found in Section 3.4.3. Figure 3-58 through Figure 3-64 show GEOS-Chem predictions for the base model (i.e., model including all anthropogenic and natural sources; labeled as GEOS-Chem in the figure) and the NA background model (i.e., model including natural sources everywhere in the world and anthropogenic sources outside the U.S., Canada, and Mexico; labeled as NA background in the figure) along with
measurements obtained from selected CASTNET sites (labeled as Measurement in the figure). Figure 3-65 shows a comparison of GEOS-Chem output with measurements at Mt. Bachelor, OR, and Trinidad Head, CA, from March-August, 2006. Figure 3-66 shows a comparison of vertical profiles (mean ± 1 standard deviation) calculated by GEOS-Chem with ozonesondes launched at Trinidad Head, California and Boulder, Colorado. Figure 3-67 and Figure 3-68 show a comparison of AM3 simulations of individual stratospheric intrusions during May-June 2010. Figure 3-69 through Figure 3-74 show box plots for measurements at CASTNET sites, GEOS-Chem predictions from Zhang et al. (2011) and CAMx predictions from Emery et al. (2012) for both the base case and NA background. Figure 3-75 shows time series of AM3 simulations at approximately 2° × 2° at Gothic, Colorado, for 2006.

Source: Zhang et al. (2011).

Figure 3-58 Comparison of time series of measurements of MDA8 O3 concentrations at four CASTNET sites in the Northeast with GEOS-Chem predictions for the base case and for the North American background case during March-August, 2006.
Figure 3-59  Comparison of time series of measurements of MDA8 O₃ concentrations at four CASTNET sites in the Southeast with GEOS-Chem predictions for the base case and for the North American background case during March-August, 2006.

Source: Zhang et al. (2011).

Figure 3-60  Comparison of time series of measurements of MDA8 O₃ concentrations at four CASTNET sites in the Upper Midwest with GEOS-Chem predictions for the base case and for the North American background case during March-August, 2006.

Source: Zhang et al. (2011).
Figure 3-61  Comparison of time series of measurements of MDA8 O<sub>3</sub> concentrations at four CASTNET sites in the Intermountain West with GEOS-Chem predictions for the base case and the North American background case during March-August, 2006.

Source: Zhang et al. (2011).

Figure 3-62  Comparison of time series of measurements of MDA8 O<sub>3</sub> concentrations at four CASTNET sites in the Intermountain West with GEOS-Chem predictions for the base case and the North American background case during March-August, 2006.

Source: Zhang et al. (2011).
Figure 3-63  Comparison of time series of measurements of MDA8 O$_3$ concentrations at four CASTNET sites in the West with GEOS-Chem predictions for the base case and the North American background case during March-August, 2006.

Figure 3-64  Comparison of time series of measurements of MDA8 O$_3$ concentrations at three CASTNET sites and the Trinidad Head site in California with GEOS-Chem predictions for the base case and the North American background case during March-August, 2006.
Figure 3-65  Comparison of MDA8 O$_3$ predicted using GEOS-Chem at 0.5° × 0.667° (and 2° × 2.5° resolution; left figure only) with measurements at Mount Bachelor, Oregon (left); and at Trinidad Head, California (right) from March to August 2006.

Note: The letter ‘n’ refers to the number of ozonesonde profiles, and the model was sampled on the same days as the ozonesonde launches. As can be seen from the figure, variability in both model and measurements increases with altitude, but variability in the model results is much smaller at high altitudes than seen in the observations at both sites.

Source: Zhang et al. (2011).

Figure 3-66  Comparison of monthly mean (± 1 standard deviation) O$_3$ calculated GEOS-Chem (in red) with ozonesondes (in black) at Trinidad Head, CA (top) and Boulder, Colorado (bottom) during April and August 2006.
Note: Shows O₃ profiles at multiple sites as observed (black) by ozonesondes and simulated (red) by the GFDL AM3 model at ~50 × 50 km resolution. Also shown are observed relative humidity (gray) and AM3 estimates of O₃ concentrations in the absence of North American anthropogenic emissions (green) and the stratospheric contribution (blue). Model results have been interpolated to sonde pressure and averaged over 0.5-km altitude bins.

Figure 3-67 A deep stratospheric O₃ intrusion over California on May 28 to May 29, 2010.
Figure 3-68  A deep stratospheric O$_3$ intrusion over California on June 7 to June 12, 2010.
Note: Stippled boxes indicate North American background. GRS = Great Smoky NP (North Carolina and Tennessee); KEF = Kane Exp. Forest (Pennsylvania); SHN = Shenandoah NP (Virginia); CTH = Connecticut Hill Management Wildlife Area (New York); HWF = Huntington Wildlife Forest (New York); ACA = Acadia NP (Maine).
Source: Adapted from Emery et al. (2012) and Zhang et al. (2011).

Figure 3-69  Box plots showing maximum, interquartile range and minimum O₃ concentrations measured at CASTNET sites (black) in the Northeast and predictions from GEOS-Chem at ~50 × 50 km resolution (green) and CAMx at 12 × 12 km resolution (blue) for May-August 2006.
Figure 3-70  Box plots showing maximum, interquartile range and minimum O₃ concentrations measured at CASTNET sites (black) in the Southeast and predictions from GEOS-Chem at ~50 × 50 km resolution (green) and CAMx at 12 × 12 km resolution (blue) for May-August 2006.

Note: Stippled boxes indicate North American background. SND = Sand Mountain (Alabama); SUM = Sumatra (Florida); GAS = Georgia Station (Georgia); IRL = Indian River Lagoon (Florida).

Source: Adapted from Emery et al. (2012) and Zhang et al. (2011)
Note: Stippled boxes indicate North American background. KNZ = Konza Prairie (Kansas); ALC = Alabama-Coushatta (Texas); CHE = Cherokee Nation (Oklahoma); VOY = Voyageurs NP (Minnesota); PRK = Perkinstown (Wisconsin); CVL = Coffeeville (Mississippi); BVL = Bondsville (Illinois); ANA = Ann Arbor (Michigan); UVL = Unionville (Michigan).

Source: Adapted from Emery et al. (2012) and Zhang et al. (2011).

**Figure 3-71** Box plots showing maximum, interquartile range and minimum \( O_3 \) concentrations measured at CASTNET sites (black) in the Central U.S. and predictions from GEOS-Chem at \( \sim 50 \times 50 \text{ km} \) resolution (green) and CAMx at \( 12 \times 12 \text{ km} \) resolution (blue) for May-August 2006.
Figure 3-72  Box plots showing maximum, interquartile range and minimum O₃ concentrations measured at CASTNET sites (black) in the Northern Rockies and predictions from GEOS-Chem at ~50 × 50 km resolution (green) and CAMx at 12 × 12 km resolution (blue) for May-August 2006.
Figure 3-73  Box plots showing maximum, interquartile range and minimum O\textsubscript{3} concentrations measured at CASTNET sites (black) in the Southern Rockies and predictions from GEOS-Chem at \( \sim 50 \times 50 \) km resolution (green) and CAMx at \( 12 \times 12 \) km resolution (blue) for May-August 2006.
Figure 3-74  Box plots showing maximum, interquartile range and minimum \( \text{O}_3 \) concentrations measured at CASTNET sites (black) in the West and predictions from GEOS-Chem at \( \sim 50 \times 50 \) km resolution (green) and CAMx at \( 12 \times 12 \) km resolution (blue) for May-August 2006.
Note: Observed (black) and simulated by the GEOS-Chem (blue; horizontal resolution is 0.5° × 0.667°) and AM3 (red; horizontal resolution is approximately 2° × 2°) global models. Also shown are the model estimates for North American background (thin lines); the spike in mid-April likely corresponds to a stratospheric intrusion. The model correlations with observations, average (over the entire March through August period) total O₃ and North American background (NAbg) O₃ estimates, and their standard deviations (shown in parentheses) are presented in the lower left.

Figure 3-75  MDA8 O₃ in surface air at Gothic, Colorado for March through August 2006.

3.9  Supplemental Figures of Observed Ambient O₃ Concentrations

3.9.1  Ozone Monitor Maps for the Urban Focus Cities

This section contains supplemental maps showing the location of O₃ monitors reporting to AQS for each of the 20 urban focus cities introduced in Section 3.6.2.1. The monitors are delineated in the maps as year-round or warm-season based on their inclusion in the year-round data set and the warm-season data set discussed in Section 3.6.2.1. The maps also include the CSA/CBSA boundary selected for monitor inclusion, the location of urban areas and water bodies, the major roadway network, as well as the population gravity center based on the entire CSA/CBSA and the individual focus city boundaries. Population gravity center is calculated from the average longitude and latitude values for the input census tract centroids and represents the mean center of the population in a given area. Census tract centroids are weighted by their population during this calculation.
Figure 3-76  Map of the Atlanta, Georgia, CSA including $O_3$ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-77  Map of the Baltimore, Maryland, CSA including $O_3$ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-78  Map of the Birmingham, Alabama, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-79  Map of the Boston, Massachusetts, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-80  Map of the Chicago, Illinois, CSA including O$_3$ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-81  Map of the Dallas, Texas, CSA including O$_3$ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-82  Map of the Denver, Colorado, CSA including $O_3$ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-83  Map of the Detroit, Michigan, CSA including $O_3$ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-84  Map of the Houston, Texas, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-85  Map of the Los Angeles, California, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-86  Map of the Minneapolis, Minnesota, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-87  Map of the New York City, New York, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-88  Map of the Philadelphia, Pennsylvania, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-89  Map of the Phoenix, Arizona, CBSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-90  Map of the Pittsburgh, Pennsylvania, CSA including $O_3$ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-91  Map of the Salt Lake City, Utah, CSA including $O_3$ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-92  Map of the San Antonio, Texas, CBSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-93  Map of the San Francisco, California, CSA including O₃ monitor locations, population gravity centers, urban areas, and major roadways.
Figure 3-94  Map of the Seattle, Washington, CSA including O$_3$ monitor locations, population gravity centers, urban areas, and major roadways.

Figure 3-95  Map of the St. Louis, Missouri, CSA including O$_3$ monitor locations, population gravity centers, urban areas, and major roadways.
3.9.2 Ozone Concentration Box Plots for the Urban Focus Cities

This section contains box plots depicting the distribution of 2007-2009 warm-season 8-h daily max O₃ data from each individual monitor in the 20 urban focus cities introduced in Section 3.6.2.1. Monitor information including the AQS site id, the years containing qualifying data between 2007 and 2009, and the number of 8-h daily max O₃ observations included in the data set are listed next to the box plot. Statistics including the mean, standard deviation (SD), median and inner quartile range (IQR) are also shown for each monitor with the site letter corresponding to the sites listed in the figures above.

![Diagram of box plots for Atlanta CSA](image-url)

**Figure 3-96** Site information, statistics and box plots for 8-h daily max O₃ from AQS monitors meeting the warm-season data set inclusion criteria within the Atlanta, Georgia, CSA.
Figure 3-97  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Baltimore, Maryland, CSA.

Figure 3-98  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Birmingham, Alabama, CSA.
Figure 3-99  Site information, statistics and box plots for 8-h daily max O₃ from AQS monitors meeting the warm-season data set inclusion criteria within the Boston, Massachusetts, CSA.

Figure 3-100  Site information, statistics and box plots for 8-h daily max O₃ from AQS monitors meeting the warm-season data set inclusion criteria within the Chicago, Illinois, CSA.
Figure 3-101  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Dallas, Texas, CSA.

Figure 3-102  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Denver, Colorado, CSA.
Figure 3-103  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Detroit, Michigan, CSA.

Figure 3-104  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Houston, Texas, CSA.
Figure 3-105 Site information, statistics and box plots for 8-h daily max \(O_3\) from AQS monitors meeting the warm-season data set inclusion criteria within the Los Angeles, California, CSA.
Figure 3-106 Site information, statistics and box plots for 8-h daily max \(O_3\) from AQS monitors meeting the warm-season data set inclusion criteria within the Minneapolis, Minnesota, CSA.

Figure 3-107 Site information, statistics and box plots for 8-h daily max \(O_3\) from AQS monitors meeting the warm-season data set inclusion criteria within the New York City, New York, CSA.
Figure 3-108 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Philadelphia, Pennsylvania, CSA.

Figure 3-109 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Phoenix, Arizona, CBSA.
Figure 3-110  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Pittsburgh, Pennsylvania, CSA.

Figure 3-111  Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Salt Lake City, Utah, CSA.
Figure 3-112 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the San Antonio, Texas, CBSA.

Figure 3-113 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the San Francisco, California, CSA.
Figure 3-114 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the Seattle, Washington, CSA.

Figure 3-115 Site information, statistics and box plots for 8-h daily max $O_3$ from AQS monitors meeting the warm-season data set inclusion criteria within the St. Louis, Missouri, CSA.
3.9.3 Ozone Concentration Relationships for the Urban Focus Cities

This section contains histograms and contour matrices of the Pearson correlation coefficient (R) and the coefficient of divergence (COD) between 8-h daily max O₃ concentrations from each monitor pair within the 20 urban focus cities discussed in Section 3.6.2.1. These figures also contain scatter plots of R and COD as a function of straight-line distance between monitor pairs.

Figure 3-116 Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Atlanta, Georgia, CSA.

Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

**Figure 3-117** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Baltimore, Maryland, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

**Figure 3-118** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Birmingham, Alabama, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of $R$.

Figure 3-119  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Boston, Massachusetts, CSA.
Figure 3-120 Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Chicago, Illinois, CSA.

Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

Figure 3-121  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Dallas, Texas, CSA.
Figure 3-122 Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Denver, Colorado, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of $R$.

**Figure 3-123** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Detroit, Michigan, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

**Figure 3-124** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Houston, Texas, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

Figure 3-125  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Los Angeles, California, CSA.
Figure 3-126 Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Minneapolis, Minnesota, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of $R$.

**Figure 3-127** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the New York City, New York, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

Figure 3-128  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Philadelphia, Pennsylvania, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

Figure 3-129  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Phoenix, Arizona, CBSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of $R$.

Figure 3-130  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Pittsburgh, Pennsylvania, CSA.
Figure 3-131  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Salt Lake City, Utah, CSA.

Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of $R$.  

**Figure 3-132** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the San Antonio, Texas, CBSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of R.

Figure 3-133  Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the San Francisco, California, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of $R$.

**Figure 3-134** Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Seattle, Washington, CSA.
Figure 3-135 Pair-wise monitor correlations expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the St. Louis, Missouri, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-136** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Atlanta, Georgia, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

Figure 3-137 Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Baltimore, Maryland, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

Figure 3-138 Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Birmingham, Alabama, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-139** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Boston, Massachusetts, CSA.
Figure 3-140 Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Chicago, Illinois, CSA.
Figure 3-141  Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Dallas, Texas, CSA.
Figure 3-142  Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Denver, Colorado, CSA.

Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-143** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Detroit, Michigan, CSA.
Figure 3-144  Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Houston, Texas, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-145** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Los Angeles, California, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-146** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Minneapolis, Minnesota, CSA.
Figure 3-147  Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the New York City, New York, CSA.

Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-148** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Philadelphia, Pennsylvania, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-149** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the Phoenix, Arizona, CBSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-150** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Pittsburgh, Pennsylvania, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

Figure 3-151 Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the Salt Lake City, Utah, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

Figure 3-152  Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O$_3$ in the San Antonio, Texas, CBSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-153** Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the San Francisco, California, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

Figure 3-154 Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max $O_3$ in the Seattle, Washington, CSA.
Note: The colors in the histogram bins correspond to the levels of the contour matrix. The histogram includes the number of monitor pairs per bin and the contour matrix includes the numeric values of COD.

**Figure 3-155**  Pair-wise monitor coefficient of divergence expressed as a histogram (top), contour matrix (middle) and scatter plot versus distance between monitors (bottom) for 8-h daily max O₃ in the St. Louis, Missouri, CSA.
3.9.4 Hourly Variations in $O_3$ for the Urban Focus Cities

This section contains diel plots of 1-h avg $O_3$ data to supplement the discussion on hourly variations in $O_3$ concentrations from Section 3.6.3.2 using data from the 20 urban focus cities first introduced in Section 3.6.2.1. Comparisons are made between cold months (October-April) and warm months (May-September), using the year-round data set, and between weekdays (Mon-Fri) and weekends (Sat-Sun) using the warm-season data set.
Note: No year-round monitors were available for the cold month/warm month comparison in the Atlanta CSA.

Figure 3-156  Diel patterns in 1-h avg $O_3$ for select CSAs between 2007 and 2009 using the year-round data set for the cold month/warm month comparison (left half) and the warm-season data set for the weekday/weekend comparison (right half).
Note: No year-round monitors were available for the cold month/warm month comparison in the Detroit CSA.

Figure 3-157  Diel patterns in 1-h avg $O_3$ for select CSAs between 2007 and 2009 using the year-round data set for the cold month/warm month comparison (left half) and the warm-season data set for the weekday/weekend comparison (right half).
Figure 3-158  Diel patterns in 1-h avg O₃ for select CSAs between 2007 and 2009 using the year-round data set for the cold month/warm month comparison (left half) and the warm-season data set for the weekday/weekend comparison (right half).
Figure 3-159  Diel patterns in 1-h avg $O_3$ for select CSAs/CBSAs between 2007 and 2009 using the year-round data set for the cold month/warm month comparison (left half) and the warm-season data set for the weekday/weekend comparison (right half).
Figure 3-160  Diel patterns in 1-h avg $O_3$ for select CSAs/CBSAs between 2007 and 2009 using the year-round data set for the cold month/warm month comparison (left half) and the warm-season data set for the weekday/weekend comparison (right half).
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4 EXPOSURE TO AMBIENT OZONE

4.1 Introduction

The 2006 O₃ AQCD (U.S. EPA, 2006b) evaluated O₃ concentrations and exposures in multiple microenvironments, discussed methods for estimating personal and population exposure via monitoring and modeling, analyzed relationships between personal exposure and ambient concentrations, and discussed the implications of using ambient O₃ concentrations as an estimate of exposure in epidemiologic studies. This chapter presents new information regarding exposure to ambient O₃ which builds upon the body of evidence presented in the 2006 O₃ AQCD. A brief summary of findings from the 2006 O₃ AQCD is presented at the beginning of each section as appropriate.

Section 4.2 presents general exposure concepts describing the relationship between ambient pollutant concentrations and personal exposure. Section 4.3 describes exposure measurement techniques and studies that measured personal, ambient, indoor, and outdoor concentrations of O₃ and related pollutants. Section 4.4 presents material on parameters relevant to exposure estimation, including activity patterns, averting behavior, and population proximity to ambient monitors. Section 4.5 describes techniques for modeling local O₃ concentrations, air exchange rates, microenvironmental concentrations, and personal and population exposure. Section 4.6 discusses the implications of using ambient O₃ concentrations to estimate exposure in epidemiologic studies, including several factors that contribute to exposure error.

4.2 General Exposure Concepts

A theoretical model of personal exposure is presented to highlight measurable quantities and the uncertainties that exist in this framework. An individual’s time-integrated total exposure to O₃ can be described based on a compartmentalization of the person’s activities throughout a given time period:

\[ E_T = \int C_j dt \]

Equation 4-1

where \( E_T \) = total exposure over a time-period of interest, \( C_j \) = airborne O₃ concentration at microenvironment \( j \), and \( dt \) = portion of the time-period spent in microenvironment \( j \). Equation 4-1 can be decomposed into a model that accounts for exposure to O₃, of ambient \( (E_a) \) and nonambient \( (E_{na}) \) origin of the form:
\[ E_T = E_a + E_{na} \]

Equation 4-2

Ambient O\textsubscript{3} is formed through photochemical reactions involving NO\textsubscript{X}, VOCs, and other compounds, as described in Chapter 3. Although nonambient sources of O\textsubscript{3} exist, such as O\textsubscript{3} generators and laser printers, these sources are specific to individuals and may not be important sources of population exposure. Ozone concentrations generated by ambient and nonambient sources are subject to spatial and temporal variability that can affect estimates of exposure and influence epidemiologic effect estimates. Exposure parameters affecting interpretation of epidemiologic studies are discussed in Section 4.6.

This assessment focuses on the ambient component of exposure because this is more relevant to the NAAQS review. Assuming steady-state outdoor conditions, \( E_a \) can be expressed in terms of the fraction of time spent in various outdoor and indoor microenvironments (U.S. EPA, 2006c; Wilson et al., 2000):

\[ E_a = \sum f_o C_o + \sum f_i F_{inf,i} C_{o,i} \]

Equation 4-3

where \( f = \) fraction of the relevant time period (equivalent to \( dt \) in Equation 4-1), subscript \( o = \) index of outdoor microenvironments, subscript \( i = \) index of indoor microenvironments, subscript \( o,i = \) index of outdoor microenvironments adjacent to a given indoor microenvironment \( i \), and \( F_{inf,i} = \) infiltration factor for indoor microenvironment \( i \). Equation 4-3 is subject to the constraint \( \Sigma f_o + \Sigma f_i = 1 \) to reflect the total exposure over a specified time period, and each term on the right hand side of the equation has a summation because it reflects various microenvironmental exposures. Here, “indoors” refers to being inside any aspect of the built environment, e.g., home, office buildings, enclosed vehicles (automobiles, trains, buses), and/or recreational facilities (movie theaters, restaurants, bars). “Outdoor” exposure can occur in parks or yards, on sidewalks, and on bicycles or motorcycles. Assuming steady state ventilation conditions, the infiltration factor is a function of the penetration (\( P \)) of O\textsubscript{3} into the microenvironment, the air exchange rate (\( a \)) of the microenvironment, and the rate of O\textsubscript{3} loss (\( k \)) in the microenvironment; \( F_{inf} = Pa/(a + k) \).

In epidemiologic studies, the central-site ambient concentration, \( C_o \), is often used in lieu of outdoor microenvironmental data to represent these exposures based on the availability of data. Thus it is often assumed that \( C_o = C_a \) and that the fraction of time spent outdoors can be expressed cumulatively as \( f_o \); the indoor terms still retain a summation because infiltration differs among different microenvironments. If an epidemiologic study employs only \( C_o \), then the assumed model of an individual’s
exposure to ambient O₃, first given in Equation 4-3, is re-expressed solely as a function of Ca:

\[ E_a = \left( f_o + \sum f_i F_{inf_i} \right) C_a \]

Equation 4-4

The spatial variability of outdoor O₃ concentrations due to meteorology, topography, varying precursor emissions and O₃ formation rates; the design of the epidemiologic study; and other factors determine whether or not Equation 4-4 is a reasonable approximation for Equation 4-3. These equations also assume steady-state microenvironmental concentrations. Errors and uncertainties inherent in use of Equation 4-4 in lieu of Equation 4-3 are described in Section 4.6 with respect to implications for interpreting epidemiologic studies. Epidemiologic studies often use concentration measured at a central site monitor to represent ambient concentration; thus \( \alpha \), the ratio between personal exposure to ambient O₃ and the ambient concentration of O₃, is defined as:

\[ \alpha = \frac{E_a}{C_a} \]

Equation 4-5

Combination of Equation 4-4 and Equation 4-5 yields:

\[ \alpha = f_o + \sum f_i F_{inf_i} \]

Equation 4-6

where \( \alpha \) varies between 0 and 1. If a person’s exposure occurs in a single microenvironment, the ambient component of a microenvironmental O₃ concentration can be represented as the product of the ambient concentration and \( F_{inf} \). The U.S.EPA (2006c) noted that time-activity data and corresponding estimates of \( F_{inf} \) for each microenvironmental exposure are needed to compute an individual’s \( \alpha \) with accuracy. In epidemiologic studies, \( \alpha \) is assumed to be constant in lieu of time-activity data and estimates of \( F_{inf} \), which can vary with building and meteorology-related air exchange characteristics. If important local outdoor sources and sinks exist that are not captured by central site monitors, then the ambient component of the local outdoor concentration may be estimated using dispersion models, land use regression models, receptor models, fine scale CTMs or some combination of these techniques. These techniques are described in Section 4.5.
4.3 Exposure Measurement

This section describes techniques that have been used to measure microenvironmental concentrations of O₃ and personal O₃ exposures as well as results of studies using those techniques. Previous studies from the 2006 O₃ AQCD are described along with newer studies that evaluate indoor-outdoor concentration relationships, associations between personal exposure and ambient monitor concentration, and multipollutant exposure to other pollutants in conjunction with O₃. Tables are provided to summarize important study results.

4.3.1 Personal Monitoring Techniques

As described in the 2006 O₃ AQCD, passive samplers have been developed and deployed to measure personal exposure to O₃ (Grosjean and Hisham, 1992; Kanno and Yanagisawa, 1992). Widely used versions of these samplers utilize a filter coated with nitrite, which is converted to nitrate by O₃ and then quantified by a technique such as ion chromatography (Koutrakis et al., 1993). This method has been licensed and marketed by Ogawa, Inc., Japan (Ogawa & Co, 2007). The cumulative sampling and the detection limit of the passive badges makes them mainly suitable for monitoring periods of 24 hours or greater, which limits their ability to measure short-term daily fluctuations in personal O₃ exposure. Longer sampling periods give lower detection limits; use of the badges for a 6-day sampling period yields a detection limit of 1 ppb, while a 24-hour sampling period gives a detection limit of approximately 5-10 ppb. This can result in a substantial fraction of daily samples being below the detection limit (Sarnat et al., 2006a; Sarnat et al., 2005), which is a limitation of past and current exposure studies. Development of improved passive samplers capable of shorter-duration monitoring with lower detection limits would enable more precise characterization of personal exposure in multiple microenvironments with relatively low participant burden.

The nitrite-nitrate conversion reaction has also been used as the basis for an active sampler consisting of a nitrite-coated glass tube through which air is drawn by a pump operating at 65 mL/min (Geyh et al., 1999; Geyh et al., 1997). The reported detection limit is 10 ppb-h, enabling the quantification of O₃ concentrations measured over a few hours rather than a full day (Geyh et al., 1999).

A portable active O₃ monitor based on the UV photometric technique used for stationary monitors (Chapter 3) has recently been approved as a FEM (75 FR 22126). This monitor includes a Naﬁon tube in the inlet line to equilibrate humidity, reducing the effect of humidity changes in different microenvironments (Wilson and Birks, 2006). Its size and weight (approximately 10×20×30 cm; 2 kg) make it suitable for use in a backpack configuration. The monitors are currently used by the U.S. National Park service as stationary monitors to measure O₃ in several national parks (Chapter 3). Future improvements and continued miniaturization of real-time O₃
monitors can yield highly time-resolved personal measurements to further evaluate O₃ exposures in specific situations, such as near roadways or while in transit.

### 4.3.2 Indoor-Outdoor Concentration Relationships

Several studies summarized in the 2006 O₃ AQCD, along with some newer studies, have evaluated the relationship between indoor O₃ concentration and the O₃ concentration immediately outside the indoor microenvironment. These studies show that the indoor concentration is often substantially lower than the outdoor concentration unless indoor sources are present. Low indoor O₃ concentrations can be explained by reactions of O₃ with surfaces and airborne constituents. Studies have shown that O₃ is deposited onto indoor surfaces where reactions produce secondary pollutants such as formaldehyde (Reiss et al., 1995b; Reiss et al., 1995a). However, the indoor-outdoor relationship is greatly affected by the air exchange rate; under conditions of high air exchange rate, such as open windows, the indoor O₃ concentration may approach the outdoor concentration. Thus, in rooms with open windows, the indoor-outdoor (I/O) ratio may approach 1.0. **Table 4-1** summarizes I/O ratios and correlations reported by older and more recent studies, with discussion of individual studies in the subsequent text. In general, I/O ratios range from about 0.1 to 0.4, with some evidence for higher ratios during the O₃ season when concentrations are higher.

Ozone concentrations near and below the monitor detection limit cause uncertainty in I/O ratios, because small changes in low concentration values cause substantial variation in resulting ratios. This problem is particularly acute in the non-ozone season when ambient O₃ concentrations are low. Further improvements in characterization of microenvironmental O₃ concentrations and I/O ratios will rely on improved monitoring. Until new monitoring techniques are available and can be used in the field, past studies summarized in the 2006 O₃ AQCD remain relevant to consider along with more recent studies in evaluating the relationship between indoor and outdoor O₃ concentrations.
### Table 4-1  Relationships between indoor and outdoor $O_3$ concentration.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Years/ Season</th>
<th>Population</th>
<th>Sample duration</th>
<th>Ratio$^a$</th>
<th>Correlation</th>
<th>Micro-environment</th>
<th>Comment</th>
<th>Concentration/ Detection limit (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Upland, Southern California</td>
<td>June - Sept 1995 and May 1996</td>
<td>Children</td>
<td>6 days</td>
<td>0.24</td>
<td></td>
<td>Home</td>
<td></td>
<td>Indoor 11.8 (9.2) Outdoor 48.2 (12.2)</td>
</tr>
<tr>
<td>Geyh et al. (2000)</td>
<td>Mountain Communities, Southern California</td>
<td>June - Sept 1995 and May 1996</td>
<td>Children</td>
<td>6 days</td>
<td>0.36</td>
<td>NR</td>
<td></td>
<td>Indoor 21.4 (14.8) Outdoor 60.1 (17.1)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Upland &amp; Mountain</td>
<td>Entire period</td>
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<td></td>
<td></td>
<td>Indoor 2.8 (4.2) Outdoor 35.7 (9.3)</td>
</tr>
<tr>
<td>Study</td>
<td>Location</td>
<td>Years/ Season</td>
<td>Population</td>
<td>Sample duration</td>
<td>Ratio&lt;sup&gt;a&lt;/sup&gt;</td>
<td>Correlation</td>
<td>Micro-environment Comment</td>
<td>Concentration/Detection limit (ppb)</td>
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<tr>
<td>Avol et al. (1998a)</td>
<td>Southern California</td>
<td>Feb-Dec 1994</td>
<td>NR</td>
<td>24 h</td>
<td>0.37</td>
<td>0.58</td>
<td>Home</td>
<td>Indoor: 13 (12)</td>
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<tr>
<td></td>
<td></td>
<td>Summer (late June – late Sept)</td>
<td>NR</td>
<td></td>
<td>0.43</td>
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<td></td>
<td>Outdoor: 37 (19)</td>
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<td></td>
<td></td>
<td>Non-summer</td>
<td>NR</td>
<td></td>
<td>0.32</td>
<td></td>
<td></td>
<td>LOD: 5</td>
<td></td>
</tr>
<tr>
<td>Romieu et al. (1998a)</td>
<td>Mexico City, Mexico</td>
<td>Sept 1993 - July 1994</td>
<td>Children</td>
<td>7 or 14 days</td>
<td>0.20</td>
<td>0.15</td>
<td>NR</td>
<td>Indoor: 7-day: 5 (4)</td>
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<td>SD: 0.18</td>
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<td>14-day: 7 (5)</td>
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<td>Range: 0.01-1.00</td>
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<td>Outdoor: 7-day: 27 (10)</td>
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<td>SD: 0.29</td>
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<td>14-day: 37 (12)</td>
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<td>IQR: 0.15</td>
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<td>LOD: 7-day: 2.4-2.9</td>
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<td>Range: 0.01-1.00</td>
<td></td>
<td>14-day: 1.2-3.5</td>
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<tr>
<td>Lee et al. (2004a)</td>
<td>Nashville, TN</td>
<td>Summer 1994</td>
<td>Children</td>
<td>1 week</td>
<td>0.1</td>
<td>0.18</td>
<td>Home</td>
<td>Indoor: Range of Weekly Means: 1.6-3.1</td>
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<td>Fraction above LOD, Range:</td>
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<td>14-87%</td>
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<td>Outdoor: Range of Weekly Means: 18.6-26.9</td>
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<td>100%</td>
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</tbody>
</table>

<sup>a</sup> Home Ratio: each pair of values

<sup>b</sup> Range: 0.01-1.00
<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Years/ Season</th>
<th>Population</th>
<th>Sample duration</th>
<th>Ratio*</th>
<th>Correlation</th>
<th>Micro-environment</th>
<th>Comment</th>
<th>Concentration/Detection limit (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Héroux et al.</td>
<td>Regina, Saskatchewan, Canada</td>
<td>Summer 2007</td>
<td>All age groups</td>
<td>5 days</td>
<td>0.13</td>
<td>NR</td>
<td>Home</td>
<td>Ratio: Indoor mean/outdoor mean</td>
<td>Mean (SD)</td>
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<td>(2010)</td>
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<td>Indoor 0.7 (0.8)</td>
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<td>Outdoor 5.4 (1.3)</td>
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<td>LOD: NR</td>
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<td>Fraction above field LOD: Indoor: &lt;50%</td>
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<td>Outdoor: NR</td>
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<tr>
<td>Liu et al.</td>
<td>Toronto, Canada</td>
<td>Winter 1992</td>
<td>All age groups</td>
<td>1 week</td>
<td>0.07</td>
<td>SD: 0.10</td>
<td>Indoor</td>
<td>Ratio: each pair of values</td>
<td>Mean (SD)</td>
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<td>(1995)</td>
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<td>SD: 0.32</td>
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<td>SD: 0.54</td>
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<td>Romieu et al. (1998a)</td>
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<td>Sept 1993 - July 1994</td>
<td>Children (during school hours)</td>
<td>5 h/day, 5 days, 10 days</td>
<td>0.30-0.40</td>
<td>NR</td>
<td>School</td>
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<td>Indoor: 22 (16.1), 10-day: 22 (16.0) LOD: 5 day: 3.6-4.1 10 day: 0.3-1.6 Outdoor: 5-day: 73 (21.5) 10-day: 56 (17.9)</td>
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*Estimated from Figure 1, showing two of eight schools
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<th>Concentration/Detection limit (ppb)</th>
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<td>July 2009</td>
<td>All age groups</td>
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<td>Dec 2009</td>
<td></td>
<td></td>
<td>0.30</td>
<td>NR</td>
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<td>No heating or air conditioning</td>
<td>Mean* Indoor: 2, Outdoor: 18.2</td>
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<td>Riediker et al. (2003)</td>
<td>North Carolina</td>
<td>Aug - Oct 2001</td>
<td>Adults</td>
<td>9 h</td>
<td>0.51</td>
<td>p-value: 0.000</td>
<td>Vehicle</td>
<td>Ratio: Indoor mean/outdoor mean</td>
<td>Mean (SD) In-vehicle 11.7 (15.9) Roadside 22.8 (13.3) LOD: 10</td>
</tr>
</tbody>
</table>

⁰Mean value unless otherwise indicated
¹Median

LOD = limit of detection; NR = not reported; SD = standard deviation
Geyh et al. (2000) measured 6-day indoor and outdoor concentrations at 116 homes in southern California, approximately equally divided between the community of Upland and several mountain communities. The extended sampling period resulted in a relatively low detection limit (1 ppb) for the passive samplers used and a large fraction of samples above the detection limit; over 80% of the indoor samples and virtually all of the outdoor samples were above the detection limit. The Upland homes were nearly all air-conditioned, while the mountain community homes were ventilated by opening windows. During the O₃ season, the indoor O₃ concentration averaged over all homes was approximately 24% of the overall mean outdoor concentration in Upland (11.8 versus 48.2 ppb), while in the mountain communities, the indoor concentration was 36% of the outdoor concentration (21.4 versus 60.1 ppb). This is consistent with the increased air exchange rate expected in homes using window ventilation. In the non-ozone season, when homes are likely to be more tightly closed to conserve heat, the ratios of indoor to outdoor concentration were 0.15 (3.2 versus 21.1 ppb) and 0.08 (2.8 versus 35.7 ppb) in Upland and the mountain communities, respectively. Avol et al. (1998a) observed a mean I/O ratio of 0.37 for 239 matched 24-hour samples collected between February and December at homes in the Los Angeles area. The I/O ratio during summer was somewhat higher than the non-summer I/O ratio (0.43 versus 0.32). The authors also reported a correlation of 0.58 between the 24-h avg indoor concentration and the outdoor concentration, which was only slightly higher than the correlation between the indoor concentration and the concentration at the neighborhood fixed-site monitor (0.49).

Substantially higher summer I/O ratios were reported in a study in Toronto, Canada (Liu et al., 1995), which found summer I/O ratios of 0.30-0.43, in comparison with a winter I/O ratio of 0.07. Romieu et al. (1998a) reported a mean I/O ratio of 0.20 in 145 homes in Mexico City, Mexico, for 7- or 14-day cumulative samples, with the highest ratios observed in homes where windows were usually open during the day and where there was no carpeting or air filters. Studies conducted in Nashville, TN, and Regina, Saskatchewan (Canada) reported mean residential I/O ratios of approximately 0.1 (Héroux et al., 2010; Lee et al., 2004a).

Investigators have also measured I/O ratios for non-residential microenvironments, including schools and vehicles. Romieu et al. (1998a) reported that O₃ concentrations measured during school hours (10-day cumulative sample, 5 h/day) were 30-40% of concentrations immediately outside the schools, while overall I/O ratios (14-day cumulative sample, 24 h/day) were approximately 15%. The authors attribute this discrepancy to increased air exchange during the school day due to opening doors and windows. Air exchange was also identified as an important factor in the I/O ratios measured at eight French schools (Blondeau et al., 2005). In this study, the I/O ratios based on simultaneous continuous measurements ranged from 0-0.45, increasing with decreasing building tightness. A historical library building in Prague, Czech Republic with no heating or air conditioning (i.e., natural ventilation) was observed to have ratios of one-month indoor and outdoor concentrations ranging from 0.10-0.30 during a nine-month sampling campaign, with the highest ratios reported in Nov-Dec 2009 and the lowest ratios during July-Aug 2009 (López-Aparicio et al., 2011). Indoor concentrations were relatively constant (approximately 3-7 µg/m³ or 2-3 ppb), while outdoor concentrations were lower in the winter.
(9-10 µg/ m³ or about 5 ppb) than in the summer (35-45 µg/ m³ or about 20 ppb). This seasonal variation in outdoor concentrations coupled with homogeneous indoor concentrations, together with increased wintertime air exchange rate due to higher indoor-outdoor temperature differences, is likely responsible for the observed seasonal pattern in I/O ratios.

Exposures in near-road, on-road and in-vehicle microenvironments are likely to be more variable and lower in magnitude than those in other microenvironments due to reaction of O₃ with NO and other combustion emissions. Depending on wind direction, O₃ concentrations near the roadway have been found to be 20-80% of ambient concentrations at sites 400 meters or more distant from roads (Section 3.6.2.1). A study on patrol cars during trooper work shifts reported in-vehicle 9-h concentrations that were approximately 51% of simultaneously measured roadside concentrations (mean of 11.7 versus 22.8 ppb) (Riediker et al., 2003).

### 4.3.3 Personal-Ambient Concentration Relationships

Several factors influence the relationship between personal O₃ exposure and ambient concentration. Due to the lack of indoor O₃ sources, along with reduction of ambient O₃ that penetrates into enclosed microenvironments, indoor and in-vehicle O₃ concentrations are highly dependent on air exchange rate and therefore vary widely in different microenvironments. Ambient O₃ varies spatially due to reactions with other atmospheric species, especially near busy roadways where O₃ concentrations are decreased by reaction with NO (Section 3.6.2.1). This is in contrast with pollutants such as CO and NOₓ, which show appreciably higher concentrations near the roadway than several hundred meters away (Karner et al., 2010). Ozone also varies temporally over multiple scales, with generally increasing concentrations during the daytime hours, and higher O₃ concentrations during summer than in winter. An example of this variability is shown in Figure 4-1, taken from a personal exposure study conducted by Chang et al. (2000).

In this figure, hourly personal exposures are seen to vary from a few ppb in some indoor microenvironments to tens of ppb in vehicle and outdoor microenvironments. The increase in ambient O₃ concentration during the day is apparent from the outdoor monitoring data. In comparison, ambient PM_{2.5} exhibits less temporal variability over the day than O₃, although personal exposure to PM_{2.5} also varies by microenvironment. This combined spatial and temporal variability for O₃ results in varying relationships between personal exposure and ambient concentration.

Correlations between personal exposure to O₃ and corresponding ambient concentrations, summarized in Table 4-2, exhibit a wide range (generally 0.3-0.8, although both higher and lower values have been reported), with higher correlations generally observed in outdoor microenvironments, high building ventilation conditions, and during the summer season. Low O₃ concentrations indoors and during the winter lead to a high proportion of personal exposures below the sampler detection limit, which may partially explain the low correlations observed in some
studies under those conditions. Studies report varying correlations over a range of averaging times, with no clear trend. Ratios of personal exposure to ambient concentration, summarized in Table 4-3, are generally lower in magnitude (typically 0.1-0.3), and are also variable, with increasing time spent outdoors associated with higher ratios. The next two subsections describe studies that have reported personal-ambient correlations and slopes for a variety of seasons, locations, and populations.

Note: the notation below each clock hour shows the location or activity during that hour.
Source: Reprinted with permission of Air and Waste Management Association (Chang et al., 2000).

**Figure 4-1** Variation in hourly personal and ambient concentrations of O$_3$ and PM$_{2.5}$ in various microenvironments during daytime hours.

Ozone concentrations near and below the passive sampler detection limit lead to uncertainty in personal-ambient correlations and ratios. Correlations are reduced in magnitude by values below the detection limit because noise obscures the underlying signal in the data, while ratios tend to fluctuate widely at low concentration since small changes in measured values cause large relative changes in resulting ratios.
As with I/O ratios, this problem is particularly acute in the non-ozone season when ambient O₃ concentrations are low. Improved characterization of the relationship between personal exposure and ambient concentration will depend on the use of recent improved monitoring techniques to accurately capture low O₃ concentrations, preferably at high time resolution to facilitate evaluation of the effect of activity pattern on exposure (Section 4.3.1). While data from studies using new monitoring techniques become available, past studies summarized in the 2006 O₃ AQCD (U.S. EPA, 2006b) remain relevant to consider along with more recent studies in evaluating personal-ambient concentration relationships.

Personal-Ambient Correlations. Correlations between personal exposure and ambient O₃ concentrations have been evaluated in several research studies, many of which were conducted prior to 2005 and are discussed in the 2006 O₃ AQCD. Some studies evaluated subject-specific, or longitudinal correlations, which describe multiple daily measurements for a single individual. These studies indicate the inter-individual variability of personal-ambient correlations. Another type of correlation is a pooled correlation, which combines data from multiple individuals over multiple monitoring periods (e.g., days), providing an overall indicator of the personal-ambient relationship for all study subjects. A third type of correlation is a community-average correlation, which correlates average exposure across all study subjects with fixed-site monitor concentrations. Community-average correlations are particularly informative for interpreting time-series epidemiologic studies, in which ambient concentrations are used as a surrogate for community-average exposure. However, few studies report this metric; this represents another opportunity for improvement of future personal exposure studies. Table 4-2 summarizes studies reporting personal-ambient correlations, and the studies in the table are discussed in the subsequent text.

The results of these studies generally indicate that personal exposures are moderately well correlated with ambient concentrations, and that the ratio of personal exposure to ambient concentration is higher in outdoor microenvironments and during the summer season. In some situations, a low correlation was observed, and this may be due in part to a high proportion of personal measurements below the detection limit of the personal sampler [e.g., Sarnat et al. (2000)]. Apart from this, correlations do not appear to be strongly dependent on concentration. The effect of season is unclear, with mixed evidence on whether higher correlations are observed during the O₃ season. Chang et al. (2000) measured hourly personal exposures in multiple microenvironments and found that the pooled correlation between personal exposure and ambient concentration was highest for outdoor microenvironments (r = 0.68 - 0.91). In-vehicle microenvironments showed moderate to high correlations (0.57-0.72). Correlations in residential indoor microenvironments were very low (r = 0.05 – 0.09), with moderate correlations (0.34-0.46) in other indoor microenvironments such as restaurants and shopping malls. Liard et al. (1999) evaluated community-average correlations based on 4-day mean personal O₃ exposure measurements for adults and children and found a relatively high correlation (r = 0.83) with ambient concentrations, even with only 18-69% of the personal measurements above the detection limit. Sarnat et al. (2000) studied a
population of older adults in Baltimore, MD, and found that longitudinal correlations between 24-h personal exposure and ambient concentration varied by subject and season, with somewhat higher correlations observed in this study during summer (mean = 0.20) than in winter (mean = 0.06). Although the fraction of samples above the detection limit was not reported separately for the older adults in this study, in the larger study of which this population is a part, only a few percent of samples were above the detection limit, with less than 1% above the detection limit during the winter (see Table 4-3) (Koutrakis et al., 2005). This may account for the low observed correlations, particularly in winter. Some evidence was presented that subjects living in well-ventilated indoor environments have higher correlations than those living in poorly ventilated indoor environments, although exceptions to this were also observed. Ramírez-Aguilar et al. (2008) measured 48- to 72-h personal exposures of four groups of asthmatic children aged 6-14 in Mexico City, Mexico, during 1998-2000. A moderate pooled correlation (r = 0.35) was observed between these exposures and corresponding ambient concentrations.
### Table 4-2 Correlations between personal and ambient O₃ concentration.

<table>
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<tr>
<th>Study</th>
<th>Location</th>
<th>Years/Season</th>
<th>Population</th>
<th>Sample duration</th>
<th>Correlation</th>
<th>Study Type</th>
<th>Comment</th>
<th>Concentration (ppb)</th>
<th>Personal Detection limit (ppb)</th>
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<td>Chang et al. (2000)</td>
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<td>Outdoor near roadway</td>
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<td>Personal 0 (1.8) Ambient 17.8 (10.3)</td>
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<td>Mean (SD) Personal 5 (3) Ambient 23 (12)</td>
<td>LOD: NR Fraction above LOD: NR</td>
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<td>Brauer and Brook (1997)</td>
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<td>Summers 1992 and 1993</td>
<td>Camp counselors</td>
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</tr>
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<td>Farm workers</td>
<td>6-14 h (work shift)</td>
<td>0.64</td>
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<td>Population</td>
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<td>Comment</td>
<td>Concentration (ppb)</td>
<td>Personal Detection limit (ppb)</td>
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<tr>
<td>Ramirez-Aguilar et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>December 1998-April 2000</td>
<td>Asthmatic children</td>
<td>48 h to 72 h</td>
<td>0.35</td>
<td>Pooled</td>
<td></td>
<td>Mean (Range) 7.8 (0.2-30.9)</td>
<td>LOD: NR</td>
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<td>Personal LOD: NR</td>
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<td>Ambient LOD: NR</td>
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<td>33.3 (12.5-64.6)</td>
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<td>Fraction above LOD: NR: NR</td>
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<td></td>
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<td>53%</td>
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<tr>
<td>Delfino et al. (1996)</td>
<td>San Diego, California</td>
<td>September and October 1993</td>
<td>Asthmatic children</td>
<td>12-h daytime</td>
<td>0.45</td>
<td>Pooled</td>
<td></td>
<td>Mean (SD) 11.6 (11.2)</td>
<td>LOD: 8.67</td>
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<td></td>
<td></td>
<td></td>
<td>Personal LOD: 53%</td>
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<td></td>
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<td></td>
<td>Ambient LOD: 53%</td>
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<td></td>
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<td>12h: 43 (17)</td>
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<td></td>
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<td>1hr max: 68 (30)</td>
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</tbody>
</table>

LOD = limit of detection; NR = not reported; SD = standard deviation
Consistent with hourly microenvironment-specific results from the Chang et al. (2000) study described above, studies have found moderate to high personal-ambient correlations for individuals spending time outdoors. A moderate pooled correlation of 0.61 was reported between 24-h avg personal and central-site measurements by Linn et al. (1996) for a population of southern California schoolchildren who spent an average of 101-136 minutes per day outdoors. The authors also report a correlation of 0.70 between central-site measurements and concentrations outside the children’s schools. Although the average school outdoor concentration (34 ppb) was higher than the average central-site concentration (23 ppb) and the average personal exposure concentration was lower (5 ppb) than the central-site value, the similarity between the correlations in this study indicate that central-site monitor concentrations can represent personal exposures in addition to representing local outdoor concentrations. A study in Vancouver, BC provided another illustration of the effect of outdoor microenvironments on personal-ambient relationships by comparing three groups spending different amounts of time outdoors: health clinic workers (0-25% of sampling period outdoors), camp counselors (7.5-45% of sampling period outdoors), and farm workers (100% of sampling period outdoors) (Brauer and Brook, 1997). Health clinic workers and camp counselors were monitored 24 h/day, while farm workers were monitored during their work shift (6-14 hours). In this study, the pooled correlations between personal exposure and fixed-site concentration were not substantially different among the groups (r = 0.60, 0.42, and 0.64, respectively), even though the farm workers experienced the highest concentrations. The ratios of personal exposure to fixed-site monitor concentration increased among the groups with increasing amount of time spent outdoors (0.35, 0.53, and 0.96, respectively). This indicates that temporal variations in personal exposure to O₃ are driven by variations in ambient concentration, even for individuals that spend little time outdoors.

**Personal-Ambient Ratios.** Studies indicate that the ratio between personal O₃ exposure and ambient concentration varies widely, depending on activity patterns, housing characteristics, and season. Higher personal-ambient ratios are generally observed with increasing time spent outside, higher air exchange rate, and in seasons other than winter. Table 4-3 summarizes the results of several such studies discussed in the 2006 O₃ AQCD together with newer studies showing the same pattern of results.

O'Neill et al. (2003) studied a population of shoe cleaners working outdoors in Mexico City, Mexico, and presented a regression model indicating a 0.56 ppb increase in 6-h personal exposure for each 1 ppb increase in ambient concentration. Regression analyses by Sarnat et al. (2005) and Sarnat et al. (2001) for 24-hour data from mixed populations of children and older adults in Baltimore, MD (Sarnat et al., 2001), and Boston, MA (Sarnat et al., 2005), found differing results between the two cities, with Baltimore subjects showing a near-zero slope (0.01) during the summertime while Boston subjects showed a positive slope of 0.27 ppb personal exposure per 1 ppb ambient concentration. In both cities, the winter slope was near zero. The low slope observed in Baltimore may have been due to differences in time spent outdoors, residential ventilation conditions, or other factors. The intercept in
Baltimore was more than half of the median personal exposure (1.84 versus 2.5 ppb), and only 6.6% of the personal samples were above the detection limit, suggesting that noise in the data may also have contributed to the low observed regression slope. However, other studies with a relatively large fraction of personal values below the detection limit reported slopes of 0.27 or above (Sarnat et al., 2006a; Brauer and Brook, 1997). Xue et al. (2005) measured 6-day personal exposure of children in southern California and found that the average ratio of personal exposure to ambient concentration was relatively stable throughout the year at 0.3. These authors also regressed personal exposures on ambient concentration after adjusting for time-activity patterns and housing characteristics and found a slope of 0.54 ppb/ppb, with the regression R² value of 0.58. Unadjusted regression slopes were not presented. It should also be noted that the ratio and slope would not be expected to be identical unless the intercept and other regression parameters were effectively zero.

A few additional studies have been published since the 2006 O₃ AQCD comparing personal exposures with ambient concentrations, and these findings generally confirm the conclusions of the 2006 O₃ AQCD that ventilation conditions, activity pattern, and season may impact personal-ambient ratios. Sarnat et al. (2006a) measured 24-hour personal exposures for a panel of older adults in Steubenville, OH during summer and fall 2000. Subjects were classified as high-ventilation or low-ventilation based on whether they spent time in indoor environments with open windows. Regression of personal exposures on ambient concentration found a higher slope for high-ventilation subjects compared with low-ventilation subjects in both summer (0.18 versus 0.08) and fall (0.27 versus 0.20). Suh and Zanobetti (2010) reported an average 24-hour personal exposure of 2.5 ppb as compared to 24-hour ambient concentration of 29 ppb for a group of individuals with either recent MI or diagnosed COPD in Atlanta. A similar result was observed in Detroit, where the mean 24-hour personal exposure across 137 participants in summer and winter was 2.1 ppb, while the mean ambient concentration on sampling days was 25 ppb (Williams et al., 2009b). Although no personal exposures were measured, McConnell et al. (2006) found that average 24-hour home outdoor O₃ concentrations were within 6 ppb of O₃ concentrations measured at central-site monitors in each of three southern California communities, with a combined average home outdoor concentration of 33 ppb compared to the central-site average of 36 ppb. In Mexico City, Mexico, Ramirez-Aguilar et al. (2008) regressed 48- to 72-hour personal exposures of four groups of asthmatic children aged 6-14 with ambient concentrations and found slope of 0.17 ppb/ppb after adjustment for distance to the fixed-site monitor, time spent outdoors, an interaction term combining these two variables, and an interaction term representing neighborhood and study group.
<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Years/Season</th>
<th>Population</th>
<th>Sample duration</th>
<th>Ratio*</th>
<th>Slope</th>
<th>Intercept</th>
<th>Study Type</th>
<th>Comment</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sarnat et al.</td>
<td>Baltimore, MD</td>
<td>Summer 1998</td>
<td>Older adults and children</td>
<td>24 h</td>
<td>NR</td>
<td>0.01</td>
<td>1.84</td>
<td>Longitudinal</td>
<td>t-value: 1.21 Median (IQR) 2.5 (0-6.4) LOD: 6.6 Fraction above LOD: 6.6% Ambient* 36 (31-43)</td>
</tr>
<tr>
<td>(2001)</td>
<td></td>
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<tr>
<td>Koutrakis et al.</td>
<td></td>
<td>Winter 1999</td>
<td>Older adults, children, and individuals with COPD</td>
<td></td>
<td>NR</td>
<td>0.00</td>
<td>0.46</td>
<td></td>
<td>t-value: 0.03 Personal* 1.1 (-0.6-1.9) LOD: 5.5 Fraction above LOD: 0.2% Ambient* 18 (8.6-26) *Estimated from Figure 1</td>
</tr>
<tr>
<td>(2005)</td>
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</tbody>
</table>

Table 4-3  Ratios of personal to ambient O₃ concentration.
<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Years/Season</th>
<th>Population</th>
<th>Sample duration</th>
<th>Ratio^a</th>
<th>Slope</th>
<th>Intercept</th>
<th>Study Type</th>
<th>Comment</th>
<th>Concentration/ Detection limit (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Brauer and Brook (1997)</td>
<td>Vancouver, Canada</td>
<td>Summers 1992 and 1993</td>
<td>Health clinic workers</td>
<td>24 h</td>
<td>0.35</td>
<td>NR</td>
<td>NR</td>
<td>Pooled</td>
<td>0-25% of time outdoors</td>
<td>Range, Personal* 0.3-10.4, Ambient* 5.3-30.1 LOD: 17</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Camp counselors</td>
<td>24 h</td>
<td>0.53</td>
<td>NR</td>
<td>NR</td>
<td>Pooled</td>
<td>7.5-45% of time outdoors</td>
<td>Personal* 0.1-13.8, Ambient* 11.2-26.1 LOD: 17</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Farm workers</td>
<td>6-14 h (work shift)</td>
<td>0.96</td>
<td>NR</td>
<td>NR</td>
<td>Pooled</td>
<td>100% of time outdoors</td>
<td>Personal* 3.2-43, Ambient* 8.6-49.0 LOD: 12 *Estimated from Figure 2</td>
</tr>
<tr>
<td>O'Neill et al. (2003)</td>
<td>Mexico City, Mexico</td>
<td>April - July 1996</td>
<td>Shoe cleaners</td>
<td>6 h</td>
<td>0.40</td>
<td>0.56</td>
<td>NR</td>
<td>Longitudinal</td>
<td>Mean (SD) Personal 34.4 (22.3) Ambient 84.0 (24.8) LOD: 21.1 (20.6)</td>
<td></td>
</tr>
<tr>
<td>Study</td>
<td>Location</td>
<td>Years/Season</td>
<td>Population</td>
<td>Sample duration</td>
<td>Ratio*</td>
<td>Slope</td>
<td>Intercept</td>
<td>Study Type</td>
<td>Comment</td>
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<tr>
<td>Sarnat et al.</td>
<td>Boston</td>
<td>Summer</td>
<td>Older adults and children</td>
<td>24 h</td>
<td>NR</td>
<td>0.27</td>
<td>95% CI: 0.18-0.37</td>
<td>NR</td>
<td>Longitudinal</td>
<td>Range of Means</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>NR</td>
<td>0.04</td>
<td>95% CI: 0.00-0.07</td>
<td>NR</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>Winter</td>
<td></td>
<td></td>
<td>NR</td>
<td>0.04</td>
<td>95% CI: 0.00-0.07</td>
<td>NR</td>
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<tr>
<td>Xue et al.</td>
<td>Southern California</td>
<td>June 1995 - May 1996</td>
<td>Children</td>
<td>6 day</td>
<td>0.3</td>
<td>0.3</td>
<td>NR</td>
<td>NR</td>
<td>Longitudinal</td>
<td>(O_3) season (May-Sept)</td>
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</tbody>
</table>

**Range of Means**
- **Personal**
  - 4.8-7.6
  - LOD: 7.0
  - Fraction above LOD: 23.8%
- **Ambient**
  - Mean (SD): 22.7-31.6

**Personal**
- 0.1-2.5
- LOD: 4.9
- Fraction above LOD: 3.2%

**Ambient**
- 14.0-21.8

**Non-\(O_3\) season (Oct-Apr)**
- **Personal**
  - Median (IQR): 6 (5-10)
- **Ambient**
  - Median (IQR): 26 (14-32)
  - LOD: NR

*Estimated from Figure 2
<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Years/Season</th>
<th>Population</th>
<th>Sample duration</th>
<th>Ratio$^a$</th>
<th>Slope</th>
<th>Intercept</th>
<th>Study Type</th>
<th>Comment</th>
<th>Concentration/ Detection limit (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sarnat et al. (2006a)</td>
<td>Steuben-ville, OH</td>
<td>Summer</td>
<td>Older adults</td>
<td>24 h</td>
<td>NR</td>
<td>0.15</td>
<td>SE: 0.02</td>
<td>t-value: 7.21</td>
<td>NR</td>
<td>All individuals</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Summer</td>
<td></td>
<td></td>
<td>NR</td>
<td>0.18</td>
<td>SE: 0.03</td>
<td>t-value: 7.34</td>
<td>NR</td>
<td>High-ventilation</td>
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<tr>
<td></td>
<td></td>
<td>Fall</td>
<td></td>
<td></td>
<td>NR</td>
<td>0.08</td>
<td>SE: 0.04</td>
<td>t-value: 1.89</td>
<td>NR</td>
<td>Low-ventilation</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Fall</td>
<td></td>
<td></td>
<td>NR</td>
<td>0.27</td>
<td>SE: 0.03</td>
<td>t-value: 8.64</td>
<td>NR</td>
<td>Longitudinal</td>
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<td>Fall</td>
<td></td>
<td></td>
<td>NR</td>
<td>0.20</td>
<td>SE: 0.05</td>
<td>t-value: 3.90</td>
<td>NR</td>
<td>Low-ventilation</td>
</tr>
<tr>
<td>Study</td>
<td>Location</td>
<td>Years/Season</td>
<td>Population</td>
<td>Sample duration</td>
<td>Ratio&lt;sup&gt;a&lt;/sup&gt;</td>
<td>Slope</td>
<td>Intercept</td>
<td>Study Type</td>
<td>Comment</td>
<td>Concentration/Detection limit (ppb)</td>
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<tr>
<td>Ramírez-Aguilar et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>Dec. 1998-Apr. 2000</td>
<td>Asthmatic children</td>
<td>48 h to 72 h</td>
<td>0.23</td>
<td>0.17</td>
<td>SE: 0.02</td>
<td>95% CI: 0.13-0.21</td>
<td>Pooled</td>
<td>Mean (Range) Personal 7.8 (0.2-30.9) Ambient 33.3 (12.5-64.6) LOD: NR</td>
</tr>
</tbody>
</table>

<sup>a</sup> Mean value unless otherwise indicated  
<sup>b</sup> Median  
IQR = interquartile range; LOD = limit of detection; NR = not reported; SD = standard deviation
4.3.4 Co-exposure to Other Pollutants and Environmental Stressors

Exposure to ambient O₃ occurs in conjunction with exposure to a complex mixture of ambient pollutants that varies over space and time. Multipollutant exposure is an important consideration in evaluating health effects of O₃ since these other pollutants have either known or potential health effects that may impact health outcomes due to O₃. The co-occurrence of high O₃ concentrations with high heat and humidity may also contribute to health effects. This section presents data on relationships between overall personal O₃ exposure and exposure to other ambient pollutants, as well as co-exposure relationships for near-road O₃ exposure.

4.3.4.1 Personal Exposure to Ozone and Copollutants

Personal exposure to O₃ shows variable correlation with personal exposure to other pollutants, with differences in correlation depending on factors such as instrument detection limit, season, city-specific characteristics, time scale, and spatial variability of the copollutant. Suh and Zanobetti (2010) reported Spearman rank correlation coefficients during spring and fall between 24-h avg O₃ measurements and copollutants of 0.14, 0.00, and -0.03 for PM₂.₅, EC, and NO₂, respectively. Titration of O₃ near roadways is likely to contribute to the low or slightly negative correlations with the traffic-related pollutants EC and NO₂. The somewhat higher correlation with PM₂.₅ may reflect the influence of air exchange rate and time spent outdoors on co-exposures to ambient PM₂.₅ and O₃. Overall, the copollutant correlations are quite small, which may be due to the very low personal exposures observed in this study (2-3 ppb), likely to be near or below the detection limit of the passive sampler over a 24-hour period. Chang et al. (2000) measured hourly personal exposures to PM₂.₅ and O₃ in summer and winter in Baltimore, Maryland. Correlations between PM₂.₅ and O₃ were 0.05 and -0.28 in summer and winter, respectively. Results indicate personal O₃ exposures were not significantly associated with personal PM₂.₅ exposures in either summer or winter. These non-significant correlations may be attributed in part to the relatively low personal O₃ exposures observed in this study; in both summer and winter, the mean personal O₃ exposure was below the calculated limit of detection.

Studies conducted in Baltimore, MD (Sarnat et al., 2001), and Boston, MA (Sarnat et al., 2005), found differing results for the correlation between 24-h avg personal O₃ and personal PM₂.₅ exposures, particularly during the winter season. Sarnat et al. (2001) found a positive slope when regressing personal exposures of both total PM₂.₅ (0.21) and PM₂.₅ of ambient origin (0.22) against personal O₃ exposures during the summer season, but negative slopes (-0.05 and -0.18, respectively) during the winter season. The summertime slope for personal PM₂.₅ exposure versus personal O₃ exposure was much higher for children (0.37) than for adults (0.07), which may be
the result of different activity patterns. This team of researchers also found a positive, although higher, summer slope between 24-h avg personal O₃ and personal PM₂.₅ in Boston (0.72) (Sarnat et al., 2005). However, the winter slope was positive (1.25) rather than negative, as in Baltimore. In both cities during both seasons, there was a wide range of subject-specific correlations between personal O₃ and personal PM₂.₅ exposures, with some subjects showing relatively strong positive correlations (>0.75) and others showing strong negative correlations (<-0.50). The median correlation in both cities was slightly positive in the summer and near zero (Boston) or slightly negative (Baltimore) in the winter. These results indicate the potential effects of city-specific characteristics, such as housing stock and building ventilation patterns, on relationships between O₃ and copollutants.

The lack of long-term exposure assessment studies limits evaluation of long-term correlations between O₃ exposure and copollutant exposure. Although some long-term epidemiologic studies have reported copollutant correlations for fixed-site monitor concentrations or city-wide averages used as exposure metrics, no clear pattern is apparent. Correlations with PM concentrations range from less than 0.2 to nearly 0.9. For example, the long-term correlation between 30-yr mean (1973-1992) PM₁₀ and the 30-yr mean of 8-h avg (9 am to 5 pm) O₃ was 0.88 for participants in the Adventist Health Smog study (AHSMOG) (McDonnell et al., 1999a). Jerrett et al. (2009) reported a moderate correlation of 0.56 between two-year average O₃ and PM₂.₅ in 86 U.S. metropolitan areas. In the Southern California Children’s Health Study, the correlation between 1994-2000 average O₃ and PM₂.₅ was 0.33 for 1-h daily max O₃, but only 0.18 for the 1994-2000 mean of 8-h avg (10 am – 6 pm) O₃ concentrations (Gauderman et al., 2004). Similar correlations were reported in this study between these O₃ metrics and PM₁₀. For children participating in the National Health Interview Survey and living in U.S. metropolitan areas, the correlation between 2000-2004 O₃ concentrations and PM₂.₅ and PM₁₀ concentrations was 0.29 and 0.55, respectively (Akinbami et al., 2010). For NO₂, near-zero or negative correlations have been reported, consistent with atmospheric chemistry involving NO₂ and O₃. Correlations with NO₂ in the Children’s Health Study were 0.10 and -0.11 for 1994-2000 mean 1-h daily max O₃ and 8-h avg O₃, respectively (Gauderman et al., 2004). This is similar to the value of -0.05 reported by Akinbami et al. (2010). However, the AHSMOG study reported a correlation of 0.61 between 30-yr averages of O₃ and NO₂, possibly reflecting similar overall levels of air pollution experienced by the participants (McDonnell et al., 1999a). This lack of a consistent pattern makes it difficult to draw conclusions regarding long-term correlations between O₃ and copollutants.

To the extent that short-term concentrations drive long-term patterns, some insight may be provided by an analysis of short-term correlations between O₃ and other criteria pollutants, such as is provided in Section 3.6.4. Warm-season 8-h daily max O₃ concentrations are generally positively correlated with co-located 24-h avg measurements of other criteria pollutants (Figure 3-57). Median correlations range from approximately 0.15 to 0.55 for CO, SO₂, NO₂, PM₁₀, and PM₂.₅, in that order. In contrast, year-round 8-h daily max O₃ data show negative median correlations with CO and NO₂, positive correlations with PM₁₀ and PM₂.₅, and essentially zero
correlation with SO₂. This reflects mostly negative correlations between O₃ and all pollutants during wintertime, as shown in Figure 3-56. Titration of O₃ near roadways also likely contributes to overall negative correlations with NO₂ and CO. Positive correlations between O₃ and PM₂.₅ during the summertime can be partly explained by meteorological conditions favoring increased formation of both secondary PM and O₃. Strong positive correlations can influence the interpretation of epidemiologic results, potentially complicating the ability to identify the independent effect of a pollutant.

4.3.4.2 Near-Road Exposure to Ozone and Copollutants

Beckerman et al. (2008) measured both 1-week and continuous concentrations of O₃, NO, NO₂, NOₓ, PM₂.₅, PM₁.₀, and several VOCs (the BTEX compounds, MTBE, hexane, and THC) in the vicinity of heavily traveled (annual average daily traffic [AADT] >340,000) roadways in Toronto, Canada. Passive samplers were deployed for one week in August 2004. Ozone concentrations were negatively correlated with all pollutants, with the exception of VOCs at one of the monitoring sites which were suspected of being influenced by small area sources. Site specific correlations are given in Figure 4-2. Correlations were -0.77 to -0.85 for NO₂, -0.48 to -0.62 for NO, and -0.55 to -0.63 for NOₓ. Pooled correlations using data from both sites were somewhat lower in magnitude. PM₂.₅ and PM₁.₀ correlations were -0.35 to -0.78 and -0.34 to -0.58, respectively. At the monitoring site not influenced by small area sources, O₃-VOC correlations ranged from -0.41 to -0.66.

Beckerman et al. (2008) also made on-road measurements of multiple pollutants with a instrumented vehicle. Concentrations were not reported, but correlations between O₃ and other pollutants were negative and somewhat greater in magnitude (i.e., more negative) than the near-road correlations. SO₂, CO, and BC were measured in the mobile laboratory, although not at the roadside, and they all showed negative correlations with O₃ when the data were controlled for site. Correlations for continuous concentrations between O₃ and copollutants were somewhat lower than the 1-week correlations, except for O₃-PM₂.₅ correlations. Correlations were -0.90, -0.66, -0.77, and -0.89 for NO₂, NO, NOₓ, and PM₁.₀ respectively. The continuous O₃-PM₂.₅ correlation was -0.62, which is in the range of the 1-week correlation.
Indoor Exposure to Ozone and Copollutants

Ambient O₃ that infiltrates indoors reacts with organic compounds and other chemicals to form oxidized products, as described in Section 3.2.3.1 as well as the 2006 O₃ AQCD. It is anticipated that individuals are exposed to these reaction products, although no evidence was identified regarding personal exposures. The reactions are similar to those occurring in the ambient air, as summarized in Chapter 3. For example, O₃ can react with terpenes and other compounds from cleaning products, air fresheners, and wood products both in the gas phase and on surfaces to form particulate and gaseous species, such as formaldehyde (Chen et al., 2011; Shu and Morrison, 2011; Aoki and Tanabe, 2007; Reiss et al., 1995b). Ozone has also been shown to react with material trapped on HVAC filters and generate airborne products (Bekö et al., 2007; Hyttinen et al., 2006). Potential oxygenated reaction products have been found to act as irritants (Anderson et al., 2007), indicating that these reaction products may have health effects separate from those of O₃ itself (Weschler and Shields, 1997). Ozone may also react to form other oxidants, which then go on to participate in additional reactions. White et al. (2010) found evidence that HONO, or other oxidants, may have been present during experiments to estimate indoor OH concentrations; indicating complex indoor oxidant chemistry. Rates of these reactions are dependent on indoor O₃ concentration, temperature, and air exchange rate, making estimation of exposures to reaction products difficult.
4.4 Exposure-Related Metrics

In this section, parameters are discussed that are relevant to the estimation of exposure, but are not themselves direct measures of exposure. Time-location-activity patterns, including behavioral changes to avoid exposure, have a substantial influence on exposure and dose. Proximity of populations to ambient monitors may influence how well their exposure is represented by measurements at the monitors, although factors other than distance play an important role as well.

4.4.1 Activity Patterns

The activity pattern of individuals is an important determinant of their exposure. Variation in O₃ concentrations among various microenvironments means that the amount of time spent in each location, as well as the level of activity, will influence an individual’s exposure to ambient O₃. The effect of activity pattern on exposure is explicitly accounted for in Equation 4-3 by the fraction of time spent in different microenvironments.

Activity patterns vary both among and within individuals, resulting in corresponding variations in exposure across a population and over time. Large-scale human activity databases, such as those developed for the National Human Activity Pattern Survey (NHAPS) (Klepeis et al., 2001) or the Consolidated Human Activity Database (CHAD) (McCurdy et al., 2000), which includes NHAPS data together with other activity study results, have been designed to characterize exposure patterns among much larger population subsets than can be examined during individual panel studies. The complex human activity patterns across the population (all ages) are illustrated in Figure 4-3 (Klepeis et al., 2001), which is presented to illustrate the diversity of daily activities among the entire population as well as the proportion of time spent in each microenvironment. For example, about 25% of the individuals reported being outdoors or in a vehicle between 2:00 and 3:00 p.m., when daily O₃ levels are peaking, although about half of this time was spent in or near a vehicle, where O₃ concentrations are likely to be lower than ambient concentrations.

Time spent in different locations has also been found to vary by age. Table 4-4 summarizes NHAPS data reported for four age groups, termed Very Young (0-4 years), School Age (5-17 years), Working (18-64 years), and Retired (65+ years) (Klepeis et al., 1996). The working population spent the least time outdoors, while the school age population spent the most time outdoors. NHAPS respondents aged 65 and over spent somewhat more time outdoors than adults aged 18-64, with a greater fraction of time spent outdoors at a residence. Children aged 0-4 also spent most of their outdoor time in a residential outdoor location. On average, the fraction of time spent outdoors by school age respondents was 2.62 percentage points higher than working respondents, corresponding to approximately 38 minutes more time outdoors per day. Although not accounting for activity level, this increased time
spent outdoors is consistent with evidence in Chapter 8 suggesting that younger and older age groups are more at risk for O₃-related health effects.

### Table 4-4 Mean fraction of time spent in outdoor locations by various age groups in the NHAPS study.

<table>
<thead>
<tr>
<th>Age Group</th>
<th>Residential-Outdoor</th>
<th>Other Outdoor</th>
<th>Total Outdoors</th>
</tr>
</thead>
<tbody>
<tr>
<td>0-4 yr</td>
<td>5.38%</td>
<td>0.96%</td>
<td>6.34%</td>
</tr>
<tr>
<td>5-17 yr</td>
<td>5.05%</td>
<td>2.83%</td>
<td>7.88%</td>
</tr>
<tr>
<td>18-64 yr</td>
<td>2.93%</td>
<td>2.33%</td>
<td>5.26%</td>
</tr>
<tr>
<td>65+ yr</td>
<td>4.48%</td>
<td>1.27%</td>
<td>5.75%</td>
</tr>
</tbody>
</table>

Source: Data from Klepeis et al. (1996).

Together with location, exertion level is an important determinant of exposure. Table 4-5 summarizes ventilation rates for different age groups at several levels of activity as presented in Table 6-2 of the EPA’s Exposure Factors Handbook (U.S. EPA, 2011b). Most of the age-related variability is seen for moderate and high intensity activities, except for individuals under 1 year. For moderate intensity, ventilation rate increases with age through childhood and adulthood until age 61, after which a moderate decrease is observed. Ventilation rate is most variable for high intensity activities. Children aged 1 to <11 years have ventilation rates of approximately 40 L/min, while children aged 11+ and adults have ventilation rates of approximately 50 L/min. The peak is observed for the 51 to <61 age group, at 53 L/min, with lower ventilation rates for older adults.

A dramatic increase in ventilation rate occurs as exercise intensity increases. For children and adults <31 years, high intensity activities result in nearly double the ventilation rate for moderate activity, which itself is nearly double the rate for light activity. Children have other important differences in ventilation compared to adults. As discussed in Chapter 5, children tend to have a greater oral breathing contribution than adults, and they breathe at higher minute ventilations relative to their lung volumes. Both of these factors tend to increase dose normalized to lung surface area.
Table 4-5  Mean ventilation rates (L/min) at different activity levels for different age groups.

<table>
<thead>
<tr>
<th>Age Group</th>
<th>Sleep or Nap</th>
<th>Sedentary/Passive</th>
<th>Light Intensity</th>
<th>Moderate Intensity</th>
<th>High Intensity</th>
</tr>
</thead>
<tbody>
<tr>
<td>Birth to &lt;1 yr</td>
<td>3.0</td>
<td>3.1</td>
<td>7.6</td>
<td>14</td>
<td>26</td>
</tr>
<tr>
<td>1 to &lt;2 yr</td>
<td>4.5</td>
<td>4.7</td>
<td>12</td>
<td>21</td>
<td>38</td>
</tr>
<tr>
<td>2 to &lt;3 yr</td>
<td>4.6</td>
<td>4.8</td>
<td>12</td>
<td>21</td>
<td>39</td>
</tr>
<tr>
<td>3 to &lt;6 yr</td>
<td>4.3</td>
<td>4.5</td>
<td>11</td>
<td>21</td>
<td>37</td>
</tr>
<tr>
<td>6 to &lt;11 yr</td>
<td>4.5</td>
<td>4.8</td>
<td>11</td>
<td>22</td>
<td>42</td>
</tr>
<tr>
<td>11 to&lt;16 yr</td>
<td>5.0</td>
<td>5.4</td>
<td>13</td>
<td>25</td>
<td>49</td>
</tr>
<tr>
<td>16 to &lt;21 yr</td>
<td>4.9</td>
<td>5.3</td>
<td>12</td>
<td>26</td>
<td>49</td>
</tr>
<tr>
<td>21 to &lt;31 yr</td>
<td>4.3</td>
<td>4.2</td>
<td>12</td>
<td>26</td>
<td>50</td>
</tr>
<tr>
<td>31 to &lt;41 yr</td>
<td>4.6</td>
<td>4.3</td>
<td>12</td>
<td>27</td>
<td>49</td>
</tr>
<tr>
<td>41 to &lt;51 yr</td>
<td>5.0</td>
<td>4.8</td>
<td>13</td>
<td>28</td>
<td>52</td>
</tr>
<tr>
<td>51 to &lt;61 yr</td>
<td>5.2</td>
<td>5.0</td>
<td>13</td>
<td>29</td>
<td>53</td>
</tr>
<tr>
<td>61 to &lt;71 yr</td>
<td>5.2</td>
<td>4.9</td>
<td>12</td>
<td>26</td>
<td>47</td>
</tr>
<tr>
<td>71 to &lt;81 yr</td>
<td>5.2</td>
<td>5.0</td>
<td>12</td>
<td>25</td>
<td>47</td>
</tr>
<tr>
<td>81+ yr</td>
<td>5.2</td>
<td>4.9</td>
<td>12</td>
<td>25</td>
<td>48</td>
</tr>
</tbody>
</table>

Source: Data from Exposure Factors Handbook (U.S. EPA, 2011b).

Longitudinal activity pattern information is also an important determinant of exposure, as different people may exhibit different patterns of time spent outdoors over time due to age, sex, employment, and lifestyle-dependent factors. These differences may manifest as higher mean exposures or more frequent high-exposure episodes for some individuals. The extent to which longitudinal variability in individuals contributes to the population variability in activity and location can be quantified by the ratio of between-person variance to total variance in time spent in different locations and activities (the intraclass correlation coefficient, ICC). Xue et al. (2004) quantified ICC values in time-activity data collected by Harvard University for 160 children aged 7–12 years in Southern California (Geyh et al., 2000). For time spent outdoors, the ICC was approximately 0.15, indicating that 15% of the variance in outdoor time was due to between-person differences. The ICC value might be different for other population groups.
The EPA’s National Exposure Research Laboratory (NERL) has consolidated many of the most important human activity databases into one comprehensive database called the Consolidated Human Activity Database (CHAD). The current version of CHAD contains data from nineteen human activity pattern studies (including NHAPS), which were evaluated to obtain over 33,000 person-days of 24-hour human activities in CHAD (McCurdy et al., 2000). The surveys include probability-based recall studies conducted by EPA and the California Air Resources Board, as well as real-time diary studies conducted in individual U.S. metropolitan areas using both probability-based and volunteer subject panels. All ages of both sexes are represented in CHAD. The data for each subject consist of one or more days of sequential activities, in which each activity is defined by start time, duration, activity type, and microenvironment classification (i.e., location). Activities vary from one minute to one hour in duration, with longer activities being subdivided into clock-hour durations to facilitate exposure modeling. CHAD also provides information on the level of exertion associated with each activity, which can be used by exposure models, including the APEX model (Section 4.5.3), to estimate ventilation rate and pollutant dose.
4.4.2 Ozone-Averting Behavior

Individuals can reduce their exposure to $O_3$ by altering their behaviors, such as by staying indoors, scheduling outdoor activity during periods of low $O_3$ concentration, and by reducing activity levels or time spent being active outdoors on high-$O_3$ days. To assist the public in avoiding exposure to air pollution on days with high pollutant concentrations, EPA has developed an information tool known as the Air Quality Index (AQI) to provide information to the public on ambient levels of pollutants and the potential for individuals to experience health effects (U.S. EPA, 2011a). The AQI describes the potential for health effects from $O_3$ (and other individual pollutants) in six color-coded categories of air-quality, ranging from good (green), moderate (yellow), unhealthy for sensitive groups (orange), unhealthy (red), very unhealthy (purple), and hazardous (maroon). The levels are associated with descriptors of the likelihood of health effects and the populations most likely to be affected. For example, the orange level indicates that the general population is not likely to be at risk, but susceptible groups may experience health effects. These advisories explicitly state that children, older adults, people with lung disease, and those who are active outdoors may be at greater risk from exposure to air pollution. Forecasted and actual conditions typically are reported to the public during high-$O_3$ months through local media outlets, using various versions of this air-quality categorization scheme. People are advised to change their behavior to reduce exposure depending on predicted $O_3$ concentrations and the likelihood of risk. Behavioral recommendations include moving outdoor activities to times when air quality is better, and reducing activity levels or the time spent being active outdoors on high-$O_3$ days. Staying indoors to reduce exposure is only recommended when the AQI is at or above the very unhealthy range.

Evidence of individual averting behaviors in response to advisories has been found in several studies, especially for potentially susceptible populations, such as children, older adults, and asthmatics. Reduced time spent outdoors was reported in an activity diary study in 35 U.S. cities (Mansfield et al., 2006), which found that asthmatic children who spent at least some time outdoors reduced their total time spent outdoors by an average of 30 min on a code red $O_3$ day relative to a code green, yellow, or orange day; however, the authors noted that there was appreciable variation in both the overall amount of time spent outdoors and the reduction in outdoor time on high $O_3$ days among asthmatic children. Bresnahan et al. (1997) examined survey data collected during 1985-86 from a panel of adults in the Los Angeles area, many of whom had compromised respiratory function, by an averting behavior model. A regression analysis indicated that individuals with smog-related symptoms spent about 12 minutes less time outdoors over a two-day period for each 10 ppb increase in $O_3$ concentration above 120 ppb. Considering that the average daily maximum $O_3$ concentration at the time was approximately 180 ppb on days when the then-current standard (1-h max of 120 ppb) was exceeded, this implies that those individuals spent about 40 minutes less time outside per day on a typical high $O_3$ day compared to days with $O_3$ concentrations below the standard. However, the behavior was not specifically linked to exceedances or air quality alerts.
The fraction of individuals who reduce time spent outdoors, or restrict their children’s outdoor activity, has been found to vary based on health status. In the Bresnahan et al. (1997) study, 40 percent of respondents reported staying indoors on days when air quality was poor. Individuals who reported experiencing smog-related symptoms were more likely to take the averting actions, although the presence of asthma or other chronic respiratory conditions did not have a statistically significant effect on behavior. A study of parents of asthmatic children (McDermott et al., 2006) suggests that parents are aware of the hazard of outdoor air pollution and the official alerts designed to protect them and their children. It also suggests that a majority of parents (55%) comply with recommendations of the alerts to restrict children’s outdoor activity, with more parents of asthmatics reporting awareness and responsiveness to alerts. However, only 7% of all parents complied with more than one-third of the advisories issued (McDermott et al., 2006). Wen et al. (2009) analyzed data from the 2005 Behavioral Risk Factor Surveillance System (BRFSS) and indicated that people with asthma are about twice as likely as people without asthma to reduce their outdoor activities based on either media alerts of poor air quality (31% versus 16%) or individual perception of air quality (26% versus 12%). Respondents who had received advice from a health professional to reduce outdoor activity when air quality is poor were more likely to report a reduction based on media alerts, both for those with and without asthma. In a study of randomly selected individuals in Houston, TX and Portland, OR, Semenza et al. (2008) found that a relatively small fraction of survey respondents (9.7% in Houston, 10.5% in Portland) changed their behaviors during poor air quality episodes. This fraction is appreciably lower than the fraction reported for people with asthma in the Wen et al. (2009) study, although it is similar to the fraction reported in that study for those without asthma. Most of the people in the Semenza et al. (2008) study reported that their behavioral changes were motivated by self-perception of poor air quality rather than an air quality advisory. It should be noted that the McDermott et al. (2006), Wen et al. (2009), and Semenza et al. (2008) studies evaluated air quality in general and therefore are not necessarily specific to O₃.

Commuting behavior does not seem to change based on air quality alerts. A study in the Atlanta area showed that advisories can raise awareness among commuters but do not necessarily result in a change in an individual’s travel behavior (Henry and Gordon, 2003). This finding is consistent with a survey for 1,000 commuters in Denver, Colorado, which showed that the majority (76%) of commuters heard and understood the air quality advisories, but did not alter their commuting behavior (Blanken et al., 2001).

Some evidence is available for other behavioral changes in response to air quality alerts. Approximately 40 percent of the respondents in the Los Angeles study by Bresnahan et al. (1997) limited or rearranged leisure activities, and 20 percent increased use of air conditioners. As with changes in time spent outdoors, individuals who reported experiencing smog-related symptoms, but not those with asthma or chronic respiratory conditions, were more likely to take the averting actions. Other factors influencing behavioral changes, such as increased likelihood of averting behavior among high school graduates, are also reported in the study. In a separate
Southern California study, attendance at two outdoor facilities (i.e., a zoo and an observatory) was reduced by 6-13% on days when smog alerts were announced, with greater decreases observed among children and older adults (Neidell, 2010, 2009).

The studies discussed in this section indicate that averting behavior is dependent on several factors, including health status and lifestage. People with asthma and those experiencing smog-related symptoms reduce their time spent outdoors and are more likely to change their behavior than those without respiratory conditions. Children and older adults appear more likely to change their behavior than the general population. Commuters, even when aware of air quality advisories, tend not to change their commuting behavior.

### 4.4.3 Population Proximity to Fixed-Site Ozone Monitors

The distribution of O₃ monitors across urban areas varies between cities (Section 3.6.2.1), and the population living near each monitor varies as well. Monitoring sites in rural areas are generally located in national or state parks and forests, and these monitors may be relevant for exposures of exercising visitors as well as those who live in similar locations. They also serve as an important source of data for evaluating ecological effects of O₃ (Chapter 9). Rural monitors tend to be less affected than urban monitors by strong and highly variable anthropogenic sources of species participating in the formation and destruction of O₃ (e.g., onroad mobile sources) and more highly influenced by regional transport of O₃ or O₃ precursors (Section 3.6.2.2). This may contribute to less diel variability in O₃ concentration than is observed in urban areas.

A variety of factors determine the siting of the O₃ monitors that are part of the SLAMS network reporting to AQS. As discussed in Section 3.5.6, the number and location of required O₃ monitors in an urban area depend on O₃ concentration and population, among other factors. Areas classified as serious, severe, or extreme nonattainment have additional monitoring requirements. Generally, high-O₃ urban areas with a population of 50,000 or greater are required to have at least one monitor; in low- or moderate-concentration areas, the minimum population for a required monitor is 350,000. Most large U.S. cities have several monitors, as shown in Figure 3-76 through Figure 3-95.

As an illustration of the location of O₃ monitors and their concentrations with respect to population density, Figure 4-4 through Figure 4-6 present this information for Atlanta, Boston, and Los Angeles, the three cities selected for detailed analysis in Chapter 3. They represent a cross-section with respect to geographic distribution, O₃ concentration, layout, geographic features, and other factors. The maps show the location of O₃ monitors, identified by the same letters as in Chapter 3 to facilitate intercomparisons, along with the 2007-2009 mean 8-h daily max O₃ concentration for perspective on the variation in O₃ concentration across the urban area. Population density at the census block group level is also presented on the maps.
Figure 4-4  Map of the Atlanta CSA including O$_3$ monitor locations and major roadways with respect to census block group population density estimates for 2009.
Figure 4-5  Map of the Boston CSA including O$_3$ monitor locations and major roadways with respect to census block group population density estimates for 2009.
Figure 4-6 Map of the Los Angeles CSA including O$_3$ monitor locations and major roadways with respect to census block group population density estimates for 2009.
Similarities and differences are apparent among the cities. The spatial distribution of monitor locations in Atlanta and Boston is similar, with one site (site A) near the high population density area and other monitors in surrounding areas of lower population density. In Atlanta, the monitors near the city all have similar concentrations, while somewhat lower concentrations are observed at sites I and J, which are located >50 km from the city center. Boston shows a different spatial concentration pattern, with some low and some high concentrations in urban and less-populated areas. The differences in spatial concentration profiles between the two cities may be due to more consistent terrain in Atlanta compared with Boston, which has a coastline, along with the downwind influence of New York and other northeastern cities contributing to concentration variability.

Los Angeles has a much more complex spatial pattern of monitors, population, and geography. There are a large number of monitors located in multiple levels of population density across the entire CSA, which includes substantial rural areas. Most monitors are near at least moderate population density areas, but there are some high-density areas without a monitor. Concentrations increase in a somewhat radial or west-east pattern from the city, with lower concentrations near the port of Long Beach (monitors B, C, and F). The highest concentrations are located near the San Bernardino forest (e.g., monitors AG, AO, and AR), which have lower population density, but more potential for ecological impacts. Low concentrations in highly populated areas near the coast likely reflect titration by NO\textsubscript{X} and other atmospheric constituents, while high downwind concentrations reflect lack of local NO\textsubscript{X} sources and increased photochemical processing time.

The location of these monitors relative to the location of dense population centers varies among urban areas. NCore sites, a subset of the overall monitoring network, are designed with population exposure as a monitoring objective, and the monitoring requirements in 40 CFR Part 58, Appendix D include population density as one of several factors that would be considered in designing the O\textsubscript{3} monitoring program for an area. At least one site for each MSA is designed to be a maximum concentration site, which could presumably represent the location with the maximum exposure potential in the city. Sites may also be required upwind and downwind of high-concentration urban areas.

All three cities have some high population density areas without an O\textsubscript{3} monitor. The siting considerations for NCore monitors generally target the neighborhood (0.5-4 km) or urban (4-50 km) scale to provide representative concentrations throughout the metropolitan area; however, a middle-scale (0.1-0.5 km) site may be acceptable in cases where the site can represent many such locations throughout a metropolitan area. In other words, a monitor could potentially represent exposures in other similar areas of the city if land use and atmospheric chemistry conditions are similar. This is supported by the correlation analyses in Chapter 3. For example, in Los Angeles, monitors H and L are located in medium-density areas and show moderately high correlation (R = 0.78), although they are some 50 km apart.

Although proximity to a monitor does not determine the degree to which that monitor represents an individual’s ambient exposure, it is one indicator. One way to calculate
monitor representativeness is to calculate the fraction of the urban population living within a certain radius of a monitor. Table 4-6 presents the fraction of the population in selected cities living within 1, 5, 10, and 20 km of an O₃ monitor. Values are presented for both total population and for those under 18 years of age, a potentially susceptible population to the effects of O₃. The data indicate that relatively few people live within 1 km of an O₃ monitor, while nearly all of the population in most cities lives within 20 km of a monitor. Looking at the results for a 5-km radius, corresponding roughly to the neighborhood scale (Section 3.5.6.1), generally 20-30% of the population lives within this distance from an O₃ monitor. Some cities have a greater population in this buffer, such as Salt Lake City, while others have a lower percentage, such as Minneapolis and Seattle. Percentages for children are generally similar to the total population, with no clear trend.

Another approach is to divide the metropolitan area into sectors surrounding each monitor such that every person in the sector lives closer to that monitor than any other. This facilitates calculation of the fraction of the city’s population represented (according to proximity) by each monitor. In Atlanta, for example, the population fraction represented by each of the 11 monitors in the city ranged from 2.9-22%. The two monitors closest to the city center (sites A and B on Figure 4-4) accounted for 16% and 8% of the population, respectively. Site B has two listed monitoring objectives, highest concentration and population exposure. The other monitor in Atlanta with a listed objective of highest concentration is Site C, which represents the largest fraction of the population (22%). The eight monitors with a primary monitoring objective of population exposure account for 2.9-17% of the population per monitor.
### Table 4-6
Fraction of the 2009 population living within a specified distance of an O$_3$ monitor in selected U.S. cities.

<table>
<thead>
<tr>
<th>City</th>
<th>Population Total</th>
<th>Within 1 km Total</th>
<th>Within 5 km Total</th>
<th>Within 10 km Total</th>
<th>Within 20 km Total</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>&lt;18 yr</td>
<td>Total</td>
<td>&lt;18 yr</td>
<td>Total</td>
<td>&lt;18 yr</td>
</tr>
<tr>
<td>Atlanta, GA, CSA</td>
<td>5,901,670</td>
<td>1,210,932</td>
<td>0.3%</td>
<td>0.3%</td>
<td>8%</td>
</tr>
<tr>
<td>Baltimore, MD, CSA</td>
<td>8,421,016</td>
<td>1,916,106</td>
<td>1.3%</td>
<td>1.1%</td>
<td>25%</td>
</tr>
<tr>
<td>Birmingham, AL, CSA</td>
<td>1,204,399</td>
<td>281,983</td>
<td>1.4%</td>
<td>1.6%</td>
<td>22%</td>
</tr>
<tr>
<td>Boston, MA, CSA</td>
<td>7,540,533</td>
<td>1,748,918</td>
<td>0.9%</td>
<td>0.9%</td>
<td>17%</td>
</tr>
<tr>
<td>Chicago, IL, CSA</td>
<td>9,980,113</td>
<td>2,502,454</td>
<td>1.5%</td>
<td>1.5%</td>
<td>28%</td>
</tr>
<tr>
<td>Dallas, TX, CSA</td>
<td>6,791,942</td>
<td>1,530,877</td>
<td>0.4%</td>
<td>0.4%</td>
<td>13%</td>
</tr>
<tr>
<td>Denver, CO, CSA</td>
<td>3,103,801</td>
<td>675,380</td>
<td>1.7%</td>
<td>1.6%</td>
<td>35%</td>
</tr>
<tr>
<td>Detroit, MI, CSA</td>
<td>5,445,448</td>
<td>1,411,875</td>
<td>0.8%</td>
<td>0.9%</td>
<td>15%</td>
</tr>
<tr>
<td>Houston, TX, CSA</td>
<td>5,993,633</td>
<td>1,387,851</td>
<td>1.5%</td>
<td>1.8%</td>
<td>26%</td>
</tr>
<tr>
<td>Los Angeles, CA, CSA</td>
<td>18,419,720</td>
<td>4,668,441</td>
<td>1.6%</td>
<td>1.7%</td>
<td>28%</td>
</tr>
<tr>
<td>Minneapolis, MN, CSA</td>
<td>3,652,490</td>
<td>872,497</td>
<td>0.3%</td>
<td>0.3%</td>
<td>5%</td>
</tr>
<tr>
<td>New York, NY, CSA</td>
<td>22,223,406</td>
<td>5,284,875</td>
<td>1.5%</td>
<td>1.7%</td>
<td>23%</td>
</tr>
<tr>
<td>Philadelphia, PA, CSA</td>
<td>6,442,836</td>
<td>1,568,878</td>
<td>0.9%</td>
<td>1.0%</td>
<td>22%</td>
</tr>
<tr>
<td>Phoenix, AZ, CBSA</td>
<td>4,393,462</td>
<td>873,084</td>
<td>2.0%</td>
<td>2.4%</td>
<td>35%</td>
</tr>
<tr>
<td>Pittsburgh, PA, CSA</td>
<td>2,471,403</td>
<td>563,309</td>
<td>1.5%</td>
<td>1.4%</td>
<td>22%</td>
</tr>
<tr>
<td>Salt Lake City, UT, CSA</td>
<td>1,717,045</td>
<td>460,747</td>
<td>3.0%</td>
<td>3.0%</td>
<td>41%</td>
</tr>
<tr>
<td>San Antonio, TX, CBSA</td>
<td>2,061,147</td>
<td>484,473</td>
<td>0.5%</td>
<td>0.5%</td>
<td>12%</td>
</tr>
<tr>
<td>San Francisco, CA, CSA</td>
<td>7,497,443</td>
<td>1,675,711</td>
<td>2.6%</td>
<td>2.9%</td>
<td>41%</td>
</tr>
<tr>
<td>Seattle, WA, CSA</td>
<td>4,181,278</td>
<td>918,309</td>
<td>0.3%</td>
<td>0.3%</td>
<td>5%</td>
</tr>
<tr>
<td>St. Louis, MO, CSA</td>
<td>2,914,754</td>
<td>720,746</td>
<td>1.3%</td>
<td>1.5%</td>
<td>17%</td>
</tr>
</tbody>
</table>

Atlanta population fractions for children (<18 years of age) are similar to those for the general population, but other populations show a different pattern of monitor representativeness. Older adults (age 65 and up) were somewhat differently distributed with respect to the monitors, with most monitors showing a difference of more than a percentage point compared to the general population. Based on 2000 population data, the fraction of older adults closest to the two city center monitors (A and B) was 4% higher and 2% lower, respectively, than the fraction for the population as a whole. Site C showed the highest differential, with 21% of the total population but only 15% of the older adult population. This indicates the potential for monitors to differentially represent potentially susceptible populations.
4.5 Exposure Modeling

In the absence of personal exposure measurements, modeling techniques are used to estimate exposures, particularly for large populations for which individual-level measurements would be impractical. Model estimates may be used as inputs to epidemiologic studies or as stand-alone assessments of the level of exposure likely to be experienced by a population under certain air quality conditions. This section describes approaches used to improve exposure estimates, including concentration surface modeling, which calculates local outdoor concentrations over a geographic area; air exchange rate modeling, which estimates building ventilation based on housing characteristics and meteorological parameters; and microenvironment-based exposure modeling, which combines air quality data with demographic information and activity pattern simulations to estimate time-weighted exposures based on concentrations in multiple microenvironments. These models each have strengths and limitations, as summarized in Table 4-7. The remainder of this section provides more detail on specific modeling approaches, as well as results of applying the models.
### Table 4-7 Characteristics of exposure modeling approaches.

<table>
<thead>
<tr>
<th>Model Type</th>
<th>Model Description</th>
<th>Strengths</th>
<th>Limitations</th>
</tr>
</thead>
<tbody>
<tr>
<td>Concentration Surface</td>
<td>Spatial Interpolation (e.g., Inverse Distance Weighing, Kriging)</td>
<td>Measured concentrations are interpolated across an area to yield local outdoor concentration estimates</td>
<td>High concentration resolution; uses available data; requires low to moderate resources for implementation; Spatial heterogeneity not fully captured; a single high-concentration monitor can skew results; no location-activity information</td>
</tr>
<tr>
<td></td>
<td>Grid-based O3 concentrations are calculated from precursor emissions, meteorology, and atmospheric chemistry and physics</td>
<td>First-principles characterization of physical and chemical processes influencing O3 formation</td>
<td>Grid cell resolution; resource-intensive; no location-activity information</td>
</tr>
<tr>
<td></td>
<td>Land-use regression (LUR)</td>
<td>Merges concentration data with local-scale variables such as land use factors to yield local concentration surface</td>
<td>High concentration resolution; Reactivity and small-scale spatial variability of O3; location-specific, limiting generalizability; no location-activity information</td>
</tr>
<tr>
<td></td>
<td>Mechanistic (LBL, LBLX)</td>
<td>Uses database on building leakage tests to predict AER based on building characteristics and meteorological variables (including natural ventilation in LBLX)</td>
<td>Physical characterization of driving forces for air exchange; Moderate resource requirement; no location-activity information</td>
</tr>
<tr>
<td></td>
<td>Empirical</td>
<td>Predicts AER based on factors such as building age and floor area</td>
<td>Low input data requirements; Cannot account for meteorology; no location-activity information</td>
</tr>
<tr>
<td></td>
<td>Population (APEX, SHEDS)</td>
<td>Stochastic treatment of air quality data, demographic variables, and activity pattern to generate estimates of microenvironmental concentrations, exposures, and doses</td>
<td>Probabilistic estimates of exposure and dose distributions for specific populations; consideration of nonambient sources; small to moderate uncertainty for exercising asthmatic children (APEX); Resource-intensive; evaluation with measured exposures; underestimation of multiple high-exposure events in an individual (APEX)</td>
</tr>
</tbody>
</table>

#### 4.5.1 Concentration Surface Modeling

One approach to improve exposure estimates in urban areas involves construction of a concentration surface over a geographic area, with the concentration at locations between monitors estimated using a model to compensate for missing data. The calculated O₃ concentration surface can then be used to estimate exposures outside residences, schools, workplaces, roadways, or other locations of interest. This technique does not estimate exposure directly because it does not account for activity patterns or concentrations in different microenvironments. This is an important consideration in the utility of these methods for exposure assessment; while improved local-scale estimates of outdoor concentrations may contribute to better assignment of exposures, information on activity patterns is needed to produce...
estimates of personal exposure. There are three main types of approaches: spatial interpolation of measured concentrations; statistical models using meteorological variables, pollutant concentrations, and other predictors to estimate concentrations at receptors in the domain; and rigorous first-principle models, such as chemistry-transport models or dispersion models incorporating O\textsubscript{3} chemistry. Some researchers have developed models that combine these techniques. The models may be applied over urban, regional, or national spatial scales, and can be used to estimate daily concentrations or longer-term averages. This discussion will focus on short-term concentrations estimated across urban areas.

The 2006 O\textsubscript{3} AQCD (U.S. EPA, 2006b) discussed concentration surface models, focusing on chemistry-transport models as well as geospatial and spatiotemporal interpolation techniques (e.g., Christakos and Vyas, 1998a, b; Georgopoulos et al., 1997). Recent research has continued to refine and extend the modeling approaches. A few recent papers have compared different approaches for the same urban area.

Marshall et al. (2008) compared four spatial interpolation techniques for estimation of O\textsubscript{3} concentrations in Vancouver, BC. The investigators assigned a daily average O\textsubscript{3} concentration to each of the 51,560 postal-code centroids using one of the following techniques: (1) the concentration from the nearest monitor within 10 km; (2) the average of all monitors within 10 km; (3) the inverse-distance-weighted (IDW) average of all monitors in the area; and (4) the IDW average of the 3 closest monitors within 50 km. Method 1 (the nearest-monitor approach) and Method 4 (IDW-50 km) had similar mean and median estimated annual- and monthly-average concentrations, although the 10th-90th percentile range was smaller for IDW-50. This is consistent with the averaging of extreme values inherent in IDW methods. The Pearson correlation coefficient between the two methods was 0.93 for monthly-average concentrations and 0.78 for annual-average concentrations. Methods 2 and 3 were considered sub-optimal and were excluded from further analysis. In the case of Method 2, a single downtown high-concentration monitor skewed the results in the vicinity, partially as a result of the asymmetric layout of the coastal city of Vancouver. Method 3 was too spatially homogenous because it assigned most locations a concentration near the regional average, except for locations immediately adjacent to a monitoring site. CMAQ concentration estimates using a 4 km\textsuperscript{2} grid were also compared to the interpolation techniques in this study. Mean and median concentrations from CMAQ were approximately 50% higher than Method 1 and Method 4 estimates for both annual and monthly average concentrations. This may be due in part to the CMAQ grid size, which was too coarse to reveal near-roadway decrements in O\textsubscript{3} concentration due to titration by NO. The IQR for the annual average was similar between CMAQ and the interpolation techniques, but the monthly average CMAQ IQR was approximately twice as large, indicating a seasonal effect.

Bell (2006) compared CMAQ estimates for northern Georgia with nearest-monitor and spatial interpolation techniques, including IDW and kriging. The area-weighted concentration estimates from CMAQ indicated areas of spatial heterogeneity that were not captured by approaches based on the monitoring network. The author
concluded that some techniques, such as spatial interpolation, were not suitable for estimation of exposure in certain situations, such as for rural areas. Using the concentration from the nearest monitor resulted in an overestimation of exposure relative to model estimates.

Land use regression (LUR) models have been developed to estimate levels of air pollutants, predominantly NO₂, as a function of several land use factors, such as land use designation, traffic counts, home heating usage, point source strength, and population density (Ryan and LeMasters, 2007; Gilliland et al., 2005; Briggs et al., 1997). LUR, initially termed regression mapping (Briggs et al., 1997), is a regression derived from monitored concentrations as a function of data from a combination of the land use factors. The regression is then used for predicting concentrations at multiple locations based on the independent variables at those particular locations without monitors. Hoek et al. (2008) warn of several limitations of LUR, including distinguishing real associations between pollutants and covariates from those of correlated copollutants, limitations in spatial resolution from monitor data, applicability of the LUR model under changing temporal conditions, and introduction of confounding factors when LUR is used in epidemiologic studies. These limitations may partially explain the lack of LUR models that have been developed for O₃ at the urban scale. Brauer et al. (2008) evaluated the use of LUR and IDW-based spatial-interpolation models in epidemiologic analyses for several different pollutants in Vancouver, BC and suggested that LUR is appropriate for directly-emitted pollutants with high spatial variability, such as NO and BC, while IDW is appropriate for secondary pollutants such as NO₂ and PM₂.₅ with less spatial variability. Although O₃ is also a secondary pollutant, its reactivity and high small-scale spatial variability near high-traffic roadways indicates this conclusion may not apply for O₃.

At a much larger spatial scale, EU-wide, Beelen et al. (2009) compared a LUR model for O₃ with ordinary kriging and universal kriging, which incorporated meteorological, topographical, and land use variables to characterize the underlying trend. The LUR model performed reasonably well at rural locations (5-km resolution), explaining a higher percentage of the variability (R² = 0.62) than for other pollutants. However, at the urban scale (1-km resolution), only one variable was selected into the O₃ LUR model (high-density residential land use), and the R² value was very low (0.06). Universal kriging was the best method for the large-scale composite EU concentration map, for O₃ as well as for NO₂ and PM₁₀, with an R² value for O₃ of 0.70. The authors noted that these methods were not designed to capture spatial variation in concentrations that are known to occur within tens of meters of roadways (Section 3.6.2.1), which could partially explain poor model performance at the urban scale.

Titration of O₃ with NO emitted by motor vehicles tends to reduce O₃ concentrations near roadways. McConnell et al. (2006) developed a regression model to predict residential O₃ concentrations in southern California using estimates of residential NOₓ calculated from traffic data with the CALINE4 line source dispersion model. The annual average model results were well-correlated (R² = 0.97) with multi-year
average monitoring data. The authors estimated that local traffic contributes 18% of NO\textsubscript{X} concentrations measured in the study communities, with the remainder coming from regional background. Their regression model indicates that residential NO\textsubscript{X} reduces residential O\textsubscript{3} concentrations by 0.51 ppb (SE 0.11 ppb) O\textsubscript{3} per 1 ppb NO\textsubscript{X}, and that a 10th-90th percentile increase in local NO\textsubscript{X} results in a 7.5 ppb decrease in local O\textsubscript{3} concentrations. This intra-urban traffic-related variability in O\textsubscript{3} concentrations suggests that traffic patterns are an important factor in the relationship between central site monitor and residential O\textsubscript{3}, and that differences in traffic density between the central site monitor and individual homes could result in either an overestimate or underestimate of residential O\textsubscript{3}.

A substantial number of researchers have used geostatistical methods and chemistry-transport models to estimate O\textsubscript{3} concentrations at urban, regional, national, and continental scales, both in the U.S. and in other countries (Section 3.3). In addition to short-term exposure assessment for epidemiologic studies, such models may also be used for long-term exposure assessment, O\textsubscript{3} forecasts, or evaluating emission control strategies. However, as discussed at the beginning of this section, caveats regarding the importance of activity pattern information in estimating personal and population exposure should be kept in mind.

### 4.5.2 Residential Air Exchange Rate Modeling

The residential air exchange rate (AER), which is the airflow into and out of a home, is an important mechanism for entry of ambient O\textsubscript{3}. As described in Section 4.3.2, the indoor-outdoor relationship is greatly affected by the AER. Since studies show that people spend approximately 66% of their time indoors at home (Leech et al., 2002; Klepeis et al., 2001), the residential AER is a critical parameter for exposure models, such as APEX, SHEDS, and EMI (discussed in Section 4.5.3) (U.S. EPA, 2011c, 2009b; Burke et al., 2001). Since the appropriate AER measurements may not be available for exposure models, mechanistic and empirical (i.e., regression-based) AER models can be used for exposure assessments. The input data for the AER models can include building characteristics (e.g., age, number of stories, wind sheltering), occupant behavior (e.g., window opening), climatic region, and meteorology (e.g., local temperature and wind speed). Mechanistic AER models use these meteorological parameters to account for the physical driving forces of the airflows due to pressure differences across the building envelope from wind and indoor-outdoor temperature differences (ASHRAE, 2009). Empirical AER models do not consider the driving forces from the wind and indoor-outdoor temperature differences. Instead, a scaling constant can be used based on factors such as building age and floor area (Chan et al., 2005b).

Single-zone mechanistic models represent a whole-building as a single, well-mixed compartment. These AER models, such as the Lawrence Berkeley Laboratory (LBL) model, can predict residential AER using input data from whole-building pressurization tests (Sherman and Grimsrud, 1980), or leakage area models (Breen et
al., 2010; Sherman and McWilliams, 2007). Recently, the LBL air infiltration model was linked with a leakage area model using population-level census and residential survey data (Sherman and McWilliams, 2007) and individual-level questionnaire data (Breen et al., 2010). The LBL model, which predicts the AER from air infiltration (i.e., small uncontrollable openings in the building envelope) was also extended to include airflow from natural ventilation (LBLX), and evaluated using window opening data (Breen et al., 2010). The AER predictions from the LBL and LBLX models were compared to daily AER measurements on seven consecutive days during each season from detached homes in central North Carolina (Breen et al., 2010). For the individual model-predicted and measured AER, the median absolute difference was 43% (0.17 h⁻¹) and 40% (0.17 h⁻¹) for the LBL and LBLX models, respectively. Given the uncertainty of the AER measurements (accuracy of 20-25% for occupied homes), these results demonstrate the feasibility of using these AER models for both air infiltration (e.g., uncontrollable openings) and natural ventilation (e.g., window opening) to help reduce the AER uncertainty in exposure models. The capability of AER models could help support the exposure modeling needs, as described in Section 4.5.3, which includes the ability to predict indoor concentrations of ambient O₃ that may be substantial for conditions of high AER such as open windows.

### 4.5.3 Microenvironment-Based Models

Population-based methods, such as the Air Pollution Exposure (APEX) and Stochastic Human Exposure and Dose Simulation (SHEDS) integrated microenvironmental exposure and dose models, involve stochastic treatment of the model inputs (U.S. EPA, 2009b; Burke et al., 2001). These are described in detail in the 2008 NOₓ ISA (U.S. EPA, 2008c), in AX3.6.1. Stochastic models utilize distributions of pollutant-related and individual-level variables, such as ambient and local O₃ concentration contributions and breathing rate respectively, to compute the distribution of individual exposures across the modeled population. The models also have the capability to estimate received dose through a dosimetry model. Using distributions of input parameters in the model framework rather than point estimates allows the models to incorporate uncertainty and variability explicitly into exposure estimates (Zidek et al., 2007). These models estimate time-weighted exposure for modeled individuals by summing exposure in each microenvironment visited during the exposure period.

The initial set of input data for population exposure models is ambient air quality data, which may come from a monitoring network or model estimates. Estimates of concentrations in a set of microenvironments are generated either by mass balance methods, which can incorporate AER models (Section 4.5.2), or microenvironmental factors. Microenvironments modeled include indoor residences; other indoor locations, such as schools, offices, and public buildings; and vehicles. The sequence of microenvironments and exertion levels during the exposure period is determined from characteristics of each modeled individual. The APEX model does this by
generating a profile for each simulated individual by sampling from distributions of demographic variables such as age, sex, and employment; physiological variables such as height and weight; and situational variables such as living in a house with a gas stove or air conditioning. Activity and location (microenvironmental) patterns from a database such as CHAD are assigned to the simulated individual in a longitudinal manner, using age, sex, and biometric characteristics (U.S. EPA, 2009a; Glen et al., 2008). Breathing rates for each individual are calculated for each activity based on predicted energy expenditures, and the corresponding dose may then be computed. APEX has an algorithm to estimate O₃ dose and changes in FEV₁ resulting from O₃ exposure. Summaries of individual- and population-level metrics are produced, such as maximum exposure or dose, number of individuals exceeding a specified exposure/dose, and number of person-days at or above benchmark exposure levels. The models also consider the nonambient contribution to total exposure. Nonambient source terms are added to the infiltration of ambient pollutants to calculate the total concentration in the microenvironment. Output from model runs with and without nonambient sources can be compared to estimate the ambient contribution to total exposure and dose.

Georgopoulos et al. (2005) used a version of the SHEDS model as the exposure component of a modeling framework known as MENTOR (Modeling Environment for Total Risk Studies) in a simulation of O₃ exposure in Philadelphia over a 2-week period in July 1999. Five hundred (500) individuals were sampled from CHAD in each of 482 census tracts to match local demographic characteristics from U.S. Census data. Outdoor concentrations over the modeling domain were calculated from interpolation of photochemical modeling results and fixed-site monitor concentrations. These concentrations were then used as input data for SHEDS. Median microenvironmental concentrations predicted by SHEDS for nine simulated microenvironments were strongly correlated with outdoor concentrations, a result consistent with the lack of indoor O₃ sources in the model. A regression of median microenvironmental concentrations against outdoor concentrations indicated that the microenvironmental concentrations were appreciably lower than outdoor concentrations (regression slope = 0.26). 95th percentile microenvironmental concentrations were also well correlated with outdoor concentrations and showed a regression slope of 1.02, although some microenvironmental concentrations were well below the outdoor values. This suggests that in most cases the high-end concentrations were associated with outdoor microenvironments. Although the authors did not report exposure statistics for the population, their dose calculations also indicated that O₃ dose due to time spent outdoors dominated the upper percentiles of the population dose distribution. They found that both the 50th and 95th percentile O₃ concentrations were correlated with census-tract level outdoor concentrations estimated by photochemical modeling combined with spatiotemporal interpolation, and attributed this correlation to the lack of indoor sources of O₃. Relationships between exposure and concentrations at fixed-site monitors were not reported.

An analysis has been conducted for the APEX model to evaluate the contribution of uncertainty in input parameters and databases to the uncertainty in model outputs.
Langstaff, 2007). The Monte Carlo analysis indicates that the uncertainty in model exposure estimates for asthmatic children during moderate exercise is small to moderate, with 95% confidence intervals of at most ± 6 percentage points at exposures above 60, 70, and 80 ppb (8-h avg). However, APEX appears to substantially underestimate the frequency of multiple high-exposure events for a single individual. The two main sources of uncertainty identified were related to the activity pattern database and the spatial interpolation of fixed-site monitor concentrations to other locations. Additional areas identified in the uncertainty analysis for potential improvement include: further information on children’s activities, including longitudinal patterns in the activity pattern database; improved information on spatial variation of O3 concentrations, including in near-roadway and indoor microenvironments; and data from personal exposure monitors with shorter averaging times to capture peak exposures and lower detection limits to capture low indoor concentrations. A similar modeling approach has been developed for panel epidemiologic studies or for controlled human exposure studies, in which activity pattern data specific to the individuals in the study can be collected. Time-activity data is combined with questionnaire data on housing characteristics, presence of indoor or personal sources, and other information to develop a personalized set of model input parameters for each individual. This model, the Exposure Model for Individuals, has been developed by EPA’s National Exposure Research Laboratory (U.S. EPA, 2011c; Zartarian and Schultz, 2010).

### 4.6 Implications for Epidemiologic Studies

Exposure measurement error, which refers to the uncertainty associated with using exposure metrics to represent the actual exposure of an individual or population, can be an important contributor to variability in epidemiologic study results. Time-series studies assess the daily health status of a population of thousands or millions of people over the course of multiple years (i.e., thousands of days) across an urban area by estimating their daily exposure using a short monitoring interval (hours to days). In these studies, the community-averaged concentration of an air pollutant measured at central-site monitors is typically used as a surrogate for individual or population ambient exposure. In addition, panel studies, which consist of a relatively small sample (typically tens) of study participants followed over a period of days to months, have been used to examine the health effects associated with short-term exposure to ambient concentrations of air pollutants (Delfino et al., 1996). Panel studies may also apply a microenvironmental model to represent exposure to an air pollutant. A longitudinal cohort epidemiologic study, such as the ACS cohort study, typically involves hundreds or thousands of subjects followed over several years or decades (Jerrett et al., 2009). Concentrations are generally aggregated over time and by community to estimate exposures.

Exposure error can under- or over-estimate epidemiologic associations between ambient pollutant concentrations and health outcomes by biasing effect estimates toward or away from the null, and tends to widen confidence intervals around those
estimates (Sheppard et al., 2005; Zeger et al., 2000). Exposure misclassification can also tend to obscure the presence of potential thresholds for health effects, as demonstrated by a simulation study of nondifferential exposure misclassification (Brauer et al., 2002). The importance of exposure misclassification varies with study design and is dependent on the spatial and temporal aspects of the design. For example, the use of a community-averaged O₃ concentration in a time-series epidemiologic study may be adequate to represent the day-to-day temporal concentration variability used to evaluate health effects, but may not capture differences in the magnitude of exposure due to spatial variability. Other factors that could influence exposure estimates include nonambient exposure, topography of the natural and built environment, meteorology, measurement errors, use of ambient O₃ concentration as a surrogate for ambient O₃ exposure, and the presence of O₃ in a mixture of pollutants. The following sections will consider various sources of error and how they affect the interpretation of results from epidemiologic studies of different designs.

### 4.6.1 Non-Ambient Ozone Exposure

For other criteria pollutants, nonambient sources can be an important contributor to total personal exposure. There are relatively few indoor sources of O₃; as a result, personal O₃ exposure is expected to be dominated by ambient O₃ in outdoor microenvironments and in indoor microenvironments with high air exchange rates (e.g., with open windows). Even in microenvironments where nonambient exposure is substantial, such as in a room with an O₃ generator, this nonambient exposure is unlikely to be temporally correlated with ambient O₃ exposure (Wilson and Suh, 1997), and therefore would not affect epidemiologic associations between O₃ and a health effect (Sheppard et al., 2005). In simulations of a nonreactive pollutant, Sheppard et al. (2005) concluded that nonambient exposure does not influence the health outcome effect estimate if ambient and nonambient concentrations are independent. Since personal exposure to ambient O₃ is some fraction of the ambient concentration, it should be noted that effect estimates calculated based on personal exposure rather than ambient concentration will be increased in proportion to the ratio of ambient concentration to ambient exposure, and daily fluctuations in this ratio can widen the confidence intervals in the ambient concentration effect estimate, but uncorrelated nonambient exposure will not bias the effect estimate (Sheppard et al., 2005; Wilson and Suh, 1997).

### 4.6.2 Spatial and Temporal Variability

Spatial and temporal variability in O₃ concentrations can contribute to exposure error in epidemiologic studies, whether they rely on central-site monitor data or concentration modeling for exposure assessment. Spatial variability in the magnitude of concentrations may affect cross-sectional and large-scale cohort studies by
undermining the assumption that intra-urban concentration and exposure differences are less important than inter-urban differences. This issue may be less important for time-series studies, which rely on day-to-day temporal variability in concentrations to evaluate health effects. Low inter-monitor correlations contribute to exposure error in time-series studies, including bias toward the null and increased confidence intervals.

4.6.2.1 Spatial Variability

Spatial variability of O₃ concentrations is highly dependent on spatial scale; in effect, O₃ is a regional pollutant subject to varying degrees of local variability. In the immediate vicinity of roadways, O₃ concentrations are reduced due to reaction with NO and other species (Section 4.3.4.2); over spatial scales of a few kilometers, O₃ may be more homogeneous due to its formation as a secondary pollutant; over scales of tens of kilometers, atmospheric processing can result in higher concentrations downwind of an urban area than in the urban core. Local-scale variations have a large impact on the relative magnitude of concentrations among urban monitors, while conditions favoring high or low rates of O₃ formation (e.g., temperature) vary over large spatial scales. This suggests that neighborhood monitors are likely to track one another temporally, but miss small-scale spatial variability in magnitude. This is supported by an analysis in Atlanta, GA, that found correlations greater than 0.8 for daily O₃ concentration metrics (1-h max, 8-h max, and 24-h avg) measured at monitors 10-60 km apart (Darrow et al., 2011a). In rural areas, a lower degree of fluctuation in O₃ precursors such as NO and VOCs is likely to make the diel concentration profile less variable than in urban areas, resulting in more sustained ambient levels. Spatial variability contributes to exposure error if the ambient O₃ concentration measured at the central site monitor is used as an ambient exposure surrogate and differs from the actual ambient O₃ concentration outside a subject’s residence and/or worksite (in the absence of indoor O₃ sources). Averaging data from a large number of samplers will dampen intersampler variability, and use of multiple monitors over smaller land areas may allow for more variability to be incorporated into an epidemiologic analysis.

Community exposure may not be well represented when monitors cover large areas with several subcommunities having different sources and topographies, such as the Los Angeles, CA, CSA (Section 3.6.2.1 and Section 4.4.3). Ozone monitors in Los Angeles had a much wider range of intermonitor correlations (-0.06 to 0.97) than Atlanta, GA, (0.61 to 0.96) or Boston, MA, (0.56 to 0.97) using 2007-2009 data. Although the negative and near-zero correlations in Los Angeles were observed for monitors located some distance apart (>150 km), some closer monitor pairs had low positive correlations, likely due to changes in land use, topography, and airflow patterns over short distances. Lower COD values, which indicate less variability among monitors in the magnitude of O₃ concentrations, were observed in Atlanta (0.05-0.13) and Boston (0.05-0.19) than Los Angeles (0.05-0.56), although a single monitor (AM) was responsible for all Los Angeles COD values above 0.40.
The spatial and temporal variability in O₃ concentration in 24 MSAs across the U.S. was also examined in the 2006 O₃ AQCD (U.S. EPA, 2006b) by using Pearson correlation coefficients, values of the 90th percentile of the absolute difference in O₃ concentrations, and CODs. No clear discernible regional differences across the U.S. were found in the ranges of parameters analyzed.

An analysis of the impact of exposure error due to spatial variability and instrument imprecision on time-series epidemiologic study results indicated that O₃ has relatively low exposure error compared to other routinely monitored pollutants, and that the simulated impact on effect estimates is minor. Goldman et al. (2011) computed population-weighted scaled semivariances and Pearson correlation coefficients for daily concentration metrics of twelve pollutants measured at multiple central-site monitors in Atlanta, GA. The 8-h daily max O₃ exhibited the lowest semivariance and highest correlation of any of the pollutants. Although this indicates some degree of urban-scale homogeneity for O₃, the analysis did not account for near-road effects on O₃ concentrations.

Studies evaluating the influence of monitor selection on epidemiologic study results have found that O₃ effect estimates are similar across different spatial averaging scales and monitoring sites. A study in Italy compared approaches for using fixed-site monitoring data in a case-crossover epidemiologic study of daily O₃ and mortality (Zauli Sajani et al., 2011). Ozone effect estimates were found to be similar whether the nearest monitor was used, or whether single-city, three-city, or six-city regional averages were used for exposure assessment. In contrast, effect estimates for PM₁₀ and NO₂ increased with increasing scale of spatial averaging. Confidence intervals increased with increasing spatial scale for all pollutants. The authors attributed the consistency of O₃ effect estimates to the relative spatial homogeneity of O₃ over multi-km spatial scales, and pointed to the high (0.85-0.95) inter-monitor correlations to support this. The use of background monitors rather than monitors influenced by local sources in this study suggests that local-scale spatial variation in O₃, such as that due to titration by traffic emissions, was not captured in the analyses. A multi-city U.S. study of asthmatic children found comparable respiratory effect estimates when restricting the analysis to the monitors closest to the child’s zip code centroid as when using the average of all monitors in the urban area (Mortimer et al., 2002), suggesting little impact of monitor selection. Sarnat et al. (2010) studied the spatial variability of O₃, along with PM₂.₅, NO₂, and CO, in the Atlanta, GA, metropolitan area and evaluated how spatial variability affects interpretation of epidemiologic results, using time-series data for circulatory disease ED visits. The authors found that associations with ambient 8-h daily max O₃ concentration were similar among all sites tested, including multiple urban sites and a rural site some 38 miles from the city center. This result was also observed for 24-hour PM₂.₅ concentrations. In contrast, hourly CO and NO₂ showed different associations for the rural site than the urban sites, although the urban site associations were similar to one another for CO. This suggests that the choice of monitor may have little impact on the results of O₃ time-series studies, consistent with the moderate to high inter-monitor correlations observed in Atlanta (Chapter 3).
One potential explanation for this finding from the study by Sarnat et al. (2010) is that although spatial variability at different scales contributes to a complicated pattern of variations in the magnitude of O₃ concentrations between near-road, urban core, and urban downwind sites, day-to-day fluctuations in concentrations may be reflected across multiple urban microenvironments. In addition, time-averaging of O₃ and PM₂.₅ concentrations may smooth out some of the intra-day spatial variability observed with the hourly CO and NO₂ concentrations. However, some uncertainty in observed effect estimates due to spatial variability and associated exposure error is expected to remain, including a potential bias toward the null.

### 4.6.2.2 Seasonality

The relationship between personal exposure and ambient concentration has been found to vary by season, with at least three factors potentially contributing to this variation: differences in building ventilation (e.g., air conditioning or heater use versus open window ventilation), higher O₃ concentrations during the O₃ season contributing to increased exposure and improved detection by personal monitors; and changes in activity pattern resulting in more time spent outside. Evidence has been presented in studies conducted in several cities regarding the effect of ventilation on personal-ambient and indoor-outdoor O₃ relationships (see Section 4.3.2 and Section 4.3.3). More limited evidence is available regarding the specific effects of O₃ detection limits and activity pattern changes on O₃ relationships.

Several studies have found increased summertime correlations or ratios between personal exposure and ambient concentration (Sarnat et al., 2005; Sarnat et al., 2000) or between indoor and outdoor O₃ concentrations (Geyh et al., 2000; Avol et al., 1998a). However, others have found higher ratios in fall than in summer (Sarnat et al., 2006a) or equivalent, near-zero ratios in winter and summer (Sarnat et al., 2001), possibly because summertime use of air conditioners decreases building air exchange rates. It should be noted that O₃ concentrations during winter are generally much lower than summertime concentrations, possibly obscuring wintertime relationships due to detection limit issues. Studies specifically evaluating the effect of ventilation conditions on O₃ relationships have found increased correlations or ratios for individuals or buildings experiencing higher air exchange rates (Sarnat et al., 2006a; Geyh et al., 2000; Sarnat et al., 2000; Romieu et al., 1998a).

Increased correlations or ratios between personal exposure and ambient concentration, or between indoor and outdoor concentration, are likely to reduce error in exposure estimates used in epidemiologic studies. This suggests that studies conducted during the O₃ season or in periods when communities are likely to have high air exchange rates (e.g., during mild weather) may be less prone to exposure error than studies conducted only during winter. Year-round studies that include both the O₃ and non-O₃ seasons may have an intermediate level of exposure error.
4.6.3 Exposure Duration

Epidemiologic studies of health effects associated with short-term and long-term exposures use different air pollution metrics and thus have different sources of exposure error. The following subsections discuss the impact of using different short-term and long-term exposure metrics on epidemiologic results.

4.6.3.1 Short-Term Exposure

The averaging time of the daily exposure metrics used to evaluate daily aggregated health data (e.g., 1-h or 8-h daily max, versus 24-h avg concentration) may also impact epidemiologic results, since different studies report different daily metrics. Correlations between 1-h daily max, 8-h daily max, and 24-h avg concentrations for U.S. monitoring sites are presented in Section 3.6.1 (Figure 3-23 and accompanying text). The two daily peak values (1-h max and 8-h max) are well correlated, with a median (IQR) correlation of 0.97 (0.96-0.98). The correlation between the 8-h max and 24-h avg are somewhat less well correlated with a median (IQR) correlation of 0.89 (0.86-0.92). While this may complicate quantitative comparisons between epidemiologic studies using different daily metrics, as well as the interpretation of studies using metrics other than the current 8-hour standard, the high inter-metric correlations suggest it is a relatively small source of uncertainty in comparing the results of studies using different metrics. This is supported by a study comparing each of these metrics in a time-series study of respiratory ED visits (Darrow et al., 2011a), which found positive associations for all metrics, with the strongest association for the 8-h daily max exposure metric (Section 6.2.7.3).

The ratios of 1-h daily max, 8-h daily max, and 24-h avg concentrations to one another have been found to differ across communities and across time within individual communities (Anderson and Bell, 2010). For example, 8:24 hour ratios ranged from 1.23-1.83, with a median of 1.53. Lower ratios were generally observed in the spring and summer compared to fall and winter. Ozone concentration was identified as the most important predictor of O₃ metric ratios, with higher overall O₃ concentrations associated with lower ratios. In communities with higher long-term O₃ concentrations, the lower 8:24 hour ratio is attributed to high baseline O₃, which results in elevated 24-h average values. Differences in the representativeness of O₃ metrics introduces uncertainty into the interpretation of epidemiologic results and complicates comparison of studies using different metrics. Preferably, studies will report results using multiple metrics. In cases where this does not occur, the results of the study by Anderson and Bell (2010) can inform the uncertainty associated with using a standard increment to adjust effect estimates based on different metrics so that they are comparable (Chapter 6).

A study compared measures of spatial and temporal variability for 1-h daily max and 24-h daily avg O₃ concentrations in Brazil (Bravo and Bell, 2011). The 1-h daily max value was found to have higher correlation between monitors (i.e., lower
temporal variability) and lower COD (a measure of spatiotemporal variability which incorporates differences in concentration magnitude, with lower values indicating lower variability; see Chapter 3) than the 24-h avg value. The range of correlation coefficients and COD values was similar between the two metrics, although the variation was lower for the 1-h daily max, as indicated by the $R^2$ value for the regression of correlation coefficient on inter-monitor distance.

### 4.6.3.2 Long-Term Exposure

A study in Canada suggests that an exposure metric based on a single year can represent exposure over a multi-decade period. The authors compared exposure assessment methods for long-term O$_3$ exposure and found that the annual average concentration in the census tract of a subject’s residence during 1980 and 1994 was well-correlated (0.76 and 0.82, respectively) with a concentration metric accounting for movement among census subdivisions during 1980-2002 (Guay et al., 2011). This may have been due in part to a relatively low rate of movement, with subjects residing on average for 71% of the 22-year period in the same census subdivision they were in during 1980.

Analysis of the exposure assessment methodology in a recent study of mortality associated with long-term O$_3$ exposure (Jerrett et al., 2009) is illustrative. In this study, the authors computed quarterly averages of the daily 1-h max O$_3$ concentration, averaged the two summer quarters together to produce an annual value, then calculated a 23-year average value for each city in the study. Producing a single value for each city enables a comparison of relatively cleaner cities with relatively more polluted cities. In this case, the average was calculated using the 1-h daily max value; if the 24-h avg value had been used, concentrations would have been lower and potentially more variable, based on analyses in Chapter 3. According to Table 3-7, the 2007-2009, 3-year average 1-h daily max value during the warm season was approximately 50% higher than the corresponding 24-avg value on a nationwide basis. Correlation between the two metrics varies by site, indicating the differential influence of the overnight period on 24-h avg concentrations. The median correlation between 1-h daily max and 24-h avg is 0.83, with an IQR of 0.78-0.88. It is not clear, however, that a different exposure assignment method would yield different results.

Long-term O$_3$ trends, as discussed in Chapter 3, show gradually decreasing concentrations. Figure 3-48 shows that concentrations have decreased most for the 90th percentile, with relatively little change among the 10th percentile monitors. The decrease has been greater in the eastern U.S. than in the western part of the country (excluding California). For the most part, the rank order of regions in terms of O$_3$ concentration has remained the same, as shown in Figure 3-50, with the Northeast, Southeast, and California exhibiting the highest concentrations. The decreasing trend is consistent across nearly all monitors in the U.S., with only 1-2% of monitors reporting an increase of more than 5 ppb between the 2001-2003...
and 2008-2010 time periods (Figure 3-52 and Figure 3-53). This figure provides some evidence that epidemiologic studies of long-term exposure are not affected by drastic changes in O₃ concentration, such as a relatively clean city becoming highly polluted or the reverse.

A few epidemiologic studies have evaluated the impact of distance to monitor on associations between long-term O₃ concentration and reproductive outcomes, as discussed in Chapter 7. It is not clear from this evidence whether using a local monitor for these multi-month concentration averages improves exposure assessment. Jalaludin et al. (2007) found somewhat higher effect estimates for women living within 5 km of a fixed-site O₃ monitor than for all women in the Sydney, Australia, metropolitan area, suggesting that increased monitor proximity reduced exposure misclassification. In contrast, Darrow et al. (2011b) found no substantial difference between effect estimates for those living within 4 mi of a fixed-site monitor and those living in the five-county area around Atlanta, GA. This result could be due to spatial variability over smaller scales than the 4-mi radius evaluated, time spent away from the residence impacting O₃ exposure, or similarity in monitor location and representativeness across the urban area (see Figure 4-4). At this time, the effect of exposure error on long-term exposure epidemiologic studies is unclear.

### 4.6.4 Relationship between Personal Exposure and Ambient Concentration

Personal exposure is generally moderately correlated with ambient O₃ concentration, although the magnitude of personal exposures is often much lower than the magnitude of ambient concentrations (Section 4.3.3). Moderate correlation between personal exposure and ambient concentration indicates that concentration-based exposure metrics are capturing the variability in exposure needed for epidemiologic studies, particularly for time-series and panel studies. Low personal-ambient correlations reported in the literature are strongly influenced by high detection limits of personal samplers. This results in a high fraction of personal samples below the detection limit that include substantial random variation and are thus unable to provide a signal that could correlate with variations in ambient concentration. Low correlations in situations with a high proportion of samples below the detection limit should not be interpreted as evidence for the lack of a relationship between personal exposure and ambient O₃ concentrations. To the extent that true correlations are less than one, epidemiologic effect estimates based on ambient concentration will be biased toward the null (Zeger et al., 2000). High detection limits are less of an issue for ratios of personal exposure to ambient concentration, for which a low personal sample value likely represents an actual low exposure, and thus appropriately leads to a low ratio. Low ratios result from low penetration and high reaction of O₃ in indoor microenvironments where people spend most of their time. This results in attenuation of the magnitude of the exposure-based effect estimate or response function relative to the ambient concentration-based response function.
(see Equation 4-5), although if the ratio is approximately constant over time, the strength of the statistical association would be similar for concentration- and exposure-based effect estimates (Sheppard, 2005; Sheppard et al., 2005).

In addition to the effect of the correlations and ratios themselves, spatial variation in their values across urban areas also impacts epidemiologic results. In this case, the exposure error is not likely to cause substantial bias, but tends more toward widening confidence intervals, thus reducing the precision of the effect estimate (Zeger et al., 2000). This loss of precision is due to the Berkson-like nature of this spatial variation, in which individual or subpopulation correlations and ratios tend to vary about the overall population mean.

Long-term O3 exposure studies are not available that permit evaluation of the relationship between long-term O3 concentrations and personal or population exposure. The value of short-term exposure data for evaluating long-term concentration-exposure relationships is uncertain. If the longer averaging time (annual, versus daily or hourly) smooths out short-term fluctuations, long-term concentrations may be well-correlated with long-term exposures. However, lower correlation between long-term exposures and ambient concentration could occur if important exposure determinants change over a period of several years, including activity pattern and residential air exchange rate.

### 4.6.5 Exposure to Copollutants and Ozone Reaction Products

Although indoor O3 concentrations are usually well below ambient concentrations, the same reactions that reduce O3 indoors form particulate and gaseous species, including other oxidants, as summarized in Section 4.3.4.3. Exposures to these reaction products would therefore be expected to be correlated with ambient O3 concentrations, although no evidence was identified regarding personal exposures. Such exposure could potentially contribute to health effects observed in epidemiologic studies.

### 4.6.6 Averting Behavior

As described in Section 4.4.2, several recent studies indicate that some lifestages and populations alter their behavior on high O3 days to avoid exposure. Such behavioral responses to information about forecasted air quality may introduce systematic measurement error in air pollution exposure, leading to biased estimates of the impact of air pollution on health. For example, studies have hypothesized that variation in time spent outdoors may be a driving factor behind the considerable heterogeneity in O3 mortality impacts across communities (Bell et al., 2004). If averting behavior reduces outdoor O3 exposure, then studies that do not account for averting behavior may produce effect estimates that are biased toward the null (Section 6.2.7.2).
This is supported by an epidemiologic study that examined the association between exposure to ambient O\textsubscript{3} concentrations and asthma hospitalizations in Southern California during 1989-1997, which indicates that controlling for avoidance behavior increases the effect estimate for both children and older adults, but not for adults aged 20-64 (Neidell and Kinney, 2010; Neidell, 2009). Figure 4-7 and Figure 4-8, reproduced from Neidell (2009), show covariate-adjusted asthma hospital admissions as a function of daily maximum 1-hour O\textsubscript{3} concentration for all days (gray line) and days when no O\textsubscript{3} alert was issued (black line). Stage 1 smog alerts were issued by the State of California for days when ambient O\textsubscript{3} concentrations were forecast to be above 0.20 ppm; however, the concentration-response functions are based on measured O\textsubscript{3} concentrations. For children aged 5-19 (Figure 4-7), hospital admissions were higher on high-O\textsubscript{3} days when no alert was issued, especially on days with O\textsubscript{3} concentrations above 0.15 ppm (150 ppb). The concentration-response curves for all days and days with no alert diverge at measured O\textsubscript{3} concentrations between 0.10 and 0.15 ppm because smog alerts begin to be issued more frequently in this range. This suggests that in the absence of information that would enable averting behavior, children experience higher O\textsubscript{3} exposure and subsequently a greater number of asthma hospital admissions than on alert days with similar O\textsubscript{3} concentrations. The lower rate of admissions observed when alert days were included in the analysis suggests that averting behavior reduced O\textsubscript{3} exposure and asthma hospital admissions. In both cases, O\textsubscript{3} was found to be associated with asthma hospital admissions, although the strength of the association is underestimated when not accounting for averting behavior. A different result was observed when examining associations for adults aged 20-64 (Figure 4-8), who had similar rates of hospital admissions on non-alert days as on all days. The lack of change for adults aged 20-64, which is primary employment age, may reflect lower response to air quality alerts due to the increased opportunity cost of behavior change. The finding that air quality information reduces the daily asthma hospitalization rate in these populations provides additional support for a link between O\textsubscript{3} and health effects.
Figure 4-7  Adjusted asthma hospital admissions by age on lagged O₃ by alert status, ages 5-19 years old.

Figure 4-8  Adjusted asthma hospital admissions by age on lagged O₃ by alert status, ages 20-64 years old.
4.6.7 Exposure Estimation Methods in Epidemiologic Studies

Epidemiologic studies use a variety of methods to assign exposure. Study design, data availability, and research objectives are all important factors for epidemiologists when selecting an exposure assessment method. Common methods for assigning exposure using monitoring data include using a single fixed-site monitor to represent population exposure, averaging concentrations from multiple monitors, and selecting the closest monitor. Investigators may also use statistical adjustment methods, such as trimming extreme values, to prepare the concentration data set. Panel or small-scale cohort studies involving dozens of individuals may use more individualized concentration measurements, such as personal exposures, residential indoor or outdoor measurements, or concentration data from local study-specific monitors. For long-term epidemiologic studies, the lack of personal exposure data or dedicated measurements means that investigators must rely on fixed-site monitoring data. These data may be used directly, averaged across counties or other geographic areas, or used to construct geospatial or regression models to assign concentrations to unmonitored locations. Longer-term averages (months to years) are typically used (e.g., in studies discussed in Section 7.3.1.1). Chapters 6 and 7 describe the exposure assessment methods used in the epidemiologic studies described therein, providing additional detail on studies with innovative or expanded techniques designed to improve exposure assessment and reduce exposure error.

The use of O₃ measurements from central ambient monitoring sites is the most common method for assigning exposure in epidemiologic studies. However, fixed-site measurements do not account for the effects of spatial variation in O₃ concentration, ambient and non-ambient concentration differences, and varying activity patterns on personal exposures (Brown et al., 2009; Chang et al., 2000; Zeger et al., 2000). Inter-individual variability in exposure error across a population will be minimal when: (1) O₃ concentrations are uniform across the region; (2) personal activity patterns are similar across the population; and (3) housing characteristics, such as air exchange rate and indoor reaction rate, are constant over the study area. To the extent that these factors vary by location and population, there will be errors in the magnitude of total exposure based solely on ambient monitoring data.

Modeled concentrations can also be used as exposure surrogates in epidemiologic studies, as discussed in Section 4.5. Geostatistical spatial interpolation techniques, such as IDW and kriging, can provide finer-scale estimates of local concentration over urban areas. A microenvironmental modeling approach simulates exposure using empirical distributions of concentrations in specific microenvironments together with human activity pattern data. The main advantage of the modeling approach is that it can be used to estimate exposures over a wide range of population and scenarios. However, this probabilistic, distribution-based approach is not well-suited to estimate exposures for specific individuals, such as might be needed for
cohort or panel epidemiologic studies. Another main disadvantage of the modeling approach is that the results of modeling exposure assessment must be compared to an independent set of measured exposure levels (Klepeis, 1999). In addition, resource-intensive development of validated and representative model inputs is required, such as human activity patterns, distributions of air exchange rate, and deposition rate. Therefore, modeled exposures are used much less frequently in epidemiologic studies.

4.7 Summary and Conclusions

This section will briefly summarize and synthesize the main points of the chapter, with particular attention to the relevance of the material for the interpretation of epidemiologic studies.

Passive badge samplers are the most widely used technique for measuring personal O₃ exposure (Section 4.3.1). The detection limit of the badges for a 24-hour sampling period is approximately 5-10 ppb, with lower detection limits at longer sampling durations. In low-concentration conditions this may result in an appreciable fraction of 24-hour samples being below the detection limit. The use of more sensitive portable active monitors, including some that have recently become available, may help overcome this issue and improve personal monitoring in the future.

Since there are relatively few indoor sources of O₃, indoor O₃ concentrations are often substantially lower than outdoor concentrations due to reactions of O₃ with indoor surfaces and airborne constituents (Section 4.3.2). Air exchange rate is a key determinant of the I/O ratio, which is generally in the range of 0.1-0.4 (Table 4-1), with some evidence for higher ratios during the O₃ season when concentrations are higher.

Personal exposure is moderately correlated with ambient O₃ concentration, as indicated by studies reporting correlations generally in the range of 0.3-0.8 (Table 4-2). Hourly concentration correlations are more variable than those averaged over 24 hours or longer, with correlations in outdoor microenvironments (0.7-0.9) much higher than those in residential indoor (0.1) or other indoor (0.3-0.4) microenvironments. Some studies report substantially lower personal-ambient correlations, a result attributable in part to low air exchange rate and O₃ concentrations below the sampler detection limit, conditions often encountered during wintertime. Low correlations may also occur for individuals or populations spending substantial time indoors.

The ratio between personal exposure and ambient concentration varies widely depending on activity patterns, housing characteristics, and season, with higher personal-ambient ratios generally observed with increasing time spent outside, higher air exchange rate, and in seasons other than winter (Table 4-3). Personal-ambient ratios are typically 0.1-0.3, although individuals spending substantial time outdoors
(e.g., outdoor workers) may have much higher ratios (0.5-0.9). Low personal-
ambient ratios result in attenuation of the magnitude of the exposure-based effect
estimate or response function relative to the concentration-based response function,
although the statistical association is similar for concentration- and exposure-based
effect estimates if the ratio is approximately constant over time.

Personal exposure to other pollutants shows variable association with personal
exposure to O₃, with differences in copollutant relationships depending on factors
such as season, city-specific characteristics, activity pattern, and spatial variability of
the copollutant (Section 4.3.4). In near-road and on-road microenvironments,
correlations between O₃ and traffic-related pollutants are moderately to strongly
negative, with the most strongly negative correlations observed for NO₂ (-0.8 to
-0.9). This is consistent with the chemistry of NO oxidation, in which O₃ is
consumed to form NO₂. The more moderate negative correlations observed for
PM₂.₅, PM₁₀, and VOC may reflect reduced concentrations of O₃ in polluted
environments due to other scavenging reactions. A similar process occurs indoors,
where infiltrated O₃ reacts with airborne or surface-associated materials to form
secondary compounds, such as formaldehyde. Although such reactions decrease
indoor O₃ exposure, they result in increasing exposure to other species which may
themselves have health effects.

Variations in ambient O₃ concentrations occur over multiple spatial and temporal
scales. Near roadways, O₃ concentrations are reduced due to reaction with NO and
other species (Section 4.3.4.2). Over spatial scales of a few kilometers and away
from roads, O₃ may be somewhat more homogeneous due to its formation as a
secondary pollutant, while over scales of tens of kilometers, additional atmospheric
processing can result in higher concentrations downwind of an urban area. Although
local-scale variability impacts the magnitude of O₃ concentrations, O₃ formation
rates are influenced by factors that vary over larger spatial scales, such as
temperature (Section 3.2), suggesting that urban monitors may track one another
temporally but miss small-scale variability in magnitude. The resulting uncertainty in
exposure contributes to exposure measurement error in epidemiologic studies.

Another factor that may influence epidemiologic results is the tendency for people to
avoid O₃ exposure by altering their behavior (e.g., reducing time spent outdoors) on
high-O₃ days. Activity pattern has a substantial effect on ambient O₃ exposure, with
time spent outdoors contributing to increased exposure (Section 4.4.2). Averting
behavior has been predominantly observed among children, older adults, and people
with respiratory problems. Such effects are less pronounced in the general
population, possibly due to the opportunity cost of behavior modification. Evidence
from one recent epidemiologic study indicates increased asthma hospital admissions
among children and older adults when O₃ alert days were excluded from the analysis
(presumably thereby eliminating averted behavior based on high O₃ forecasts).
The lower rate of admissions observed when alert days were included in the analysis
suggests that estimates of health effects based on concentration-response functions
which do not account for averted behavior may be biased toward the null.

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The range of personal-ambient correlations reported by most studies (0.3-0.8) is similar to that for NO\textsubscript{2} (U.S. EPA, 2008c) and somewhat lower than that for PM\textsubscript{2.5} (U.S. EPA, 2009d). To the extent that relative changes in central-site monitor concentration are associated with relative changes in exposure concentration, this indicates that ambient monitor concentrations are representative of day-to-day changes in average total personal exposure and in personal exposure to ambient O\textsubscript{3}. The lack of indoor sources of O\textsubscript{3}, in contrast to NO\textsubscript{2} and PM\textsubscript{2.5}, is partly responsible for low indoor-outdoor ratios (generally 0.1-0.4) and low personal-ambient ratios (generally 0.1-0.3), although it contributes to increased personal-ambient correlations. The lack of indoor sources also suggests that fluctuations in ambient O\textsubscript{3} may be primarily responsible for changes in personal exposure, even under low-ventilation, low-concentration conditions. Nevertheless, low personal-ambient correlations are a source of exposure error for epidemiologic studies, tending to obscure the presence of potential thresholds, bias effect estimates toward the null, and widen confidence intervals, and this impact may be more pronounced among populations spending substantial time indoors. The impact of this exposure error may tend more toward widening confidence intervals than biasing effect estimates, since epidemiologic studies evaluating the influence of monitor selection indicate that effect estimates are similar across different spatial averaging scales and monitoring sites.
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5  DOSIMETRY, MODE OF ACTION, AND SPECIES HOMOLOGY

5.1  Introduction

This chapter has three main purposes. The first is to describe the principles that underlie the dosimetry of O₃ and to discuss factors that influence it. The second is to describe the modes of action leading to the health effects that will be presented in Chapters 6 and 7. The third is to discuss the homology of responses in animals and humans exposed to O₃ and the interspecies differences that may affect these responses. This chapter is not intended to be a comprehensive overview, but rather, it updates the basic concepts derived from O₃ literature presented in previous documents (U.S. EPA, 2006b, 1996a) and introduces the recent relevant literature.

In Section 5.2, particular attention is given to dosimetric factors influencing individual risk of developing effects from O₃ exposure. As there have been few O₃ dosimetry studies published since the last AQCD, the reader is referred to previous documents (U.S. EPA, 2006b, 1996a) for more detailed discussion of the past literature. Evaluation of the progress in the interpretation of past dosimetry studies, as well as studies published since 2005, in the areas of uptake, reactions, and models for O₃ dosimetry, is discussed.

Section 5.3 highlights findings of studies published since the 2006 O₃ AQCD, which provide insight into the biological pathways by which O₃ exerts its actions. Since common mechanisms lead to health effects from both short- and long-term exposure to O₃, these pathways are discussed in Chapter 5 rather than in later chapters. The related sections of Chapters 6 and 7 are indicated. Earlier studies that represent the current state of the science are also discussed. Studies conducted at more environmentally-relevant concentrations of O₃ are of greater interest, since mechanisms responsible for effects at low O₃ concentrations may not be identical to those occurring at high O₃ concentrations. Some studies at higher concentrations are included if they were early demonstrations of key mechanisms or if they are recent demonstrations of potentially important new mechanisms. The topics of dosimetry and mode of action are bridged by reactions of O₃ with components of the extracellular lining fluid (ELF), which play a role in both O₃ uptake and biological responses (Figure 5-1).

In addition, this chapter discusses interindividual variability in responses, and issues related to species comparison of doses and responses (Section 5.4 and Section 5.5). These topics are included in this chapter because they are influenced by both dosimetric and mechanistic considerations.
5.2 Human and Animal Ozone Dosimetry

5.2.1 Introduction

Dosimetry refers to the measurement or estimation of the quantity of or rate at which a chemical and/or its reaction products are absorbed and retained at target sites. Figure 5-1 illustrates the transport of O₃ or its reaction products from exposure to dose to the development of health effects. Ozone exposure has been defined in Section 4.2 and consists of contact between the human or animal and O₃ at a specific concentration for a specified period of time (i.e., exposure = concentration × time). The amount of O₃ present in a given volume of air for which animals and individuals are exposed is termed exposure concentration. Ozone exposure will result in some amount (dose) of O₃ crossing an exposure surface to enter a target area. The initial measure of dose after O₃ enters the respiratory tract (RT) is inhaled dose and is the amount or rate of O₃ that crosses the outer RT surface before crossing the ELF and is effectively \( C \times t \times \dot{V}_E \), where \( C \) is concentration, \( t \) is time, and \( \dot{V}_E \) is minute ventilation. Ozone may then cross from the gas phase across the ELF interface where net dose may be measured. Net dose is the amount or rate of entry of O₃ across the gas/ELF interface. In modeling studies, the dose rate is often expressed as a flux per unit of surface area of a region of respiratory epithelium. Finally, O₃ or its reaction products may reach the tissues and tissue dose of O₃ can be reported. Tissue dose is the amount of O₃ or its reaction products absorbed and available for reacting with tissues and is difficult and rarely measured. In the literature, the exposure concentration and various measures of dose (i.e., net dose and inhaled dose) are often used as surrogates for tissue dose. However, ambient or exposure concentrations are not a true measure of dose so understanding the relationship between ambient...
concentrations and tissue dose allows for a greater appreciation of the dose-response from $O_3$ exposure.

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**Figure 5-2**  
Representation of respiratory tract regions in humans.

Ozone is a highly reactive, though poorly water soluble, gas at physiological temperature. The latter feature is believed to be the reason why it is able to penetrate into targets in the lower respiratory tract (LRT). **Figure 5-2** presents the basic structure of the human RT. The lung can be divided into three major regions: the extrathoracic (ET) region or upper respiratory tract (URT, from the nose/mouth to the end of the larynx); the tracheobronchial (TB) tree (from trachea to the terminal bronchioles); and the alveolar or pulmonary region (from the respiratory bronchioles to the terminal alveolar sacs). The latter two regions comprise the LRT. Although the structure varies, the illustrated anatomic regions are common to all mammalian
species with the exception of the respiratory bronchioles. Respiratory bronchioles, the transition region between ciliated and fully alveolated airways, are found in humans, dogs, ferrets, cats, and monkeys. Respiratory bronchioles are absent in rats and mice and abbreviated in hamsters, guinea pigs, sheep, and pigs. The branching structure of the ciliated bronchi and bronchioles also differs between species from being a rather symmetric and dichotomous branching network of airways in humans to a more monopodial branching network in other mammals.

*Figure 5-3* illustrates the structure of the LRT with progression from the large airways in the TB region to the alveolus in the alveolar region. The fact that O₃ is so chemically reactive has suggested to some that its tissue dose at the target sites exists in the form of oxidation products such as aldehydes and peroxides (see *Section 5.2.3*). Reaction products are formed when O₃ interacts with components of the ELF such as lipids and antioxidants. The ELF varies throughout the length of the RT with the nasal airways through the bronchial tree lined with a thicker layer of ELF than the alveolar region (*Figure 5-3b*). Ozone dose is directly related to the coupled diffusion and chemical reactions occurring in ELF, a process termed “reactive absorption.” Thus, the O₃ dose depends on both the concentration of O₃ as well as the availability of substrates within the ELF.

Ozone dose is affected by complex interactions between a number of other major factors including RT morphology, breathing route, frequency, and volume, physicochemical properties of the gas, physical processes of gas transport, as well as the physical and chemical properties of the ELF and tissue layers (*Figure 5-3c*). The role of these processes varies throughout the length of the RT and as O₃ moves from the gas to liquid compartments of the RT.
Figure 5-3  Structure of lower airways with progression from the large airways to the alveolar region.

Two types of measurements have been used to arrive at the O₃ dose to target sites during breathing: (1) measurement of removal of O₃ from the air stream (termed “uptake”); and (2) measurement of chemical reactions in tissues or with biomolecules known to be present in tissues (termed “reactants”). The results of the above measurements have been incorporated into mathematical models for the purpose of explaining, predicting, and extrapolating O₃ dose in different exposure conditions.
scenarios. Few new studies have investigated the uptake of O$_3$ in the RT since the last O$_3$ assessment (U.S. EPA, 2006b). The studies that have been conducted generally agree with the results presented in the past and do not change the dosimetry conclusions of the last document.

5.2.2 Ozone Uptake

Past AQCDs provide information on the majority of literature relevant to understanding the state of the science in O$_3$ dosimetry. Measurements of O$_3$ dose have been inferred from simultaneous measurements of airflow and O$_3$ concentration at the airway opening of the nose or mouth (Nodelman and Ultman, 1999; Wiester et al., 1996a) as well as at internal sampling catheters (Gerrity et al., 1995; Gerrity et al., 1988). One method of quantifying O$_3$ dose is to measure the amount of O$_3$ removed from the air stream during breathing (termed “uptake”). The difference in the amount of O$_3$ inhaled and exhaled relative to the amount of inhaled O$_3$ is termed fractional absorption. Uptake efficiency is also reported and refers to the O$_3$ absorbed in a region expressed as a fraction of the total amount of O$_3$ entering the given region. Uptake studies have utilized bolus and continuous O$_3$ breathing techniques as well as modeling to investigate these measures of uptake and the distribution of O$_3$ uptake between the URT and LRT. A number of the studies that have measured the fractional absorption and uptake efficiency of O$_3$ in the human RT, URT, and LRT are presented in Table 5-1. For studies that reported fractional absorption of O$_3$ boluses, the equivalent fractional absorption of a continuous inhalation of O$_3$ was estimated as the sum of the products of the experimental bolus absorption and incremental volume of a bolus into a breath divided by the tidal volume of the breath, or, where available, was taken from Table 1A of Schlesinger et al. (1997).
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\( a \) OR = oral exposure during spontaneous breathing; N = nasal exposure during spontaneous breathing; OR/N = pooled data from oral and nasal exposure; mouthpiece = exposure by mouthpiece.

\( b \) \( f_B \) is either measured or is computed from flows and \( V_T \).

\( c \) \( F_{URT} \) from Santiago et al. (2001) represents nasal absorption (\( F_{nose} \)).

\( d \) \( V_T \) is computed from flow and \( f_B \).

### 5.2.2.1 Gas Transport Principles

The three-dimensional transport of \( \text{O}_3 \) in the lumen of an airway is governed by diffusion associated with the Brownian motion of gas molecules and convection that depends on local velocity patterns. Simultaneously, \( \text{O}_3 \) is absorbed from the gas stream into the ELF where it undergoes simultaneous radial diffusion and chemical reaction.

When air flows through an airway, \( \text{O}_3 \) located near the tube center moves faster than \( \text{O}_3 \) near the tube wall where frictional forces retard the flow. This non-uniformity in the radial profile of velocity gives rise to an axial spreading or dispersion of the \( \text{O}_3 \) that operates in parallel with bulk flow and axial diffusion. The shape of the velocity profile is affected by the flow direction through bifurcating airway branches (Schroter and Sudlow, 1969). The velocity profile is nearly parabolic during inhalation but quite flat during exhalation. Thus, there tends to be greater axial dispersion during inhalation than during exhalation. Dispersion also depends on the nature of the flow, that is, whether it is laminar (i.e., streamlined) or turbulent (i.e., possessing random velocity fluctuations). Because turbulent flow flattens velocity profiles, it may actually diminish dispersion. In humans, turbulent flow persists only a few generations into the RT. The persistence of turbulence into the RT also varies by species and flow rates. For example, airflow is nonturbulent in the rat nose at any physiologic flow rate but may be highly turbulent in the human nose during exercise (Miller, 1995).
The relative importance of axial convection, diffusion, and dispersion varies among RT regions for a given level of ventilation. In the URT and major bronchi, axial convection and dispersion tend to be the predominant mechanisms. Moving into more distal areas of the RT, the summed cross-sectional area of the airways rapidly increases and linear velocities decrease, leading to a greater role for molecular diffusion. The principal mechanism of gas mixing in the lung periphery is molecular diffusion (Engel, 1985).

Absorption of O₃ at the airway wall depends on a concentration boundary layer on the gas side of the airway wall as well as simultaneous radial diffusion and chemical reaction within the ELF (Figure 5-3c) (Miller, 1995). The boundary layer caused by slowly moving gas near the airway wall can be an important component of the radial diffusion resistance to O₃ absorption. This diffusive resistance increases with distal penetration into the RT with one study reporting that the gas boundary layer contributes 53% of the overall diffusive resistance in the URT, 78% in the proximal LRT, and 87% in the distal LRT (Hu et al., 1994). The geometry of airway surfaces also affects local O₃ absorption. For example, nasal and lung regions receive different O₃ exposures or doses (Miller and Kimbell, 1995); and larger surface-to-volume ratio of the smaller airways in women enhances local O₃ uptake and reduces the distal penetration volume of O₃ into the RT of women relative to men (Ultman et al., 2004).

### 5.2.2.2 Target Sites for Ozone Dose

A primary uptake site of O₃ delivery to the lung epithelium is believed to be the centriacinar region (CAR). The CAR refers to the zone at the junction of the TB airways and the gas exchange region. This area is also termed the proximal alveolar region (PAR) and is defined as the first generation distal to the terminal bronchioles. Contained within the CAR, the respiratory bronchioles were confirmed as the site receiving the greatest O₃ dose (¹⁸O mass/lung weight) in resting O₃ exposed rhesus monkeys, when not considering the nose (Plopper et al., 1998). Furthermore, the greatest cellular injury occurred in the vicinity of the respiratory bronchioles and was dependent on the delivered O₃ dose to these tissues (see also Section 5.4.1). However, ¹⁸O label was detected to a lesser extent in other regions of the TB airway tree, showing that O₃ is delivered to these compartments as well, although in a smaller dose. These studies agree with earlier model predictions showing that the tissue O₃ dose (O₃ flux to liquid-tissue interface) was low in the trachea, increased to a maximum in the terminal bronchioles and the CAR, and then rapidly decreased in the alveolar region (Miller et al., 1985). It was also predicted that the net O₃ dose (O₃ flux to air-liquid interface) gradually decreased with distal progression from the trachea to the end of the TB region and then rapidly decreased in the alveolar region. Despite the exclusion of the URT and appreciable O₃ reactions with ELF constituents after the 16th generation, the results from the model agree with experimental results showing that the greatest O₃ tissue dose was received in the CAR (Miller et al., 1985).
Inhomogeneity in the RT structure may affect the dose delivered to this target site. Models have predicted that the farther the PAR is from the trachea, the less the O\textsubscript{3} tissue dose to the region. Ultman and Anjilvel (1990) and Overton and Graham (1989) predicted approximately a 50 to 300\% greater PAR dose for the shortest path relative to the longest path in humans and rats, respectively. In addition, Mercer et al. (1991) found that both path distance and ventilatory unit size affected dose. The variation of O\textsubscript{3} dose among anatomically equivalent ventilatory units was predicted to vary as much as 6-fold, as a function of path length from the trachea. This could have implications in regional damage to the LRT, such that even though the average LRT dose may be at a level where health effects would not be predicted, local regions of the RT may receive considerably higher than average doses and therefore be at greater risk of effects.

Since the URT is the first part of the RT to be exposed to O\textsubscript{3}, the nasal membranes are another target site at risk of injury from inhaled O\textsubscript{3}. Injury to the nasal epithelium has been shown to be site-specific (see Section 5.3.7) and studies have found that the location of reactive gas-induced nasal lesions may be attributable to the local dose of gas reaching that area (Garcia et al., 2009a; Morgan and Monticello, 1990). Similar to the LRT, inhomogeneity of the nasal anatomy, nasal fluid composition, and ventilation and airflow patterns affects the uptake of O\textsubscript{3} into the nasal passageways.

### 5.2.2.3 Upper Respiratory Tract Ozone Removal and Dose

Total O\textsubscript{3} uptake in the entire RT in rats and guinea pigs is 40-54\% efficient (Hatch et al., 1989; Wiester et al., 1988; Wiester et al., 1987), while in humans at rest it ranges from 80-95\% efficient (Hu et al., 1992). The URT provides a defense against O\textsubscript{3} entering the lungs by removing half of the O\textsubscript{3} that will be absorbed from the airstream. In both animals and humans, about 50\% of the O\textsubscript{3} that was absorbed in the RT was removed in the head (nose, mouth, and pharynx), about 7\% in the larynx/trachea, and about 43\% in the lungs (Hu et al., 1992; Hatch et al., 1989; Miller et al., 1979). However, experimental studies in dogs have reported 75-100\% uptake in the URT (Yokoyama and Frank, 1972; Vaughan et al., 1969). The fraction of O\textsubscript{3} taken up was inversely related to flow rate and to inlet O\textsubscript{3} concentration (Yokoyama and Frank, 1972; Vaughan et al., 1969). URT absorption is relatively high due in part to the large surface area of the nasal airways. The limiting factors in nasal O\textsubscript{3} uptake were simultaneous diffusion and chemical reaction of O\textsubscript{3} in the nasal ELF (Santiago et al., 2001). The ELF layer in the nose is thicker than in the rest of the RT, and mathematical estimates predicted that O\textsubscript{3} penetrates less than the thickness of the ELF layer; reaction products are likely the agents damaging the nasal tissue and not O\textsubscript{3} itself. It was hypothesized that the nasal non-linear kinetics of O\textsubscript{3} uptake fraction result from the depleting substrates in the nasal ELF becoming the limiting factor of the reaction (Santiago et al., 2001).

Uptake efficiencies have been measured for various segments of the URT (Table 5-1). Gerrity et al. (1995) reported unidirectional uptake efficiencies of O\textsubscript{3}.
inhaled from a mouthpiece; of 0.18 from the mouth to vocal cords, 0.095 from the vocal cords to the upper trachea (totaling 0.27), 0.084 from the upper trachea to the main bifurcation carina (total cumulative efficiency from the mouth of 0.36), and essentially zero between the carina and the bronchus intermedius (total cumulative efficiency from the mouth of 0.33). These values are lower than those calculated by Hu et al. (1992) that reported cumulative uptake efficiencies of 0.21, 0.36, 0.44, and 0.46 during a complete breath in which an O₃ bolus penetrated between the mouth and the vocal cords, the upper trachea, the main bifurcation carina, and the bronchus intermedius, respectively. The lower efficiencies seen in Gerrity et al. (1995) may have resulted because these investigators measurements were based on inhalation alone or were caused by O₃ scrubbing by the mouthpiece.

Past studies investigating nasal uptake of O₃ have shown that the nose partially protects the LRT from damage from inspired O₃ (Santiago et al., 2001; Gerrity et al., 1988). Sawyer et al. (2007) further investigated nasal uptake of O₃ in healthy adults during exercise. Fractional O₃ uptake, acoustic rhinometry (AR), and nasal NO measurements were taken in ten adults (8 women, 2 men) exposed to 200 ppb O₃ before and after moderate exercise at two flow rates (10 and 20 L/min). The percent nasal uptake of O₃ was ~50% greater at 10 L/min compared to 20 L/min both pre- and post-exercise. However, the inhaled O₃ dose delivered to the LRT (i.e., flow rate × exposure concentration × (1 - nasal absorbed fraction)) was 1.6-fold greater at the higher flow than at the lower flow (2.5 compared to 0.9 ppm·L/min). Prior exercise did not affect O₃ uptake at either flow rate, but did significantly increase nasal volume (Vn) and AR measurements of nasal cross-sectional area (minimum cross-sectional area (MCA) that corresponds to the nasal valve, CSA2 that corresponds to the anterior edge of the nasal turbinates, and CSA3 that corresponds to the posterior edge of the nasal turbinates) (p ≤ 0.05) (Sawyer et al., 2007). Conversely, exercise decreased nasal resistance (Rn) (p < 0.01) and NO production (nonsignificant, p >0.05). The change in Vn and CSA2:MCA ratio was correlated with the percent change in nasal uptake, however the overall effect was small and sensitive to elimination of outliers and sex segregation.

Overall, the majority of studies suggest that the URT removes about half of the O₃ that will be absorbed by reactions in the nasal ELF. The exact uptake efficiency is dependent on variations in flow rate and inhaled concentration.

### 5.2.2.4 Lower Respiratory Tract Ozone Uptake and Dose

Approximately 43% of the O₃ absorption occurs in the LRT of both humans and animals. Models predicted that the net O₃ dose decreases distally from the trachea toward the end of the TB region and then rapidly decreases in the alveolar region (Miller et al., 1985). Further, these models predicted low tissue O₃ dose in the trachea and large bronchi.

Uptake efficiency depends on a number of variables, including O₃ exposure concentration, exposure time, and breathing pattern. For breaths of similar
waveforms, respiratory patterns are uniquely described by breathing frequency ($f_B$) and tidal volume ($V_T$); by minute ventilation ($\dot{V}_E = f_B \times V_T$) and $f_B$; or by $\dot{V}_E$ and $V_T$. Simulations from the Overton et al. (1996) single-path anatomical respiratory tract model, where the upper and lower respiratory tracts were modeled but uptake by the URT was not considered, predicted that fractional uptake and PAR O$_3$ dose increased with $V_T$ when $f_B$ was held constant. Likewise, experimental studies found that O$_3$ uptake was positively correlated with changes in $V_T$ (Ultman et al., 2004; Gerrity et al., 1988). Also, O$_3$ exposure led to a reflex mediated increase in $f_B$ and reduction in $V_T$, hypothesized to be protective by decreasing the dose delivered to the lung at a particular $\dot{V}_E$ (Gerrity et al., 1994). Nasal O$_3$ uptake efficiency was inversely proportional to flow rate (Santiago et al., 2001), so that an increase in $\dot{V}_E$ will increase O$_3$ delivery to the lower airways. At a fixed $\dot{V}_E$, increasing $V_T$ (corresponding to decreasing $f_B$) drove O$_3$ deeper into the lungs and increased total respiratory uptake efficiency (Figure 5-4) (Ultman et al., 2004; Wiester et al., 1996a; Gerrity et al., 1988). Modeling predicted a decrease in fractional uptake with increased $f_B$ when $V_T$ was held constant, but an increase in PAR dose with increased $f_B$ (Overton et al., 1996). Similarly, increased $f_B$ (80 - 160 bpm) and shallow breathing in rats decreased midlevel tracheal $^{18}$O content and an increased $^{18}$O content in the mainstem bronchi (Alfaro et al., 2004). This dependence may be a result of frequency-induced alterations in contact time that affects the first-order absorption rate for O$_3$ (Postlethwait et al., 1994). Also, an association of O$_3$ uptake efficiency was found with $\dot{V}_E$ and exposure time.

Increasing flow leads to deeper penetration of O$_3$ into the lung, such that a smaller fraction of O$_3$ is absorbed in the URT and uptake shifts to the TB airways and respiratory airspaces (Nodelman and Ultman, 1999; Hu et al., 1994; Ultman et al., 1994). Hu et al. (1994) and Ultman et al. (1994) found that O$_3$ absorption increased with volumetric penetration ($V_p$) of a bolus of O$_3$ into the RT. Ozone uptake efficiency and $V_p$ were not affected by bolus O$_3$ concentration (Kabel et al., 1994; Hu et al., 1992), indicating that under these experimental conditions O$_3$ uptake was a linear absorption process, where the diffusion and chemical reaction rates of O$_3$ were proportional to the O$_3$ concentration. The absorption relationship would not be linear once interfacial mass transfer was saturated. As mentioned above, a weak negative relationship between O$_3$ concentration and uptake efficiency was reported for the nasal cavities by Santiago et al. (2001). Rigas et al. (2000) also found a weak but significant negative dependence of O$_3$ concentration on RT uptake efficiency in exercising individuals. This study also found that exposure time had a small but significant influence on uptake efficiency; however, this negative dependence may be an artifact of progressive depletion of reactive substrates from the ELF.
Note: Subjects breathed 250 ppb O₃ oronasally via a breathing mask. The uptake efficiency was well correlated with breathing frequency ($r = -0.723$, $p < 0.001$) and tidal volume (not illustrated; $r = 0.490$, $p < 0.001$).

Source: Reprinted with permission of Health Effects Institute (Ultman et al., 2004).

**Figure 5-4** Total O₃ uptake efficiency as a function of breathing frequency at a constant minute ventilation of 30 L/min.

Past studies have shown that O₃-induced epithelial damage to the lung occurs with a reproducible pattern of severity between daughter branches of individual bifurcations that is dependent on the O₃ concentration-time profile of the inhaled gas. A 3-D computational fluid dynamics model was created to investigate the O₃ transport in a single airway bifurcation (Taylor et al., 2007). The model consisted of one parent branch and two symmetrical daughter branches with a branching angle of 90° and a sharp carinal ridge. Various flow scenarios were simulated using Reynolds numbers (Re) ranging from 100 to 500. The Reynolds number that corresponds to a certain airway generation is dependent upon both lung size and $\dot{V}_E$, such that the range in Reynolds numbers, from 100-500, would encompass generations 1-5, 3-7, and 6-10 for an adult during quiet breathing, light exertion, and heavy exercise, respectively, whereas the same Reynolds number range corresponds to generations 0-4, 1-6, and 4-8 for a 4-year-old child. This model predicted velocity distributions that were consistent with earlier work of Schroter and Sudlow (1969), and also reported O₃ concentration and wall uptake distributions. The model predicted that during inspiration, the velocity and O₃ concentration distribution were axisymmetric throughout the parent branch, but skewed toward the inner wall within the daughter branches. During expiration, the model predicted that the velocity and O₃ concentration distribution was slightly skewed toward the outer walls of the daughter branches. Hot spots of wall flux existed at the carina during inspiration and
expiration with Re >100. Additional hot spots were found during expiration on the parent branch wall downstream of the branching region.

Overall O₃ inhalation uptake in humans is over 80% efficient, but the exact efficiency that determines how much O₃ is available at longitudinally distributed compartments in the lung is sensitive to changes in Vₖ, f_b, and to a minor extent, exposure time.

### 5.2.2.5 Mode of Breathing

Ozone uptake and distribution is sensitive to the mode of breathing. Variability in TB airways volume had a weaker influence on O₃ absorption during nasal breathing compared to oral breathing. This could be a result of O₃ scrubbing in the nasal passageways that are bypassed by oral breathing. Studies by Ultman and colleagues, using bolus inhalation in humans, demonstrated that O₃ uptake fraction into the upper airways was greater during nasal breathing than during oral breathing (e.g., 0.90 during nasal breathing and 0.80 during oral breathing at 150 mL/sec and 0.45 during nasal breathing and 0.25 during oral breathing at 1,000 mL/sec) (Nodelman and Ultman, 1999; Kabel et al., 1994; Ultman et al., 1994). Therefore, oral breathing results in deeper penetration of O₃ into the RT with a higher absorbed fraction in the TB and alveolar airways (Nodelman and Ultman, 1999). Similar results were also obtained from O₃ uptake studies in dogs (Yokoyama and Frank, 1972). Earlier human studies suggested that oral or oronasal breathing results in a higher O₃ uptake efficiency than nasal breathing (Wiester et al., 1996a; Gerrity et al., 1988). Overall, the mode of breathing may have a seemingly small effect on the RT uptake efficiency; however, it does play an important role in the distribution of O₃ deposited in the distal airways.

### 5.2.2.6 Interindividual Variability in Dose

Similarly exposed individuals vary in the amount of actual dose delivered to the LRT (Santiago et al., 2001; Rigas et al., 2000; Bush et al., 1996). Interindividual variability accounted for between 10-50% of the absolute variability in O₃ uptake measurements (Santiago et al., 2001; Rigas et al., 2000). When concentration, time, and ̇VE were held constant, fractional absorption ranged from 0.80 to 0.91 (Rigas et al., 2000). It has been hypothesized that interindividual variation in O₃ induced responses such as FEV₁ is the result of interindividual variation in net dose or regional O₃ uptake among exposed individuals.

Recent studies have reiterated the importance of intersubject variation in O₃ uptake. The intersubject variability in nasal O₃ uptake determined by Sawyer et al. (2007) ranged from 26.8 to 65.4% (pre- and post-exercise). A second study investigating the use of the CO₂ expirogram to quantify pulmonary responses to O₃ found that
intersubject variability accounted for 50% of the overall variance in the study (Taylor et al., 2006).

Variability in net or tissue dose may be attributed to differences in the pulmonary physiology, anatomy, and biochemistry. Since the URT and TB airways remove the majority of inhaled O₃ before it reaches the gas exchange region, the volume and surface area of these airways will influence O₃ uptake. Models predicted that fractional O₃ uptake and PAR dose (flux of O₃ to the PAR surfaces divided by exposure concentration) increase with decreasing TB volume and decreasing TB region expansion. On the contrary, alveolar expansion had minimal effect on uptake efficiency as relatively little O₃ reaches the peripheral lung (Bush et al., 2001; Overton et al., 1996). Ozone uptake was virtually complete by the time O₃ reaches the alveolar spaces of the lung (Postlethwait et al., 1994). Experimental studies have found that differences in TB volumes may account for 75% of the variation in absorption between subjects (Ultman et al., 2004). In support of this concept, regression analysis showed that O₃ absorption was positively correlated with anatomical dead space (V_D) and TB volume (i.e., V_D minus V_URT), but not total lung capacity (TLC), forced vital capacity (FVC), or functional residual capacity (FRC) (Ultman et al., 2004; Bush et al., 1996; Hu et al., 1994; Postlethwait et al., 1994). Variability in V_D was correlated more with the variability in the TB volume than the URT volume. Similarly, uptake was correlated with changes in individual bronchial cross-sectional area, indicating that changes in cross-sectional area available for gas diffusion are related to overall O₃ retention (Reeser et al., 2005; Ultman et al., 2004). When coupled, these results suggest that the larger surface-to-volume ratio associated with the smaller airways in women enhances local O₃ uptake, thereby reducing the distal penetration volume of O₃ into the female respiratory system. When absorption data were normalized to Vp/V_D, variability attributed to sex differences were not distinguishable (Bush et al., 1996). These studies provide support to the RT anatomy, especially the TB volume and surface area, playing a key role in variability of O₃ uptake between individuals.

In addition, variability between individuals is influenced by age. Overton and Graham (1989) predicted that the total mass of O₃ absorbed per minute (in units of: μg/min per [μg/m³ of ambient O₃]) increased with age from birth to adulthood. This model predicted that during quiet breathing the LRT distribution of absorbed O₃ and the CAR O₃ tissue dose were not sensitive to age. However, during heavy exercise or work O₃ uptake was dependent on age. A physiologically based pharmacokinetic model simulating O₃ uptake predicted that regional extraction of O₃ was relatively insensitive to age, but extraction per unit surface area was 2-fold to 8-fold higher in infants compared to adults, due to the fact that children under age 5 have much a much smaller airway surface area in the extrathoracic (nasal) and alveolar regions (Sarangapani et al., 2003). Additionally, children tend to have a greater oral breathing contribution than adults at rest and during exercise (Bennett et al., 2008; Becquemin et al., 1999; James et al., 1997). Even after adjusting for differences in surface area, the dose rate to the lower airways of children compared to adults is increased further because children breathe at higher minute ventilations relative to their lung volumes.
Smoking history, with its known increase in mucus production, was not found to affect the fractional uptake of a bolus of O₃ in apparently healthy smokers with limited smoking history (Bates et al., 2009). Despite similar internal O₃ dose distribution, the smokers exhibited greater pulmonary responses to O₃ bolus exposures, measured as FEV₁ decrements and increases in the normalized slope of the alveolar plateau (Sₙ). This was contrary to previous studies conducted in smokers with a greater smoking history that found decreased O₃ induced decrements in FEV₁ in smokers during continuous O₃ exposure (Frampton et al., 1997a; Emmons and Foster, 1991).

### 5.2.2.7 Physical Activity

Exercise increases the overall exposure of the lung to inhaled contaminants due, in most part, to the increased intake of air. Thus, human studies have used exercise, at a variety of activity levels, to enhance the effects of O₃ (Table 5-2). Further explanation of the effects of physical activity on ventilation can be found in Chapters 4 and 6. Table 4-5 presents the mean ventilation rates at different activity levels for different age groups. Table 6-1 provides activity levels as detailed in specific human exposure studies.

<table>
<thead>
<tr>
<th>Activity Level</th>
<th>Inhalation Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>Light</td>
<td>2 to 3 × resting $\dot{V}_E$ *</td>
</tr>
<tr>
<td>Moderate</td>
<td>4 to 6 × resting $\dot{V}_E$</td>
</tr>
<tr>
<td>Heavy</td>
<td>7 to 8 × resting $\dot{V}_E$</td>
</tr>
<tr>
<td>Very Heavy</td>
<td>&gt;9 × resting $\dot{V}_E$</td>
</tr>
</tbody>
</table>

*Resting $\dot{V}_E$ approximates 8 L/min


As exercise increases from a light to moderate level, $V_T$ increases. This increase in $V_T$ is achieved by encroaching upon both the inspiratory and expiratory reserve volumes of the lung (Dempsey et al., 1990). After $V_T$ reaches about 50% of the vital capacity, generally during heavy exercise, further increases in ventilation are achieved by increasing $f_b$. Ventilatory demands of very heavy exercise require airway flow rates that often exceed 10 times resting levels and $V_T$ that approach 5 times resting levels (Dempsey et al., 2008).

In addition to increasing the bulk transport of O₃ into the lung, exercise also leads to a switch from nasal to oronasal breathing. Higher ventilatory demand necessitates a lower-resistance path through the mouth. The contribution of nasal breathing to the $\dot{V}_E$ varies as a function of age, sex, and race. Children tended to have a lesser nasal contribution to breathing than adults at rest and during exercise at matched percent
maximum work (Bennett et al., 2008). Males had less nasal contribution to breathing at rest and during exercise at matched percent maximum work compared with females (Bennett et al., 2003). The difference between the sexes may be explained by the difference in $\dot{V}_E$ at a given percent maximal workload. Females had a lower $\dot{V}_E$ than males so had to augment breathing orally at higher work efforts. Caucasians had a lesser nasal contribution than African-Americans at rest and during exercise at matched percent maximum work (Bennett et al., 2003).

This increase in $V_T$ and flow associated with exercise in humans shifts the net $O_3$ dose further into the periphery of the RT causing a disproportionate increase in distal lung tissue dose. Modeling heavy exercise by increasing ventilatory parameters from normal respiration levels predicted a 10-fold increase in total mass uptake of $O_3$ (Miller et al., 1985). This model also predicted that as exercise and ventilatory demand increased, the maximum tissue dose, the $O_3$ reaching the tissues, moved distally into the RT (Figure 5-5). When the flow was increased to what is common in moderate or heavy exercise (respiratory flow = 45-60 L/min compared to 15 L/min), the URT absorbed a smaller fraction of the $O_3$ (0.10 at high flow rate to ~0.50 at low flow rate); however, the trachea and more distal TB airways received higher doses during higher flow rates than at lower flow rates (0.65 absorbed in the lower TB airways, and 0.25 absorbed in the alveolar zone with high flow compared to 0.5 in the TB with almost no $O_3$ reaching the alveolar zone at low flow) (Hu et al., 1994). The same shift in the $O_3$ dose distribution more distally in the lung occurred in other studies mimicking the effects of exercise (Nodelman and Ultman, 1999). Also, LRT uptake efficiency was sensitive to age only under exercise conditions (Overton and Graham, 1989). The total mass of $O_3$ absorbed per minute ($\mu$g/min per $[\mu$g/m$^3$ of ambient $O_3]$) was predicted to increase with age during heavy work or exercise. A recent study by Sawyer et al. (2007) approximated that doubling minute ventilation led to only a 1.6-fold higher delivered dose rate of $O_3$ to the lung (delivered dose was calculated as: flow rate $\times [O_3$ ppm] $\times$ (100-percent nasal $O_3$ uptake)) due to a decrease in URT uptake with increasing flow rate. Past models have predicted the increase in uptake during exercise is distributed unevenly in the RT compartments and regions. Tissue and net dose in the TB region increased $\sim$1.4-fold during heavy exercise compared to resting conditions, whereas the alveolar surface layer and tissue uptake increased by factors of 5.2 and 13.6, respectively (Miller et al., 1985).
5.2.2.8 Summary

In summary, O₃ uptake is affected by complex interactions between a number of factors including RT morphology, breathing route, frequency, and volume, physicochemical properties of the gas, physical processes of gas transport, as well as the physical and chemical properties of the ELF and tissue layers. The role of these processes varies throughout the length of the RT and as O₃ moves from the gas into liquid compartments of the RT.

About half of the O₃ that will be absorbed from the airstream is removed in the URT, which provides a defense against O₃ entering the lungs. However, the local dose to the URT tissue is site-specific and dependent on the nasal anatomy, nasal fluid composition, and ventilation and airflow patterns of the nasal passageways.
The primary uptake site of O$_3$ delivery to the LRT epithelium is believed to be the CAR, however, similar to the URT, inhomogeneity in the RT structure may affect the dose delivered to this target site with larger path lengths leading to smaller locally delivered doses. This could have implications in regional damage to the RT, such that even though the average RT dose may be at a level where health effects would not be predicted, local regions of the RT may receive considerably higher than average doses and therefore be at greater risk of effects. Recent studies have provided evidence for hot spots of O$_3$ flux around bifurcations in airways.

Experimental studies and models have suggested that the net O$_3$ dose gradually decreases distally from the trachea toward the end of the TB region and then rapidly decreases in the alveolar region. However, the tissue O$_3$ dose is low in the trachea, increases to a maximum in the terminal bronchioles and the CAR, and then rapidly decreases distally into the alveolar region.

Ozone uptake efficiency is sensitive to a number of factors. Fractional absorption will decrease with increased flow and increase proportional to $V_T$, so that at a fixed $\dot{V}_E$, increasing $V_T$ (or decreasing $f_B$) drives O$_3$ deeper into the lungs and increases total respiratory uptake efficiency. Individual total airway O$_3$ uptake efficiency is also sensitive to large changes in O$_3$ concentration, exposure time, and $\dot{V}_E$. Major sources of variability in absorption of O$_3$ include O$_3$ concentration, exposure time, $f_B$, $\dot{V}_E$, and $V_T$, but the interindividual variation is the greatest source of variability uptake efficiency. The majority of this interindividual variability is due to differences in TB volume and surface area.

An increase in $V_T$ and $f_B$ are both associated with increased physical activity. These changes and a switch to oronasal breathing during exercise results in deeper penetration of O$_3$ into the lung with a higher absorbed fraction in the ET, TB, and alveolar airways. For these reasons, increased physical activity acts to move the maximum tissue dose of O$_3$ distally into the RT and into the alveolar region.

### 5.2.3 Ozone Reactions and Reaction Products

Ozone dose is affected by the chemical reactions or the products of these reactions that result from O$_3$ exposure. The process by which O$_3$ moves from the airway lumen into the ELF is related to the coupled diffusion and chemical reactions occurring in ELF and is called “reactive absorption.” Ozone is chemically reactive with a wide spectrum of biomolecules and numerous studies have evaluated the loss of specific molecules such as GSH and the appearance of plausible products such as nonanal. Both in vitro and in vivo studies contribute to the understanding of O$_3$ reactions and reaction products.

Ozone may interact with many of the components in the ELF including phospholipids, neutral lipids like cholesterol, free fatty acids, proteins, and low molecular weight antioxidants as has been demonstrated in in vitro studies (Perez-Gil, 2008; Uppu et al., 1995). It was estimated that 88% of the O$_3$ that does not come in contact with antioxidants will react with unsaturated fatty acids in the ELF.
including those contained within phospholipids or neutral lipids (Uppu et al., 1995). Ozone reacts with the double bond of unsaturated fatty acids to form stable and less reactive ozonide, aldehyde, and hydroperoxide reaction products via chemical reactions such as the Criegee ozonolysis mechanism (Figure 5-6) (Pryor et al., 1991). Lipid ozonation products, such as the aldehydes hexanal, heptanal, and nonanal, have been recovered after O₃ exposure in human BAL fluid (BALF), rat BALF, isolated rat lung, and in vitro systems (Frampton et al., 1999; Postlethwait et al., 1998; Pryor et al., 1996). Adducts of the aldehyde 4-hydroxynonenal were found in human alveolar macrophages after O₃ exposure in vivo (Hamilton et al., 1998).

Polyunsaturated fatty acid (PUFA) reactions are limited by the availability of O₃ since lipids are so abundant in the ELF. Yields of O₃-induced aldehydes were increased by the decrease in other substrates such as ascorbic acid (AH₂) (Postlethwait et al., 1998). Free radicals are also generated during O₃-mediated oxidation reactions with PUFA (Pryor, 1994). These reactions are reduced by the presence of the lipid-soluble free radical scavenger α-tocopherol (α-TOH) (Pryor, 1994; Fujita et al., 1987; Pryor, 1976). PUFA reactions may not generate sufficient bioactive materials to account for acute cell injury, however only modest amounts of products may be necessary to induce cytotoxicity (Postlethwait and Ultman, 2001; Postlethwait et al., 1998).

Note: Not all secondary reaction products are shown.


**Figure 5-6** Schematic overview of O₃ interaction with PUFA in ELF and lung cells.

Cholesterol is the most abundant neutral lipid in human ELF. Reaction of cholesterol with O₃ results in biologically active cholesterol products such as the oxysterols, β-epoxide and 6-oxo-3,5-diol (Murphy and Johnson, 2008; Pulfer et al., 2005; Pulfer and Murphy, 2004). Product yields depend on ozonolysis conditions, however cholesterol ozonolysis products form in similar abundance to phospholipid-derived ozonolysis products in rat ELF (Pulfer and Murphy, 2004).
The ELF also contains proteins derived from blood plasma as well as proteins secreted by surface epithelial cells. Ozone reactions with proteins have been studied by their in vitro reactions as well as reactions of their constituent amino acids (the most reactive of which are cysteine, histidine, methionine, tyrosine, and tryptophan). Ozone preferentially reacts with biomolecules in the following order: thiosulfate > ascorbate > cysteine ≈ methionine > glutathione (Kanofsky and Sima, 1995). Rate constants for the reaction of amino acids with O₃ vary between studies due to differing reaction conditions and assumptions; however aliphatic amino acids were consistently very slow to react with O₃ (e.g., alanine: 25-100 moles/L/sec) (Kanofsky and Sima, 1995; Ignatenko and Cherenkevich, 1985; Pryor et al., 1984; Hoigné and Bader, 1983). Uppu et al. (1995) predicted that 12% of inhaled O₃ that does not react with antioxidants will react with proteins in the ELF.

Reactions of O₃ with low molecular weight antioxidants have been extensively studied. The consumption of antioxidants such as uric acid (UA), ascorbate (AH₂), and reduced glutathione (GSH) by O₃ was linear with time and positively correlated with initial substrate concentration and O₃ concentration (Mudway and Kelly, 1998; Mudway et al., 1996). Endogenous antioxidants are present in relatively high concentrations in the ELF of the human airways (obtained as BALF) and display high (but not equal) intrinsic reactivities toward O₃. In individual and in limited composite mixtures, UA was the most reactive antioxidant tested, followed by AH₂ (Mudway and Kelly, 1998). GSH was consistently less reactive than UA or AH₂ (Mudway and Kelly, 1998; Mudway et al., 1996; Kanofsky and Sima, 1995). To quantify these reactions, Kermani et al. (2006) evaluated the interfacial exposure of aqueous solutions of UA, AH₂, and GSH (50-200 µM) with O₃ (1-5 ppm). Similar to the results of Mudway and Kelly (1998), this study found the hierarchy in reactivity between O₃ and these antioxidants to be UA ≥ AH₂ >> GSH. UA and AH₂ shared a 1:1 stoichiometry with O₃, whereas 2.5 moles of GSH were consumed per mole of O₃. Using these stoichiometries, reaction rate constants were derived (5.8×10⁴ M⁻¹ sec⁻¹, 5.5×10⁴ M⁻¹ sec⁻¹, and 57.5 M⁰.₇₅ sec⁻¹ [20.9 M⁻¹ sec⁻¹] for the reaction of O₃ with UA, AH₂, and GSH, respectively). Other studies report reactive rate constants that are two to three orders of magnitude larger, however these studies used higher concentrations of O₃ and antioxidants under less physiologically relevant experimental conditions (Kanofsky and Sima, 1995; Giamalva et al., 1985; Pryor et al., 1984). Since O₃ acts through competition kinetics, the effective concentration of the reactants present in the ELF will determine the reactions that occur in vivo. For example, the pKa of GSH is about 8.7 so that at physiological pH very little is in the reactive form of thiolate (GS⁻). On the other hand, ascorbic acid has a pKa of about 4.2 so it exists almost entirely as ascorbate (AH⁻) in the ELF. Thus, the effective concentration of GSH that is available to react with O₃ will be much lower than that of ascorbate in ELF.

A series of studies used new techniques to investigate the reaction products resulting from initial air-liquid interface interactions of O₃ with ELF components (e.g., antioxidants and proteins) in ~1 millisecond (Enami et al., 2009a, b, c, 2008a, b). Solutions of aqueous UA, AH₂, GSH, α-TOH, and protein cysteines (CyS) were sprayed as microdroplets in O₃/N₂ mixtures at atmospheric pressure and analyzed by
electrospray mass spectrometry. These recent studies in which the large surface to volume ratio of microdroplets promote an interfacial reaction demonstrated different reactivity toward AH$_2$, UA, and GSH by O$_3$ compared to previous studies using bulk liquid phase bioreactors. This artificial system does not recapitulate the lung surface so caution must be taken in translating the results of these studies to in vivo conditions.

As was seen in previous studies (Kermani et al., 2006; Kanofsky and Sima, 1995), the hierarchy of reactivity of these ELF components with O$_3$ was determined to be AH$_2$ ≈ UA > CyS > GSH. There was some variance between the reaction rates and product formation of UA, AH$_2$, and GSH with O$_3$ as investigated by Enami et al. (2009a, b, c, 2008a, b) versus O$_3$ reacting with bulk liquid phase bioreactors as described previously. UA was more reactive than AH$_2$ toward O$_3$ in previous studies, but in reactions with O$_3$ using microdroplets, these antioxidants had equivalent reactivity (Enami et al., 2008b). As O$_3$ is a kinetically slow one-electron acceptor but very reactive O-atom donor, products of the interaction of O$_3$ with UA, AH$_2$, GSH, CyS, and α-TOH result from addition of $n$ O-atoms ($n = 1-4$). These products included epoxides (e.g., U-O’), peroxides (e.g., U-O$_2$’), and ozonides (e.g., U-O$_3$’). For instance, GSH was oxidized to sulfonates (GSO$_3^-$/GSO$_3^{2-}$), not glutathione disulfide (GSSG) by O$_3$ (Enami et al., 2009b). However, it is possible that other oxidative species are oxidizing GSH in vivo, since sulfonates are not detected in O$_3$ exposed ELF whereas GSSG is. This is also supported by the fact that O$_3$ is much less reactive with GSH than other antioxidants, such that <3% of O$_3$ will be scavenged by GSH when in equimolar amounts with AH$_2$ (Enami et al., 2009b).

This series of studies also demonstrated that ozonolysis product yields and formation were affected by pH. Acidified conditions (pH ≈ 3-4), such as those that may result from acidic particulate exposure or pathological conditions like asthma (pH ≈ 6), decreased the scavenging ability of UA and GSH for O$_3$; such that at low pH, the scavenging of O$_3$ must be taken over by other antioxidants, such as AH$_2$ (Enami et al., 2009b, 2008b). Also, under acidic conditions (pH ≈ 5), the ozonolysis products of AH$_2$ shifted from the innocuous dehydro-ascorbic acid to the more persistent products, AH$_2$ ozonide and threonic acid (Enami et al., 2008a). It is possible that the acidification of the ELF by acidic copollutant exposure will increase the toxicity of O$_3$ by preventing some antioxidant reactions and shifting the reaction products to more persistent compounds.

Since ELF exists as a complex mixture, it is important to look at O$_3$ reactivity in substrate mixtures. Individual antioxidant consumption rates decreased as the substrate mixture complexity increased (e.g., antioxidant mixtures and albumin addition) (Mudway and Kelly, 1998). However, O$_3$ reactions with AH$_2$ predominated over the reaction with lipids, when exposed to substrate solution mixtures (Postlethwait et al., 1998). It was suggested that O$_3$ may react with other substrates once AH$_2$ concentrations within the reaction plane fall sufficiently. Additionally, once AH$_2$ was consumed, the absorption efficiency diminished, allowing inhaled O$_3$ to be distributed to more distal airways (Postlethwait et al., 1998). Multiple studies have concluded O$_3$ is more reactive with AH$_2$ and UA than
with the weakly reacting GSH (or cysteine or methionine) or with amino acid residues and protein thiols (Kanofsky and Sima, 1995; Cross et al., 1992).

In a red blood cell (RBC) based system, AH₂ augmented the in vitro uptake of O₃ by 6-fold, as computed by the mass balance across the exposure chamber (Ballinger et al., 2005). However, estimated in vitro O₃ uptake was not proportional to the production of O₃-derived aldehydes from exposing O₃ to RBC membranes (Ballinger et al., 2005). In addition, O₃ induced cell membrane oxidation that required interactions with AH₂ and GSH, but not UA or the vitamin E analog Trolox. Further, aqueous phase reactions between O₃ and bovine serum albumin did not result in membrane oxidation (Ballinger et al., 2005). The presence of UA or bovine serum albumin protected against lipid and protein oxidation resulting from the reaction of O₃ and AH₂ (Ballinger et al., 2005). This study provided evidence that antioxidants may paradoxically facilitate O₃-mediated damage. This apparent contradiction should be viewed in terms of the concentration-dependent role of the ELF antioxidants. Reactions between O₃ and antioxidant species exhibited a biphasic concentration response, with oxidation of protein and lipid occurring at lower, but not higher, concentrations of antioxidant. In this way, endogenous reactants led to the formation of secondary oxidation products that were injurious and also led to quenching reactions that were protective. Moreover, the formation of secondary oxidation products mediated by some antioxidants was opposed by quenching reactions involving other antioxidants.

Alterations in ELF composition can result in alterations in O₃ uptake. Bolus O₃ uptake in human subjects can be decreased by previous continuous O₃ exposure (120-360 ppb), possibly due to depletion of compounds able to react with O₃ (Rigas et al., 1997; Asplund et al., 1996). Conversely, O₃ (360 ppb) bolus uptake was increased with prior NO₂ (360-720 ppb) or SO₂ (360 ppb) exposure (Rigas et al., 1997). It was hypothesized that this increased fractional absorption of O₃ could be due to increased production of reactive substrates in the ELF due to oxidant-induced airway inflammation.

Besides AH₂, GSH and UA, the ELF contains numerous antioxidant substances that appear to be an important cellular defense against O₃ including α-TOH, albumin, ceruloplasmin, lactoferrin, mucins, and transferrin (Mudway et al., 2006; Freed et al., 1999). The level and type of antioxidant present in ELF varies between species, regions of the RT, and can be altered by O₃ exposure. Mechanisms underlying the regional variability are not well-understood. It is thought that both plasma ultrafiltrate and locally secreted substances contribute to the antioxidant content of the ELF (Mudway et al., 2006; Freed et al., 1999). In the case of UA, the major source appears to be the plasma (Peden et al., 1995). Repletion of UA in nasal lavage fluid was demonstrated during sequential nasal lavage in human subjects (Mudway et al., 1999a). When these subjects, exercising at a moderate level, were exposed to 200 ppb O₃ for 2 hours, nasal lavage fluid UA was significantly decreased while plasma UA levels were significantly increased (Mudway et al., 1999a). The finding that UA, but not AH₂ or GSH, was depleted in nasal lavage fluid indicated that UA was the predominant antioxidant with respect to O₃ reactivity in the nasal cavity.
However, in human BALF samples, the mean consumption of AH$_2$ was greater than UA (Mudway et al., 1996). In addition, concentrations of UA were increased by cholinergic stimulation of the airways in human subjects, which suggested that increased mucosal gland secretions were an important source (Peden et al., 1993). Using the O$_3$-specific antioxidant capacity assay on human nasal lavage samples, Rutkowski et al. (2011) concluded that about 30% of the antioxidant capacity of the nasal ELF was attributed to UA activity. Additionally, more than 50% of the subject-to-subject differences in antioxidant capacity were driven by differences in UA concentration. However, day-to-day within-subject variations in measured antioxidant capacity were not related to the corresponding variations in UA concentration in the nasal lavage fluid. Efforts to identify the predominant antioxidant(s) in other RT regions besides the nasal cavity have failed to yield definitive results.

Regulation of AH$_2$, GSH and α-TOH concentrations within the ELF is less clear than that of UA (Mudway et al., 2006). In a sequential nasal lavage study in humans, wash-out of AH$_2$ and GSH occurred, indicating the absence of rapidly acting repletion mechanisms (Mudway et al., 1999a). Other studies demonstrated increases in BALF GSH and decreases in BALF and plasma AH$_2$ levels several hours following O$_3$ exposure (200 ppb for 2 h, while exercising at a moderate level) (Mudway et al., 2001; Blomberg et al., 1999; Mudway et al., 1999b). Studies with rats exposed to 0.4-1.1 ppm O$_3$ for 1-6 hours have shown consumption of AH$_2$ that correlates with O$_3$ exposure (Gunnison and Hatch, 1999; Gunnison et al., 1996; Vincent et al., 1996b). Further, cellular uptake of oxidized AH$_2$ by several cell types followed by intracellular reduction and export of reduced AH$_2$ has been demonstrated in vitro (Welch et al., 1995).

A body of evidence suggests that reaction of O$_3$ within the ELF limits its diffusive transport through the ELF; direct contact of O$_3$ with the apical membranes of the underlying epithelial cells therefore might be negligible in many regions of the RT (Ballinger et al., 2005; Connor et al., 2004; Postlethwait and Ultman, 2001; Pryor, 1992). This conclusion is based on computational analyses and in vitro studies. Direct confirmation using in vivo studies is limited. Nevertheless, when predicting exposure-related outcomes across species and anatomic sites, whether O$_3$ directly contacts the apical membranes of the epithelial cells is an important consideration, given that the extracellular surface milieu of the RT appreciably varies in terms of the types and concentrations of the substrates present and the thickness of the ELF.

For O$_3$ or its reaction products to gain access to the underlying cellular compartments, O$_3$ must diffuse at the air-liquid interface of the airway surface and travel through the ELF layer. In vitro experiments have shown that O$_3$ disappearance from the gas phase depends on the characteristics of the ELF substrates (Postlethwait et al., 1998; Hu et al., 1994). The ELF is comprised of the airway surface lining, which includes the periciliary sol layer and overlying mucus gel layer, and the alveolar surface lining, which includes the subphase of liquid and vesicular surfactant and the continuous surfactant monolayer (Bastacky et al., 1995). There is a progressive decrease in ELF thickness and increase in interfacial surface with
progression from the TB region to the alveolus (Mercer et al., 1992). The progressive thinning of the ELF while moving further down the RT decreases the radial distance O₃ or its reaction products must travel to reach the cells lining the RT.

Taking into account the high reactivity and low water solubility of O₃, Pryor (1992) estimated the distance that O₃ can penetrate into an ELF layer before it reacts with endogenous substrates to form other more long-lived reactive species, thus initiating a reaction cascade. These calculations utilize the Einstein-Smoluchowski equation to compare the time (t_{diff}) for O₃ to diffuse a distance (d) to the half-life (t_{rx}) of O₃ in its simultaneous reaction with substrates (Equation 5-1).

\[ t_{diff} = \frac{d^2}{2D_{O₃}} \text{ and } t_{rx} = \frac{\ln 2}{k_s C_s} \]

Equation 5-1

where \( D_{O₃} \) is the O₃ diffusion coefficient in ELF, \( k_s \) is the bimolecular reaction rate constant of O₃ with a reactive substrate (s) in ELF, and \( C_s \) is the molar substrate concentration. Importantly, it is assumed in the derivation of \( t_{rx} \) that the substrate is far in excess of O₃ so that \( C_s \) is spatially uniform in the ELF. To within some proportionality constant, the distance that O₃ penetrates can be estimated by equating \( t_{diff} \) to \( t_{rx} \) such that

\[ d \propto \left( \frac{D_{O₃}}{k_s C_s} \right)^{1/2} \]

Equation 5-2

There is reasonable certainty that the O₃ diffusion coefficient anywhere in the ELF is in the range of \( D_{O₃} \sim 10^{-5} – 10^{-6} \text{ cm}^2/\text{sec} \), but values of the \( k_s C_s \) product for the reaction of O₃ with specific substrates are much less reliable. Moreover, it is unknown which substrates make the most important contributions to \( k_s C_s \) and how these contributions vary from airway region to airway region. By asserting that polyunsaturated fatty acids are the primary reactive substrate, Miller et al. (1985) estimated that \( k_s C_s = 1,198 \text{ sec}^{-1} \) in airway surface lining fluid and \( k_s C_s = 21.4 \text{ sec}^{-1} \) in alveolar surface lining fluid. Pryor (1992) estimated the value of \( k_s C_s = 10^{-6} \text{ sec}^{-1} \), by assuming reduced glutathione is the primary substrate in airway surface lining fluid. A value of \( k_s C_s = 2.5 \times 10^5 \text{ sec}^{-1} \) was extracted from in vivo measurements of O₃ uptake into the airway surface lining fluid of the nasal cavities (Santiago et al., 2001). These studies suggest that there is an uncertainty in the magnitude of \( k_s C_s \) within airway surface lining fluid by a factor of 1,000, and that \( k_s C_s \) may be more than 100 times greater in airway surface lining fluid than in alveolar surface lining fluid.

With their estimates of \( k_s C_s = 10^6 \text{ sec}^{-1} \) and \( D_{O₃} = 10^{-6} \text{ cm}^2/\text{sec} \), Pryor (1992) concluded that O₃ could not penetrate an airway surface lining layer even as thin as 0.1 µm. Comparable computations with the \( k_s C_s = 1,198 \text{ sec}^{-1} \) value from Miller et
al. (1985) would indicate that O₃ penetrates an airway surface lining layer as thick as 3 µm. Since airway surface lining layer thickness is on the order of 10 µm in large airways and 0.1 µm in small airways, results using different estimates of \( k_sC_s \) have entirely different implications regarding the direct role of O₃ in damage to underlying epithelium versus the role of toxic reaction products.

In the nasal passages, in particular, a diffusion analysis of in vivo O₃ uptake measurements made at different air flows indicated that the O₃ penetration distance (0.5 µm) is considerably less than the thickness of the nasal surface lining layer (10 µm) (Santiago et al., 2001). A computational fluid dynamics model was able to predict experimentally measured O₃ uptake when the presences of a nasal surface lining layer thickness was considered (Cohen-Hubal et al., 1996), further indicating the need to properly account for the reaction-diffusion processes in the surface lining layer.

Despite calculations and in vitro studies suggesting that reactions of O₃ with underlying epithelial cells may be negligible in some regions of the RT, there is some evidence that suggests direct interaction of O₃ with epithelial cells is possible. While moving distally in the lung, the ELF thickness decreases and becomes ultra thin in the alveolar region, possibly allowing for direct interaction of O₃ with the underlying epithelial cells. One definitive study conducted in excised rat lung measured alveolar surface lining layer thickness over relatively flat portions of the alveolar wall to be 0.14 µm, to be 0.89 µm at the alveolar wall junctions, and 0.09 µm over the protruding features (Bastacky et al., 1995). The area-weighted average thickness of the alveolar surface lining fluid was found to be about 0.2 µm and the alveolar surface lining layer was continuous over the entire alveolar surface measured. The surface appeared smooth; and no epithelial surface features or macrophage features protruded above the air-liquid interface. It was noted that measurements of alveolar surface lining layer thickness were made in lungs prepared in a state of roughly 80% of total lung capacity, and as a result, the values reported would be approaching the lowest values possible during the respiratory cycle. However, 4% of the surface area in the alveolar compartment was covered by alveolar lining fluid layer of less than 20 nm (Bastacky et al., 1995), suggesting the possibility that unreacted O₃ could penetrate to the cell layer in this region. Further it remains a possibility that airways macrophages may protrude into the gas phase, allowing for direct contact between O₃ and airways epithelial cells.

Still, direct reaction of O₃ with alveolar epithelial cells or macrophages may be limited by the presence of dipalmitoyl phosphatidylcholine (DPPC), the major component of surfactant, which has been shown in vitro to inhibit uptake of O₃ into an aqueous compartment containing ascorbate, glutathione, and uric acid (Connor et al., 2004). Further, the amount of O₃ available to the alveolar compartment may be limited by uptake of O₃ in nasal and TB compartments. In fact, the amount of \(^{18}\text{O}\) reaction product was lower in the alveolar tissues than in TB tissues of rhesus monkeys immediately following a 2 hour exposure to \(^{18}\text{O}\)-labeled O₃ (0.4 and 1 ppm) (Plopper et al., 1998). These considerations illustrate the difficulty in determining whether O₃ reacts directly with cells in the alveolar compartment.
In some cases, however, with regard to the initiating mechanisms of cellular perturbations, the precise reactive species that encounters the epithelia might or might not have specificity to \( \text{O}_3 \) per se or to its secondary oxidants. Many of the measurable products formed as a consequence of \( \text{O}_3 \) exposure have limited specificity to \( \text{O}_3 \), such as 4-hydroxynonenal that is formed by autoxidation, an event that can be initiated by \( \text{O}_3 \) but also by a multitude of other oxidants. Although some classes of lipid oxidation products (e.g., specific aldehydes, cholesterol products) are specific to \( \text{O}_3 \), measurement in either BALF or in tissue does not necessarily provide insight regarding the compartment in which they were formed (i.e., the ELF, cell membrane, intracellular space) because the ELF is a dynamic compartment and, once formed, hydrophobic species can partition. Oxidation of membrane components might produce similar cellular outcomes regardless of the initiating oxidant. Lipid ozonides, which could be generated either within the ELF or from ozonation of cell membrane unsaturated lipids, could bind to receptors, activate signaling cascades, and act in other ways, making differences between pure extracellular reaction and direct membrane reaction indistinguishable. Thus, in some cases documenting whether \( \text{O}_3 \) per se reacts directly with cellular constituents might be essential (despite the challenges of in vivo demonstrations), while in other cases precisely where \( \text{O}_3 \) reacts might be of less concern with regard to characterizing mechanisms of health outcomes.

Thus, components of the ELF are major targets for \( \text{O}_3 \) and the resulting secondary oxidation products are key mediators of toxicity in the airways. The role of reaction products in \( \text{O}_3 \)-induced toxicity is discussed in Section 5.3. The reaction cascade resulting from the interaction of \( \text{O}_3 \) with ELF substrates can then carry the oxidative burden deeper into cells lining the RT to elicit the health effects observed.

### 5.2.3.1 Summary

The ELF is a complex mixture of lipids, proteins, and antioxidants that serve as the first barrier and target for inhaled \( \text{O}_3 \) (Figure 5-7). The thickness of the airways and alveolar surface lining layers is an important determinant of the dose of \( \text{O}_3 \) to the tissues. The progressive decrease in ELF thickness and increase in interfacial surface with progression from the TB region to the alveolus decreases the radial distance \( \text{O}_3 \) or its reaction products must travel to reach the cells lining the RT. The antioxidant substances present in the ELF appear in most cases to limit interaction of \( \text{O}_3 \) with underlying tissues and to prevent penetration of \( \text{O}_3 \) deeper into the lung. However, as the ELF thickness decreases and becomes ultra thin in the alveolar region, it may be possible for direct interaction of \( \text{O}_3 \) with the underlying epithelial cells to occur. The formation of secondary oxidation products is likely related to the concentration of antioxidants present and the quenching ability of the lining fluid. Mechanisms are present to replenish the antioxidant substrate pools as well as to remove secondary reaction products and prevent tissue interactions. Important differences exist in the reaction rates for \( \text{O}_3 \) and these ELF biomolecules and the reactivity of the resulting products. Overall, studies suggest that UA and AH\(_2\) are more reactive with \( \text{O}_3 \) than
GSH, proteins, or lipids. In addition to contributing to the driving force for O₃ uptake, formation of secondary oxidation products may lead to increased cellular injury and cell signaling (discussed in Section 5.3). Studies indicate that the antioxidants might be participating in reactions where the resulting secondary oxidation products might penetrate into the tissue layer and lead to perturbations.

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**Figure 5-7** Details of the O₃ interaction with the airway ELF to form secondary oxidation products.
5.3 Possible Pathways/Modes of Action

5.3.1 Introduction

Mode of action refers to a sequence of key events and processes that result in a given toxic effect (U.S. EPA, 2005). Elucidation of mechanisms provides a more detailed understanding of these key events and processes (U.S. EPA, 2005). Moreover, toxicity pathways describe the processes by which perturbation of normal biological processes produce changes sufficient to lead to cell injury and subsequent events such as adverse health effects (U.S. EPA, 2009f). The purpose of this section of Chapter 5 is to describe the key events and toxicity pathways that contribute to health effects resulting from short-term and long-term exposures to O₃. The extensive research carried out over several decades in humans and in laboratory animals has yielded numerous studies on mechanisms by which O₃ exerts its effects. This section will discuss some of the representative studies with particular emphasis on studies published since the 2006 O₃ AQCD (U.S. EPA, 2006b) and on studies in humans that inform biological mechanisms underlying responses to O₃.

It is well-appreciated that secondary oxidation products, which are formed as a result of O₃ exposure, initiate numerous responses at the cellular, tissue and whole organ level of the respiratory system. These responses include the activation of neural reflexes, initiation of inflammation, alteration of epithelial barrier function, sensitization of bronchial smooth muscle, modification of innate/adaptive immunity and airways remodeling, as will be discussed below. These have the potential to result in effects on other organ systems such as the cardiovascular, central nervous, hepatic and reproductive systems or result in developmental effects. It has been proposed that lipid ozonides and other secondary oxidation products, which are bioactive and cytotoxic in the respiratory system, are responsible for systemic effects. However it is not known whether they gain access to the vascular space (Chuang et al., 2009). Recent studies in animal models show that inhalation of O₃ results in systemic oxidative stress. The following subsections describe the current understanding of potential pathways and modes of action responsible for the pulmonary and extrapulmonary effects of O₃ exposure.

5.3.2 Activation of Neural Reflexes

Acute O₃ exposure results in reversible effects on lung function parameters through activation of neural reflexes. The involvement of bronchial C-fibers, a type of nociceptive sensory nerve, has been demonstrated in dogs exposed through an endotracheal tube to 2-3 ppm O₃ for 20-70 minutes (Coleridge et al., 1993; Schelegle et al., 1993). This vagal afferent pathway was found to be responsible for O₃-mediated rapid shallow breathing and other changes in respiratory mechanics in O₃-exposed dogs (Schelegle et al., 1993). Ozone also triggers neural reflexes that
stimulate the autonomic nervous system and alter electrophysiologic responses of the heart. For example, bradycardia, altered HRV and arrhythmia have been demonstrated in rodents exposed for several hours to 0.1-0.6 ppm O₃ (Hamade and Tankersley, 2009; Watkinson et al., 2001; Arito et al., 1990). Another effect is hypothermia, which in rodents occurred subsequent to the activation of neural reflexes involving the parasympathetic nervous system (Watkinson et al., 2001). Vagal afferent pathways originating in the RT may also be responsible for O₃-mediated activation of nucleus tractus solitarius neurons that resulted in neuronal activation in stress-responsive regions of the central nervous system (CNS) (rats, 0.5-2.0 ppm O₃ for 1.5-120 hours) (Gackière et al., 2011).

Recent studies in animals provide new information regarding the effects of O₃ on reflex responses mediated by bronchopulmonary C-fibers. In ex vivo mouse lungs, O₃ exposure (30 µM solubilized) selectively activated a subset of C-fiber receptors that are TRPA1 ion channels (Taylor-Clark and Undem, 2010). TRPA1 ion channels are members of the TRP family of ion channels, which are known to mediate the responses of sensory neurons to inflammatory mediators (Caceres et al., 2009). In addition to TRPA1 ion channels possibly playing a key role in O₃-induced decrements in pulmonary function, they may mediate allergic asthma (Caceres et al., 2009). Activation of TRPA1 ion channels following O₃ exposure is likely initiated by secondary oxidation products such as aldehydes and prostaglandins (Taylor-Clark and Undem, 2010) through covalent modification of cysteine and lysine residues (Trevisani et al., 2007). Ozonation of unsaturated fatty acids in the ELF was found to result in the generation of aldehydes (Frampton et al., 1999) such as 4-hydroxynonenal and 4-oxononenal (Taylor-Clark et al., 2008; Trevisani et al., 2007). 4-oxononenal is a stronger electrophile than 4-hydroxynonenal and exhibits greater potency toward the TRPA1 channels (Taylor-Clark et al., 2008; Trevisani et al., 2007). In addition, PGE₂ is known to sensitize TRPA1 channels (Bang et al., 2007).

In humans exercising at a moderate level, the response to O₃ (500 ppb for 2 h) was characterized by substernal discomfort, especially on deep inspiration, accompanied by involuntary truncation of inspiration (Hazucha et al., 1989). This latter response led to decreased inspiratory capacity and to decreased forced vital capacity (FVC) and forced expiratory volume in one second (FEV₁), as measured by spirometry. These changes, which occurred during O₃ exposure, were accompanied by decreased Vₜ and increased respiratory frequency in human subjects. Spirometric changes in FEV₁ and FVC were not due to changes in respiratory muscle strength (Hazucha et al., 1989). In addition, parasympathetic involvement in the O₃-mediated decreases in lung volume was minimal (Mudway and Kelly, 2000), since changes in FVC or symptoms were not modified by treatment with bronchodilators such as atropine in human subjects exposed to 400 ppb O₃ for 2 hours while exercising at a heavy level (Beckett et al., 1985). However, the loss of vital capacity was reversible with intravenous administration of the rapid-acting opioid agonist, sufentanil, in human subjects exercising at a moderate level and exposed to 420 ppb O₃ for 2 hours, which indicated the involvement of opioid receptor-containing nerve fibers and/or more central neurons (Passannante et al., 1998). The effects of sufentanil may be
attributed to blocking C-fiber stimulation by O₃ since activation of opioid receptors downregulated C-fiber function (Belvisi et al., 1992). Thus, nociceptive sensory nerves, presumably bronchial C-fibers, are responsible for O₃-mediated responses in humans (Passannante et al., 1998). This vagal afferent pathway is responsible for pain-related symptoms and inhibition of maximal inspiration in humans (Hazucha et al., 1989).

There is some evidence that eicosanoids (see Section 5.3.3) play a role in the neural reflex since cyclooxygenase inhibition with indomethacin (Alexis et al., 2000; Schelegle et al., 1987) or ibuprofen, which also blocks some lipoxygenase activity (Hazucha et al., 1996), before exposure to O₃ significantly blunted the spirometric responses. These studies involved exposures of 1-2 hours to 350-400 ppb O₃ in human subjects exercising at light, moderate, and heavy levels. In the latter study, ibuprofen treatment resulted in measurable decreases in BALF levels of PGE₂ and TXB₂ at 1-hour postexposure (Hazucha et al., 1996). Although an earlier study demonstrated that PGE₂ stimulated bronchial C-fibers (Coleridge et al., 1993; Coleridge et al., 1976) and suggested that PGE₂ mediated O₃-induced decreases in pulmonary function, no correlation was observed between the degree of ibuprofen-induced inhibition of BALF PGE₂ levels and blunting of the spirometric response to O₃ (Hazucha et al., 1996). These results point to the involvement of a lipoxygenase product. Further, as noted above, PGE₂ may play a role in the neural reflex by sensitizing TRPA1 channels. A recent study in human subjects exercising at a moderate to high level and exposed for 1 hour to 350 ppb O₃ also provided evidence that arachidonic acid metabolites, as well as oxidative stress, contribute to human responsiveness to O₃ (Alfaro et al., 2007).

In addition to the spirometric changes, mild airways obstruction occurred in human subjects exercising at a moderate level during O₃ exposure (500 ppb for 2 hours) (Hazucha et al., 1989). This pulmonary function decrement is generally measured as specific airway resistance (sRaw) which is the product of airway resistance and thoracic gas volume. In several studies involving human subjects exercising at a moderate to heavy level and exposed for 1-4 hours to 200-300 ppb O₃, changes in sRaw correlated with changes in inflammatory and injury endpoints measured 18-hours postexposure, but did not follow the same time course or change to the same degree as spirometric changes (i.e., FEV₁, FVC) measured during exposure (Balmes et al., 1996; Aris et al., 1993; Schelegle et al., 1991). In addition, a small but persistent increase in airway resistance associated with narrowing of small peripheral airways (measured as changes in isovolumetric FEF₂₅₋₇₅) was demonstrated in O₃-exposed human subjects (350 ppb for 130 minutes, moderate exercise level) (Weinmann et al., 1995c; Weinmann et al., 1995b). A similar study (400 ppb O₃ for 2 hours in human subjects exercising at a heavy level) found decreases in FEF₂₅₋₇₅ concomitant with increases in residual volume, which is suggestive of small airways dysfunction (Kreit et al., 1989). In separate studies, a statistically significant increase in residual volume (500 ppb for 2 hours) (Hazucha et al., 1989) and a statistically significant decrease in FEF₂₅₋₇₅ (160 ppb for 7.6 hours) (Horstman et al., 1995) were observed following O₃ exposure in human subjects exercising at moderate and light
levels, respectively, providing further support for an $O_3$-induced effect on small airways.

Mechanisms underlying this rapid increase in airway resistance following $O_3$ exposure are incompletely understood. Pretreatment with atropine decreased baseline $sRaw$ and prevented $O_3$-induced increases in $sRaw$ in human subjects exercising at a heavy level (400 ppb for 0.5 hours) (Beckett et al., 1985), indicating the involvement of muscarinic cholinergic receptors of the parasympathetic nervous system. Interestingly, atropine pretreatment partially blocked the decrease in $FEV_1$, but had no effect on the decrease in $FVC$, breathing rate, tidal volume or respiratory symptoms (Beckett et al., 1985). Using a $\beta$-adrenergic agonist, it was shown that smooth muscle contraction, not increased airway mucus secretion, was responsible for $O_3$-induced increases in airway resistance (Beckett et al., 1985). Thus, pulmonary function decrements measured as $FEV_1$ may reflect both restrictive (such as decreased inspiratory capacity) and obstructive (such as bronchoconstriction) type changes in airway responses. This is consistent with findings of McDonnell et al. (1983) who observed a relatively strong correlation between $sRaw$ and $FEV_1$ ($r = -0.31$, $p = 0.001$) and a far weaker correlation between $sRaw$ and $FVC$ ($r = -0.16$, $p = 0.10$) in human subjects exercising at a heavy level and exposed for 2.5 hours to 120-400 ppb $O_3$.

Furthermore, tachykinins may contribute to $O_3$-mediated increases in airway resistance. In addition to stimulating CNS reflexes, bronchopulmonary C-fibers mediate local axon responses by releasing neuropeptides such as substance P (SP), neurokinin (NK) A and calcitonin gene-related peptide (CGRP). Tachykinins bind to NK receptors resulting in responses such as bronchoconstriction. Recent studies in animals demonstrated that NK-1 receptor blockade had no effect on $O_3$-stimulated physiologic responses such as $V_T$ and $f_B$ in rats over the 8 hour exposure to 1 ppm $O_3$ (Oslund et al., 2008). However, SP and NK receptors contributed to vagally-mediated bronchoconstriction in guinea pigs 3 days after a single 4-hour exposure to 2 ppm $O_3$ (Verhein et al., 2011). In one human study in which bronchial biopsies were performed and studied by immunohistochemistry, SP was substantially diminished in submucosal sensory nerves 6 hours following $O_3$ exposure (200 ppb for 2 hours, light exercise) (Krishna et al., 1997). A statistically significant correlation was observed between loss of SP immunoreactivity from neurons in the bronchial mucosa and changes in $FEV_1$ measured 1-hour postexposure (Krishna et al., 1997). Another study found that SP was increased in lavage fluid of human subjects immediately after $O_3$ challenge (250 ppb for 1 hour, heavy exercise) (Hazbun et al., 1993). These results provide evidence that the increased airway resistance observed following $O_3$ exposure is due to vagally-mediated responses and possibly by local axon reflex responses through bronchopulmonary C-fiber-mediated release of SP.

A role for antioxidant defenses in modulating neural reflexes has been proposed given the delay in onset of $O_3$-induced pulmonary function responses that has been noted in numerous studies. Recently, this delay was characterized in terms of changes in $f_B$ (Schelegle et al., 2007). In humans exposed for 1-2 hours to...
120-350 ppb O₃ while exercising at a high level, no change in f_B was observed until a certain cumulative inhaled dose of O₃ had been reached. Subsequently, the magnitude of the change in f_B was correlated with the inhaled dose rate (Schelegle et al., 2007). These investigators proposed that initial reactions of O₃ with ELF resulted in a time-dependent depletion of ELF antioxidants, and that activation of neural reflexes occurred only after the antioxidant defenses were overwhelmed (Schelegle et al., 2007).

5.3.3 Initiation of inflammation

As described previously (Section 5.2.3), O₃ mainly reacts with components of the ELF and cellular membranes resulting in the generation of secondary oxidation products. Higher concentrations of these products may directly injure RT epithelium. Subsequent airways remodeling may also occur (Section 5.3.7) (Mudway and Kelly, 2000). Lower concentrations of secondary oxidation products may initiate cellular responses including cytokine generation, adhesion molecule expression, and modification of tight junctions leading to inflammation and increased permeability across airway epithelium (Section 5.3.4) (Dahl et al., 2007; Mudway and Kelly, 2000).

An important hallmark of acute O₃ exposure in humans and animals is neutrophilic airways inflammation. Neutrophil influx into nasal airways has been demonstrated in human subjects (400 ppb O₃ 2 hours, heavy exercise) (Graham and Koren, 1990) and in rats (0.8 ppm O₃, 6 hours) (Hotchkiss et al., 1989). Many studies of neutrophil influx have focused on the lower airways (Hazucha et al., 1996; Aris et al., 1993). The time course of this response in the lower airways and its resolution appears to be slower than that of the decrements in pulmonary function in exercising human subjects (Hazucha et al., 1996). In several studies, airways neutrophilia was observed by 1-3 hours, peaked by 6 hours and was returning to baseline levels at 18-24 hours in human subjects exercising at a heavy level and exposed for 1-2 hours to 300-400 ppb O₃ (Schelegle et al., 1991; Koren et al., 1989; Seltzer et al., 1986). Neutrophils are thought to be injurious and a study in guinea pigs demonstrated that the influx and persistence of neutrophils in airways following O₃ exposure correlated with the temporal profile of epithelial injury (0.26-1 ppm O₃, 72 hours) (Hu et al., 1982). However, neutrophils have also been shown to contribute to repair of O₃-injured epithelium in rats exposed for 8 hours to 1 ppm O₃, possibly by removing necrotic epithelial cells (Mudway and Kelly, 2000; Vesely et al., 1999). Nonetheless, the degree of airways inflammation due to O₃ is thought to have more important long-term consequences than the more quickly resolving changes in pulmonary function since airways inflammation is often accompanied by tissue injury (Balmes et al., 1996).

Ozone exposure results in alterations in other airways inflammatory cells besides neutrophils, including lymphocytes, macrophages, monocytes and mast cells. Influx of some of these cells accounts for the later (i.e., 18-20 hours) phase of inflammation.
following O\textsubscript{3} exposure. Numbers of lymphocytes and total cells in BALF were decreased early after O\textsubscript{3} exposure in human subjects exercising at a light to moderate level and exposed for 2 hours to 200 ppb O\textsubscript{3}, which preceded the neutrophil influx (Mudway and Kelly, 2000; Blomberg et al., 1999; Krishna et al., 1997). The decrease in total cells was thought to reflect decreases in macrophages, although it was not clear whether the cells were necrotic or whether membrane adhesive properties were altered making them more difficult to obtain by lavage (Mudway and Kelly, 2000; Blomberg et al., 1999; Mudway et al., 1999b; Frampton et al., 1997b; Pearson and Bhalla, 1997). A recent study in human subjects exercising at a moderate level and exposed for 6.6 hours to 80 ppb O\textsubscript{3} demonstrated an increase in numbers of sputum monocytes and dendritic-like cells with increased expression of innate immune surface proteins and antigen presentation markers (Alexis et al., 2010). An increase in submucosal mast cells was observed 1.5 hours after a 2 hour-exposure to 200 ppb O\textsubscript{3} (Blomberg et al., 1999) and an increase in BAL mast cell number was observed 18 hours after a 4-hour exposure to 220 ppb O\textsubscript{3} exposure in human subjects exercising at a moderate level (Frampton et al., 1997b). Mast cells may play an important role in mediating neutrophil influx since they are an important source of several pro-inflammatory cytokines and since their influx preceded that of neutrophils in human subjects exercising at a moderate level and exposed for 2 hours to 200 ppb O\textsubscript{3} (Stenfors et al., 2002; Blomberg et al., 1999). Further, a study using mast cell-deficient mice demonstrated decreased neutrophilic inflammation in response to O\textsubscript{3} (1.75 ppm, 3 hours) compared with wild type mice (Kleeberger et al., 1993). Influx of these inflammatory cell types in the lung is indicative of O\textsubscript{3}-mediated activation of innate immunity as will be discussed in Section 5.3.6.

Much is known about the cellular and molecular signals involved in inflammatory responses to O\textsubscript{3} exposure (U.S. EPA, 2006b). Eicosanoids are one class of secondary oxidation products that may be formed rapidly following O\textsubscript{3} exposure and that may mediate inflammation. Eicosanoids are metabolites of arachidonic acid—a 20-carbon PUFA—that are released from membrane phospholipids by phospholipase A\textsubscript{2}-mediated catalysis. Activation of phospholipase A\textsubscript{2} occurs by several cell signaling pathways and may be triggered by O\textsubscript{3}-mediated lipid peroxidation of cellular membranes (Rashba-Step et al., 1997). Additionally, cellular phospholipases A\textsubscript{2}, C and D may be activated by lipid ozonation products (Kafoury et al., 1998). While the conversion of arachidonic acid to prostaglandins, leukotrienes and other eicosanoid products is generally catalyzed by cyclooxygenases and lipoxygenases, non-enzymatic reactions also occur during oxidative stress leading to the generation of a wide variety of eicosanoids and reactive oxygen species. Further, the release of arachidonic acid from phospholipids is accompanied by the formation of lysophospholipids that are precursors for platelet activating factors (PAFs). Thus, formation of eicosanoids, reactive oxygen species and PAFs accompanies O\textsubscript{3}-mediated lipid peroxidation.

In addition, secondary reaction products may stimulate macrophages to produce cytokines such as IL-1, IL-6, and TNF-\textalpha that in turn activate IL-8 production by epithelial cells. Although IL-8 has been proposed to play a role in neutrophil chemotaxis, measurements of IL-8 in BALF from humans exposed to O\textsubscript{3} found
The time-course profiles of PGE₂ and IL-6 responses suggest that they may play a role in neutrophil chemotaxis in humans (Mudway and Kelly, 2000). However, pretreatment with ibuprofen attenuated O₃-induced increases in BALF PGE₂ levels, but had no effect on neutrophilia in human subjects exercising at a heavy level and exposed for 2 hour to 400 ppb O₃ (Hazucha et al., 1996).

One set of studies in humans focused on the earliest phase of airways inflammation (1-2 hours following exposure). Human subjects, exercising at a moderate level, were exposed to 200 ppb O₃ for 2 hours and bronchial biopsy tissues were obtained 1.5 and 6 hours after exposure (Bosson et al., 2009; Bosson et al., 2003; Stenfors et al., 2002; Blomberg et al., 1999). Results demonstrated upregulation of vascular endothelial adhesion molecules P-selectin and ICAM-1 at both 1.5 and 6 hours (Stenfors et al., 2002; Blomberg et al., 1999). Submucosal mast cell numbers were increased at 1.5 hours in the biopsy samples without an accompanying increase in neutrophil number (Blomberg et al., 1999). Pronounced neutrophil infiltration was observed at 6 hours in the bronchial mucosa (Stenfors et al., 2002). Surprisingly, suppression of the NF-κB and AP-1 pathways at 1.5 hours and a lack of increased IL-8 at 1.5 or 6 hours in bronchial epithelium were observed (Bosson et al., 2009). The authors suggested that vascular endothelial adhesion molecules, rather than redox sensitive transcription factors, play key roles in early neutrophil recruitment in response to O₃.

Increases in markers of inflammation occurred to a comparable degree in human subjects with mild (least sensitive) and more remarkable (more sensitive) spirometric responses to O₃ (200 ppb, 4 hours, moderate exercise) (Balmes et al., 1996). Two other studies (200 ppb for 4 hours with moderate exercise and 300 ppb for 1 hour with heavy exercise) found that acute spirometric changes were not positively correlated with cellular and biochemical indicators of inflammation (Aris et al., 1993; Schelegle et al., 1991). However inflammation was correlated with changes in sRaw (Balmes et al., 1996). In another study, pretreatment with ibuprofen had no effect on neutrophilia although it blunted the spirometric response in human subjects exercising at heavy level and exposed for 2 hours to 400 ppb O₃ (Hazucha et al., 1996). Taken together, results from these studies indicate different mechanisms underlying the spirometric and inflammatory responses to O₃.

A common mechanism underlying both inflammation and impaired pulmonary function was suggested by Krishna et al. (1997). This study, conducted in human subjects exercising at a light level and exposed to 200 ppb O₃ for 2 hours, demonstrated a correlation between loss of SP immunoreactivity from neurons in the bronchial mucosa and numbers of neutrophils and epithelial cells (shed epithelial cells are an index of injury) in the BALF 6-hours postexposure. Furthermore, the loss of SP immunoreactivity was correlated with the observed changes in FEV₁. Another study found that SP was increased in lavage fluid of exercising human subjects immediately after O₃ challenge (250 ppb, 1 hour, heavy exercise) (Hazbun et al., 1992). SP is a neuropeptide released by sensory nerves which mediates neurogenic edema and bronchoconstriction (Krishna et al., 1997). Collectively, these findings
suggest that O₃-mediated stimulation of sensory nerves that leads to activation of central and local axon reflexes is a common effector pathway leading to impaired pulmonary function and inflammation.

Studies in animal models have confirmed many of these findings and provided evidence for additional mechanisms involved in O₃-induced inflammation. A study in mice (2 ppm O₃, 3 hours) demonstrated that PAF may be important in neutrophil chemotaxis (Longphre et al., 1999), while ICAM-1 and macrophage inflammatory protein-2 (MIP-2), the rodent IL-8 homologue, have been implicated in a rat model (1 ppm O₃, 3 hours) (Bhalla and Gupta, 2000). Another study found that TNF receptor, NF-κB and JNK1 mediated lung inflammation induced by O₃ in mice (0.3 ppm O₃, 6 and 24 hours) (Cho et al., 2007). Key roles for CXCR2, a receptor for keratinocyte-derived chemokine (KC) and MIP-2, and for IL-6 in O₃-mediated neutrophil influx were demonstrated in mice (1 ppm O₃, 3 hours) (Johnston et al., 2005a; Johnston et al., 2005b). Activation of JNK and p38 pathways and cathepsin-S were also found to be important in this response (3 ppm O₃, 3 hours) (Williams et al., 2009a; Williams et al., 2008a; Williams et al., 2007a). Matrix metalloproteinase-9 (MMP-9) appeared to confer protection against O₃-induced airways inflammation and injury in mice (0.3 ppm O₃, 6-72 hours) (Yoon et al., 2007). Interleukin-10 (IL-10) also appeared to be protective since IL-10 deficient mice responded to O₃ exposure (0.3 ppm, 24-72 hours) with enhanced numbers of BAL neutrophils, enhanced NF-κB activation and MIP-2 levels compared with IL-10 sufficient mice (Backus et al., 2010).

In addition, lung epithelial cells may release ATP in response to O₃ exposure (Ahmad et al., 2005). ATP and its metabolites (catalyzed by ecto-enzymes) can bind to cellular purinergic receptors resulting in activation of cell signaling pathways (Picher et al., 2004). One such metabolite, adenine, is capable of undergoing oxidation leading to the formation of UA which, if present in high concentrations, could activate inflammasomes and result in caspase 1 activation and the maturation and secretion of IL-1β and IL-18 (Dostert et al., 2008). A recent study in human subjects exercising at a moderate level and exposed for 2 hours to 400 ppb O₃ demonstrated a correlation between ATP metabolites and inflammatory markers (Esther et al., 2011), which provides some support for this mechanism.

Several recent studies have focused on the role of Toll-like receptor (TLR) and its related adaptor protein MyD88 in mediating O₃-induced neutrophilia. Hollingsworth et al. (2004) demonstrated airways neutrophilia that was TLR4-independent following acute (2 ppm, 3 hours) and subchronic (0.3 ppm, 72 hours) O₃ exposure in a mouse model. However, Williams et al. (2007b) found that MyD88 was important in mediating O₃-induced neutrophilia in mice (3 ppm, 3 hours), with TLR4 and TLR2 contributing to the speed of the response. Moreover, MyD88, TLR2 and TLR4 contributed to inflammatory gene expression in this model and O₃ upregulated MyD88, TLR4 and TLR4 gene expression (Williams et al., 2007a). Neutrophilic inflammation was also found to be partially dependent on MyD88 in mice exposed to 1 ppm O₃ for 3 hours (Li et al., 2011).
Hyaluronan was found to mediate a later phase (24 hours) of O₃-induced inflammation in mice (Garantziotis et al., 2010; Garantziotis et al., 2009). Hyaluronan is an extracellular matrix component that is normally found in the ELF as a large polymer. Exposure to 2 ppm O₃ for 3 hours resulted in elevated levels of soluble low molecular weight hyaluronan in the BALF 24-hours postexposure (Garantziotis et al., 2010; Garantziotis et al., 2009). Similar results were found in response to 3 hour exposure to 1 ppm O₃ (Li et al., 2011). Ozone may have caused the depolymerization of hyaluronan to soluble fragments that are known to be endogenous ligands of the CD44 receptor and TLR4 in the macrophage (Jiang et al., 2005). Binding of hyaluronan fragments to the CD44 receptor activates hyaluronan clearance, while binding to TLR4 results in signaling through MyD88 to produce chemokines that stimulate the influx of inflammatory cells (Jiang et al., 2005). Activation of NF-κB occurred in both airway epithelia and alveolar macrophages 24-hours postexposure to O₃. Increases in BALF pro-inflammatory factors KC, IL-1β, MCP-1, TNF-α and IL-6 observed 24 hours following O₃ exposure were found to be partially dependent on TLR4 (Garantziotis et al., 2010) while increases in BAL inflammatory cells, which consisted mainly of macrophages, were dependent on CD44 (Garantziotis et al., 2009). BAL inflammatory cells number and injury markers following O₃ exposure were similar in wild-type and TLR4-deficient animals (Garantziotis et al., 2010).

Since exposure to O₃ leads to airways inflammation characterized by neutrophilia, and since neutrophil-derived oxidants often consume ELF antioxidants, concentrations of ELF antioxidants have been examined during airways neutrophilia (Long et al., 2001; Gunnison and Hatch, 1999; Mudway et al., 1999b). In human subjects exercising at a moderate level and exposed to 200 ppb O₃ for 2 hours, UA, GSH and α-TOH levels remained unchanged in BALF 6-hours postexposure while AH₂ was decreased significantly in both BALF and plasma (Mudway et al., 1999b). A second study involving the same protocol reported a loss of AH₂ from bronchial wash fluid and BALF, representing proximal and distal airway ELF respectively, as well as an increase in oxidized GSH in both compartments (Mudway et al., 2001). No change was observed in ELF UA levels in response to O₃ (Mudway et al., 2001). Further, O₃ exposure (0.8 ppm, 4 hours) in female rats resulted in a 50% decrease in BALF AH₂ immediately postexposure (Gunnison and Hatch, 1999). These studies suggested a role for AH₂ and GSH in protecting against the oxidative stress associated with inflammation.

The relationship between inflammation, antioxidant status and O₃ dose has also been investigated. The degree of inflammation in rats has been correlated with ¹⁸O-labeled O₃ dose markers in the lower lung. In female rats exposed to 0.8 ppm O₃ for 4 hours, BAL neutrophil number and ¹⁸O reaction product were directly proportional (Gunnison and Hatch, 1999). Kari et al. (1997) observed that a 3-week caloric restriction (75%) in rats abrogated the toxicity of O₃ (2 ppm, 2 hours), measured as BALF increases in protein, fibronectin and neutrophils, that was seen in normally fed rats. Accompanying this resistance to O₃ toxicity was a reduction (30%) in the accumulation of ¹⁸O reaction product in the lungs. These investigations also demonstrated an inverse relationship between AH₂ levels and O₃ dose and provided
evidence for AH$_2$ playing a protective role following O$_3$ exposure in these studies. Pregnant and lactating rats had lower AH$_2$ content in BALF and exhibited a greater increase in accumulation of $^{18}$O reaction products compared with pre-pregnant rats in response to O$_3$ exposure (Gunnison and Hatch, 1999). In the calorie restricted model, a 30% higher basal BALF AH$_2$ concentration and a rapid accumulation of AH$_2$ into the lungs to levels 60% above normal occurred as result of O$_3$ exposure (Kari et al., 1997). However, this relationship between AH$_2$ levels and O$_3$ dose did not hold up in every study. Aging rats (9 and 24 months old) had 49% and 64% lower AH$_2$ in lung tissue compared with month-old rats but the aging-induced AH$_2$ loss did not increase the accumulation of $^{18}$O reaction products following O$_3$ exposure (0.4-0.8 ppm, 2-6 hours) (Vincent et al., 1996b).

A few studies have examined the dose- or concentration-responsiveness of airways neutrophilia in O$_3$-exposed humans (Holz et al., 1999; Devlin et al., 1991). No concentration-responsiveness was observed in healthy human subjects exposed for 1 hour to 125-250 ppb O$_3$ while exercising at a light level and a statistically significant increase in sputum neutrophilia was observed only at the higher concentration (Holz et al., 1999). However, concentration-dependent and statistically significant increases in BAL neutrophils and the inflammatory mediator IL-6 were reported following exposure to 80 and 100 ppb O$_3$ for 6.6 hours in human subjects exercising at a moderate level (Devlin et al., 1991). Additional evidence is provided by a meta-analysis of the O$_3$ dose-inflammatory response in controlled human exposure studies involving exposure to 80-600 ppb O$_3$ for 60-396 minutes and exercise levels ranging from light to heavy (Mudway and Kelly, 2004b). Results demonstrated a linear relationship between inhaled O$_3$ dose (determined as the product of concentration, ventilation and time) and BAL neutrophils at 0-6 hours and 18-24 hours following O$_3$ exposure (Mudway and Kelly, 2004b).

### 5.3.4 Alteration of Epithelial Barrier Function

Following O$_3$ exposure, injury and inflammation can lead to altered airway barrier function. Histologic analysis has demonstrated damage to tight junctions between epithelial cells, suggesting an increase in epithelial permeability. In addition, the presence of shed epithelial cells in the BALF and increased epithelial permeability, which is measured as the flux of small solutes, have been observed and are indicative of epithelial injury. This could potentially lead to the loss of ELF solutes that could diffuse down their concentration gradient from the lung to the blood. Increases in vascular permeability, as measured by BALF protein and albumin, have also been demonstrated (Costa et al., 1985; Hu et al., 1982).

An early study in sheep measured changes in airway permeability as the flux of inhaled radiolabeled histamine into the plasma (Abraham et al., 1984). Exposure of sheep to 0.5 ppm O$_3$ for 2 hours via an endotracheal tube resulted in an increased rate of histamine appearance in the plasma at 1 day postexposure. Subsequently, numerous studies have measured epithelial permeability as the flux of the small
solute $^{99}$Tc-DTPA that was introduced into the air spaces in different regions of the RT. Increased pulmonary epithelial permeability, measured as the clearance of $^{99}$Tc-DTPA from lung to blood, was demonstrated in humans 1-2 hours following a 2-hour exposure to 400 ppb O$_3$ while exercising at a heavy level (Kehrl et al., 1987). Another study in human subjects found increased epithelial permeability 19-hours postexposure to 240 ppb O$_3$ for 130 minutes while exercising at moderate level (Foster and Stetkiewicz, 1996). Increased bronchial permeability was also observed in dogs 1-day postexposure (0.4 ppm O$_3$ by endotracheal tube for 6 hours) and did not resolve for several days (Foster and Freed, 1999).

A role for tachykinins in mediating airway epithelial injury and decreased barrier function has been suggested. Nishiyama et al. (1998) demonstrated that capsaicin, which depletes nerve fibers of substance P, blocked the O$_3$-induced increase in permeability of guinea pig tracheal mucosa (0.5-3 ppm O$_3$, 0.5 hours). Pretreatment with propranolol or atropine failed to inhibit this response, suggesting that adrenergic and cholinergic pathways were not involved. In another study, tachykinins working through NK-1 and CGRP receptors were found to contribute to airway epithelial injury in O$_3$-exposed rats (1 ppm, 8 hours) (Oslund et al., 2009, 2008).

Kleeberger et al. (2000) evaluated genetic susceptibility to O$_3$-induced altered barrier function in recombinant inbred strains of mice. Lung hyperpermeability, measured as BALF protein, was evaluated 72 hours after exposure to 0.3 ppm O$_3$ and found to be associated with a functioning Tlr4 gene. This study concluded that Tlr4 was a strong candidate gene for susceptibility to hyperpermeability in response to O$_3$ (Kleeberger et al., 2000). A subsequent study by these same investigators found that Tlr4 modulated mRNA levels of the Nos2 genes and suggested that the protein product of Nos2, iNOS, plays an important role in O$_2$-induced lung hyperpermeability (0.3 ppm, 72 hours) (Kleeberger et al., 2001). More recently, HSP70 was identified as part of the TLR4 signaling pathway (0.3 ppm, 6-72 hours) (Bauer et al., 2011).

Antioxidants have been shown to confer resistance to O$_3$-induced injury. In a recent study, lung hyperpermeability in response to O$_3$ (0.3 ppm, 48 hours) was unexpectedly reduced in mice deficient in the glutamate-cysteine ligase modifier subunit gene compared with sufficient mice (Johansson et al., 2010). Since the lungs of these mice exhibited 70% glutathione depletion, protection against O$_3$-induced injury was unexpected (Johansson et al., 2010). However it was found that several other antioxidant defenses, including metallothionein, were upregulated in response to O$_3$ to a greater degree in the glutathione-deficient mice compared with sufficient mice (Johansson et al., 2010). The authors suggested that resistance to O$_3$-induced lung injury was due to compensatory augmentation of antioxidant defenses (Johansson et al., 2010). Antioxidant effects have also been attributed to Clara cell secretory protein (CCSP) and surfactant protein A (SP-A). CCSP was found to modulate the susceptibility of airway epithelium to injury in mice exposed to O$_3$ (0.2 or 1 ppm for 8 hours) by an unknown mechanism (Plopper et al., 2006). SP-A appeared to confer protection against O$_3$-induced airways inflammation and injury in mice (2 ppm, 3 hours) (Haque et al., 2007).
Increased epithelial permeability has been proposed to play a role in allergic sensitization (Matsumura, 1970), in activation of neural reflexes and in stimulation of smooth muscle receptors (Dimeo et al., 1981). Abraham et al. (1984) reported a correlation between airway permeability and airways hyperresponsiveness (AHR) in O₃-exposed sheep. However a recent study in human subjects exposed to 220 ppb O₃ for 135 minutes while exercising at a light to moderate level did not find a relationship between O₃-induced changes in airway permeability and AHR (Que et al., 2011).

5.3.5 Sensitization of Bronchial Smooth Muscle

Bronchial reactivity is generally determined in terms of a response to a challenge agent. Non-specific bronchial reactivity in humans is assessed by measuring the effect of inhaling increasing concentrations of a bronchoconstrictive drug on lung mechanics (sRaw or FEV₁). Methacholine is most commonly employed but histamine and other agents are also used. Specific bronchial reactivity is assessed by measuring effects in response to an inhaled allergen in individuals (or animals) already sensitized to that allergen. An increase in sRaw in response to non-specific or specific challenge agents indicates AHR.

In addition to causing mild airways obstruction as discussed above, acute O₃ exposure results in reversible increases in bronchial reactivity by mechanisms that are not well understood. In one study, bronchial reactivity of healthy subjects was significantly increased 19-hours postexposure to O₃ (120-240 ppb O₃ for 2 hours with moderate exercise) (Foster et al., 2000). These effects may be more considerable in human subjects with already compromised airways (Section 5.4.2.2).

Ozone may sensitize bronchial smooth muscle to stimulation through an exposure-related effect on smooth muscle or through effects on the sensory nerves in the epithelium or on the motor nerves innervating the smooth muscle (O'Byrne et al., 1984; O'Byrne et al., 1983; Holtzman et al., 1979). It is also recognized that increased bronchial reactivity can be both a rapidly occurring and a persistent response to O₃ (Foster and Freed, 1999). Tachykinins and secondary oxidation products of O₃ have been proposed as mediators of the early response and inflammation-derived products have been proposed as mediators of the later response (Foster and Freed, 1999). Furthermore, bronchial reactivity may be increased as a result of O₃-mediated generation of ROS.

Ozone-induced increases in epithelial permeability, which could improve access of agonist to smooth muscle receptors, may be one mechanism of sensitization through a direct effect on bronchial smooth muscle (Holtzman et al., 1979). As noted above, a correlation between airway permeability and AHR has been reported in O₃-exposed sheep (Abraham et al., 1984) but not in O₃-exposed human subjects (Que et al., 2011).
Neurally-mediated sensitization has been demonstrated. In human subjects exposed for 2 hours to 600 ppb O₃ while exercising at a light level, pretreatment with atropine inhibited O₃-induced AHR, suggesting the involvement of cholinergic postganglionic pathways (Holtzman et al., 1979). Animal studies have demonstrated that O₃-induced AHR involved vagally-mediated responses (rabbits, 0.2 ppm O₃, 72 hours) (Freed et al., 1996) and local axon reflex responses through bronchopulmonary C-fiber-mediated release of SP (guinea pigs, 0.8 ppm O₃, 2 hours) (Joad et al., 1996). Further, pretreatment with capsaicin to deplete nerve fibers of SP blocked O₃-mediated AHR (guinea pigs, 1-2 ppm O₃, 2-2.25 hours) (Tepper et al., 1993). Other investigators demonstrated that SP released from airway nociceptive neurons in ferrets contributed to O₃-induced AHR (2 ppm O₃, 3 hours) (Wu et al., 2008c; Wu et al., 2003).

Some evidence suggests the involvement of arachidonic acid metabolites and neutrophils in mediating O₃-induced AHR (Seltzer et al., 1986; Fabbri et al., 1985). Increased BAL neutrophils and cyclooxygenase products were found in one study demonstrating AHR in human subjects exercising at a heavy level immediately postexposure to 600 ppb O₃ for 2 hours (Seltzer et al., 1986). Another study found that ibuprofen pretreatment had no effect on AHR in human subjects exercising at a heavy level following exposure to 400 ppb O₃ for 2 hours, although spirometric responses were blunted (Hazucha et al., 1996). This study measured arachidonic acid metabolites and provided evidence that the arachidonic acid metabolites whose generation was blocked by ibuprofen, (i.e., prostaglandins, thromboxanes and some leukotrienes) did not play a role in AHR. Experiments in dogs exposed for 2 hours to 2.1 ppm O₃ demonstrated a close correlation between O₃-induced AHR and airways neutrophilic inflammation measured in tissue biopsies (Holtzman et al., 1983). Furthermore, the increased AHR observed in dogs following O₃ exposure (3 ppm, 2 hours) was inhibited by neutrophil depletion (O'Byrne et al., 1983) and by pretreatment with inhibitors of arachidonic acid metabolism. In one of these studies, indomethacin pre-treatment did not prevent airways neutrophilia in response to O₃ (3 ppm, 2 hours) providing evidence that the subset of arachidonic acid metabolites whose generation was inhibitable by the cyclooxygenase inhibitor indomethacin (i.e., prostaglandins and thromboxanes) was not responsible for neutrophil influx (O'Byrne et al., 1984). It should be noted that these studies did not measure whether the degree to which the inhibitor blocked arachidonic acid metabolism and thus their results should be interpreted with caution. Taken together, these findings suggest that arachidonic acid metabolites may be involved in the AHR response following O₃ exposure in dogs. Studies probing the role of neutrophils in mediating the AHR response have provided inconsistent results (Al-Hegelan et al., 2011).

Evidence for cytokine and chemokine involvement in the AHR response to O₃ has been described. Some studies have suggested a role for TNF-α (mice, 0.5 and 2 ppm O₃, 3 hours) (Cho et al., 2001; Shore et al., 2001) and IL-1 (mice and ferrets, 2 ppm O₃, 3 hours) (Wu et al., 2008c; Park et al., 2004). The latter study found that SP expression in airway neurons was upregulated by IL-1 that was released in response to O₃. Other studies in mice have demonstrated a key role for CXCR2, the chemokine receptor for the neutrophil chemokines KC and MIP-2, but not for IL-6 in
O3-mediated AHR (1 ppm O3, 3 hours) (Johnston et al., 2005a; Johnston et al., 2005b). In contrast, CXCR2 and IL-6 were both required for neutrophil influx in this model (Johnston et al., 2005a; Johnston et al., 2005b), as discussed above. Williams et al. (2008b) demonstrated that the Th2 cytokine IL-13 contributed to AHR, as well as to airways neutrophilia, in mice (3 ppm O3, 3 hours).

Other studies have focused on the role of TLR4. Hollingsworth et al. (2004) measured AHR, as well as airways neutrophilia, in mice 6 and 24 hours following acute (2 ppm O3 for 3 hours) and subchronic (0.3 ppm for 3 days) exposure to O3. TLR4 is a key component of the innate immune system and is responsible for the immediate inflammatory response seen following challenge with endotoxin and other pathogen-associated substances. In this study, a functioning TLR4 was required for the full AHR response following O3 exposure but not for airways neutrophilia (Hollingsworth et al., 2004). These findings are complemented by an earlier study demonstrating that O3 effects on lung hyperpermeability required a functioning TLR4 (mice, 0.3 ppm O3, 72 hours) (Kleeberger et al., 2000). Williams et al. (2007b) found that TLR2, TLR4 and the TLR adaptor protein MyD88 contributed to AHR in mice (3 ppm O3, 3 hours). Ozone was also found to upregulate MyD88, TLR4 and TLR4 gene expression in this model (Williams et al., 2007b). Furthermore, a recent study reported O3-induced AHR that required TLR4 and MyD88 in mice exposed to 1 ppm O3 for 3 hours (Li et al., 2011).

A newly recognized mechanistic basis for O3-induced AHR is provided by studies focusing on the role of hyaluronan following O3 exposure in mice (Garantziotis et al., 2010; Garantziotis et al., 2009). Hyaluronan is an extracellular matrix component that is normally found in the ELF as a large polymer. Briefly, TLR4 and CD44 were found to mediate AHR in response to O3 and hyaluronan. Exposure to 2 ppm O3 for 3 hours resulted in enhanced AHR and elevated levels of soluble low molecular weight hyaluronan in the BALF 24-hours postexposure (Garantziotis et al., 2010; Garantziotis et al., 2009). Ozone may have caused the depolymerization of hyaluronan to soluble fragments that are known to be endogenous ligands of the CD44 receptor and TLR4 in the macrophage (Jiang et al., 2005). In the two recent studies, O3-induced AHR was attenuated in CD44 and TLR4-deficient mice (Garantziotis et al., 2010; Garantziotis et al., 2009). Hyaluronan fragment-mediated stimulation of AHR was found to require functioning CD44 receptor and TLR4 (Garantziotis et al., 2010; Garantziotis et al., 2009). In contrast, high-molecular-weight hyaluronan blocked AHR in response to O3 (Garantziotis et al., 2009). In another study high-molecular-weight hyaluronan enhanced repair of epithelial injury (Jiang et al., 2005). These studies provide a link between innate immunity and the development of AHR following O3 exposure, and indicate a role for TLR4 in increasing airways responsiveness. While TLR4-dependent responses usually involve activation of NF-κB and the upregulation of proinflammatory factors, the precise mechanisms leading to AHR are unknown (Al-Hegelan et al., 2011).

In guinea pigs, AHR was found to be mediated by different pathways at 1- and 3-days postexposure to a single exposure of O3 (2 ppm for 4 hours) (Verhein et al., 2011; Yost et al., 2005). At 1 day, AHR was due to activation of airway
parasympathetic nerves rather than to an exposure-related effect on smooth muscle (Yost et al., 2005). This effect occurred as a result of O₃-stimulated release of major basic protein from eosinophils (Yost et al., 2005). Major basic protein is known to block inhibitory M2 muscarinic receptors that normally dampen acetylcholine release from parasympathetic nerves (Yost et al., 2005). The resulting increase in acetylcholine release caused an increase in smooth muscle contraction following O₃ exposure (Yost et al., 2005). Eosinophils played a different role 3-days postexposure to O₃ in guinea pigs (Yost et al., 2005). Ozone-mediated influx of eosinophils into lung airways resulted in a different population of cells present 3-days postexposure compared to those present at 1 day (Yost et al., 2005). At this time point, eosinophil-derived major basic protein increased smooth muscle responsiveness to acetylcholine which also contributed to AHR (Yost et al., 2005). However, the major effect of eosinophils was to protect against vagal hyperreactivity (Yost et al., 2005). The authors suggested that these beneficial effects were due to the production of nerve growth factor (Yost et al., 2005). Further work by these investigators demonstrated a key role for IL-1β in mediating AHR 3-days postexposure to O₃ (Verhein et al., 2011). In this study, IL-1β increased nerve growth factor and SP that acted through the NK1 receptor to cause vagally-mediated bronchoconstriction (Verhein et al., 2011). The mechanism by which SP caused acetylcholine release from parasympathetic nerves following O₃ exposure was not determined (Verhein et al., 2011). Taken together, the above study results indicate that mechanisms involved in O₃-mediated AHR can vary over time postexposure and that eosinophils and SP can play a role. Results of this animal model may provide some insight into allergic airways disease in humans that is characterized by eosinophilia (Section 5.4.2.2).

5.3.6 Modification of Innate/Adaptive Immune System Responses

Host defense depends on effective barrier function and on innate immunity and adaptive immunity (Al-Hegelan et al., 2011). The effects of O₃ on barrier function in the airways were discussed above (Section 5.3.4). This section focuses on the mechanisms by which O₃ impacts innate and adaptive immunity. Both tissue damage and foreign pathogens are triggers for the activation of the innate immune system. This results in the influx of inflammatory cells such as neutrophils, mast cells, basophils, eosinophils, monocytes and dendritic cells and the generation of cytokines such as TNF-α, IL-1, IL-6, KC and IL-17. Further, innate immunity encompasses the actions of complement and collections, and the phagocytic functions of macrophages, neutrophils and dendritic cells. Airway epithelium also contributes to innate immune responses. Innate immunity is highly dependent on cell signaling networks involving TLR4. Adaptive immunity provides immunologic memory through the actions of B and T-cells. Important links between the two systems are provided by dendritic cells and antigen presentation. Recent studies demonstrate that exposure to O₃ modifies cells and processes which are required for innate immunity, contributes to innate-adaptive immune system interaction and primes pulmonary immune responses to endotoxin.
Ozone exposure of human subjects resulted in recruitment of activated innate immune cells to the airways. Healthy individuals were exposed to 80 ppb O₃ for 6.6 hours while exercising at a moderate level and airways inflammation was characterized in induced sputum 18-hours postexposure (Alexis et al., 2010). Previous studies demonstrated that induced sputum contains liquid and cellular constituents of the ELF from central conducting airways (Alexis et al., 2001b) and also identified these airways as a site of preferential O₃ absorption during exercise (Hu et al., 1994). Ozone exposure resulted in increased numbers of neutrophils, airway monocytes and dendritic-like cells in sputum (Alexis et al., 2010). In addition, increased expression of cell surface markers characteristic of innate immunity and antigen presentation (i.e., CD-14 and HLA-DR) was demonstrated on airway monocytes (Alexis et al., 2010). Enhanced antigen presentation contributes to exaggerated T-cell responses and promotes Th2 inflammation and an allergic phenotype (Lay et al., 2007). Upregulation of pro-inflammatory cytokines was also demonstrated in sputum of O₃-exposed subjects (Alexis et al., 2010). One of these cytokines, IL-12p70, correlated with numbers of dendritic-like cells in the sputum, and is an indicator of dendritic cell activation (Alexis et al., 2010). These authors have previously reported that exposure of human subjects exercising at a light to moderate level to 400 ppb O₃ for 2 hours resulted in activation of monocytes and macrophages (Lay et al., 2007), which could play a role in exacerbating existing asthma by activating allergen-specific memory T-cells. The current study confirms these findings and extends them by suggesting a potential mechanism whereby O₃-activated dendritic cells could stimulate naïve T-cells to promote the development of asthma (Alexis et al., 2010). A companion study by these same investigators (described in detail in Section 5.4.2.1) provides evidence of dendritic cell activation, measured as increased expression of HLA-DR, in a subset of the human subjects (GSTM1 null) exposed to 400 ppb O₃ for 2 hours while exercising at a light to moderate level (Alexis et al., 2009). Since dendritic cells are a link between innate and adaptive immunity, these studies provide evidence for an O₃-mediated interaction between the innate and adaptive immune systems.

Another recent study linked O₃-mediated activation of the innate immune system to the development of non-specific AHR in a mouse model (Pichavant et al., 2008). Repeated exposure to 1 ppm O₃ for 3 hours (3 days over a 5 day period) induced non-specific AHR measured 24 hours following the last exposure (Pichavant et al., 2008). This response was found to require NKT-cells, which are effector lymphocytes of innate immunity, as well as IL-17 and airways neutrophilia (Pichavant et al., 2008). Since glycolipids such as galactosyl ceramide are ligands for the invariant CD1 receptor on NKT-cells and serve as endogenous activators of NKT-cells, a role for O₃-oxidized lipids in activating NKT-cells was proposed (Pichavant et al., 2008). The authors contrasted this innate immunity pathway with that of allergen-provoked specific AHR which involves adaptive immunity, the cytokines IL-4, IL-13, IL-17, and airways eosinophilia (Pichavant et al., 2008). Interestingly, NKT-cells were required for both the specific AHR provoked by allergen and the non-specific AHR provoked by O₃ (Pichavant et al., 2008). Different cytokine profiles of the NKT-cells from allergen and O₃-exposed mice were proposed to account for the different pathways (Pichavant et al., 2008). More
recently, NKT-cells have been found to function in both innate and adaptive immunity (Vivier et al., 2011).

An interaction between allergen and O₃ in the induction of nonspecific AHR was shown in another animal study (Larsen et al., 2010). Mice were sensitized with the aerosolized allergen OVA on 10 consecutive days followed by exposure to O₃ (0.1-0.5 ppm for 3 hours) (Larsen et al., 2010). While allergen sensitization alone did not alter airways responsiveness to a nonspecific challenge, O₃ exposure of sensitized mice resulted in nonspecific AHR at 6- and 24-hours postexposure (Larsen et al., 2010). The effects of O₃ on AHR were independent of airways eosinophilia and neutrophilia (Larsen et al., 2010). However, OVA pretreatment led to goblet cell metaplasia which was enhanced by O₃ exposure (Larsen et al., 2010). It should be noted that OVA sensitization using only aerosolized antigen in this study is less common than the usual procedure for OVA sensitization achieved by one or more initial systemic injections of OVA and adjuvant followed by repeated inhalation exposure to OVA. This study also points to an interaction between innate and adaptive immune systems in the development of the AHR response.

Furthermore, O₃ was found to act as an adjuvant for allergic sensitization (Hollingsworth et al., 2010). Oropharyngeal aspiration of OVA on day 0 and day 6 failed to lead to allergic sensitization unless mice were first exposed to 1 ppm O₃ for 2 hours (Hollingsworth et al., 2010). The O₃-mediated response involved Th2 (IL-4, IL-5 and IL-9) and Th17 cytokines (IL-17) and was dependent on a functioning TLR4 (Hollingsworth et al., 2010). Ozone exposure also activated OVA-bearing dendritic cells in the thoracic lymph nodes, as measured by the presence of the CD86 surface marker, which suggests naïve T-cell stimulation and the involvement of Th2 pathways (Hollingsworth et al., 2010). Thus the adjuvant effects of O₃ may be due to activation of both innate and adaptive immunity.

Priming of the innate immune system by O₃ was reported by Hollingsworth et al. (2007). In this study, exposure of mice to 2 ppm O₃ for 3 hours led to nonspecific AHR at 24- and 48-hours postexposure, an effect which subsided by 72 hours (Hollingsworth et al., 2007). However, in mice treated with aerolized endotoxin immediately following O₃ exposure, AHR was greatly enhanced at 48-and 72-hours postexposure (Hollingsworth et al., 2007). In addition, O₃ pre-exposure was found to reduce the number of inflammatory cells in the BALF, to increase cytokine production and total protein in the BALF and to increase systemic IL-6 following exposure to endotoxin (Hollingsworth et al., 2007). Furthermore, O₃ stimulated the apoptosis of alveolar macrophages 24-hours postexposure, an effect which was greatly enhanced by endotoxin treatment. Apoptosis of circulating blood monocytes was also observed in response to the combined exposures (Hollingsworth et al., 2007). Ozone pre-exposure enhanced the response of lung macrophages to endotoxin (Hollingsworth et al., 2007). Taken together, these findings demonstrated that O₃ exposure increased innate immune responsiveness to endotoxin. The authors attributed these effects to the increased surface expression of TLR4 and increased signaling in macrophages observed in the study (Hollingsworth et al., 2007). It was proposed that the resulting decrease in airway inflammatory cells could account for
O$_3$-mediated decreased clearance of bacterial pathogens observed in numerous animal models (Hollingsworth et al., 2007).

More recently, these authors demonstrated that hyaluronan contributed to the O$_3$-primed response to endotoxin (Li et al., 2010). In this study, exposure of mice to 1 ppm O$_3$ for 3 hours resulted in enhanced responses to endotoxin, which was mimicked by intratracheal instillation of hyaluronan fragments (Li et al., 2010). Hyaluronan, like O$_3$, was also found to induce TLR4 receptor peripheralization in the macrophage membrane (Li et al., 2010; Hollingsworth et al., 2007), an effect which is associated with enhanced responses to endotoxin. This study and previous ones by the same investigators showed elevation of BALF hyaluronan in response to O$_3$ exposure (Garantziotis et al., 2010; Li et al., 2010; Garantziotis et al., 2009), providing evidence that the effects of O$_3$ on innate immunity are at least in part mediated by hyaluronan fragments. The authors note that excessive TLR4 signaling can lead to lung injury and suggest that O$_3$ may be responsible for an exaggerated innate immune response which may underlie lung injury and decreased host defense (Li et al., 2010).

Activation or upregulation of the immune system has not been reported in all studies. Impaired antigen-specific immunity was demonstrated following subacute O$_3$ exposure (0.6 ppm, 10 h/day for 15 days) in mice (Feng et al., 2006). Specifically, O$_3$ exposure altered the lymphocyte subset and cytokine profile and impacted thymocyte early development leading to immune dysfunction. Further, recent studies demonstrated SP-A oxidation in mice exposed for 3-6 hours to 2 ppm O$_3$. SP-A is an important innate immune protein which plays a number of roles in host defense including acting as opsonin for the recognition of some pathogens (Haque et al., 2009). These investigations found that O$_3$-mediated carbonylation of purified SP-A was associated with impaired macrophage phagocytosis in vitro (Mikerov et al., 2008c). In addition, O$_3$ exposure (2 ppm for 3 hours) in mice was found to increase susceptibility to pneumonia infection in mice through an impairment of SP-A dependent phagocytosis (Mikerov et al., 2008b; Mikerov et al., 2008a). Furthermore, early life exposure to O$_3$ in infant monkeys followed by a recovery period led to hyporesponsiveness to endotoxin (Maniar-Hew et al., 2011), as discussed below and in Section 5.4.2.4 and Section 7.2.3.2.

Taken together, results of recent studies provide evidence that O$_3$ alters host immunologic response and leads to immune system dysfunction through its effects on innate and adaptive immunity.

### 5.3.7 Airways Remodeling

The nasal airways, conducting airways and distal airways (i.e., respiratory bronchioles or CAR depending on the species) have all been identified as sites of O$_3$-mediated injury and inflammation (Mudway and Kelly, 2000). At all levels of the RT, loss of sensitive epithelial cells, degranulation of secretory cells, proliferation of resistant epithelial cells and neutrophilic influx have been observed as a result of O$_3$.
exposure (Mudway and Kelly, 2000; Cho et al., 1999). An important study (Plopper et al., 1998) conducted in adult rhesus monkeys (0.4 and 1.0 ppm O₃ for 2 hours at rest) found that 1 ppm O₃ resulted in the greatest epithelial injury in the respiratory bronchioles immediately postexposure although injury was observed at all of the RT sites studied except for the lung parenchyma. Exposure to 0.4 ppm O₃ resulted in epithelial injury only in the respiratory bronchioles. Initial cellular injury correlated with site-specific O₃ dose since the respiratory bronchioles received the greatest O₃ dose (¹⁸O mass/lung weight) and sustained the greatest initial cellular injury. The respiratory bronchioles were also the site of statistically significant GSH reduction. In addition, a study in isolated perfused rat lungs found greater injury in conducting airways downstream of bifurcations where local doses of O₃ were higher (Postlethwait et al., 2000).

In addition to the degree of initial injury, the degree of airways inflammation due to O₃ may have important long-term consequences since airways inflammation may lead to tissue injury (Balmes et al., 1996). Persistent inflammation and injury, observed in animal models of chronic and intermittent exposure to O₃, are associated with airways remodeling, including mucous cell metaplasia of nasal transitional epithelium (Harkema et al., 1999; Hotchkiss et al., 1991) and bronchiolar metaplasia of alveolar ducts (Mudway and Kelly, 2000). In a nonhuman primate model, hyperplasia of both URT and LRT epithelium resulted from chronic exposure to O₃ concentrations as low as 0.15 ppm and 0.3 ppm (Harkema et al., 1993, 1987b). Fibrotic changes such as deposition of collagen in the airways and sustained lung function decrements especially in small airways have also been demonstrated as a response to chronic O₃ exposure (Mudway and Kelly, 2000; Chang et al., 1992). These effects, described in detail in Section 7.2.3.2, have been demonstrated in rats exposed to levels of O₃ as low as 0.25 ppm. Mechanisms responsible for the resolution of inflammation and the repair of injury remain to be clarified and there is only a limited understanding of the biological processes underlying long-term morphological changes. However, a recent study in mice demonstrated a key role for the TGF-β signaling pathway in the deposition of collagen in the airways wall following chronic intermittent exposure to 0.5 ppm O₃ (Katre et al., 2011). Studies in infant monkeys have also documented effects of chronic intermittent exposure to 0.5 ppm O₃ on the developing lung and immune system. Extensive discussion of this topic is found in Section 5.4.2.4 (Lifestage) and in Section 7.2.3.2.

It should be noted that repeated exposure to O₃ results in attenuation of some O₃-induced responses, including those associated with the activation of neural reflexes (e.g., decrements in pulmonary function), as discussed in Section 5.3.2. However, numerous studies demonstrate that some markers of injury and inflammation remain increased during multi-day exposures to O₃. Mechanisms responsible for attenuation, or the lack thereof, are incompletely understood.
5.3.8 Systemic Inflammation and Oxidative/Nitrosative Stress

Extrapulmonary effects of O$_3$ have been noted for decades (U.S. EPA, 2006b). It has been proposed that lipid oxidation products resulting from reaction of O$_3$ with lipids in the ELF are responsible for systemic effects, however it is not known whether they gain access to the vascular space (Chuang et al., 2009). Alternatively, extrapulmonary release of diffusible mediators may initiate or propagate inflammatory responses in the vascular or systemic compartments (Cole and Freeman, 2009). A role for O$_3$ in modulating endothelin, a potent vasoconstrictor, has also been proposed. Studies in rats found that exposure to 0.4 and 0.8 ppm O$_3$ induced endothelin system genes in the lung and increased circulating levels of endothelin (Thomson et al., 2006; Thomson et al., 2005). Systemic oxidative stress may be suggested by studies in humans which reported associations between O$_3$ exposure and levels of plasma 8-isoprostanes and the presence of peripheral blood lymphocyte micronuclei (Chen et al., 2007a; Chen et al., 2006a). However, plasma isoprostanes are not a direct measure of systemic oxidative stress since they are stable and can be generated in any compartment before diffusion into the vascular space. Evidence of O$_3$-mediated systemic oxidative stress is better provided by new animal studies described below.

Ozone-induced perturbations of the cardiovascular system were recently investigated in young mice and monkeys (Chuang et al., 2009) and in rats (Kodavanti et al., 2011; Perepu et al., 2010) (see Section 6.3.3 and Section 7.3.1.2). These are the first studies to suggest that systemic oxidative stress and inflammation play a mechanistic role in O$_3$-induced effects on the systemic vasculature and heart. Exposure to 0.5 ppm O$_3$ for 5 days resulted in oxidative/nitrosative stress, vascular dysfunction and mitochondrial DNA damage in the aorta (Chuang et al., 2009). Chronic exposure to 0.8 ppm O$_3$ resulted in an enhancement of inflammation and lipid peroxidation in the heart following an ischemia-reperfusion challenge (Perepu et al., 2010). In addition, chronic intermittent exposure to 0.4 ppm O$_3$ increased aortic levels of mRNA for biomarkers of oxidative stress, thrombosis, vasoconstriction and proteolysis and aortic lectin-like oxidized-low density lipoprotein receptor-1 (LOX-1) mRNA and protein levels (Kodavanti et al., 2011). The latter study suggests a role for circulating oxidized lipids in mediating the effects of O$_3$.

Two recent controlled human exposure studies also demonstrated cardiovascular effects in response to short-term O$_3$ exposure (see Section 6.3.1). Changes in high frequency heart rate variability (HRV) were reported, albeit effects were in opposing directions (Devlin et al., 2012; Fakhri et al., 2009). Differences in study design may account for this discrepancy; the increase in high frequency HRV was observed following relatively low O$_3$ exposure (120 ppb for 2 hours during rest) and the decrease in high frequency HRV was observed following higher O$_3$ exposures (300 ppb for 2 hours while exercising at a high rate). These changes in cardiac function provide evidence of O$_3$-induced modulation of the autonomic nervous system, potentially through the activation of neural reflexes in the lung. Devlin et al. (2012) also demonstrated O$_3$-induced increases in markers of systemic inflammation and a pro-thrombogenic environment (Devlin et al., 2012). Older controlled human
exposure studies have reported increased myocardial work (Gong et al., 1998), and increased markers of oxidative stress (Liu et al., 1999; Liu et al., 1997) following a single O₃ exposure and reduced serum tocopherol levels following repeated O₃ exposures (Foster et al., 1996) (see Section 6.3.1). These findings in humans, together with findings from animal toxicological studies, provide evidence of O₃-mediated cardiovascular effects that may involve changes in autonomic tone, systemic and/or vascular oxidative stress and inflammation, and activation of the fibrinolytic system.

Systemic inflammation and oxidative/nitrosative stress may similarly affect other organ systems as well as the plasma compartment. Circulating cytokines have the potential to enter the brain through diffusion and active transport and to contribute to neuroinflammation, neurotoxicity, cerebrovascular damage and a break-down of the blood brain barrier (Block and Calderón-Garcidueñas, 2009) (see Section 6.4 and Section 7.5). They can also activate neuronal afferents (Block and Calderón-Garcidueñas, 2009). Vagal afferent pathways originating in the RT may also be responsible for O₃-mediated activation of nucleus tractus solitarius neurons which resulted in neuronal activation in stress-responsive regions of the CNS in rats (0.5 or 2 ppm O₃ for 1.5-120 hours) (Gackière et al., 2011). Recent studies have demonstrated O₃-induced brain lipid peroxidation, cytokine production in the brain and upregulated expression of VEGF in rats (0.5 ppm O₃, 3 hours or 0.25-0.5 ppm O₃, 4 h/day, 15-60 days) (Guevara-Guzmán et al., 2009; Araneda et al., 2008; Pereyra-Muñoz et al., 2006). Further, O₃-induced oxidative stress resulted in increased plasma lipid peroxides (0.25 ppm, 4h/day, 15-60 days) (Santiago-López et al., 2010), which was correlated with damage to specific brain regions (Pereyra-Muñoz et al., 2006).

Oxidative stress is one mechanism by which testicular and sperm function may be disrupted (see Section 7.4.1). Studies in Leydig cells in vitro have demonstrated that steroidogenesis is blocked by oxidative stress (Diemer et al., 2003). It has been proposed that lipid peroxidation of sperm plasma membrane may lead to impaired sperm mobility and decreased sperm quality (Agarwal et al., 2003). Further, it has been proposed that oxidative stress may damage DNA in the sperm nucleus and lead to apoptosis and a decline in sperm counts (Agarwal et al., 2003). One study reported an association between O₃ exposure and semen quality and suggested oxidative stress as an underlying mechanism (Sokol et al., 2006). Additional evidence is required to substantiate this link.

A role for plasma antioxidants in modulating O₃-induced respiratory effects was suggested by a recent study (Aibo et al., 2010). In this study, pretreatment of rats with a high dose of acetaminophen resulted in increased levels of plasma cytokines and the influx of inflammatory cells into the lung following O₃ exposure (0.25-0.5 ppm, 6 hours) (Aibo et al., 2010). These effects were not observed in response to O₃ alone. Furthermore, acetaminophen-induced liver injury was exacerbated by O₃ exposure. A greater increase in hepatic neutrophil accumulation and greater alteration in gene expression profiles was observed in mice exposed to O₃ and acetaminophen compared with either exposure alone (Aibo et al., 2010).
Although not measured in this study, glutathione depletion in the liver is known to occur in acetaminophen toxicity. Since liver glutathione is the source of plasma glutathione, acetaminophen treatment may have lowered plasma glutathione levels and altered the redox balance in the vascular compartment. These findings indicate interdependence between RT, plasma and liver responses to O₃, possibly related to glutathione status.

5.3.9 Impaired Alveolar-Arterial Oxygen Transfer

Ozone may impair alveolar-arterial oxygen transfer and reduce the supply of arterial oxygen to the myocardium. This may have a greater impact in individuals with compromised cardiopulmonary systems. Gong et al. (1998) provided evidence of a small decrease in arterial oxygen saturation in human subjects exposed for 3 hours to 300 ppb O₃ while exercising at a light to moderate level. In addition, Delaunois et al. (1998) demonstrated pulmonary vasoconstriction in O₃-exposed rabbits (0.4 ppm, 4 hours). Although of interest, the contribution of this pathway to O₃-induced cardiovascular effects remains uncertain.

5.3.10 Summary

This section summarizes the modes of action and toxicity pathways resulting from O₃ inhalation (Figure 5-8). These pathways provide a mechanistic basis for the health effects which are described in detail in Chapters 6 and 7. However the precise sequence by which the key events lead to health effects is not entirely clear. Three distinct short-term responses have been well-characterized in humans challenged with O₃: decreased pulmonary function, airways inflammation, and increased bronchial reactivity. In addition, O₃ exposure exacerbates, and possibly also causes, asthma and allergic airways disease in humans. Animal studies have demonstrated airways remodeling and fibrotic changes in response to chronic and intermittent O₃ exposures and a wide range of other responses. While the RT is the primary target tissue, cardiovascular and other organ effects occur following short- and long-term exposures of animals to O₃.

The initial key event in the toxicity pathway of O₃ is the formation of secondary oxidation products in the RT. This mainly involves direct reactions with components of the ELF. The resulting secondary oxidation products transmit signals to the epithelium, nociceptive sensory nerve fibers and, if present, dendritic cells, mast cells and eosinophils. Thus, the effects of O₃ are mediated by components of ELF and by the multiple cell types found in the RT. Further, oxidative stress is an implicit part of this initial key event.

Another key event in the toxicity pathway of O₃ is the activation of neural reflexes which lead to decrements in pulmonary function (see Section 6.2.1). Evidence is accumulating that secondary oxidation products are responsible for this effect.
Eicosanoids have been implicated in humans, while both eicosanoids and aldehydes are effective in animal models. Different receptors on bronchial C-fibers have been shown to mediate separate effects of O₃ on pulmonary function. Nociceptive sensory nerves are involved in the involuntary truncation of inspiration which results in decreases in FVC, FEV₁, tidal volume and pain upon deep inspiration. Opioids block these responses while atropine has only a minimal effect. New evidence in an animal model suggests that TRPA1 receptors on bronchial C-fibers mediate this pathway. Ozone exposure also results in activation of vagal sensory nerves and a mild increase in airway obstruction measured as increased sRaw. Atropine and β-adrenergic agonists greatly inhibit this response in humans indicating that the airways obstruction is due to bronchoconstriction. Other studies in humans implicated SP release from bronchial C-fibers resulting in airway narrowing due to either neurogenic edema or bronchoconstriction. New evidence in an animal model suggests that the SP-NK receptor pathway caused bronchoconstriction following O₃ exposure. Activation of neural reflexes also results in extrapulmonary effects such as bradycardia.

Figure 5-8  The modes of action/possible pathways underlying the health effects resulting from inhalation exposure to O₃.
Initiation of inflammation is also a key event in the toxicity pathway of O$_3$. Secondary oxidation products, as well as chemokines and cytokines elaborated by airway epithelial cells and macrophages, have been implicated in the initiation of inflammation. Vascular endothelial adhesion molecules may also play a role. Work from several laboratories using human subjects and animal models suggest that O$_3$ triggers the release of tachykinins such as SP from airway sensory nerves which could contribute to downstream effects including inflammation (see Section 6.2.3 and Section 7.2.4). Airways neutrophilia has been demonstrated in BALF, mucosal biopsy and induced sputum samples. Influx of mast cells, monocytes and macrophages also occur. Inflammation further contributes to O$_3$-mediated oxidative stress. Recent investigations show that O$_3$ exposure leads to the generation of hyaluronan fragments from high molecular weight polymers of hyaluronan normally found in the ELF in mice. Hyaluronan activates TLR4 and CD44-dependent signaling pathways in macrophages, and results in an increased number of macrophages in the BALF. Activation of these pathways occurs later than the acute neutrophilic response suggesting that they may contribute to longer-term effects of O$_3$. The mechanisms involved in clearing O$_3$-provoked inflammation remain to be clarified. It should be noted that inflammation, as measured by airways neutrophilia, is not correlated with decrements in pulmonary function as measured by spirometry.

A fourth key event in the toxicity pathway of O$_3$ is alteration of epithelial barrier function. Increased permeability occurs as a result of damage to tight junctions between epithelial cells subsequent to O$_3$-induced injury and inflammation. It may play a role in allergic sensitization and in AHR (see Section 6.2.2, Section 6.2.6, and Section 7.2.5). Tachykinins mediate this response while antioxidants may confer protection. Genetic susceptibility has been associated with functioning $\text{Tlr}4$ and $\text{Nos}2$ genes.

A fifth key event in the toxicity pathway of O$_3$ is the sensitization of bronchial smooth muscle. Increased bronchial reactivity can be both a rapidly occurring and a persistent response. The mechanisms responsible for early and later AHR are not well-understood (see Section 6.2.2). One proposed mechanism of sensitization, O$_3$-induced increases in epithelial permeability, would improve access of agonist to smooth muscle receptors. The evidence for this mechanism is not consistent. Another proposed mechanism, for which there is greater evidence, is neurally-mediated sensitization. In humans exposed to O$_3$, atropine blocked the early AHR response indicating the involvement of cholinergic postganglionic pathways. Animal studies demonstrated that O$_3$-induced AHR involved vagally-mediated responses and local axon reflex responses through bronchopulmonary C-fiber-mediated release of SP. Later phases of increased bronchial reactivity may involve the induction of IL-1$\beta$ which in turn upregulates SP production. In guinea pigs, eosinophil-derived major basic protein contributed to the stimulation of cholinergic postganglionic pathways. A novel role for hyaluronan in mediating the later phase effects O$_3$-induced AHR has recently been demonstrated. Hyaluronan fragments stimulated AHR in a TLR4- and CD44 receptor-dependent manner. Tachykinins and secondary oxidation products of O$_3$ have been proposed as mediators of the early response and inflammation-derived products have been proposed as mediators of the later response. Inhibition of
arachidonic acid metabolism was ineffective in blocking O₃-induced AHR in humans while in animal models mixed results were found. Other cytokines and chemokines have been implicated in the AHR response to O₃ in animal models.

A sixth key event in the toxicity pathway of O₃ is the modification of innate/adaptive immunity. While the majority of evidence for this key event comes from animal studies, there are several studies suggesting that this pathway may also be relevant in humans. Ozone exposure of human subjects resulted in recruitment of activated innate immune cells to the airways. This included macrophages and monocytes with increased expression of cell surface markers characteristic of innate immunity and antigen presentation, the latter of which could contribute to exaggerated T-cell responses and the promotion of an allergic phenotype. Evidence of dendritic cell activation was observed in GSTM1 null human subjects exposed to O₃, suggesting O₃-mediated interaction between the innate and adaptive immune systems. Animal studies further linked O₃-mediated activation of the innate immune system to the development of nonspecific AHR, demonstrated an interaction between allergen and O₃ in the induction of nonspecific AHR, and found that O₃ acted as an adjuvant for allergic sensitization through the activation of both innate and adaptive immunity. Priming of the innate immune system by O₃ was reported in mice. This resulted in an exaggerated response to endotoxin which included enhanced TLR4 signaling in macrophages. Ozone-mediated impairment of the function of SP-A, an innate immune protein, has also been demonstrated. Taken together these studies provide evidence that O₃ can alter host immunologic response and lead to immune system dysfunction. These mechanisms may underlie the exacerbation and induction of asthma (see Section 6.2.6 and Section 7.2.1), as well as decreases in host defense (see Section 6.2.5 and Section 7.2.6).

Another key event in the toxicity pathway of O₃ is airways remodeling. Persistent inflammation and injury, which are observed in animal models of chronic and intermittent exposure to O₃, are associated with morphologic changes such as mucous cell metaplasia of nasal epithelium, bronchiolar metaplasia of alveolar ducts and fibrotic changes in small airways (see Section 7.2.3). Mechanisms responsible for these responses are not well-understood. However a recent study in mice demonstrated a key role for the TGF-β signaling pathway in the deposition of collagen in the airway wall following chronic intermittent exposure to O₃. Chronic intermittent exposure to O₃ has also been shown to result in effects on the developing lung and immune system.

Systemic inflammation and vascular oxidative/nitrosative stress are also key events in the toxicity pathway of O₃. Extrapulmonary effects of O₃ occur in numerous organ systems, including the cardiovascular, central nervous, reproductive, and hepatic systems (see Sections 6.3 to 6.5 and Sections 7.3 to 7.5). It has been proposed that lipid oxidation products resulting from reaction of O₃ with lipids and/or cellular membranes in the ELF are responsible for systemic responses; however, it is not known whether they gain access to the vascular space. Alternatively, release of diffusible mediators from the lung into the circulation may initiate or propagate inflammatory responses in the vascular or in systemic compartments.

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5.4 Interindividual Variability in Response

Responses to O₃ exposure are variable within the population (Mudway and Kelly, 2000). Some studies have shown a large range of pulmonary function responses to O₃ among healthy young adults (i.e., 4 hours to 200 ppb O₃ or for 1.5 hours to 420 ppb O₃ while exercising at a moderate level) (Hazucha et al., 2003; Balmes et al., 1996). Since individual responses were relatively consistent across time in these studies, it was thought that responsiveness reflected an intrinsic characteristic of the subject (Mudway and Kelly, 2000). Other studies have shown that age and body mass index may influence responsiveness to O₃. In human subjects exercising moderately and exposed to 420 ppb O₃ for 1.5 hours, older adults were generally not responsive to O₃ (Hazucha et al., 2003), while obese young women appeared to be more responsive than lean young women (Bennett et al., 2007). In another study, the observed lack of spirometric responsiveness in one group of human subjects was not attributable to the presence of endogenous endorphins, which could vary between individuals and which could potentially block C-fiber stimulation by O₃ (420 ppb, 2 hours, moderate exercise (Passannante et al., 1998)). Other responses to O₃ have also been characterized by a large degree of interindividual variability. For example, interindividual variability in the neutrophilic response has been noted in human subjects (Holz et al., 1999; Devlin et al., 1991; Schelegle et al., 1991). One study demonstrated a 3-fold difference in airways neutrophilia, measured as percent of total cells in proximal BALF, among human subjects exposed to 300 ppb O₃ for 1 hour while exercising at a heavy level (Schelegle et al., 1991). Another study reported a 20-fold difference in BAL neutrophils following exposure to 80-100 ppb O₃ for 6.6 hours in human subjects exercising at a moderate level (Devlin et al., 1991). In contrast, reproducibility of intraindividual responses to 1-hour exposure to 250 ppb O₃ in human subjects exercising at a light level, measured as sputum neutrophilia, was demonstrated by Holz et al. (1999). While the basis for the observed interindividual variability in responsiveness to O₃ is not clear, both dosimetric and mechanistic factors are likely to contribute and will be discussed below.

5.4.1 Dosimetric Considerations

Two studies have investigated the correlation of O₃ uptake with the pulmonary function responses to O₃ exposure (Reeser et al., 2005; Gerrity et al., 1994). These studies found that the large subject-to-subject variability in %ΔFEV₁ response to O₃ does not appear to have a dosimetric explanation. Reeser et al. (2005) found no significant relationship between %ΔFEV₁ and fractional absorption of O₃ using the bolus method. Contrary to previous findings, the percent change in dead space volume of the respiratory tract (%ΔV_D) did not correlate with O₃ uptake, possibly due to the contraction of dead space caused by airway closure. Gerrity et al. (1994) found that intersubject variability in FEV₁ and airway resistance was not related to differences in the O₃ dose delivered to the lower airways, whereas minute ventilation
was predictive of FEV\textsubscript{1} decrement. No study has yet demonstrated that subjects show a consistent pattern of O\textsubscript{3} retention when re-exposed over weeks of time, as has been shown to be the case for the FEV\textsubscript{1} response, or that within-subject variation in FEV\textsubscript{1} response is related to fluctuations in O\textsubscript{3} uptake. However, these studies did not control for the differences in conducting airway volume between individuals. By controlling for conducting airway volume, it may be possible to estimate how much of the intersubject variation in FEV\textsubscript{1} response at a given O\textsubscript{3} exposure is due to actual inter-individual variability in dose.

5.4.2 Mechanistic Considerations

This section considers mechanistic factors that may contribute to variability in responses between individuals. It has been proposed that some of the variability may be genetically determined (Yang et al., 2005a). Besides gene-environment interactions, other factors such as pre-existing diseases and conditions, nutritional status, lifestage, attenuation, and co-exposures may also contribute to inter-individual variability in responses to O\textsubscript{3} and are discussed below.

5.4.2.1 Gene-environment Interactions

The pronounced interindividual variation in responses to O\textsubscript{3} infers that genetic background may play an important role in responsiveness to O\textsubscript{3} (Cho and Kleeberger, 2007; Kleeberger et al., 1997) (see also Section 8.4). Strains of mice which are prone or resistant to O\textsubscript{3}-induced effects have been used to systematically identify candidate susceptibility genes. Using these recombinant inbred strains of mice and exposures to 0.3 ppm O\textsubscript{3} for up to 72 hours, genome wide linkage analyses (also known as positional cloning) demonstrated quantitative trait loci for O\textsubscript{3}-induced lung inflammation and hyperpermeability on chromosome 17 (Kleeberger et al., 1997) and chromosome 4 (Kleeberger et al., 2000), respectively. More specifically, these studies found that Tnf\textsubscript{c}, whose protein product is the inflammatory cytokine TNF-\alpha, and Tlr4, whose protein product is TLR4, were candidate susceptibility genes (Kleeberger et al., 2000; Kleeberger et al., 1997). Other studies, which used targeted deletion, identified genes encoding iNOS and heat shock proteins as TLR4 effector genes (Bauer et al., 2011; Kleeberger et al., 2001) and found that IL-10 protects against O\textsubscript{3}-induced pulmonary inflammation (Backus et al., 2010). Investigations in inbred mouse strains found that differences in expression of certain proteins, such as CCSP (1.8 ppm O\textsubscript{3} for 3 hours) (Broeckaert et al., 2003) and MARCO (0.3 ppm O\textsubscript{3} for up to 48 hours) (Dahl et al., 2007), were responsible for phenotypic characteristics, such as epithelial permeability and scavenging of oxidized lipids, respectively, which confer sensitivity to O\textsubscript{3}.

Genetic polymorphisms have received increasing attention as modulators of O\textsubscript{3}-mediated effects. Functionally relevant polymorphisms in candidate susceptibility genes have been studied at the individual and population level in humans, and also in
animal models. Genes whose protein products are involved in antioxidant defense/oxidative stress and xenobiotic metabolism, such as glutathione-S-transferase M1 (GSTM1) and NADPH:quinone oxidoreductase 1 (NQO1), have also been a major focus of these efforts. This is because oxidative stress resulting from O₃ exposure is thought to contribute to the pathogenesis of asthma, and because xenobiotic metabolism detoxifies secondary oxidation products formed by O₃ which contribute to oxidative stress (Islam et al., 2008). TNF-α is of interest since it is linked to a candidate O₃ susceptibility gene and since it plays a key role in initiating airways inflammation (Li et al., 2006d). Polymorphisms of genes coding for GSTM1, NQO1 and TNF-α have been associated with altered risk of O₃-mediated effects (Li et al., 2006d; Yang et al., 2005a; Romieu et al., 2004b; Corradi et al., 2002; Bergamaschi et al., 2001). Additional studies have focused on functional variants in other genes involved in antioxidant defense such as catalase (CAT), myeloperoxidase, heme oxygenase (HMOX-1) and manganese superoxide dismutase (MnSOD) (Wenten et al., 2009; Islam et al., 2008). These studies are discussed below.

GSTM1 is a phase II antioxidant enzyme which is transcriptionally regulated by NF-e2-related factor 2-antioxidant response element (Nrf2-ARE) pathway. A large proportion (40-50%) of the general public (across ethnic populations) has the GSTM1-null genotype, which has been linked to an increased risk of health effects due to exposure to air pollutants (London, 2007). A role for GSTs in metabolizing electrophiles such as 4-hydroxynonenal, which is a secondary oxidation product resulting from O₃ exposure, has been demonstrated (Awasthi et al., 2004). A recent study found that the GSTM1 genotype modulated the time course of the neutrophilic inflammatory response following acute O₃ exposure (400 ppb for 2 hours with light to moderate exercise) in healthy adults (Alexis et al., 2009). In GSTM1-null and -sufficient subjects, O₃-induced sputum neutrophilia was similar at 4 hours. However, neutrophilia resolved by 24 hours in sufficient subjects but not in GSTM1-null subjects. In contrast, no differences in 24 hour sputum neutrophilia were observed between GSTM1-null and -sufficient human subjects exposed to 60 ppb O₃ for 6.6 hours with moderate exercise (Kim et al., 2011). It is not known whether the effect seen at the higher exposure level (Alexis et al., 2009) was due to the persistence of pro-inflammatory stimuli, impaired production of downregulators or impaired neutrophil apoptosis and clearance. However, a subsequent in vitro study by these same investigators found that GSTM1 deficiency in airway epithelial cells enhanced IL-8 production in response to 0.4 ppm O₃ for 4 hours (Wu et al., 2011). Furthermore, NF-κB activation was required for O₃-induced IL-8 production (Wu et al., 2011). Since IL-8 is a potent neutrophil activator and chemotaxin, this study provides additional evidence for the role of GSTM1 as a modulator of inflammatory responses due to O₃ exposure.

In addition, O₃ exposure increased the expression of the surface marker CD14 in airway neutrophils of GSTM1-null subjects to a greater extent than in sufficient subjects (Alexis et al., 2009). Furthermore, differences in airway macrophages were noted between the GSTM1-sufficient and -null subjects. Numbers of airway macrophages were decreased at 4 and 24 hours following O₃ exposure in
GSTM1-sufficient subjects (Alexis et al., 2009). Airway macrophages in GSTM1-null subjects were greater in number and found to have greater oxidative burst and phagocytic capability than those of sufficient subjects. Airway macrophages and dendritic cells from GSTM1-null subjects exposed to O3 expressed higher levels of the surface marker HLA-DR, suggesting activation of the innate immune system (Alexis et al., 2009). These differences in inflammatory responses between the GSTM1-null and -sufficient subjects may provide biological plausibility for the differences in O3-mediated effects reported in controlled human exposure studies (Corradi et al., 2002; Bergamaschi et al., 2001). It should also be noted that GSTM1 genotype did not affect the acute pulmonary function (i.e., spirometric) response to O3 which provides additional evidence for separate mechanisms underlying the effects of O3 on pulmonary function and inflammation in adults (Alexis et al., 2009). However, GSTM1-null asthmatic children were previously found to be more at risk of O3-induced effects on pulmonary function than GSTM1-sufficient asthmatic children (Romieu et al., 2004b).

Another enzyme involved in the metabolism of secondary oxidation products is NQO1. NQO1 catalyzes the 2-electron reduction by NADPH of quinones to hydroquinones. Depending on the substrate, it is capable of both protective detoxification reactions and redox cycling reactions resulting in the generation of reactive oxygen species. A recent study using NQO1-null mice demonstrated that NQO1 contributes to O3-induced oxidative stress, AHR and inflammation following a 3-hour exposure to 1 ppm O3 (Voynow et al., 2009). These experimental results may provide biological plausibility for the increased biomarkers of oxidative stress and increased pulmonary function decrements observed in O3-exposed individuals bearing both the wild-type NQO1 gene and the null GSTM1 gene (Corradi et al., 2002; Bergamaschi et al., 2001).

Besides enzymatic metabolism, other mechanisms participate in the removal of secondary oxidation products formed as a result of O3 inhalation. One involves scavenging of oxidized lipids via the macrophage receptor with collagenous structure (MARCO) expressed on the cell surface of alveolar macrophages. A recent study demonstrated increased gene expression of MARCO in the lungs of an O3-resistant C3H mouse strain (HeJ) but not in an O3-sensitive, genetically similar strain (OuJ) (Dahl et al., 2007). Upregulation of MARCO occurred in mice exposed to 0.3 ppm O3 for 24-48 hours; inhalation exposure for 6 hours at this concentration was insufficient for this response. Animals lacking the MARCO receptor exhibited greater inflammation and injury, as measured by BAL neutrophils, protein and isoprostanes, following exposure to 0.3 ppm O3 (Dahl et al., 2007). MARCO also protected against the inflammatory effects of oxidized surfactant lipids (Dahl et al., 2007). Scavenging of oxidized lipids may limit O3-induced injury since ozonized cholesterol species formed in the ELF (mice, 0.5-3 ppm O3, 3 hours) (Pulfer et al., 2005; Pulfer and Murphy, 2004) stimulated apoptosis and cytotoxicity in vitro (Gao et al., 2009b; Sathishkumar et al., 2009; Sathishkumar et al., 2007b; Sathishkumar et al., 2007a).
Two studies reported relationships between TNF promoter variants and O₃-induced effects in humans. In one study, O₃-induced change in lung function was significantly lower in adult subjects with TNF promoter variants -308A/A and -308G/A compared with adult subjects with the variant -308G/G (Yang et al., 2005a). This response was modulated by a specific polymorphism of LTA (Yang et al., 2005a), a previously identified candidate susceptibility gene whose protein product is lymphotoxin-α (Kleeberger et al., 1997). In the second study, an association between the TNF promoter variant -308G/G and decreased risk of asthma and lifetime wheezing in children was found (Li et al., 2006d). The protective effect on wheezing was modulated by ambient O₃ levels and by GSTM1 and GSTP1 polymorphisms. The authors suggested that the TNF-308 G/G genotype may have a protective role in the development of childhood asthma (Li et al., 2006d).

Similarly, a promoter variant of the gene HMOX-1, consisting of a smaller number of (GT)n repeats, was associated with a reduced risk for new-onset asthma in non-Hispanic white children (Islam et al., 2008). The number of (GT)n repeats in this promoter has been shown to be inversely related to the inducibility of HMOX-1. A modulatory effect of O₃ was demonstrated since the beneficial effects of this polymorphism were seen only in children living in low O₃ communities (Islam et al., 2008). This study also identified an association between a polymorphism of the CAT gene and increased risk of new-onset asthma in Hispanic children; however no modulation by O₃ was seen (Islam et al., 2008). No association was observed in this study between a MnSOD polymorphism and asthma (Islam et al., 2008).

Studies to date indicate that some variability in individual responsiveness to O₃ may be accounted for by functional genetic polymorphisms. Further, the effects of gene-environment interactions may be different in children and adults.

5.4.2.2 Pre-existing Diseases and Conditions

Pre-existing diseases and conditions can alter the response to O₃ exposure. For example, responsiveness to O₃, as measured by spirometry, is decreased in smokers and individuals with COPD (U.S. EPA, 2006b). Asthma and allergic diseases are of major importance in this discussion. In individuals with asthma, there is increased responsiveness to bronchoconstrictor challenge. This results from a combination of structural and physiological factors including increased airway inner-wall thickness, smooth muscle responsiveness and mucus secretion. Although inflammation is likely to contribute, its relationship to AHR is not clear (U.S. EPA, 2006b). However, some asthmatics have higher baseline levels of neutrophils, lymphocytes, eosinophils and mast cells in bronchial washes and bronchial biopsy tissue (Stenfors et al., 2002). It has been proposed that enhanced sensitivity to O₃ is conferred by the presence of greater numbers of resident airway inflammatory cells in disease states such as asthma (Mudway and Kelly, 2000).

In order to determine whether asthmatics exhibit greater responses to O₃, several earlier studies compared pulmonary function in asthmatic and non-asthmatic subjects
following O₃ exposure. Some also probed mechanisms which could account for enhanced sensitivity. While the majority focused on measurements of FEV₁ and FVC and found no differences between the two groups following exposures of 2-4 hours to 125-250 ppb O₃ or to a 30-minute exposure to 120-180 ppb O₃ by mouthpiece in human subjects exercising at a light to moderate level (Stenfors et al., 2002; Mudway et al., 2001; Holz et al., 1999; Scannell et al., 1996; Koenig et al., 1987; Linn et al., 1978), there were notable exceptions. In one study, greater airways obstruction in asthmatics compared with non-asthmatic subjects was observed immediately following a 2-hour exposure to 400 ppb O₃ while exercising at a heavy level (Kreit et al., 1989). These changes were measured as statistically significant greater decreases in FEV₁ and in FEF₂₅₋₇₅ (but not in FVC) in the absence of a bronchoconstrictor challenge (Kreit et al., 1989). These results suggest that this group of asthmatics responded to O₃-exposure with a greater degree of vagally-mediated bronchoconstriction compared with the non-asthmatics. A second study demonstrated a statistically significant greater decrease in FEV₁ and in FEV₁/FVC (but not in FVC) in asthmatics compared with non-asthmatics exposed to 160 ppb O₃ for 7.6 hours with light exercise (Horstman et al., 1995). These responses were accompanied by wheezing and inhaler use in the asthmatics (Horstman et al., 1995). Aerosol bolus dispersion measurements demonstrated a statistically significant greater change in asthmatics compared with non-asthmatics, which was suggestive of O₃-induced small airway dysfunction (Horstman et al., 1995). Furthermore, a statistically significant correlation was observed between the degree of baseline airway status and the FEV₁ response to O₃ in the asthmatic subjects (Horstman et al., 1995). A third study found similar decreases in FVC and FEV₁ in both asthmatics and non-asthmatics exposed to 400 ppb O₃ for 2 hours with light exercise (Alexis et al., 2000). However, a statistically significant decrease in FEF₇₅, a measure of small airway function, was observed in asthmatics but not in non-asthmatics (Alexis et al., 2000). Taken together, these latter studies indicate that while the magnitude of restrictive type spirometric decline was similar in asthmatics and non-asthmatics, that obstructive type changes (i.e., bronchoconstriction) were greater in asthmatics. Further, asthmatics exhibited greater sensitivity to O₃ in terms of small airways function.

Since asthma exacerbations occur in response to allergens and/or other triggers, some studies have focused on O₃-induced changes in AHR following a bronchoconstrictor challenge. No difference in sensitivity to methacholine bronchoprovocation was observed between asthmatics and non-asthmatics exposed to 400 ppb O₃ for 2 hours while exercising at a heavy level (Kreit et al., 1989). However, increased bronchial reactivity to inhaled allergens was demonstrated in mild allergic asthmatics exposed to 160 ppb for 7.6 hours, 250 ppb for 3 hours and 120 ppb for 1 hour while exercising at a light level or at rest (Kehrl et al., 1999; Jorres et al., 1996; Molfino et al., 1991) and in allergen-sensitized guinea pigs following O₃ exposure (1 ppm, 1 hour) (Sun et al., 1997). Similar, but modest, responses were reported for individuals with allergic rhinitis (Jorres et al., 1996). Further, the contractile response of isolated airways from human donor lung tissue, which were sensitized and challenged with allergen, was increased by pre-exposure to 1 ppm O₃ for 20 (Roux et
These studies provide support for O3-mediated enhancement of responses to allergens in allergic subjects.

In terms of airways neutrophilia, larger responses were observed in asthmatics compared to non-asthmatics subjects, who were exercising at a light to moderate level and exposed to O3, in some (Balmes et al., 1997; Scannell et al., 1996; Basha et al., 1994) but not all (Mudway et al., 2001) of the earlier studies. While each of these studies involved exposure of exercising human subjects to 200 ppb O3, the duration of exposure was longer (i.e., 4-6 hours) in the former studies than in the latter study (2 hours). Further, statistically significantly increases in myeloperoxidase levels (an indicator of neutrophil activation) in bronchial washes was observed in mild asthmatics compared with non-asthmatics, despite no difference in O3-stimulated neutrophil influx between the 2 groups following exposure to 200 ppb O3 for 2 hours with moderate exercise (Stenfors et al., 2002). A more recent study found that atopic asthmatic subjects exhibited an enhanced inflammatory response to O3 (400 ppb, 4 hours, with light to moderate exercise) (Hernandez et al., 2010). This response was characterized by greater numbers of neutrophils, higher levels of IL-6, IL-8 and IL-1β and greater macrophage cell-surface expression of TLR4 and IgE receptors in induced sputum compared with healthy subjects. This study also reported a greater increase in hyaluronan in atopic subjects and atopic asthmatics compared with healthy subjects following O3 exposure. Animal studies have previously reported that hyaluronic acid activates TLR4 signaling and results in AHR (see Section 5.3.5).

Furthermore, levels of IL-10, a potent anti-inflammatory cytokine, were greatly reduced in atopic asthmatics compared to healthy subjects. These results provide evidence that innate immune and adaptive responses are different in asthmatics and healthy subjects exposed to O3.

Eosinophils may be an important modulator of responses to O3 in asthma and allergic airways disease. Eosinophils and associated proteins are thought to affect muscarinic cholinergic receptors which are involved in vagally-mediated bronchoconstriction (Mudway and Kelly, 2000). Studies described in Section 5.3.5 which demonstrated a key role of eosinophils in O3-mediated AHR may be relevant to human allergic airways disease which is characterized by airways eosinophilia (Yost et al., 2005). Furthermore, O3 exposure sometimes results in airways eosinophilia in allergic subjects or animal models. For example, eosinophilia of the nasal and other airways was observed in individuals with pre-existing allergic disease following O3 inhalation (160 ppb, 7.6 hours with light exercise and 270 ppb, 2 hours with moderate exercise) (Vagaggini et al., 2002; Peden et al., 1997). Further, O3 exposure (0.5 ppm, 8 hours/day for 1-3 days) increased allergic responses, such as eosinophilia and augmented intraepithelial mucosubstances, in the nasal airways of ovalbumin (OVA)-sensitized rats (Wagner et al., 2002). In contrast, Stenfors et al. (2002) found no stimulation of eosinophil influx measured in bronchial washes and BALF of mild asthmatics following exposure to a lower concentration (200 ppb, 2 hours, with moderate exercise) of O3.

The role of mast cells in O3-mediated asthma exacerbations has been investigated. Mast cells are thought to play a key role in O3-induced airways inflammation, since
airways neutrophilia was decreased in mast cell-deficient mice exposed to O₃ (Kleeberger et al., 1993). However, another study found that mast cells were not involved in the development of increased bronchial reactivity in O₃-exposed mice (Noviski et al., 1999). Nonetheless, mast cells release a wide variety of important inflammatory mediators which may lead to asthma exacerbations (Stenfors et al., 2002). A large increase in mast cell number in bronchial submucosa was observed in non-asthmatics and a significant decrease in mast cell number in bronchial epithelium was observed in mild asthmatics 6 hours following exposure to 200 ppb O₃ for 2 hours during mild exercise (Stenfors et al., 2002). While these results point to an O₃-mediated flux in bronchial mast cell populations which differed between the non-asthmatics and mild asthmatics, interpretation of these findings is difficult. Furthermore, mast cell number did not change in airway lavages in either group in response to O₃ (Stenfors et al., 2002).

Cytokine profiles in the airways have been investigated as an indicator of O₃ sensitivity. Differences in epithelial cytokine expression were observed in bronchial biopsy samples in non-asthmatic and asthmatic subjects both at baseline and 6-hours postexposure to 200 ppb O₃ for 2 hours with moderate exercise (Bosson et al., 2003). The asthmatic subjects had a higher baseline expression of IL-4 and IL-5 compared to non-asthmatics. In addition, expression of IL-5, IL-8, GM-CSF, and ENA-78 in asthmatics was increased significantly following O₃ exposure compared to non-asthmatics (Bosson et al., 2003). Some of these (IL-4, IL-5 and GM-CSF) are Th2-related cytokines or neutrophil chemoattractants, and play a role in IgE production, airways eosinophilia and suppression of Th1-cytokine production (Bosson et al., 2003). These findings suggest a link between adaptive immunity and enhanced responses of asthmatics to O₃.

A further consideration is the compromised status of ELF antioxidants in disease states such as asthma (Mudway and Kelly, 2000). This could possibly be due to ongoing inflammation which causes antioxidant depletion or to abnormal antioxidant transport or synthesis (Mudway and Kelly, 2000). For example, basal levels of AH₂ were significantly lower and basal levels of oxidized GSH and UA were significantly higher in bronchial wash fluid and BALF of mild asthmatics compared with healthy control subjects (Mudway et al., 2001). Differences in ELF antioxidant content have also been noted between species. These observations have led to the suggestion that the amount and composition of ELF antioxidants, the capacity to replenish antioxidants in the ELF or the balance between beneficial and injurious interactions between antioxidants and O₃ may contribute to O₃ sensitivity, which varies between individuals and species (Mudway et al., 2006; Mudway and Kelly, 2000; Mudway et al., 1999a). The complexity of these interactions was demonstrated by a study in which a 2-hour exposure to 200 ppb O₃, while exercising at a moderate level, resulted in similar increases in airway neutrophils and decreases in pulmonary function in both mild asthmatics and healthy controls, despite differences in ELF antioxidant concentrations prior to O₃ exposure (Mudway et al., 2001). Further, the O₃-induced increase in oxidized GSH and decrease in AH₂ observed in ELF of healthy controls was not observed in mild asthmatics (Mudway et al., 2001). While the authors concluded that basal AH₂ and oxidized GSH concentrations were not
predictive of responsiveness to O₃, they also suggested that the increased basal UA concentrations in the mild asthmatics may have played a protective role (Mudway et al., 2001). Thus compensatory mechanisms resulting in enhanced total antioxidant capacity may play a role in modulating responses to O₃.

Collectively these older and more recent studies provide insight into mechanisms which may contribute to enhanced responses of asthmatic and atopic individuals following O₃ exposure. Greater airways inflammation and/or greater bronchial reactivity have been demonstrated in asthmatics compared to non-asthmatics. This pre-existing inflammation and altered baseline bronchial reactivity may contribute to the enhanced bronchoconstriction seen in asthmatics exposed to O₃. Furthermore, O₃-induced inflammation may contribute to O₃-mediated AHR. An enhanced neutrophilic response to O₃ has been demonstrated in some asthmatics. A recent study in humans provided evidence for differences in innate immune responses related to TLR4 signaling between asthmatics and healthy subjects. Animal studies have demonstrated a role for eosinophil-derived proteins in mediating the effects of O₃. Since airways eosinophilia occurs in both allergic humans and allergic animal models, this pathway may underlie the exacerbation of allergic asthma by O₃.

In addition, differences have been noted in epithelial cytokine expression in bronchial biopsy samples of healthy and asthmatic subjects. A Th2 phenotype, indicative of adaptive immune system activation and enhanced allergic responses, was observed before O₃ exposure and was increased by O₃ exposure in asthmatics. These findings support links between innate and adaptive immunity and sensitivity to O₃-mediated effects in asthmatics and allergic airways disease.

In addition to asthma and allergic diseases, obesity may alter responses to O₃. While O₃ is a trigger for asthma, obesity is a known risk factor for asthma (Shore, 2007). The relationship between obesity and asthma is not well understood but recent investigations have focused on alterations in endocrine function of adipose tissue in obesity. It is thought that the increases in serum levels of factors produced by adipocytes (i.e., adipokines), such as cytokines, chemokines, soluble cytokine receptors and energy regulating hormones, may contribute to the relationship between obesity and asthma. Some of these same mechanisms may be relevant to insulin resistant states such as metabolic syndrome.

In a reanalysis of the data of Hazucha et al. (2003), increasing body mass index in young women was associated with increased O₃ responsiveness, as measured by spirometry following a 1.5-hour exposure to 420 ppb O₃ while exercising at a moderate level (Bennett et al., 2007). In several mouse models of obesity, airways were found to be innately more hyperresponsive and responded more vigorously to acute O₃ exposure than lean controls (Shore, 2007). Pulmonary inflammatory and injury in response to O₃ were also enhanced (Shore, 2007). It was postulated that oxidative stress resulting from obesity-related hyperglycemia could account for these effects (Shore, 2007). However, responses to O₃ in the different mouse models are somewhat variable and depend on whether exposures are acute or subacute. For example, diet-induced obesity augmented inflammation and injury, as measured by BALF markers, and enhanced AHR in mice exposed acutely to O₃ (2 ppm, 3 hours).
In contrast, the inflammatory response following sub-acute exposure to O$_3$ was dampened by obesity in a different mouse model (0.3 ppm, 72 hours) (Shore et al., 2009). It is not known whether differences in responsiveness to O$_3$ are due to differences in lung development in genetically-modified animals which result in smaller lungs and thus to differences in inhaled dose because of the altered body mass to lung size ratio.

### 5.4.2.3 Nutritional Status

A further consideration is the compromised status of ELF antioxidants in nutritional deficiencies. Thus, many investigations have focused on antioxidant deficiency and supplementation as modulators of O$_3$-mediated effects. One study in mice found that vitamin A deficiency enhanced lung injury induced by exposure to 0.3 ppm O$_3$ for 72 hours (Paquette et al., 1996). Ascorbate deficiency was shown to increase the effects of acute (0.5-1 ppm for 4 hours), but not subacute (0.2-0.8 ppm for 7 days), O$_3$ exposure in guinea pigs (Kodavanti et al., 1995; Slade et al., 1989).

Supplementation with AH$_2$ and α-TOH was protective in healthy adults who were on an AH$_2$-deficient diet and exposed to 400 ppb O$_3$ for 2 hours while exercising at a moderate level (Samet et al., 2001). In this study, the protective effect consisted of a smaller reduction in FEV$_1$ following O$_3$ exposure (Samet et al., 2001). However the inflammatory response (influx of neutrophils and levels of IL-6) measured in BALF 1 hour after O$_3$ exposure was not different between supplemented and non-supplemented subjects (Samet et al., 2001). Other investigators found that AH$_2$ and α-TOH supplementation failed to ameliorate the pulmonary function decrements or airways neutrophilia observed in humans exposed to 200 ppb O$_3$ for 2 hours while exercising at a moderate level (Mudway et al., 2006). It was suggested that supplementation may be ineffective in the absence of antioxidant deficiency (Mudway et al., 2006).

In asthmatic adults, these same dietary antioxidants reduced O$_3$-induced bronchial hyperresponsiveness (120 ppb, 45 min, light exercise) (Trenga et al., 2001). Furthermore, supplementation with AH$_2$ and α-tocopherol protected against pulmonary function decrements and nasal inflammatory responses which were associated with high levels of ambient O$_3$ in asthmatic children living in Mexico City, Mexico (Sienra-Monge et al., 2004; Romieu et al., 2002). Similarly, supplementation with ascorbate, α-tocopherol and β-carotene improved pulmonary function in Mexico City street workers (Romieu et al., 1998b).

Protective effects of supplementation with α-tocopherol alone have not been observed in humans experimentally exposed to O$_3$ (Mudway and Kelly, 2000). Alpha-TOH supplementation also failed to protect against O$_3$-induced effects in animal models of allergic rhinosinusitis and lower airways allergic inflammation (rats, 1 ppm O$_3$ for 2 days) (Wagner et al., 2007). However, protection in these same animal models was reported using γ-TOH supplementation (Wagner et al., 2009; Wagner et al., 2007). Whether or not this effect was due to enhanced antioxidant
status or to activated signaling pathways is not known. Other investigators found that α-TOH deficiency led to an increase in liver lipid peroxidation following O₃ exposure (rats, 0.3 ppm 3 hours/day for 7 months) (Sato et al., 1980) and a drop in liver α-TOH levels following O₃ exposure (mice, 0.5 ppm, 6 hours/day for 3 days) (Vasu et al., 2010). A recent study used α-TOH transfer protein null mice as a model of α-TOH deficiency and demonstrated an altered adaptive response of the lung genome to O₃ exposure (Vasu et al., 2010). Taken together, these studies provide evidence that the tocopherol system modulates O₃-induced responses.

5.4.2.4 Lifestage

Responses to O₃ are modulated by factors associated with lifestage. On one end of the lifestage spectrum is aging. The spirometric response to O₃ appears to be lost in humans as they age, as was demonstrated in two studies involving exposures of human subjects exercising at levels ranging from light to heavy to 420-450 ppb O₃ for 1.5-2 hours (Hazucha et al., 2003; Drechsler-Parks, 1995). In mice, physiological responses to O₃ (600 ppb, 2 hours) were diminished with age (Hamade et al., 2010). Mechanisms accounting for this effect have not been well-studied but could include altered number and sensitivity of receptors, altered signaling pathways involved in neural reflexes or compromised status of ELF antioxidants.

On the other side of the lifestage spectrum is pre/postnatal development. Critical windows of development during the pre/postnatal period are associated with an enhanced sensitivity to environmental toxicants. Adverse birth outcomes and developmental disorders may occur as a result (Section 7.4).

Adverse birth outcomes may result from stressors which impact transplacental oxygen and nutrient transport by a variety of mechanisms including oxidative stress, placental inflammation and placental vascular dysfunction (Kannan et al., 2006). These mechanisms may be linked since oxidative/nitrosative stress is reported to cause vascular dysfunction in the placenta (Myatt et al., 2000). As described earlier in this chapter and in Section 7.4, systemic inflammation and oxidative/nitrosative stress and modification of innate and adaptive immunity are key events underlying the health effects of O₃ and as such they may contribute to adverse birth outcomes. An animal toxicology study showing that exposure to 2 ppm O₃ led to anorexia (Kavlock et al., 1979) (see Section 7.4.2) in exposed rat dams provide an additional mechanism by which O₃ exposure could lead to diminished transplacental nutrient transport. Disturbances of the pituitary-adrenocortico-placental system (Ritz et al., 2000) may also impact normal intrauterine growth and development. Further, restricted fetal growth may result from pro-inflammatory cytokines which limit trophoblast invasion during the early stages of pregnancy (Hansen et al., 2008). Direct effects on maternal health, such as risk of infection, and on fetal health, such as DNA damage, have also been proposed as mechanisms underlying adverse birth outcomes (Ritz et al., 2000). In addition to restricted fetal growth, preterm birth may contribute to adverse birth outcomes. Preterm birth may result from the development
of premature contractions and/or premature rupture of membranes as well as from disrupted implantation and placentation which results in suboptimal placental function (Darrow et al., 2009; Ritz et al., 2000). Genetic mutations are thought to be an important cause of placental abnormalities in the first trimester, while vascular alterations may be the main cause of placental abnormalities in later trimesters (Jalaludin et al., 2007). Ozone-mediated systemic inflammation and oxidative stress/nitrosative stress may possibly be related to these effects although there is no firm evidence.

Enhanced sensitivity to environmental toxicants during critical windows of development may also result in developmental disorders. For example, normal migration and differentiation of neural crest cells are important for heart development and are particularly sensitive to toxic insults (Ritz et al., 2002). Further, immune dysregulation and related pathologies are known to be associated with pre/postnatal environmental exposures (Dietert et al., 2010). Ozone exposure is associated with developmental effects in several organ systems. These include the lung and immune system (see below) and neurobehavioral changes which could reflect the effect of O₃ on CNS plasticity or the hypothalamic-pituitary axis (Auten and Foster, 2011) (see Section 7.4.9).

The majority of developmental effects due to O₃ have been described for the respiratory system (see Sections 7.2.3 and 7.4.8). Since its growth and development take place during both the prenatal and early postnatal periods, both prenatal and postnatal exposures to O₃ have been studied. Maternal exposure to 0.4-1.2 ppm O₃ during gestation resulted in developmental health effects in the RT of mice (Sharkhuu et al., 2011). Recent studies involving postnatal exposure to O₃ have focused on differences between developing and adult animals in antioxidant defenses, respiratory physiology and sensitivity to cellular injury, and on mechanisms, such as lung structural changes, antigen sensitization, interaction with nitric oxide signaling, altered airway afferent innervation and loss of alveolar repair capacity, by which early O₃ exposure could lead to asthma pathogenesis or exacerbations in later life (Auten and Foster, 2011).

An interesting set of studies conducted over the last 10 years in the infant rhesus monkey has identified numerous O₃-mediated perturbations in the developing lung and immune system (Plopper et al., 2007). These investigations were prompted by the dramatic rise in the incidence of childhood asthma and focused on the possible interaction of O₃ and allergens in promoting remodeling of the epithelial-mesenchymal trophic unit during postnatal development of the tracheobronchial airway wall. In humans, airways growth during the 8-12 year period of postnatal development is not well understood. Rhesus monkeys were used in these studies because the branching pattern and distribution of airways in this model are more similar to humans than those of rodents are to humans. In addition, a model of allergic airways disease, which exhibits the main features of human asthma, had already been established in the adult rhesus monkey. Studies in infant monkeys were designed to determine whether repeated exposure to O₃ altered postnatal growth and development, and if so, whether such effects were reversible. In addition, exposure to
O₃ was evaluated for its potential to increase the development of allergic airways disease. Exposures were to cyclic episodic O₃ over 5 months, which involved 5 biweekly cycles of alternating filtered air and O₃ – 9 consecutive days of filtered air and 5 consecutive days of 0.5 ppm O₃, 8 h/day – and to house dust mite allergen (HDMA) for 2 hours per day for 3 days on the last 3 days of O₃ exposure followed by 11 days of filtered air.

Key findings were numerous. First, baseline airway resistance and AHR in the infant monkeys were dramatically increased by combined exposure to both HDMA and O₃ (Joad et al., 2006; Schelegle et al., 2003). Secondly, O₃ exposure led to a large increase in BAL eosinophils (Schelegle et al., 2003) while HDMA exposure led to a large increase of eosinophils in airways tissue (Joad et al., 2006; Schelegle et al., 2003). Thirdly, the growth pattern of distal airways was changed to a large extent by exposure to O₃ alone and in combination with HDMA. More specifically, longer and narrower airways resulted and the number of conducting airway generations between the trachea and the gas exchange area was decreased (Fanucchi et al., 2006). This latter effect was not ameliorated by a recovery period of 6 months. Fourthly, exposure to both HDMA and O₃ altered the abundance and distribution of CD25+ lymphocytes in the airways (Miller et al., 2009). Lastly, several effects were seen at the level of the epithelial mesenchymal trophic unit in response to O₃. These include altered organization of the airways epithelium (Schelegle et al., 2003), increased abundance of mucous goblet cells (Schelegle et al., 2003), disruption of the basement membrane zone (Evans et al., 2003), reduced innervation (Larson et al., 2004), increased neuroendocrine-like cells (Joad et al., 2006), and altered orientation and abundance of smooth muscle bundles (Plopper et al., 2007; Tran et al., 2004). Six months of recovery in filtered air led to reversal of some but not all of these effects (Kajekar et al., 2007; Plopper et al., 2007; Evans et al., 2004). The authors concluded that cyclic challenge of infant rhesus monkeys to allergen and O₃ during the postnatal period compromised airway growth and development and resulted in changes which favor allergic airways responses and persistent effects on the immune system (Plopper et al., 2007). A more recent study in this same infant rhesus monkey model reported that early life exposure to O₃ resulted in decreased total peripheral blood leukocyte numbers and increased blood eosinophils as well as persistent effects on pulmonary and systemic innate immunity (Maniar-Hew et al., 2011).

Furthermore, the effect of cyclic episodic O₃ exposure on nasal airways was studied in the infant rhesus monkey model. The three-dimensional detail of the nasal passages was analyzed for developing predictive dosimetry models and exposure-dose-response relationships (Carey et al., 2007). The authors reported that the relative amounts of the five epithelial cell types in the nasal airways of monkeys remained consistent between infancy and adulthood (comparing to (Gross et al., 1987; Gross et al., 1982)). Cyclic episodic O₃ exposure (as described in the previous paragraphs) resulted in 50-80% decreases in epithelial thickness and epithelial cell volume of the ciliated respiratory and transitional epithelium, confirming that these cell types in the nasal cavity were the most sensitive to O₃ exposure. The character and location of nasal lesions resulting from O₃ exposure were similar in the infant monkeys and adult monkeys similarly exposed. However, the nasal epithelium of
infant monkeys did not undergo nasal airway epithelial remodeling or adaptation which occurs in adult animals following O₃-mediated injury and which may protect against subsequent O₃ challenge. The authors suggested that infant monkeys may be prone to developing persistent necrotizing rhinitis following episodic longer-term exposures.

### 5.4.2.5 Attenuation of Responses

Repeated daily exposure to O₃ often results in a reduction in the degree of a response, i.e., an attenuation of response. This phenomenon may reflect compensatory mechanisms and is not necessarily beneficial. Furthermore, there is variability among the different O₃-related endpoints in terms of response attenuation, as will be described below. As a result, attenuation of some responses occurs concomitantly with the exacerbation of others.

In responsive individuals, a striking degree of attenuation of the FEV₁ response occurred following repeated daily exposures to O₃. Generally, the young O₃ responder was no longer responsive on the fourth or fifth day of consecutive daily O₃ exposure (200-500 ppb O₃ for 2-4 hours with light to heavy levels of exercise) and required days to weeks of nonexposure in order for the subject to regain O₃ responsiveness (Christian et al., 1998; Devlin et al., 1997; Linn et al., 1982b; Horvath et al., 1981; Hackney et al., 1977b). This phenomena has been reported for both lung function and symptoms such as upper airway irritation, nonproductive cough, substernal discomfort and pain upon deep inspiration (Linn et al., 1982b; Horvath et al., 1981; Hackney et al., 1977b). Repeated daily exposures also led to an attenuation of the sRaw response in moderately exercising human subjects exposed for 4 hours to 200 ppb O₃ (Christian et al., 1998) and to a dampened AHR response compared with a single day exposure in light exercising human subjects exposed for 2 hours to 400 ppb O₃ (Dimeo et al., 1981). However, one group reported persistent small airway dysfunction despite attenuation of the FEV₁ response on the third day of consecutive O₃ exposure (250 ppb, 2 hours, with moderate exercise) (Frank et al., 2001).

Studies in rodents also indicated an attenuation of the physiologic response measured by breathing patterns and tidal volume following five consecutive days of exposure to 0.35-1 ppm O₃ for 2.25 hours (Tepper et al., 1989). Attenuation of O₃-induced bradycardic responses, which also result from activation of neural reflexes, has been reported in rodents (0.5-0.6 ppm O₃, 2-6 h/day, 3-5 days (Hamade and Tankersley, 2009; Watkinson et al., 2001).

Multi-day exposure to O₃ has been found to decrease some markers of inflammation compared with a single day exposure (Christian et al., 1998; Devlin et al., 1997). For example, in human subjects exposed for 4 hours to 200 ppb O₃ during moderate exercise, decreased numbers of BAL neutrophils and decreased levels of BALF fibronectin and IL-6 were observed after 4 days of consecutive exposure compared with responses after 1 day (Christian et al., 1998). Results indicated an attenuation of
the inflammatory response in both proximal airways and distal lung. However, markers of injury, such as lactate dehydrogenase (LDH) and protein in the BALF, were not attenuated in this study (Christian et al., 1998). Other investigators found that repeated O₃ exposure (200 ppb O₃ for 4 hours on 4 consecutive days with light exercise) resulted in increased numbers of neutrophils in bronchial mucosal biopsies despite decreased BAL neutrophilia (Jorres et al., 2000). Other markers of inflammation, including BALF protein and IL-6 remained elevated following the multi-day exposure (Jorres et al., 2000).

In rats, the increases in BALF levels of proteins, fibronectin, IL-6 and inflammatory cells observed after one day of exposure to 0.4 ppm O₃ for 12 hours were no longer observed after 5 consecutive days of exposure (Van Bree et al., 2002). A separate study in rats exposed to 0.35-1 ppm O₃ for 2.25 hours for 5 consecutive days demonstrated a lack of attenuation of the increase in BALF protein, persistence of macrophages in the centriacinar region and histological evidence of progressive tissue injury (Tepper et al., 1989). Findings that injury, measured by BALF markers or by histopathology, persist in the absence of BAL neutrophilia or pulmonary function decrements suggested that repeated exposure to O₃ may have serious long-term consequences such as airway remodeling. In particular, the small airways were identified as a site where cumulative injury may occur (Frank et al., 2001).

Some studies examined the recovery of responses which were attenuated by repeated O₃ exposure. In a study of humans undergoing heavy exercise who were exposed for 2 hours to 400 ppb O₃ for five consecutive days (Devlin et al., 1997), recovery of the inflammatory responses which were diminished by repeated exposure required 10-20 days following the exposure (Devlin et al., 1997). In an animal study conducted in parallel (Van Bree et al., 2002), full susceptibility to O₃ challenge following exposure to O₃ for five consecutive days required 15-20 days recovery.

Several mechanisms have been postulated to explain the attenuation of some responses observed in human subjects and animal models following repeated exposure to O₃. First, the upregulation of antioxidant defenses (or conversely, a decrease in critical O₃-reactive substrates) may protect against O₃-mediated effects. Increases in antioxidant content of the BALF have been demonstrated in rats exposed to 0.25 and 0.5 ppm O₃ for several hours on consecutive days (Devlin et al., 1997; Wiester et al., 1996b; Tepper et al., 1989). Second, IL-6 was demonstrated to be an important mediator of attenuation in rats exposed to 0.5 ppm for 4 hours on two consecutive days (Mckinney et al., 1998). Third, a protective role for increases in mucus producing cells and mucus concentrations in the airways has been proposed (Devlin et al., 1997). Fourth, epithelial hyperplasia or metaplasia may decrease further effects due to subsequent O₃ challenge (Carey et al., 2007; Harkema et al., 1987a; Harkema et al., 1987b). These morphologic changes have been observed in nasal and lower airways in monkeys exposed chronically to 0.15-0.5 ppm O₃ and reflect a persistent change in epithelial architecture which may lead to other long-term sequelae. Although there is some evidence to support these possibilities, there is no consensus on mechanisms underlying response attenuation. Recent studies demonstrating that O₃ activates TRP receptors suggest that modulation of TRP
receptor number or sensitivity by repeated O₃ exposures may also contribute to the attenuation of responses.

In summary, the attenuation of pulmonary function responses by repeated exposure to O₃ has been linked to exacerbation of O₃-mediated injury. Enhanced exposure to O₃ due to a dampening of the O₃-mediated truncation of inspiration may be one factor which contributes to this relationship.

5.4.2.6 Co-exposures with Particulate Matter

Numerous studies have investigated the effects of co-exposure to O₃ and PM because of the prevalence of these pollutants in ambient air. Results are highly variable and depend on whether exposures are simultaneous or sequential, the type of PM employed and the endpoint examined. Additive and interactive effects have been demonstrated. For example, simultaneous exposure to O₃ (120 ppb for 2 hours at rest) and concentrated ambient particles (CAPs) in human subjects resulted in a diminished systemic IL-6 response compared with exposure to CAPs alone (Urch et al., 2010). However, exposure to O₃ alone did not alter blood IL-6 levels (Urch et al., 2010). The authors provided evidence that O₃ mediated a switch to shallow breathing which may have accounted for the observed antagonism (Urch et al., 2010). Further, simultaneous exposure to O₃ (114 ppb for 2 hours at rest) and CAPs but not exposure to either alone, resulted in increased diastolic blood pressure in human subjects (Fakhri et al., 2009). Mechanisms underlying this potentiation of response were not explored. In some strains of mice, pre-exposure to O₃ (0.5 ppm for 2 hours) modulated the effects of carbon black PM on heart rate, HRV and breathing patterns (Hamade and Tankersley, 2009). Another recent study in mice demonstrated that treatment with carbon nanotubes followed 12 hours later by O₃ exposure (0.5 ppm for 3 hours) resulted in a dampening of some of the pulmonary effects of carbon nanotubes measured as markers of inflammation and injury in the BALF (Han et al., 2008). Further, Harkema and Wagner (2005) found that epithelial and inflammatory responses in the airways of rats were enhanced by co-exposure to O₃ (0.5 ppm for 3 days) and LPS (used as a model of biogenic PM) or to O₃ (1 ppm for 2 days) and OVA (used as a model of an aeroallergen). Lastly, a recent study demonstrated that maternal exposure to particulate matter (PM) resulted in augmented lung inflammation, airway epithelial mucous metaplasia and AHR in young mice exposed chronically and intermittently to 1 ppm O₃ (Auten et al., 2009).

In summary, many of the demonstrated responses to co-exposure were more than additive. These findings are hard to interpret but demonstrate the complexity of responses following combined exposure to PM and O₃.
5.4.2.7 Summary

Collectively, these earlier and more recent studies provide some evidence for mechanisms that may underlie the variability in responsiveness seen among individuals (Figure 5-9). Certain functional genetic polymorphisms, pre-existing conditions and diseases, nutritional status, lifestage and co-exposures contribute to altered risk of O$_3$-induced effects. Attenuation of responses may also be important, but it is incompletely understood, both in terms of the pathways involved and the resulting consequences.

Figure 5-9 Some factors, illustrated in yellow, that likely contribute to the interindividual variability in responses resulting from inhalation of O$_3$. 
5.5 Species Homology and Interspecies Sensitivity

The previous O₃ AQCDs discussed the homology of responses in animals and humans exposed to O₃ and the interspecies differences that may affect these responses and concluded that the acute and chronic functional responses of laboratory animals to O₃ appear qualitatively homologous to human responses. Thus, animal studies can provide important data in determining cause-effect relationships between exposure and health outcome that would be impossible to collect in human studies. Furthermore, animal studies add to a better understanding of the full range of potential O₃-mediated effects.

Still, care must be taken when comparing quantitative dose-response relationships in animal models to humans due to obvious interspecies differences. This section will qualitatively describe basic concepts in species homology concerning both dose and response to O₃ exposure. Overall, there have been few new publications examining interspecies differences in dosimetry and response to O₃ since the last AQCD. These studies do not overtly change the conclusions discussed in the previous document.

5.5.1 Interspecies Dosimetry

As discussed in Section 5.2.1, O₃ uptake depends on complex interactions between RT morphology, breathing route, rate, and depth, physicochemical properties of the gas, physical processes of gas transport, as well as the physical and chemical properties of the ELF and tissue layers. Understanding differences in these variables between humans and experimental animals is important to interpreting delivered doses in animal and human toxicology studies.

Physiological and anatomical differences exist between experimental species. The structure of the URT is vastly different between rodents and humans but scales according to body mass. The difference in the cross-sectional shape and size of the nasal passages affects bulk airflow patterns, particularly the shape of major airflow streams. The nasal epithelium is lined by squamous, respiratory, transitional, or olfactory cells, depending on location. The differences in airflow patterns in the URT mean that not all nasal surfaces and cell types receive the same exposure to inhaled O₃ leading to differences in local absorption and potential for site-specific tissue damage. The morphology of the LRT also varies within and among species. Rats and mice do not possess respiratory bronchioles; however, these structures are present in humans, dogs, ferrets, cats, and monkeys. Respiratory bronchioles are abbreviated in hamsters, guinea pigs, sheep, and pigs. The branching structure of the ciliated bronchi and bronchioles also differs between species from being a rather symmetric and dichotomous branching network of airways in humans and primates to a more monopodial branching network in other mammals. In addition, rodents have fewer terminal bronchioles due to a smaller lung size compared to humans or canines (McBride, 1992). The cellular composition in the pulmonary region is similar across mammalian species; at least 95% of the alveolar epithelial tissue is composed of
Type I cells. However, considerable differences exist between species in the number and type of cells in the TB airways. Differences also exist in breathing route and rate. Primates are oronasal breathers, while rodents are obligate nasal breathers. Past studies of the effect of body size on resting oxygen consumption also suggest that rodents inhale more volume of air per lung mass than primates. These distinctions as well as differences in nasal structure between primates and rodents affect the amount of O₃ uptake.

As O₃ absorption and reactivity relies on ELF antioxidant substances (see Section 5.2.3), variability in antioxidant concentrations and metabolism between species may affect dose and O₃-induced health outcomes. The thickness of the ELF in the TB airways varies among species. Mercer et al. (1992) found that the human ELF thickness in bronchi and bronchioles was 6.9 and 1.8 µm, respectively, compared to 2.6 and 1.9 µm for the same locations in the rat. Guinea pigs and mice have a lower basal activity of GSH transferase and GSH peroxidase, and lower α-TOH levels in the lung compared to rats (Ichinose et al., 1988; Sagai et al., 1987). Nasal lavage fluid analysis shows that humans have a higher proportion of their nasal antioxidants as UA and low levels of AH₂ whereas mice, rats, or guinea pigs have high levels of AH₂ and undetectable levels of UA. GSH is not detected in the nasal lavage fluid of most of these species, but is present in monkey nasal lavage fluid. Guinea pigs and rats have a higher antioxidant to protein ratio in nasal lavage fluid and BALF than humans (Hatch, 1992). The BALF profile differs from the nasal lavage fluid. Humans have a higher proportion of GSH and less AH₂ making up their BALF content compared to the guinea pigs and rats (Slade et al., 1993; Hatch, 1992). Similar to the nose, rats have the highest antioxidant to protein mass ratio found in BALF (Slade et al., 1993). Antioxidant defenses also vary with age (Servais et al., 2005) and exposure history (Duan et al., 1996; Duan et al., 1993) reported that differences in antioxidant levels between species and lung regions did not appear to be the primary factor in O₃ induced tissue injury. However, a close correlation between site-specific O₃ dose, the degree of epithelial injury, and reduced glutathione depletion was observed in monkeys (Plopper et al., 1998).

Even with these differences, humans and animals are similar in the pattern of regional O₃ dose distribution. As discussed for humans in Section 5.2.2, O₃ flux to the air-liquid interface of the ELF slowly decreases distally in the TB region and then rapidly decreases distally in the alveolar region (Miller et al., 1985). Modeled tissue dose in the human RT, representing O₃ flux to the liquid-tissue interface, is very low in the trachea, increases to a maximum in the CAR, and then rapidly decreases distally in the alveolar region (Figure 5-10a). Similar patterns of O₃ tissue dose profiles normalized to inhaled O₃ concentration were predicted for rat, guinea pig, and rabbit (Miller et al., 1988) (Figure 5-10a). Overton et al. (1987) modeled rat and guinea pig O₃ dose distribution and found that after comparing two different morphometrically based anatomical models for each species, considerable difference in predicted percent RT and alveolar region uptakes were observed. This was due to the variability between the two anatomical models in airway path distance to the first alveolated duct. As a result, the overall dose profile was similar between species however the O₃ uptake efficiency varied due to RT size and path length.
A similar pattern of O\textsubscript{3} dose distribution was measured in monkeys exposed to 0.4 and 1.0 ppm \textsuperscript{18}O\textsubscript{3} (Plopper et al., 1998) (Figure 5-10b). Less \textsuperscript{18}O was measured in the trachea, proximal bronchus, and distal bronchus than was observed in the respiratory bronchioles. Again indicating the highest concentration of O\textsubscript{3} tissue dose is localized to the CAR, which are the respiratory bronchioles in nonhuman primates. In addition, the lowest \textsuperscript{18}O detected in the RT was in the parenchyma (i.e., alveolar region), reflecting the rapid decrease in tissue O\textsubscript{3} dose predicted by models for the alveolar regions of humans and other animals.

Humans and animal models are similar in the pattern of regional O\textsubscript{3} dose, but absolute values differ. Hatch et al. (1994) reported that exercising humans exposed to oxygen-18 labeled O\textsubscript{3} (400 ppb) accumulated 4-5 times higher concentrations of O\textsubscript{3} reaction product in BAL cells, surfactant and protein fractions compared to resting rats similarly exposed (400 ppb) (Figure 5-11). The use of \textsuperscript{18}O was specifically employed in an attempt to accurately measure O\textsubscript{3} dose to BALF fractions and lung tissue and was normalized to the dried mass of lavaged constituents. It was necessary to expose resting rats to 2 ppm O\textsubscript{3} to achieve the same BALF accumulation of \textsuperscript{18}O reaction product that was observed in humans exposed to 400 ppb with intermittent heavy exercise (\textit{\dot{V}}\textsubscript{E} \approx 60 L/min). The concentration of \textsuperscript{18}O reaction product in BALF paralleled the accumulation of BALF protein and cellular effects of the O\textsubscript{3} exposure observed such that these responses to 2.0 ppm O\textsubscript{3} were similar to those of the 400 ppb O\textsubscript{3} in exercising humans. This suggests that animal data obtained in resting conditions would underestimate the reaction of O\textsubscript{3} with cells in the RT and presumably the resultant risk of effect for humans. However these results should be interpreted with caution given an important limitation in the \textsuperscript{18}O labeling technique when used for interspecies comparisons. The reaction between O\textsubscript{3} and some reactants such as ascorbate produce \textsuperscript{18}O-labeled products that are lost during sample processing. When levels of ascorbate or other such reactants vary between species, this lost portion of the total \textsuperscript{18}O-reaction products will also vary, leading to uncertainty in interspecies comparisons.
Figure 5-10  Humans and animals are similar in the regional pattern of O₃ tissue dose distribution.
Note: The excess $^{18}$O in each fraction is expressed relative to the dry weight of that fraction. Fractions assayed include cells, high speed pellet (HSP), high speed supernatant (HSS), and lavaged lung homogenates.

Source: Hatch et al. (1994)

**Figure 5-11** Oxygen-18 incorporation into different fractions of BALF from humans and rats exposed to 0.4 and 2.0 ppm $^{18}$O$_3$.

Recently, a quantitative comparison of O$_3$ transport in the airways of rats, dogs, and humans was conducted using a three-compartment airways model, based on upper and lower airway casts and mathematical calculation for alveolar parameters (Tsujino et al., 2005). This one-dimensional gas transport model examined how interspecies anatomical and physiological differences affect intra-airway O$_3$ concentrations and the amount of gas absorbed. The morphological model consisted of cylindrical tubes with constant volume and no airway branching patterns. Peak, real-time, and mean O$_3$ concentrations were higher in the upper and lower airways of humans compared to rats and dogs, but lowest in the alveoli of humans. The amount of O$_3$ absorbed was lowest in humans when normalized by body weight. The intra-airway concentration decreased distally in all species. Sensitivity analysis demonstrated that
VT, \( f_B \), and upper and lower airways surface area had a statistically significant impact on model results. The model is limited in that it did not account for chemical reactions in the ELF or consider gas diffusion as a driving force for O\(_3\) transport. Also, the model was run at a respiratory rate of 16/min simulating a resting individual, however exercise may cause a further deviation from animal models as was seen in Hatch et al. (1994).

Overall, animal models exhibit qualitatively similar patterns of O\(_3\) net and tissue dose distribution with the largest tissue dose delivered to the CAR. However, due to anatomical and biochemical RT differences the absolute values of O\(_3\) dose delivered differs. Past results suggest that animal data obtained in resting conditions would underestimate the O\(_3\) reactions with cells in the BALF and presumably the resultant risk of effect for humans, especially for humans during exercise.

### 5.5.2 Interspecies Homology of Response

Biological response to O\(_3\) exposure broadly shows commonalities in many species. Among rodents, non-human primates, and humans, for example, ample data suggest that O\(_3\) induces oxidative stress, cell injury, upregulation of cytokines/chemokines, inflammation, alterations in lung function, and disruption of normal lung growth and development (see Chapters 6 and 7).

The effects related to early life exposures can differ appreciably across species due to the maturation stage of the lung and immune systems at birth. Evidence from non-human primate studies shows that early life O\(_3\) exposure disrupts lung development producing physiologic perturbations that are similar to those observed in children exposed to urban air pollution (Fanucchi et al., 2006; Joad et al., 2006). Studies of O\(_3\) effects on lung surface chemistry also show some degree of homology. Lipid oxidation products specific to O\(_3\) reactions with unsaturated fatty acids have been reported, for example, in lavage fluids from both rodents and humans (Frampton et al., 1999; Pryor et al., 1996). In humans, the extent to which systemic effects occur is less well studied; plasma indices of lipid oxidation such as isoprostanes unfortunately do not pinpoint the compartment(s) where oxidative stress has transpired. That oxidative stress occurs systemically in both rodents and non-human primates (Chuang et al., 2009), nevertheless, suggests that it likely also occurs in humans.

Despite the overall similarities in responses to O\(_3\) among species, studies have reported variability in the responsiveness to O\(_3\) between and within species, as well as between endpoints. Rodents appear to have a slightly higher tachypneic response to O\(_3\) and are less sensitive to changes in pulmonary function responses than humans (U.S. EPA, 1996a). However, rats experience attenuation of pulmonary function and tachypneic ventilatory responses, similar to humans (Wiester et al., 1996b). Hatch et al. (1986) reported that guinea pigs were the most responsive to O\(_3\)-induced inflammatory cell and protein influx. Rabbits were the least responsive and rats, hamsters, and mice were intermediate responders. Further analysis of this study by Miller et al. (1988) found that the protein levels in BALF from guinea pigs increased
more rapidly with predicted pulmonary tissue dose than in rats and rabbits. Alveolar macrophages isolated from guinea pigs and humans mounted similar qualitative and quantitative cytokine responses to in vitro \( \text{O}_3 \) (0.1-1.0 ppm for 60 minutes) exposure (Arsalane et al., 1995).

Also, because of their higher body surface to volume ratio, rodents can rapidly lower body temperature during exposure leading to lowered \( \text{O}_3 \) dose and toxicity (Watkinson et al., 2003; Iwasaki et al., 1998; Slade et al., 1997). In addition to lowering the \( \text{O}_3 \) dose to the lungs, this hypothermic response may cause: (1) lower metabolic rate, (2) altered enzyme kinetics, and (3) altered membrane function. The thermoregulatory mechanisms also may disrupt heart rate that may lead to: (1) decreased cardiac output, (2) lowered blood pressure, and (3) decreased tissue perfusion (Watkinson et al., 2003). These responses have not been observed in humans except at very high exposures, thus further complicating extrapolation of effects from animals to humans.

The degree to which \( \text{O}_3 \) induces injury and inflammation responses appears to be variable between species. However, the majority of those studies did not normalize the response to the dose received to account for species differences in \( \text{O}_3 \) absorption. For example, Dormans et al. (1999) found that rats, mice, and guinea pigs all exhibited \( \text{O}_3 \)-induced (0.2 - 0.4 ppm for 3-56 days) inflammation; however, guinea pigs were the most responsive with respect to alveolar macrophage elicitation and pulmonary cell density in the centriacinar region. Mice were the most responsive in terms of bronchiolar epithelial hypertrophy and biochemical changes (e.g., LDH, glutathione reductase, glucose-6-phosphate dehydrogenase activity), and had the slowest recovery from \( \text{O}_3 \) exposure. All species displayed increased collagen in the ductal septa and large lamellae bodies in Type II pneumocytes at the longest exposure and highest concentration; whereas this response occurred in the rat and guinea pig at lower \( \text{O}_3 \) levels (0.2 ppm) as well. Overall, the authors rated mice as most responsive, followed by guinea pigs, then rats (Dormans et al., 1999). Rats were also less responsive in terms of epithelial necrosis and inflammatory responses as a result of \( \text{O}_3 \) exposure (1.0 ppm for 8 hours) compared with monkeys and ferrets, which manifested a similar response (Sterner-Kock et al., 2000). Results of this study should be interpreted with caution since no dose metric was used to normalize the total inhaled dose or local organ dose between species.

To further understand the genetic basis for age-dependent differential response to \( \text{O}_3 \), adult (15 week old) and neonatal (15-16 day old) mice from 8 genetically diverse strains were examined for \( \text{O}_3 \)-induced (0.8 ppm for 5 hours) pulmonary injury and lung inflammation (Vancza et al., 2009). Ozone exposure increased polymorphonuclear leukocytes (PMN) influx in all strains of neonatal mice tested, but significantly greater PMNs occurred in neonatal compared to adult mice for only some sensitive strains, suggesting a genetic background effect. This strain difference was not due to differences in delivered dose of \( \text{O}_3 \) to the lung, evidenced by \(^{18}\text{O} \) lung enrichment. The sensitivity of strains for \( \text{O}_3 \)-induced increases in BALF protein and PMNs was different for different strains of mice suggesting that genetic factors contributed to heightened responses. Interestingly, adult mice accumulated more than
twice the levels of $^{18}$O reaction product of O$_3$ than corresponding strain neonates. Thus, it appeared that the infant mice showed a 2-fold to 3-fold higher response than the adults when expressed relative to the accumulated O$_3$ reaction product in their lungs. The apparent decrease in delivered O$_3$ dose in neonates could be a result of a more rapid loss of body temperature in infant rodents incident to maternal separation and chamber air flow.

In animal studies, inhaled O$_3$ concentration and exposure history rarely reflect actual human environmental exposures. Generally, very high exposure concentrations are used to induce murine AHR, which in some human subjects is observed at far more relevant concentrations. This calls into question whether the differences in airway reactivity are simply a function of differential nasopharyngeal scrubbing or whether the complexities encompassing a variety of contributory biological pathways show species divergence. Furthermore, in non-human primates exposed during early life, eosinophil trafficking occurs, which has not been observed in rodents (unless sensitized) (Maniar-Hew et al., 2011). This response has been shown to be persistent when O$_3$ challenges are administered after a recovery period of $\geq 9$ months during which no exposure transpired.

Quantitative extrapolation is challenging due to a number of uncertainties. Unfortunately, many input parameters needed to conduct quantitative extrapolations across species have not been obtained or currently remain undefined. It is not clear whether characterization of the ELF provides the information needed to compute a profile of reaction products or whether environmentally relevant exposure has altered the physicochemical interactions that occur within the RT surface compartment (e.g., O$_3$ diffusion through regions where the ELF is thin). That systemic effects have been documented in both rodents and non-human primates leads to the question of whether reaction products, cytokines/chemokines, or both enter the nasopharyngeal or bronchial circulation, both of which show species-dependent differences (Chuang et al., 2009; Cole and Freeman, 2009).

In addition, the response to O$_3$ insult across species and more recent health effects such as immune system development are uncertain. Non-human primate studies have shown hypo-responsiveness to endotoxin challenge as a consequence of exposure; whether this occurs in rodents and humans is largely unknown (Maniar-Hew et al., 2011). In addition, structural changes (e.g., airways remodeling, fibrogenesis) might differ appreciably across species. Moreover, whether the upper airways differentially contribute to either distal lung or systemic impacts has not been explored.

Some outcomes (e.g., inflammation) support the conclusion of homologous responses across species. However, factors such as age, exposure history, diet, endogenous substrate generation and homeostatic regulation, the cellular machinery that regulates inflammatory cell trafficking, responses to other environmental challenges, and the precise chemical species (whether ELF or cell membrane-derived) that account for exposure-related initiation of pathophysiologic sequelae might differ across species, but the extent of species-specific contributing factors remains unknown. Consequently, some level of uncertainty cannot be dismissed. Nonetheless, if experimental animals show pathophysiological consequences of
exposure, the overall weight of the toxicological evidence supports the likelihood that qualitatively similar effects occur in humans given appropriate exposure scenarios.

5.5.3 Summary

In summary, biological response to O₃ exposure broadly shows commonalities in many species and thus supports the use of animal studies in determining mechanistic and cause-effect relationships and as supporting evidence that similar effects could occur in humans if O₃ exposure is sufficient. However, there is uncertainty regarding the similarity of response to O₃ across species for some recently described endpoints. Differences exist between species in a number of factors that influence O₃ dosimetry and responses, such as RT anatomy, breathing patterns, and ELF antioxidant concentrations and chemical species. While humans and animals are similar in the pattern of regional O₃ dose distribution, these differences will likely result in differences in the absolute values of O₃ dose delivered throughout the RT. Thus, these considerations can impact quantitative comparison between species.

5.6 Chapter Summary

Ozone is a highly reactive gas and a powerful oxidant with a short half-life. Both O₃ uptake and responses are dependent upon the formation of secondary reaction products in the ELF; however more complex interactions occur. Total RT uptake in humans at rest is 80-95% efficient and it is influenced by a number of factors including RT morphology, breathing route, frequency, and volume, physicochemical properties of the gas, physical processes of gas transport, as well as the physical and chemical properties of the ELF and tissue layers. In fact, even though the average RT dose may be at a level where health effects would not be predicted, local regions of the RT may receive considerably higher than average doses due to RT inhomogeneity and differences in the pathlengths, and therefore be at greater risk of effects. About half of the O₃ that will be absorbed from the airstream is removed in the URT, which provides a defense against O₃ entering the lungs. However, the local dose to the URT tissue is site-specific and dependent on the nasal anatomy, nasal fluid composition, and ventilation and airflow patterns of the nasal passageways. The primary uptake site of O₃ delivery to the LRT epithelium is believed to be the CAR, however changes in a number of factors (e.g., physical activity) can alter the distribution of O₃ uptake in the RT. Ozone uptake is chemical reaction-dependent and the substances present in the ELF appear in most cases to limit interaction of O₃ with underlying tissues and to prevent penetration of O₃ distally into the RT. Still, reactions of O₃ with soluble ELF components or possibly plasma membranes result in distinct products, some of which are highly reactive and can injure and/or transmit signals to RT cells.
Thus, in addition to contributing to the driving force for \( O_3 \) uptake, formation of secondary oxidation products initiates pathways that provide the mechanistic basis for health effects that are described in detail in Chapters 6 and 7 and that involve the RT as well as extrapulmonary systems. These pathways include activation of neural reflexes, initiation of inflammation, alteration of epithelial barrier function, sensitization of bronchial smooth muscle, modification of innate and adaptive immunity, airways remodeling, and systemic inflammation and oxidative/nitrosative stress. With the exception of airways remodeling, these pathways have been demonstrated in both animals and human subjects in response to the inhalation of \( O_3 \).

Both dosimetric and mechanistic factors contribute to the understanding of interindividual variability in responses to \( O_3 \). This variability is influenced by differences in RT volume and surface area, certain genetic polymorphisms, pre-existing conditions and disease, nutritional status, lifestages, attenuation, and co-exposures. Some of these factors also underlie differences in species homology and sensitivity. Qualitatively, animal models exhibit similar patterns of \( O_3 \) net and tissue dose distribution with the largest tissue dose of \( O_3 \) delivered to the CAR. However, due to anatomical and biochemical RT differences, the absolute value of delivered \( O_3 \) dose differs, with animal data obtained in resting conditions underestimating the dose to the RT and presumably the resultant risk of effect for humans, especially humans during exercise. Even though interspecies differences can complicate quantitative comparison between species, many short-term responses of laboratory animals to \( O_3 \) appear qualitatively homologous to those of the human. Furthermore, animal studies add to a better understanding of the full range of potential \( O_3 \)-mediated effects. Given the commonalities in many responses across species, animal studies that observe \( O_3 \)-induced effects may be used as supporting evidence that similar effects could occur in humans or in determining mechanistic and cause-effect relationships if \( O_3 \) exposure is sufficient.
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6 INTEGRATED HEALTH EFFECTS OF SHORT-TERM OZONE EXPOSURE

6.1 Introduction

This chapter reviews, summarizes, and integrates the evidence for various health outcomes associated with short-term (i.e., hours, days, or weeks) exposures to O₃. Numerous controlled human exposure, epidemiologic, and toxicological studies have permitted evaluation of the relationships between short-term O₃ exposure and a range of endpoints related to respiratory effects (Section 6.2), cardiovascular effects (Section 6.3), and mortality (Section 6.2, Section 6.3, and Section 6.6). A smaller number of studies were available to assess the effects of O₃ exposure on other physiological systems such as the central nervous system (Section 6.4), liver and metabolism (Section 6.5.1), and cutaneous and ocular tissues (Section 6.5.2). This chapter evaluates the majority of recent [i.e., published since the completion of the 2006 O₃ AQCD (U.S. EPA, 2006b)] short-term exposure studies; however, those for birth outcomes and infant mortality are evaluated in Chapter 7 (Section 7.4), because they compare associations among overlapping short- and long-term exposure windows that are difficult to distinguish.

Within each individual section of this chapter, a brief summary of conclusions from the 2006 O₃ AQCD is included along with an evaluation of recent evidence that is intended to build upon the body of evidence from previous reviews. The studies evaluated are organized by health endpoint (e.g., lung function, pulmonary inflammation) then by scientific discipline (e.g., controlled human exposure, epidemiology, and toxicology). Each major section (e.g., respiratory, cardiovascular, mortality) concludes with an integrated summary of the findings and a conclusion regarding causality based upon the framework described in the Preamble to this ISA. The causal determinations are presented for a broad health effect category, such as respiratory effects, with coherence and plausibility based on the total evidence available across disciplines and across the suite of related health endpoints, including cause-specific mortality.

6.2 Respiratory Effects

Based on evidence integrated across controlled human exposure, epidemiologic, and toxicological studies, the 2006 O₃ AQCD concluded “that acute O₃ exposure is causally associated with respiratory system effects” (U.S. EPA, 2006b). Contributing to this conclusion were the consistency and coherence across scientific disciplines for the effects of short-term O₃ exposure on a variety of respiratory outcomes including “pulmonary function decrements, respiratory symptoms, lung inflammation, and increased lung permeability, airway hyperresponsiveness.” Collectively, these findings provided biological plausibility for associations in epidemiologic studies
observed between short-term increases in ambient O₃ concentration and increases in respiratory symptoms and respiratory-related hospitalizations and emergency department (ED) visits.

Controlled human exposure studies have provided strong and quantifiable exposure-response data on the human health effects of O₃. The most salient observations from studies reviewed in the 1996 and 2006 O₃ AQCDs (U.S. EPA, 2006b, 1996a) included: (1) young healthy adults exposed to O₃ concentrations ≥ 80 ppb develop significant reversible, transient decrements in pulmonary function and symptoms of breathing discomfort if minute ventilation (V̇E) or duration of exposure is increased sufficiently; (2) relative to young adults, children experience similar spirometric responses but lower incidence of symptoms from O₃ exposure; (3) relative to young adults, O₃-induced spirometric responses are decreased in older individuals; (4) there is a large degree of intersubject variability in physiologic and symptomatic responses to O₃, but responses tend to be reproducible within a given individual over a period of several months; (5) subjects exposed repeatedly to O₃ for several days experience an attenuation of spirometric and symptomatic responses on successive exposures, which is lost after about a week without exposure; and (6) acute O₃ exposure initiates an inflammatory response that may persist for at least 18 to 24 hours postexposure.

Substantial evidence for biologically plausible O₃-induced respiratory morbidity has been derived from the coherence between toxicological and controlled human exposure study findings for parallel endpoints. For example, O₃-induced lung function decrements and increased airway hyperresponsiveness have been observed in both animals and humans. Airway hyperresponsiveness could be an important consequence of exposure to ambient O₃ because the airways are then predisposed to narrowing upon inhalation of a variety of ambient stimuli. Additionally, airway hyperresponsiveness tends to resolve more slowly and appears less subject to attenuation with repeated O₃ exposures than lung function decrements. Increased permeability and inflammation have been observed in the airways of humans and animals alike after O₃ exposure, although these processes are not necessarily associated with immediate changes in lung function or hyperresponsiveness. Furthermore, the potential relationship between repetitive bouts of acute inflammation and the development of chronic respiratory disease is unknown. Another feature of O₃-related respiratory morbidity is impaired host defense and reduced resistance to lung infection, which has been strongly supported by toxicological evidence and, to a limited extent, by evidence from controlled human exposure studies. Recurrent respiratory infection in early life is associated with increased incidence of asthma in humans.

In concordance with experimental studies, epidemiologic studies have provided clear evidence for decrements in lung function related to short-term ambient O₃ exposure. These effects have been demonstrated in healthy children attending camps, adults exercising or working outdoors, and children with and without asthma (U.S. EPA, 2006b, 1996a). In addition to lung function decrements, short-term increases in ambient O₃ concentration have been associated with increases in respiratory symptoms (e.g., cough, wheeze, shortness of breath), notably in large U.S. panel
studies of children with asthma (Gent et al., 2003; Mortimer et al., 2000). The evidence across disciplines for O$_3$ effects on a range of respiratory endpoints collectively provides support for epidemiologic studies that have demonstrated consistent associations between short-term increases in ambient O$_3$ concentration and increases in respiratory hospital admissions and ED visits, specifically during the summer or warm months. In contrast with other respiratory health endpoints, previous epidemiologic evidence has not clearly supported a relationship between short-term O$_3$ exposure and respiratory mortality. Although O$_3$ has been consistently associated with nonaccidental and cardiopulmonary mortality, the contribution of respiratory causes to these findings was uncertain as the few studies that have examined mortality specifically from respiratory causes reported inconsistent associations with ambient O$_3$ concentrations.

As will be discussed throughout this section, consistent with the strong body of evidence presented in the 2006 O$_3$ AQCD, recent studies continue to support associations between short-term O$_3$ exposure and respiratory effects, in particular, lung function decrements in controlled human exposure studies, airway inflammatory responses in toxicological studies, and respiratory-related hospitalizations and ED visits. Recent epidemiologic studies contribute new evidence for potentially at-risk populations and associations linking ambient O$_3$ concentrations with biological markers of airway inflammation and oxidative stress, which is consistent with the extensive evidence from controlled human exposure and toxicological studies. Furthermore, extending the potential range of well-established O$_3$-associated respiratory effects, recent multicity studies and a multicontinent study demonstrate associations between short-term increases in ambient O$_3$ concentration and respiratory-related mortality.

### 6.2.1 Lung Function

#### 6.2.1.1 Controlled Human Exposure

This section focuses on studies examining O$_3$ effects on lung function and respiratory symptoms in volunteers exposed, for periods of up to 8 hours, to O$_3$ concentrations ranging from 40 to 500 ppb, while at rest or during exercise of varying intensity. Responses to acute O$_3$ exposures in the range of ambient concentrations include decreased inspiratory capacity; mild bronchoconstriction; rapid, shallow breathing patterns during exercise; and symptoms of cough and pain on deep inspiration (PDI). Reflex inhibition of inspiration results in a decrease in forced vital capacity (FVC) and total lung capacity (TLC) and, in combination with mild bronchoconstriction, contributes to a decrease in the forced expiratory volume in 1 second (FEV$_1$).

In studies that have exposed subjects during exercise, the majority of shorter duration (≤ 4-hour exposures) studies utilized an intermittent exercise protocol in which
subjects rotated between 15-minute periods of exercise and rest. A limited number of 1- to 2-hour studies, mainly focusing on exercise performance, have utilized a continuous exercise regime. A quasi continuous exercise protocol is common to prolonged exposure studies where subjects complete 50-minute periods of exercise followed by 10-minute rest periods.

The majority of controlled human exposure studies have been conducted within exposure chambers, although a smaller number of studies used a facemask to expose subjects to O₃. Little effort has been made herein to differentiate between facemask and chamber exposures since FEV₁ and respiratory symptom responses appear minimally differentially affected by these exposure modalities. Similar responses between facemask and chamber exposures have been reported for exposures to 80 and 120 ppb O₃ (6.6-hour, moderate quasi continuous exercise, 40 L/min) and 300 ppb O₃ (2 hours, heavy intermittent exercise, 70 L/min) (Adams, 2003a, b, 2002).

The majority of controlled human exposure studies investigating the effects O₃ are of a randomized, controlled, crossover design in which subjects were exposed, without knowledge of the exposure condition and in random order to clean filtered air (FA; the control) and, depending on the study, to one or more O₃ concentrations. The FA control exposure provides an unbiased estimate of the effects of the experimental procedures on the outcome(s) of interest. Comparison of responses following this FA exposure to those following an O₃ exposure allows for estimation of the effects of O₃ itself on an outcome measurement while controlling for independent effects of the experimental procedures. As individuals may experience small changes in various health endpoints from exercise, diurnal variation, or other effects in addition to those of O₃ during the course of an exposure, the term “O₃-induced” is used herein to designate effects that have been corrected or adjusted for such extraneous responses as measured during FA exposures.

Spirometry, viz., FEV₁, is a common health endpoint used to assess effects of O₃ on respiratory health in controlled human exposure studies. In considering 6.6-hour exposures to FA, group mean FEV₁ changes have ranged from -0.7% (McDonnell et al., 1991) to 2.7% (Adams, 2006a). On average, across ten 6.6-hour exposure studies, there has been a 1.0% (n = 279) increase in FEV₁ (Kim et al., 2011; Schelegle et al., 2009; Adams, 2006a, 2003a, 2002; Adams and Ollison, 1997; Folinsbee et al., 1994; McDonnell et al., 1991; Horstman et al., 1990; Folinsbee et al., 1988). Regardless of the reason for small changes in FEV₁ over the course of FA exposures, whether biologically based or a systematic effect of the experimental procedures, the use of FA responses as a control for the assessment of responses following O₃ exposure in randomized exposure studies serves to eliminate alternative explanations other than those of O₃ itself in causing the measured responses.

With respect to FEV₁ responses in young healthy adults, an O₃-induced change in FEV₁ is typically the difference between the decrement observed with O₃ exposure and the improvement observed with FA exposure. Noting that some healthy individuals experience small improvements while others have small decrements in FEV₁ following FA exposure, investigators have used the randomized, crossover
design with each subject serving as their own control (exposure to FA) to discern relatively small effects with certainty since alternative explanations for these effects are controlled for by the nature of the experimental design. The utility of intraindividual FA control exposures becomes more apparent when considering individuals with respiratory disease. The occurrence of exercise-induced bronchospasm is well recognized in patients with asthma and chronic obstructive pulmonary disease (COPD) and may be experienced during both FA and O$_3$ exposures. Absent correction for FA responses, exercise-induced changes in FEV$_1$ could be mistaken for responses due to O$_3$. This biological phenomenon serves as an example to emphasize the need for a proper control exposure in assessing the effects of O$_3$ as well as the role of this control in eliminating the influence of other factors on the outcomes of interest.

**Pulmonary Function Effects of Ozone Exposure in Healthy Subjects**

**Acute Exposure of Healthy Subjects**

The majority of controlled human exposure studies have investigated the effects of exposure to O$_3$ in young healthy nonsmoking adults (18-35 years of age). These studies typically use fixed concentrations of O$_3$ under carefully regulated environmental conditions and subject activity levels. The magnitude of respiratory effects (decrements in spirometry measurements and increases in symptomatic responses) in these individuals is a function of O$_3$ concentration (C), minute ventilation ($\dot{V}_E$), and exposure duration (time). Any physical activity will increase minute ventilation and therefore the dose of inhaled O$_3$. Dose of inhaled O$_3$ to the lower airways is also increased due to a shift from nasal to oronasal breathing with a consequential decrease in O$_3$ scrubbing by the upper airways. Thus, the intensity of physiological response following an acute exposure will be strongly associated with minute ventilation.

The product of C × $\dot{V}_E$ × time is commonly used as a surrogate for O$_3$ dose to the respiratory tract in controlled human exposure studies. A large body of data regarding the interdependent effects of C, $\dot{V}_E$, and time on pulmonary responses was assessed in the 1986 and 1996 O$_3$ AQCDs (U.S. EPA, 1996a, 1986). Acute responses were modeled as a function of total inhaled dose (C × $\dot{V}_E$ × time) which was found to be a better predictor of response to O$_3$ than C, $\dot{V}_E$, or time of exposure, alone, or as a combination of any two of these factors. However, intake dose (C × $\dot{V}_E$ × time) did not adequately capture the temporal dynamics of pulmonary responses in a comparison between a constant (square-wave) and a variable (triangular) O$_3$ exposure (average 120 ppb O$_3$; moderate exercise, $\dot{V}_E = 40$ L/min; 8 hour duration) conducted by Hazucha et al. (1992). Recent nonlinear statistical models clearly describe the temporal dynamics of FEV$_1$ responses as a function of C, $\dot{V}_E$, time, and age of the exposed subject (McDonnell et al., 2010, 2007).

For healthy young adults exposed at rest for 2 hours, 500 ppb is the lowest O$_3$ concentration reported to produce a statistically significant O$_3$-induced group mean
FEV$_1$ decrement of 6.4% (n = 10) (Folinsbee et al., 1978) to 6.7% (n = 13) (Horvath et al., 1979). Airway resistance was not clearly affected during at-rest exposure to these O$_3$ concentrations. For exposures of 1-2 hours to ≥ 120 ppb O$_3$, statistically significant symptomatic responses and effects on FEV$_1$ are observed when $\dot{V}_E$ is sufficiently increased by exercise (McDonnell et al., 1999b). For instance, 5% of young healthy adults exposed to 400 ppb O$_3$ for 2 hours during rest experienced pain on deep inspiration. Respiratory symptoms were not observed at lower exposure concentrations (120-300 ppb) or with only 1 hour of exposure even at 400 ppb. However, when exposed to 120 ppb O$_3$ for 2 hours during light-to-moderate intermittent exercise ($\dot{V}_E$ of 22 - 35 L/min), 9% of individuals experienced pain on deep inspiration, 5% experienced cough, and 4% experienced shortness of breath. With very heavy continuous exercise ($\dot{V}_E$ = 89 L/min), an O$_3$-induced group mean decrement of 9.7% in FEV$_1$ has been reported for healthy young adults exposed for 1 hour to 120 ppb O$_3$ (Gong et al., 1986). Symptoms are present and decrements in forced expiratory volumes and flows occur at 160-240 ppb O$_3$ following 1 hour of continuous heavy exercise ($\dot{V}_E$ ≈ 55 to 90 L/min (Gong et al., 1986; Avol et al., 1984; Folinsbee et al., 1984; Adams and Schelegle, 1983) and following 2 hours of intermittent heavy exercise ($\dot{V}_E$ ≈ 65-68 L/min) (Linn et al., 1986; Kulle et al., 1985; McDonnell et al., 1983). With heavy intermittent exercise (15-min intervals of rest and exercise [$\dot{V}_E$ = 68 L/min]), symptoms of breathing discomfort and a group mean O$_3$-induced decrement of 3.4% in FEV$_1$ occurred in young healthy adults exposed for 2 hours to 120 ppb O$_3$ (McDonnell et al., 1983). Table 6-1 provides examples of typical exercise protocols utilized in controlled human exposures to O$_3$. The $\dot{V}_E$ rates in this table are per body surface area (BSA) which is, on average, about 1.7 m$^2$ and 2.0 m$^2$ for young healthy adult females and males, respectively, who participated in controlled O$_3$ exposure studies.

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1 In total, subjects were exposed to O$_3$ for 2.5 hours. Intermittent exercise periods, however, were only conducted for the first 2 hours of exposure and FEV$_1$ was determined 5 minutes after the exercise was completed.
Table 6-1  Activity levels used in controlled exposures of healthy young adults to O$_3$.

<table>
<thead>
<tr>
<th>Activity$^{a,b}$</th>
<th>Study Duration (hours)</th>
<th>$\dot{V}_E$ (L/min per m$^2$ BSA)</th>
<th>Heart Rate (bpm)</th>
<th>Treadmill Speed (mph)</th>
<th>Treadmill Grade (%)</th>
<th>Cycle (watts)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rest</td>
<td>2</td>
<td>4</td>
<td>70</td>
<td>n.a.</td>
<td>n.a.</td>
<td>n.a.</td>
</tr>
<tr>
<td>Light quasi-continuous exercise</td>
<td>6.6-7.6</td>
<td>15</td>
<td>110</td>
<td>3.5-4.4</td>
<td>0</td>
<td>42</td>
</tr>
<tr>
<td>Moderate quasi-continuous exercise</td>
<td>6.6</td>
<td>17-23</td>
<td>115-130</td>
<td>3.3-3.5</td>
<td>4-5</td>
<td>72</td>
</tr>
<tr>
<td>Heavy intermittent exercise</td>
<td>1-2</td>
<td>27-33</td>
<td>160</td>
<td>3.5-5</td>
<td>10-12</td>
<td>100</td>
</tr>
<tr>
<td>Very heavy continuous exercise</td>
<td>1</td>
<td>45</td>
<td>160</td>
<td>n.a.</td>
<td>n.a.</td>
<td>260</td>
</tr>
</tbody>
</table>

$^a$Based on group mean exercise specific data provided in the individual studies. On average, subjects were 23 years of age. For exercise protocols, the minute ventilation and heart rate are for the exercise periods. Quasi-continuous exercise consists of 50 minutes of exercise periods followed by 10 minutes of rest. Intermittent exercise consists of alternating periods of 15 minutes of exercise and 15 minutes of rest.

Source: Brown et al. (2008).

Top, panel A: All studies exposed subjects to a constant (square-wave) concentration in a chamber, except Adams (2002) where a facemask was used. All responses at and above 0.06 ppm were statistically significant. The McDonnell et al. (2007) curve illustrates the predicted FEV₁ decrement at 6.6 hours as a function of O₃ concentration for a 23 year-old (the average age of subjects that participated in the illustrated studies). Note that this curve was not “fitted” to the plotted data. Error bars (where available) are the standard error of responses.

Bottom, panel B: All studies used constant (square-wave) exposures in a chamber unless designated as triangular (t) and/or facemask (m) exposures. All responses at and above 0.07 ppm were statistically significant. At 0.06 ppm, Adams (2006a) and Kim et al. (2011) responses to square-wave chamber exposures were statistically significant. During each hour of the exposures, subjects were engaged in moderate quasi continuous exercise (40 L/min) for 50 minutes and rest for 10 minutes. Following the third hour, subjects had an additional 35-minute rest period for lunch. The data at 0.06, 0.08 and 0.12 ppm have been offset for illustrative purposes.


Figure 6-1  Cross-study comparison of mean O₃-induced FEV₁ decrements following 6.6 hours of exposure to O₃.
For prolonged (6.6 hours) exposures relative to shorter exposures, significant pulmonary function responses and symptoms have been observed at lower O₃ concentrations and at a moderate level of exercise (\( \dot{V}_E = 40 \) L/min). The 6.6-hour experimental protocol was intended to simulate the performance of heavy physical labor for a full workday (Folinsbee et al., 1988). The results from studies using 6.6 hours of constant or square-wave exposures to between 40 and 120 ppb O₃ are illustrated in Figure 6-1(A). Figure 6-1(B) focuses on the range from 40 to 80 ppb and includes triangular exposure protocols as well as facemask exposures. Exposure to 40 ppb O₃ for 6.6 hours produces small, statistically nonsignificant changes in FEV₁ that are relatively similar to responses from FA exposure (Adams, 2002).

Volunteers exposed to 60 ppb O₃ experience group mean O₃-induced FEV₁ decrements of about 3% (Kim et al., 2011; Brown et al., 2008; Adams, 2006a); those exposed to 80 ppb have group mean decrements that range from 6 to 8% (Adams, 2006a, 2003a; McDonnell et al., 1991; Horstman et al., 1990); at 100 ppb, group mean decrements range from 8 to 14% (McDonnell et al., 1991; Horstman et al., 1990); and at 120 ppb, group mean decrements of 13 to 16% are observed (Adams, 2002; Horstman et al., 1990; Folinsbee et al., 1988). As illustrated in Figure 6-1, there is a smooth intake dose-response curve without evidence of a threshold for exposures between 40 and 120 ppb O₃. This is consistent with Hazucha and Lefohn (2007), who suggested that a randomly selected group of healthy individuals of sufficient size would include hypo-, normo-, and hyper-responsive individuals such that the average response would show no threshold for any spirometric endpoint. Taken together, these data indicate that mean FEV₁ is clearly decreased by 6.6-hour exposures to 60 ppb O₃ and higher concentrations in subjects performing moderate exercise. Discussed later in this subsection, the recent McDonnell et al. (2012) and Schelegle et al. (2012) studies analyzed large datasets and fit compartmental models, which included the concept of a response threshold.

The time course of responses during prolonged (6.6 hours) square-wave O₃ exposures with moderate exercise (\( \dot{V}_E = 40 \) L/min) depends on O₃ concentration. At 120 ppb O₃, Folinsbee et al. (1988) observed that somewhat small FEV₁ decrements and symptoms of breathing discomfort become apparent in healthy subjects following the second hour of exposure with a more rapid change in responses between the 3rd and 5th hour of exposure and a diminishing response or plateau in responses over the last hour of exposure. Relative to FA, the change in FEV₁ at 120 ppb O₃ became statistically significant after 4.6 hours. Following the same exposure protocol, Horstman et al. (1990) observed a linear increase in FEV₁ responses with time following 2 hours of exposure to 120 ppb O₃ that was statistically different from FA responses after 3 hours. At 100 ppb O₃, FEV₁ responses diverged from FA after 3 hours and were statistically different at 4.6 hours (Horstman et al., 1990). At 80 ppb O₃, FEV₁ responses diverged from FA after 4.6 hours and were statistically different from FA at 5.6 hours (Horstman et al., 1990). Subsequently, Adams (2006a) observed FEV₁ decrements and total respiratory

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1 Adams (2006a) did not find effects on FEV₁ at 60 ppb to be statistically significant. In an analysis of the Adams (2006a) data, even after removal of potential outliers, Brown et al. (2008) found the average effect on FEV₁ at 60 ppb to be small, but highly statistically significant (p < 0.002) using several common statistical tests.
symptoms at 80 ppb O₃ to diverge from FA responses after 3 hours, but did not become statistically different until 6.6 hours. At 60 ppb O₃, FEV₁ responses generally tracked responses in FA for the first 4.6 hours of exposure and diverged after 5.6 hours (Adams, 2006a). FEV₁ responses, but not symptomatic responses, become statistically different between 60 ppb O₃ and FA at 6.6 hours (Kim et al., 2011; Brown et al., 2008). At 40 ppb, FEV₁ and symptomatic responses track FA for 5.6 hours of exposure and may begin to diverge after 6.6 hours (Adams, 2002). In prolonged (6.6 hours) square-wave O₃ exposures between 40 and 120 ppb with moderate exercise (Vₑ = 40 L/min), the time required for group mean responses to differ between O₃ and FA exposures increases with decreasing O₃ concentration.

As opposed to constant (i.e., square-wave) concentration patterns used in the studies described above, many studies conducted at the levels of 40-80 ppb have used variable (i.e., triangular) O₃ concentration patterns. It has been suggested that a triangular exposure profile can potentially lead to higher FEV₁ responses than square-wave profiles despite having the same average O₃ concentration over the exposure period. Hazucha et al. (1992) were the first to investigate the effects of variable versus constant concentration exposures on responsiveness to O₃. In their study, volunteers were randomly exposed to a triangular concentration profile (averaging 120 ppb over the 8-hour exposure) that increased linearly from 0-240 ppb for the first 4 hours of the 8-hour exposure, then decreased linearly from 240 to 0 ppb over the next 4 hours of the 8-hour exposure, and to an square-wave exposure of 120 ppb O₃ for 8 hours. While the total inhaled O₃ doses at 4 hours and 8 hours for the square-wave and the triangular concentration profile were almost identical, the FEV₁ responses were dissimilar. For the square-wave exposure, FEV₁ declined ~5% by the fifth hour and then remained at that level. With the triangular O₃ profile, there was minimal FEV₁ response over the first 3 hours followed by a rapid decrease in FEV₁ to a decrement of 10.3% over the next 3 hours. During the seventh and eighth hours, mean FEV₁ decrement improved to 6.3% as the O₃ concentration decreased from 120 to 0 ppb (mean = 60 ppb). These findings illustrate that the severity of symptoms and the magnitude of spirometric responses are time-dependent functions of O₃ delivery rate with periods of both effect development and recovery during the course of an exposure.

Subsequently, others have also demonstrated that variable concentration exposures can elicit greater FEV₁ and symptomatic responses than do square-wave exposures to O₃ (Adams, 2006a, b, 2003a). Adams (2006b) reproduced the findings of Hazucha et al. (1992) at 120 ppb. However, Adams (2006a, 2003a) found that responses from an 80 ppb O₃ (average) triangular exposure did not differ significantly from those observed with the 80 ppb O₃ square-wave exposure at 6.6 hours. Nevertheless, FEV₁ and symptoms were significantly different from pre-exposure at 4.6 hours (when the O₃ concentration was 150 ppb) in the triangular exposure, but not until 6.6 hours in the square-wave exposure. At the lower O₃ concentration of 60 ppb, no temporal pattern differences in FEV₁ responses could be discerned between square-wave and triangular exposure profiles (Adams, 2006a). However, both total symptom scores and pain on deep inspiration tended to be greater following the 60 ppb triangular than the 60 ppb square-wave exposure. At 80 ppb O₃, respiratory symptoms tended to
increase more rapidly during the triangular than square-wave exposure protocol, but then decreased during the last hour of exposure to be less than that observed with the square-wave exposure at 6.6 hours. Both total symptom scores and pain on deep inspiration were significantly increased following exposures to 80 ppb O₃ relative to all other exposure protocols, i.e., FA, 40, and 60 ppb exposures. Following the 6.6-hour exposures, respiratory symptoms at 80 ppb were roughly 2-3 times greater than those observed at 60 ppb. At 40 ppb, triangular and square-wave patterns produced spirometric and subjective symptom responses similar to FA exposure (Adams, 2006a, 2002).

For O₃ exposures of 60 ppb and greater, studies (Adams, 2006a, b, 2003a; Hazucha et al., 1992) demonstrate that during triangular exposure protocols, volunteers exposed during moderate exercise ($\dot{V}_E = 40$ L/min) may develop greater spirometric and/or symptomatic responses during and following peak O₃ concentrations as compared to responses over the same time interval of square-wave exposures. This observation is not unexpected since the inhaled dose rate during peaks of the triangular protocols approached twice that of the square-wave protocols, e.g., 150 ppb versus 80 ppb peak concentration. At time intervals toward the end of an exposure, O₃ delivery rates for the triangular protocols were less than those of square-wave. At these later time intervals, there is some recovery of responses during triangular exposure protocols, whereas there is a continued development of or a plateau of responses in the square-wave exposure protocols. Thus, responses during triangular protocols relative to square-wave protocols may be expected to diverge and be greater following peak exposures and then converge toward the end of an exposure. Subsequent discussion will focus on exposures between 40 and 80 ppb where FEV₁ pre-to-post responses are similar (although not identical) between triangular and square-wave protocols having equivalent average exposure concentrations.

Schelegle et al. (2009) recently investigated the effects of 6.6-hour variable O₃ exposure protocols at mean concentrations of 60, 70, 80, and 87 ppb on respiratory symptoms and pulmonary function in young healthy adults (16 F, 15 M; 21.4 ± 0.6 years) exposed during moderate quasi continuous exercise ($\dot{V}_E = 40$ L/min). The mean FEV₁ ($\pm$ standard error) decrements at 6.6 hours (end of exposure relative to pre-exposure) were: −0.80 ± 0.90%, 2.72 ± 1.48%, 5.34 ± 1.42%, 7.02 ± 1.60%, and 11.42 ± 2.20% for exposure to FA, 60, 70, 80, and 87 ppb O₃, respectively. Statistically significant decrements in FEV₁ and increases in total subjective symptom scores ($p <0.05$) were found following exposure to mean concentrations of 70, 80, and 87 ppb O₃ relative to FA. Statistically significant effects were not found at 60 ppb. One of the expressed purposes of the Schelegle et al. (2009) study was to determine the minimal mean O₃ concentration that produces a statistically significant decrement in FEV₁ and respiratory symptoms in healthy individuals completing 6.6-hour exposure protocols. At 70 ppb, Schelegle et al. (2009) observed a statistically significant O₃-induced FEV₁ decrement of 6.1% at 6.6 hours and a significant increase in total subjective symptoms at 5.6 and 6.6 hours. A re-analysis found the FEV₁ responses at 70 ppb to be significantly different from FA responses beginning at 4.6 hours of exposure (Lefohn et al., 2010a). At 60 ppb,
an O₃-induced 3.5% FEV₁ decrement was not found to be statistically significant. However, this effect is similar in magnitude to the 2.9% FEV₁ decrement at 60 ppb observed by Adams (2006a), which was found to be statistically significant by Brown et al. (2008).

More recently, Kim et al. (2011) investigated the effects of a 6.6-hour exposure to 60 ppb O₃ during moderate quasi continuous exercise (\( \dot{V}_E = 40 \) L/min) on pulmonary function and respiratory symptoms in young healthy adults (32 F, 27 M; 25.0 ± 0.5 years) who were roughly half glutathione S-transferase μ-1 (GSTM1)-null genetically and half GSTM1-positive. Sputum neutrophil levels were also measured in a subset of the subjects (13 F, 11 M). The mean FEV₁ (± standard error) decrements at 6.6 hours (end of exposure relative to pre-exposure) were significantly different (p = 0.008) between the FA (0.002 ± 0.46%) and O₃ (1.76 ± 0.50%) exposures. The inflammatory response following O₃ exposure was also significantly (p <0.001) increased relative to the FA exposure. Respiratory symptoms were not affected by O₃ exposure. There was also no significant effect of GSTM1 genotype on FEV₁ or inflammatory responses to O₃.

Consideration of the minimal O₃ concentration producing statistically significant effects on FEV₁ and respiratory symptoms (e.g., cough and pain on deep inspiration) following 6.6-hour exposures warrants additional discussion. As discussed above, numerous studies have demonstrated statistically significant O₃-induced group mean FEV₁ decrements of 6-8% and an increase in respiratory symptoms at 80 ppb. Schelegle et al. (2009) have now reported a statistically significant O₃-induced group mean FEV₁ decrement of 6%, as well as increased respiratory symptoms, at 70 ppb. At 60 ppb, there is information available from 4 separate studies (Kim et al., 2011; Schelegle et al., 2009; Adams, 2006a, 2002).¹ The group mean O₃-induced FEV₁ decrements observed in these studies were 3.6% (facemask, square-wave) by Adams (2006a, 2002)², 2.8% (triangular exposure) and 2.9% (square-wave exposure) by Adams (2006a), 3.5% (triangular exposure) by Schelegle et al. (2009), and 1.8% (square-wave exposure) by Kim et al. (2011). Based on data from these studies, at 60 ppb, the weighted-average group mean O₃-induced FEV₁ decrement (i.e., adjusted for FA responses) is 2.7% (n = 150). Although not consistently statistically significant, these group mean changes in FEV₁ at 60 ppb are consistent among studies, i.e., none observed an average improvement in lung function following a 6.6-hour exposure to 60 ppb O₃. Indeed, as was illustrated in Figure 6-1, the group mean FEV₁ responses at 60 ppb fall on a smooth intake dose-response curve for exposures between 40 and 120 ppb O₃. Furthermore, in a re-analysis of the 60 ppb square-wave data from Adams (2006a), Brown et al. (2008) found the mean effects on FEV₁ to be highly statistically significant (p <0.002) using several common statistical tests even after removal of 3 potential outliers. A statistically

¹ Adams (2006a, 2002) both provide data for an additional group of 30 healthy subjects that were exposed via facemask to 60 ppb (square-wave) O₃ for 6.6 hours with moderate exercise (\( \dot{V}_E = 23 \) L/min per m² BSA). These subjects are described on page 133 of Adams (2006a) and pages 747 and 761 of Adams (2002). The FEV₁ decrement may be somewhat increased due to a target \( \dot{V}_E \) of 23 L/min per m² BSA relative to other studies with which it is listed having the target \( \dot{V}_E \) of 20 L/min per m² BSA. Based on Adams (2003a, b, 2002) the facemask exposure is not expected to affect the FEV₁ responses relative to a chamber exposure.

² This group average FEV₁ response is for a set of subjects exposed via facemask to 60 ppb O₂, see page 133 of Adams (2006a).
significant increase in total respiratory symptoms at 60 ppb has only been reported by Adams (2006a) for a triangular exposure protocol at 5.6 hours and 6.6 hours relative to baseline (not FA). Although not statistically significant, there was a tendency for an increase in total symptoms and pain on deep inspiration following the 60 ppb exposures (triangular and square-wave) relative to those following both FA and 40 ppb exposures. The time-course and magnitude of FEV₁ responses at 40 ppb resemble those occurring during FA exposures (Adams, 2006a, 2002). In both of these studies, there was a tendency (not statistically significant) for a small increase in total symptoms and pain on deep inspiration following the 40 ppb exposures relative to those following FA. Taken together, the available evidence shows that detectable effects of O₃ on group mean FEV₁ persist down to 60 ppb, but not 40 ppb in young healthy adults exposed for 6.6 hours during moderate exercise. Although group mean FEV₁ responses at 60 ppb are relatively small (2-3% mean FEV₁ decrement), it should be emphasized that there is considerable intersubject variability, with some responsive individuals consistently experiencing larger than average FEV₁ responses.

In addition to overt effects of O₃ exposure on the large airways indicated by spirometric responses, O₃ exposure also affects the function of the small airways and parenchymal lung. Foster et al. (1997); (1993) examined the effect of O₃ on ventilation distribution. In healthy adult males (n = 6; and 26.7 ± 7 years old) exposed to O₃ (330 ppb with light intermittent exercise for 2 hours), there was a significant reduction in ventilation to the lower lung (31% of lung volume) and significant increases in ventilation to the upper- and middle-lung regions (Foster et al., 1993). In a subsequent study of healthy males (n = 15; and 25.4 ± 2 years old) exposed to O₃ (350 ppb with moderate intermittent exercise for 2.2 hours), O₃ exposure caused a delayed gas washout in addition to a 14% FEV₁ decrement (Foster et al., 1997). The pronounced slow phase of gas washout following O₃ exposure represented a 24% decrease in the washout rate. A day following O₃ exposure, 50% of the subjects still had (or developed) a delayed washout relative to the pre-O₃ maneuver. These studies suggest a prolonged O₃ effect on the small airways and ventilation distribution in healthy young individuals.

There is a rapid recovery of O₃-induced spirometric responses and symptoms; 40 to 65% recovery appears to occur within about 2 hours following exposure (Folinsbee and Hazucha, 1989). For example, following a 2-hour exposure to 400 ppb O₃ with intermittent exercise, Nightingale et al. (2000) observed a 13.5% mean decrement in FEV₁. By 3 hours postexposure, however, only a 2.7% FEV₁ decrement persisted. Partial recovery also occurs following cessation of exercise despite continued exposure to O₃ (Folinsbee et al., 1977) and at low O₃ concentrations during exposure (Hazucha et al., 1992). A slower recovery phase, especially after exposure to higher O₃ concentrations, may take at least 24 hours to complete (Folinsbee and Hazucha, 2000; Folinsbee et al., 1993). Repeated daily exposure studies at higher concentrations typically show that FEV₁ response to O₃ is enhanced on the second day of exposure. This enhanced response suggests a residual effect of the previous exposure, about 22 hours earlier, even though the pre-exposure spirometry may be the same as on the previous day. The absence of the enhanced response with
repeated exposure at lower O₃ concentrations may be the result of a more complete recovery or less damage to pulmonary tissues (Folinsbee et al., 1994).

**Predicted Responses in Healthy Subjects**

Studies analyzing large data sets (hundreds of subjects) provide better predictive ability of acute changes in FEV₁ at low levels of O₃ and \( \dot{V}_E \) than is possible via comparisons between smaller studies. A few such studies described in the 2006 O₃ AQCD (U.S. EPA, 2006b) analyzed FEV₁ responses in healthy young adults (18-35 years of age) recruited from the area around Chapel Hill, NC and exposed for 2 hours to O₃ concentrations of up to 400 ppb at rest or with intermittent exercise (McDonnell et al., 1997; Seal et al., 1996; Seal et al., 1993). McDonnell et al. (1999b) examined changes in respiratory symptoms with O₃ exposure in a subset of the Chapel Hill data. In general, these studies showed that FEV₁ and respiratory symptom responses increase with increasing O₃ concentration and \( \dot{V}_E \) and decrease with increasing subject age. More recent studies expand upon these analyses of FEV₁ responses to also include longer duration (up to 8 hours) studies and periods of recovery following exposure.

McDonnell et al. (2007) provided a nonlinear empirical model for predicting group average FEV₁ responses as a function of O₃ concentration, exposure time, \( \dot{V}_E \), and age of the exposed individual. The model predicts temporal dynamics of FEV₁ change in response to any set of O₃ exposure conditions that might reasonably be experienced in the ambient environment. The model substantially differs from earlier statistical models in that it effectively considers the concurrent processes of damage and repair, i.e., the model allows effects on FEV₁ to accumulate during exposure at the same time they are reduced due to the reversible nature of the effects. The model was based on response data of healthy, nonsmoking, white males (n = 541), 18-35 years old, from 15 studies conducted at the U.S. EPA Human Studies Facility in Chapel Hill, NC.

McDonnell et al. (2010) tested the predictive ability of the model (McDonnell et al., 2007) against independent data (i.e., data that were not used to fit the model) of Adams (2006a, b, 2003a, 2002, 2000), Hazucha et al. (1992), and Schelegle et al. (2009). The model generally captured the dynamics of group average FEV₁ responses within about a one percentage point of the experimental data. Consistent with Bennett et al. (2007), an increased body mass index (BMI) was found to be associated with enhanced FEV₁ responses to O₃ by McDonnell et al. (2010). The BMI effect is of the same order of magnitude but in the opposite direction of the age effect whereby FEV₁ responses diminish with increasing age. Although the effects of age and BMI are relatively strong, these characteristics account for only a small amount of the observed variability in individual responses.

Alternatively, Lefohn et al. (2010a) proposed that FEV₁ responses to O₃ exposure might be described by a cumulative integrated exposure index with a sigmoidal weighting function similar to the W126 used for predicting vegetation effects (see Section 9.5). The integrated exposure index is the sum of the hourly average O₃
concentrations times their respective weighing factors. Based on a limited number of studies, the authors assumed weighting factors ranged from near zero at 50 ppb up to approximately 1.0 for concentrations at ≥ 125 ppb. The concentrations of 60, 70, and 80 ppb correspond to the assumed weights of 0.14, 0.28, and 0.50, respectively, and apply only to the case of exposure during moderate exercise ($\dot{V}_E = 20 \text{ L/min per m}^2 \text{ BSA}$). Lefohn et al. (2010a) calculated the cumulative exposure index for the protocols used by Adams (2006a, 2003a) and Schelegle et al. (2009). They found statistically significant O$_3$ effects after 4 hours on FEV$_1$ at 105 ppb-hour based on Schelegle et al. (2009) and at 235 ppb-hour based on Adams (2006a, 2003a). Based on this analysis, the authors recommended a 5-hour accumulation period to protect against O$_3$ effects on lung function.

More recently the McDonnell et al. (2007) model, as well as a variant containing a response threshold (described in more detail below), was fit to a larger dataset consisting of the FEV$_1$ responses of 741 young healthy adults (104 F, 637 M; mean age 23.8 yrs) from 23 individual controlled exposure studies conducted in either Chapel Hill, NC or Davis, CA (McDonnell et al., 2012). Concentrations across individual studies ranged from 40 ppb to 400 ppb, activity level ranged from rest to heavy exercise, duration of exposure was from 2 to 7.6 hours, and some studies provided data during periods of recovery following exposure. The resulting empirical models can estimate the frequency distribution of individual responses for any exposure scenario as well as summary measures of the distribution such the mean or median response and the proportions of individuals with FEV$_1$ decrements > 10%, 15%, and 20%. Predictions were found to be close agreement with the experimental data. The responses of males and females were, on average, approximately equal when activity level was controlled by normalizing $\dot{V}_E$ to BSA. Thus, any effects of sex upon FEV$_1$ responses to O$_3$ exposure can be accounted for by utilizing $\dot{V}_E$/BSA. In this large dataset, the coefficient of BMI was not statistically significantly different from zero, although its magnitude was similar to that estimated by of the earlier study (McDonnell et al., 2010). The threshold model fit the experimental data better than the non-threshold model, particularly at the earliest time points of low concentration exposures.

Schelegle et al. (2012) also analyzed a large dataset with substantial overlap to that used by McDonnell et al. (2012). From an initial dataset consisting of the FEV$_1$ responses of 704 young healthy adults (76 F, 628 M; mean age 23.8 yrs) from 21 individual controlled exposure studies conducted in either Chapel Hill, NC or Davis, CA, their model was fit to the FEV$_1$ responses of 220 young healthy adults (51 F, mean age 22 yrs; 169 M, mean age 24 yrs). Eighty-one of the excluded individuals appeared to be the result of inherent variability of repeated FEV$_1$ measurements in certain individuals and were present in both FA and O$_3$ exposure protocols. The resulting model unrealistically overestimated the FEV$_1$ responses of 11 individuals that participated in short-duration exposures (2.5 hours) with heavy exercise ($\dot{V}_E = 35 \text{ L/min per BSA}$) and high O$_3$ concentrations (240, 300, and 400 ppb). However, in general, for most exposure scenarios, the authors concluded that their model coefficients based on 220 individuals reliably predicted the mean FEV$_1$ decrements for the full dataset of 704 individuals.
Both McDonnell et al. (2012) and Schelegle et al. (2012) developed two compartment models that considered a dose of onset in response or a threshold of response. The first compartment in the McDonnell et al. (2012) model considers the level of oxidant stress in response to O₃ exposure to increase over time as a function of dose rate (C×\(\dot{V}_E\)) and decrease by clearance or metabolization over time according to first order reaction kinetics. In the second compartment of the threshold model, once oxidant stress reaches some threshold level, the decrement in FEV₁ increases as a sigmoid-shaped function of the oxidant stress with age. In the Schelegle et al. (2012) model, a first compartment acts as a reservoir in which oxidant stress builds up until the dose of onset at which time it spills over into a second compartment. The second compartment is identical to the first compartment in McDonnell et al. (2012) model. The oxidant levels in the second compartment were multiplied by a responsiveness coefficient to predict FEV₁ responses for the Schelegle et al. (2012) model.

Exposures predicted to not reach threshold in the McDonnell et al. (2012) dataset were those with moderate, near continuous exercise for 1 hour to 60 and 80 ppb O₃, and for 2 hours to 40 ppb; and those at rest for 1 hour to 180 and 240 ppb O₃, and for 2 hours to 120 ppb O₃. However, there were also exposures above the threshold having small predicted responses due to the sigmoid shape of the exposure-response function. Schelegle et al. (2012) reported an average predicted dose of onset in response was 1,080 μg O₃. For a prolonged (6.6 hours) O₃ exposure with moderate quasi continuous exercise (\(\dot{V}_E = 20\) L/min per BSA), this dose of onset would not be reached until between 4-5 hrs of exposure to 60 ppb or 3-4 hrs of exposure to 80 ppb. However, 14% of the individuals in the Schelegle et al. (2012) study had a dose of onset of less than 400 μg O₃. More consistent with the threshold in response reported by McDonnell et al. (2012), this dose of onset (i.e., 400 μg O₃) would be reached in 1-2 hrs of exposure to 50-80 ppb O₃ with moderate quasi continuous exercise.

**Intersubject Variability in Response of Healthy Subjects**

Consideration of group mean changes is important in discerning if observed effects are due to O₃ exposure rather than chance alone. Inter-individual variability in responses is, however, considerable and pertinent to assessing the fraction of the population that might actually be affected during an O₃ exposure. Hackney et al. (1975) first recognized a wide range in the sensitivity of subjects to O₃. The range in the subjects’ ages (29 to 49 years) and smoking status (0 to 50 pack years) in the Hackney et al. (1975) study are now understood to affect the spirometric and symptomatic responses to O₃. Subsequently, DeLucia and Adams (1977) examined responses to O₃ in six healthy non-smokers and found that two exhibited notably greater sensitivity to O₃. Since that time, numerous studies have documented considerable variability in responsiveness to O₃ even in subjects recruited to assure homogeneity in factors recognized or presumed to affect responses.

An individual’s FEV₁ response to a 2 hour O₃ exposure is generally reproducible over several months and presumably reflects the intrinsic responsiveness of the individual to O₃ (Hazucha et al., 2003; McDonnell et al., 1985c). The frequency
distribution of individual \( FEV_1 \) responses following these relatively short exposures becomes skewed as the group mean response increases, with some individuals experiencing large reductions in \( FEV_1 \) (Weinmann et al., 1995a; Kulle et al., 1985). For 2-hour exposures with intermittent exercise causing a predicted average \( FEV_1 \) decrement of 10%, individual decrements ranged from approximately 0 to 40% in white males aged 18-36 years (McDonnell et al., 1997). For an average \( FEV_1 \) decrement of 13%, Ultman et al. (2004) reported \( FEV_1 \) responses ranging from a 4% improvement to a 56% decrement in young healthy adults (32 M, 28 F) exposed for 1 hour to 250 ppb \( O_3 \). One-third of the subjects had \( FEV_1 \) decrements of >15%, and 7% of the subjects had decrements of >40%. The differences in \( FEV_1 \) responses did not appear to be explained by intersubject differences in the fraction of inhaled \( O_3 \) retained in the lung (Ultman et al., 2004).

Note: During each hour of the exposures, subjects were engaged in moderate quasi continuous exercise (40 L/min) for 50 minutes and rest for 10 minutes. Following the third hour, subjects had an additional 35 minute rest period for lunch. Subjects were exposed to a triangular \( O_3 \) concentration profile having the average \( O_3 \) concentration provided in each panel. As average \( O_3 \) concentration increased, the distribution of responses became asymmetric with a few individuals exhibiting large \( FEV_1 \) decrements. The percentage indicated in each panel is the portion of subjects having a \( FEV_1 \) decrement in excess of 10%.

Source: Adapted with permission of American Thoracic Society (Schelegle et al., 2009).

Figure 6-2 Frequency distributions of \( FEV_1 \) decrements observed by Schelegle et al. (2009) in young healthy adults (16 F, 15 M) following 6.6-hour exposures to \( O_3 \) or filtered air.

Consistent with the 1- to 2-hour studies, the distribution of individual responses following 6.6-hour exposures becomes skewed with increasing \( O_3 \) exposure concentration and magnitude of the group mean \( FEV_1 \) response (McDonnell, 1996). Figure 6-2 illustrates frequency distributions of individual \( FEV_1 \) responses observed in 31 young healthy adults following 6.6-hour exposures between 0 and 80 ppb \( O_3 \). Schelegle et al. (2009) found >10% \( FEV_1 \) decrements in 16, 19, 29, and 42% of individuals exposed for 6.6 hours to 60, 70, 80, and 87 ppb \( O_3 \), respectively. Just as there are differences in mean decrements between studies having similar exposure scenarios (Figure 6-1 at 80 and 120 ppb), there are differences in the proportion of individuals affected with >10% \( FEV_1 \) decrements. At 80 ppb \( O_3 \), the proportion affected with >10% \( FEV_1 \) decrements was 17% (n = 30) by Adams (2006a)\(^1\), 26% Not assessed by Adams (2006a), the proportion was provided in Figure 8-1B of the 2006 \( O_3 \) AQCD (U.S. EPA, 2006b).
(n = 60) by McDonnell (1996), and 29% (n = 31) by Schelegle et al. (2009).
At 60 ppb O₃, the proportion with >10% FEV₁ decrements was 20% (n = 30) by Adams (2002)¹, 3% (n = 30) by Adams (2006a)¹, 16% (n = 31) by Schelegle et al. (2009), and 5% (n = 59) by Kim et al. (2011). Based on these studies, the weighted average proportion of individuals with >10% FEV₁ decrements is 10% following exposure to 60 ppb O₃ and 25% following exposure to 80 ppb O₃. Due to limited data within the published papers, these proportions were not corrected for responses to FA exposure during which lung function typically improves in healthy adults. For example, uncorrected versus O₃-induced (i.e., adjusted for response during FA exposure) proportions of individuals having >10% FEV₁ decrements in the Adams (2006a)² study were, respectively, 3% versus 7% at 60 ppb and 17% versus 23% at 80 ppb. Thus, uncorrected proportions may underestimate the actual fraction of healthy individuals affected in some studies.

In addition to examining individual responses on a study-by-study basis, the recently published McDonnell et al. (2012) model can also be utilized to directly calculate the proportion of individuals expected to experience O₃-induced (i.e., adjusted for response during FA exposure) FEV₁ decrements of a given magnitude under a variety of exposure conditions and demographic characteristics. This model was fit to the data of young healthy adults (104 F, 637 M; 18-36 yrs of age) that participated in controlled O₃ exposure studies conducted in Chapel Hill, NC and Davis, CA. Figure 6-3 illustrates the proportions of individuals predicted to have greater than 10%, 15%, and 20% O₃-induced FEV₁ decrements following a 6.6 hour exposure to O₃ with moderate exercise. Consistent with the observed responses of individual studies cited above, the model predicts that >10% FEV₁ decrements occur in 9% of the individuals exposed to 60 ppb and 22% of those exposed to 80 ppb O₃.

¹ This information is from page 761 of Adams (2002). Adams (2006a, 2002) both provide data for a group of 30 healthy subjects that were exposed via facemask to 60 ppb (square-wave) O₃ for 6.6 hours with moderate exercise (V̇E = 23 L/min per m² BSA). These subjects are described on page 133 of Adams (2006a) and pages 747 and 761 of Adams (2002). The FEV₁ decrement may be somewhat increased due to a target V̇E of 23 L/min per m² BSA relative to other studies with which it is listed having the target V̇E of 20 L/min per m² BSA. Based on Adams (2003a, b, 2002), similar FEV₁ responses are expected between facemask and chamber exposures.

² Not assessed by Adams (2006a), uncorrected and O₃-induced proportions are from Figures 8-1B and 8-2, respectively, of the 2006 O₃ AQCD (U.S. EPA, 2006b).
Note: Predictions based threshold model of McDonnell et al. (2012) for healthy 23.8 year old adults exposed to a constant concentration of O₃ for 6.6 hrs. During each hour of the exposures, subjects were presumed engaged in moderate quasi continuous exercise (20 L/min/BSA) for 50 minutes and rest for 10 minutes. Following the third hour, subjects had an additional 35 minute rest period for lunch.

Source: Adapted from McDonnell et al. (2012).

**Figure 6-3** Proportion of individuals predicted to have greater than 10%, 15%, and 20% O₃-induced FEV₁ decrements following 6.6-hour exposure to O₃ with moderate exercise.

Given considerable inter-individual variability in responses, the interpretation of biologically small group mean decrements requires careful consideration. Following prolonged 6.6-hour exposures to an average level of 60 ppb O₃, data available from four studies yield a weighted-average group mean O₃-induced FEV₁ decrement (i.e., adjusted for FA responses) of 2.7% (n = 150) (Kim et al., 2011; Schelegle et al., 2009; Adams, 2006a, 1998). The data from these studies also yield a weighted-average proportion (uncorrected for FA responses) of subjects with >10% FEV₁ decrements of 10% (n = 150) (Kim et al., 2011; Schelegle et al., 2009; Adams, 2006a, 1998). In an individual with relatively “normal” lung function, with recognition of the technical and biological variability in measurements, confidence can be given that within-day changes in FEV₁ of ≥ 5% are clinically meaningful (Pellegrino et al., 2005; ATS, 1991). Here focus is given to individuals with >10% decrements in FEV₁ since some individuals in the Schelegle et al. (2009) study experienced 5-10% FEV₁ decrements following exposure to FA. A 10% FEV₁ decrement is also generally accepted as an abnormal response and a reasonable criterion for assessing exercise-induced bronchoconstriction (Dryden et al., 2010; ATS, 2000a). The data are not available in the published papers to determine the O₃-induced proportion for either the Adams (1998) or Schelegle et al. (2009) studies. As already stated, however, this uncorrected proportion likely underestimates the actual proportion of healthy individuals experiencing O₃-induced FEV₁ decrements.
in excess of 10%. Therefore, by considering uncorrected responses and those individuals having >10% decrements, 10% is an underestimate of the proportion of healthy individuals that are likely to experience clinically meaningful changes in lung function following exposure for 6.6 hours to 60 ppb O₃ during moderate exercise. Of the studies conducted at 60 ppb, only Kim et al. (2011) reported FEV₁ decrements at 60 ppb to be statistically significant. However, Brown et al. (2008) found those from Adams (2006a) to be highly statistically significant. Though group mean decrements are biologically small and generally do not attain statistical significance, a considerable fraction of exposed individuals experience clinically meaningful decrements in lung function.

Factors Modifying Responsiveness to Ozone

Physical activity increases $\dot{V}_E$ and therefore the dose of inhaled O₃. Consequently, the intensity of physiological response during and following an acute O₃ exposure will be strongly associated with $\dot{V}_E$. Apart from inhaled O₃ dose and related environmental factors (e.g., repeated daily exposures), individual-level factors, such as health status, age, sex, race/ethnicity, race, smoking habit, diet, and socioeconomic status (SES) have been considered as potential modifiers of a physiologic response to such exposures.

Responses in Individuals with Pre-existing Disease

Individuals with respiratory disease are of primary concern in evaluating the health effects of O₃ because a given change in function is likely to have more impact on a person with pre-existing function impairment and reduced reserve.

Possibly due to the age of subjects studied, patients with COPD performing light to moderate exercise do not generally experience statistically significant pulmonary function decrements following 1- and 2-hour exposures to ≤ 300 ppb O₃ (Kehrl et al., 1985; Linn et al., 1983; Linn et al., 1982a; Solic et al., 1982). Following a 4-hour exposure to 240 ppb O₃ during exercise, Gong et al. (1997b) found an O₃-induced FEV₁ decrement of 8% in COPD patients which was not statistically different from the decrement of 3% in healthy subjects. Demonstrating the need for control exposures and the presumed effect of exercise, four of the patients in the Gong et al. (1997b) study had FEV₁ decrements of >14% following both the FA and O₃ exposures. Although the clinical significance is uncertain, small transient decreases in arterial blood oxygen saturation have also been observed in some of these studies.

Based on studies reviewed in the 1996 and 2006 O₃ AQCDs, subjects with asthma appear to be at least as sensitive to acute effects of O₃ as healthy subjects. Horstman et al. (1995) found the O₃-induced FEV₁ decrement in 17 subjects with mild-to-moderate asthma to be significantly larger than that in 13 healthy subjects (19% versus 10%, respectively) exposed to 160 ppb O₃ during light exercise ($\dot{V}_E$ of 15 L/min per m² BSA) for a 7.6-hour exposure. In subjects with asthma, a significant positive correlation between O₃-induced spirometric responses and baseline lung
function was observed, i.e., responses increased with severity of disease. In the shorter duration study by Kreit et al. (1989), 9 subjects with asthma also showed a considerably larger average O3-induced FEV1 decrement than 9 healthy controls (25% versus 16%, respectively) following exposure to 400 ppb O3 for 2 hours with moderate-heavy exercise (\( \dot{V}_E = 54 \) L/min). Alexis et al. (2000) [400 ppb; 2 hours; exercise, \( \dot{V}_E = 30 \) L/min] and Jorres et al. (1996) [250 ppb; 3 hours; exercise, \( \dot{V}_E = 30 \) L/min] reported a tendency for slightly greater FEV1 decrements in subjects with asthma than healthy subjects. Several studies reported similar responses between individuals with asthma and healthy individuals (Scannell et al., 1996; Hiltermann et al., 1995; Basha et al., 1994). The lack of differences in the Hiltermann et al. (1995) [400 ppb; 2 hours; exercise, \( \dot{V}_E = 20 \) L/min] and Basha et al. (1994) [200 ppb; 6 hours; exercise, \( \dot{V}_E = 25 \) L/min] studies was not surprising, however, given extremely small sample sizes (5-6 subjects per group) and corresponding lack of statistical power. Power was not likely problematic for Scannell et al. (1996) [200 ppb; 4 hours; exercise, \( \dot{V}_E \approx 44 \) L/min] with 18 subjects with mild asthma and 81 age-matched healthy controls from companion studies (Balmes et al., 1996; Aris et al., 1995). Of note, Mudway et al. (2001) reported a tendency for subjects with asthma to have smaller O3-induced FEV1 decrements than healthy subjects (3% versus 8%, respectively) when exposed to 200 ppb O3 for 2 hours during exercise. However, the subjects with asthma in Mudway et al. (2001) also tended to be older than the healthy subjects, which could partially explain their smaller response since FEV1 responses to O3 diminish with age.

In a study published since the 2006 O3 AQCD, Stenfors et al. (2010) exposed subjects with persistent asthma (n = 13; aged 33 years) receiving chronic inhaled corticosteroid therapy to 200 ppb O3 for 2 hours with moderate exercise. An average O3-induced FEV1 decrement of 8.4% was observed, whereas, only a 3.0% FEV1 decrement is predicted for similarly exposed age-matched healthy controls (McDonnell et al., 2007). Vagaggini et al. (2010) exposed subjects with mild-to-moderate asthma (n = 23; 33 ± 11 years) to 300 ppb O3 for 2 hours with moderate exercise. Although the group mean O3-induced FEV1 decrement was only 4%, eight subjects were categorized as “responders” with >10% FEV1 decrements. Baseline lung function did not differ between the responders and nonresponders suggesting that, in contrast to Horstman et al. (1995), O3-induced FEV1 responses were not associated with disease severity.

**Lifestyle**

Children, adolescents, and young adults (<18 years of age) appear, on average, to have nearly equivalent spirometric responses to O3, but have greater responses than middle-aged and older adults when similarly exposed to O3 (U.S. EPA, 1996a). Symptomatic responses to O3 exposure, however, appear to increase with age until early adulthood and then gradually decrease with increasing age (U.S. EPA, 1996a). For example, healthy children (n=22; mean age 10 yrs) exposed to FA and 120 ppb O3 (2.5 hours; heavy intermittent exercise, \( \dot{V}_E=32-35 \) L/min per m² BSA) experienced similar spirometric responses, but lesser symptoms than similarly exposed young healthy adults (n=21-22; mean age 22 yrs)(McDonnell et al., 1985a).
For subjects aged 18-36 years, McDonnell et al. (1999b) reported that symptom responses from O₃ exposure also decrease with increasing age. Diminished symptomatic responses in children and the elderly might put these groups at increased risk for continued O₃ exposure, i.e., a lack of symptoms may result in their not avoiding or ceasing exposure. Once lung growth and development reaches the peak (18-20 years of age in females and early twenties in males), pulmonary function, which is at its maximum as well, begins to decline progressively with age as does O₃ sensitivity.

A couple analyses of large datasets have assessed the effects of age on responsiveness to O₃ exposure. McDonnell et al. (2007) found O₃-induced FEV₁ responses to decrease significantly with increasing age in an analysis of data from studies of healthy, nonsmoking, white males (n = 541), 18-35 years old (mean age 24.1), conducted in Chapel Hill, NC. Using this same dataset, McDonnell et al. (2010) reported that O₃-induced FEV₁ responses, while decreasing significantly with age, also increased significantly with BMI. In a larger dataset of 741 young healthy adults (104 F, 637 M; mean age 23.8 yrs) from studies conducted in either Chapel Hill, NC or Davis, CA, McDonnell et al. (2012) did not find a statistically significant effect of either age or BMI on the FEV₁ responses. Analysis of the Davis data alone showed a tendency for increases in O₃-induced FEV₁ responses with increases in both age and BMI, whereas FEV₁ responses in the Chapel Hill data decreased with age and increased with BMI. The authors speculated that the lack of a significant age effect may be, in part, due to a significant correlation (r = 0.23) between age and BMI in the 142 subjects from studies conducted at Davis. No correlation (r = 0.03) between age and BMI was observed in the Chapel Hill data.

In healthy individuals, the fastest rate of decline in O₃ responsiveness appears between the ages of 18 and 35 years (Passannante et al., 1998; Seal et al., 1996), more so for females than males (Hazucha et al., 2003). During the middle age period (35-55 years), O₃ sensitivity continues to decline, but at a much lower rate. Beyond this age (>55 years), acute O₃ exposure elicits minimal spirometric changes. Whether the same age-dependent pattern of O₃ sensitivity decline also holds for nonspirometric pulmonary function, airway reactivity or inflammatory endpoints has not been determined. Although there is considerable evidence that spirometric and symptomatic responses to O₃ exposure decrease with age beyond young adulthood, this evidence comes from cross-sectional analyses and has not been confirmed by longitudinal studies of the same individuals.

**Sex**

Several studies have suggested that physiological differences between sexes may predispose females to greater O₃-induced health effects. In females, lower plasma and nasal lavage fluid (NLF) levels of uric acid (the most prevalent antioxidant), the initial defense mechanism of O₃ neutralization in airway surface liquid, may be a contributing factor (Housley et al., 1996). Consequently, reduced absorption of O₃ in the upper airways may promote its deeper penetration. Dosimetric measurements have shown that the absorption distribution of O₃ is independent of sex when
absorption is normalized to anatomical dead space (Bush et al., 1996). Thus, a sex-related differential removal of O₃ by uric acid seems to be minimal. In general, the physiologic response of young healthy females to O₃ exposure appears comparable to the response of young males (Hazucha et al., 2003). Based on analysis of a large dataset (104 F, 637 M; 18-36 yrs of age), FEV₁ responses of males and females to O₃ exposure were, on average, approximately equal when activity level was controlled by normalizing \( \dot{V}_E \) to BSA (McDonnell et al., 2012). Additionally, with activity level assessed as \( \dot{V}_E/\text{BSA} \), Schelegle et al. (2012) observed no significant difference in the average dose to onset of responses between males and females.

Several studies have investigated the effects of the menstrual cycle on responses to O₃ in healthy young women. In a study of 9 women exposed during exercise to 300 ppb O₃ for an hour, Fox et al. (1993) found lung function responses to O₃ significantly enhanced during the follicular phase relative to the luteal phase. However, Weinmann et al. (1995c) found no difference in responses between the follicular and luteal phases as well as no significant differences between 12 males and 12 females exposed during exercise to 350 ppb O₃ for 2.15 hours. Seal et al. (1996) also reported no effect of menstrual cycle phase in their analysis of responses of 150 women (n = 25 per exposure group; 0, 120, 240, 300, and 400 ppb O₃). Seal et al. (1996) conceded that the methods used by Fox et al. (1993) more precisely defined menstrual cycle phase.

**Race/Ethnicity**

Only two controlled human exposure studies have assessed differences in lung function responses between races. Seal et al. (1993) compared lung function responses of whites (93 M, 94 F) and blacks (undefined ancestry; 92 M, 93 F) exposed to a range of O₃ concentrations (0-400 ppb). The main effects of the sex-race group and O₃ concentration were statistically significant (both at p <0.001), although the interaction between sex-race group and O₃ concentration was not significant (p = 0.13). These findings indicate some overall difference between the sex-race groups that is independent of O₃ concentration, i.e., the concentration-response (C-R) curves for the four sex-race groups are parallel. In a multiple comparison procedure on data collapsed across all O₃ concentrations for each sex-race group, both black men and black women had significantly larger decrements in FEV₁ than did white men. The authors noted that the O₃ dose per unit of lung tissue would be greater in blacks and females than whites and males, respectively. It cannot be ruled out that this difference in tissue dose might have affected responses to O₃.

The college students recruited for the Seal et al. (1993) study were noted by the authors as probably being from better educated and SES advantaged families, thus reducing the potential influence of these variables on results. In a follow-up analysis, Seal et al. (1996) reported that, of three SES categories, individuals in the middle SES category showed greater concentration-dependent decline in percent-predicted FEV₁ (4.5% at 400 ppb O₃) than low and high SES groups. The authors did not have an “immediately clear” explanation for this finding.
More recently, Que et al. (2011) assessed pulmonary responses in blacks of African American ancestry (22 M, 24 F) and Caucasians (55 M, 28 F) exposed to 220 ppb O₃ for 2.25 hours (alternating 15 min periods of rest and brisk treadmill walking). On average, the black males experienced a 16.8% decrement in FEV₁ following O₃ exposure which was significantly larger than mean FEV₁ decrements of 6.2, 7.9, and 8.3% in black females and Caucasian males and Caucasian females, respectively. In the study by Seal et al. (1993), there was potential that the increased FEV₁ decrements in blacks relative to whites were due to increased O₃ tissue doses since exercise rates were normalized to BSA. Differences in O₃ tissue doses between the races should not have occurred in the Que et al. (2011) study because exercise rates were normalized to lung volume (viz., 6-8 times FVC). Thus, the increased mean FEV₁ decrement in black males is not likely attributable to systematically larger O₃ tissue doses in blacks relative to whites.

**Smoking**

Smokers are less responsive to O₃ for some (but not all) health endpoints than nonsmokers. Spirometric and plethysmographic pulmonary function decline, respiratory symptoms, and nonspecific airway hyperreactivity of smokers to O₃ were all weaker than data reported for nonsmokers. However, the time course of development and recovery of these effects as well their reproducibility in smokers were not different from nonsmokers (Frampton et al., 1997a). Another similarity between smokers and nonsmokers is that, the inflammatory response to O₃ does not appear to depend on smoking status or the responsiveness of individuals to changes in lung function (Torres et al., 1997). Chronic airway inflammation with desensitization of bronchial nerve endings and an increased production of mucus may plausibly explain the reduced responses to O₃ in smokers relative to nonsmokers (Frampton et al., 1997a; Torres et al., 1997).

**Antioxidant supplementation**

The first line of defense against oxidative stress is antioxidants-rich ELF which scavenges free radicals and limits lipid peroxidation. Exposure to O₃ depletes the antioxidant level in nasal ELF probably due to scrubbing of O₃ (Mudway et al., 1999a); however, the concentration and the activity of antioxidant enzymes either in ELF or plasma do not appear to be related to O₃ responsiveness (Samet et al., 2001; Avissar et al., 2000; Blomberg et al., 1999). Carefully controlled studies of dietary antioxidant supplementation have demonstrated some protective effects of α-tocopherol and ascorbate on spirometric lung function from O₃ but not on the intensity of subjective symptoms or inflammatory response including cell recruitment, activation and a release of mediators (Samet et al., 2001; Trenga et al., 2001). Dietary antioxidants have also been reported to attenuate O₃-induced bronchial hyperresponsiveness in asthmatics (Trenga et al., 2001).

**Genetic polymorphisms**

Some studies (e.g., Corradi et al., 2002; Bergamaschi et al., 2001) reviewed in the 2006 O₃ AQCD reported that genetic polymorphisms of antioxidant enzymes may
modulate pulmonary function and inflammatory response to O₃ challenge. It was suggested that healthy carriers of NAD(P)H:quinone oxidoreductase wild type (NQO1wt) in combination with GSTM1 null were more responsive to O₃. Bergamaschi et al. (2001) reported that subjects having NQO1wt and GSTM1 null genotypes had increased O₃ responsiveness (FEV₁ decrements and epithelial permeability), whereas subjects with other combinations of these genotypes were less affected. A subsequent study from the same laboratory reported a positive association between O₃ responsiveness, as characterized by the level of oxidative stress and inflammatory mediators (8-isoprostane, LTB₄ and TBARS) in exhaled breath condensate and the NQO1wt and GSTM1null genotypes (Corradi et al., 2002). However, none of the spirometric endpoints (e.g., FEV₁) were affected by O₃ exposure.

In a controlled exposure of subjects with mild-to-moderate asthma (n = 23; 33 ± 11 years) to 300 ppb O₃ for 2 hours with moderate exercise, Vagaggini et al. (2010) found that six of the subjects had a NQO1wt and GSTM1 null, but this genotype was not associated with the changes in lung function or inflammatory responses to O₃. Kim et al. (2011) also recently reported that GSTM1 genotype was not predictive of FEV₁ responses to O₃ in young healthy adults (32 F, 27 M; 25.0 ± 0.5 year) who were roughly half GSTM1-null and half GSTM1-sufficient. Sputum neutrophil levels, measured in a subset of the subjects (13 F, 11 M), were also not significantly associated with GSTM1 genotype.

In a study of healthy volunteers with GSTM1 sufficient (n = 19; 24 ± 3) and GSTM1 null (n = 16; 25 ± 5) genotypes exposed to 400 ppb O₃ for 2 hours with exercise, Alexis et al. (2009) found that inflammatory responses but not lung function responses to O₃ were dependent on genotype. At 4 hours post-O₃ exposure, both GSTM1 genotype groups had significant increases in sputum neutrophils with a tendency for a greater increase in GSTM1 sufficient than null subjects. At 24 hours postexposure, sputum neutrophils had returned to baseline levels in the GSTM1 sufficient individuals. In the GSTM1 null subjects, however, sputum neutrophil levels increased from 4 hours to 24 hours and were significantly greater than both baseline levels and levels at 24 hours in the GSTM1 sufficient individuals. Since there was no FA control in the Alexis et al. (2009) study, effects of the exposure other than O₃ itself cannot be ruled out. In general, the findings between studies are inconsistent.

**Body Mass Index**

In a retrospective analysis of data from 541 healthy, nonsmoking, white males between the ages of 18-35 years from 15 studies conducted at the U.S. EPA Human Studies Facility in Chapel Hill, NC, McDonnell et al. (2010) found that increased BMI was associated with enhanced FEV₁ responses to O₃. The BMI effect was of the same order of magnitude but in the opposite direction of the age effect whereby FEV₁ responses diminish with increasing age. In a similar retrospective analysis, Bennett et al. (2007) found enhanced FEV₁ decrements following O₃ exposure with increasing BMI in a group of 75 healthy, nonsmoking, women (age 24 ± 4 years;
BMI range 15.7 to 33.4), but not 122 healthy, nonsmoking, men (age 25 ± 4 years; BMI range 19.1 to 32.9). In the women, greater O₃-induced FEV₁ decrements were seen in overweight (BMI >25) than in normal weight (BMI from 18.5 to 25), and in normal weight than in underweight (BMI <18.5) (P trend ≤ 0.022). Together, these results indicate that higher BMI may be a risk factor for pulmonary effects associated with O₃ exposure.

**Repeated Ozone Exposure Effects**

The attenuation of responses observed after repeated consecutive O₃ exposures in controlled human exposure studies has also been referred to in the literature as “adaptation” or “tolerance” (e.g., Linn et al., 1988). In animal toxicology studies, however, the term tolerance has more classically been used to describe the phenomenon wherein a prior exposure to a low, nonlethal concentration of O₃ provides some protection against death and lung edema at a higher, normally lethal exposure concentration (see Section 9.3.5 of U.S. EPA, 1986). The term “attenuation” will be used herein to refer to the reduction in responses to O₃ observed with repeated O₃ exposures in controlled human exposure studies. Neither tolerance nor attenuation should be presumed to imply complete protection from the biological effects of inhaled O₃, because continuing injury still occurs despite the desensitization to some responses.

The attenuation of responses due to ambient O₃ exposure was first investigated by Hackney et al. (1977a; 1976). Experiencing frequent ambient O₃ exposures, Los Angeles residents were compared to groups having less ambient O₃ exposure. Following a controlled laboratory exposure to 370-400 ppb O₃ for 2 hours with light intermittent exercise (2-2.5 times resting \( \dot{V}_E \)), the Los Angeles residents exhibited minimal FEV₁ responses relative to groups having less ambient O₃ exposure. Subsequently, Linn et al. (1988) examined the seasonal variation in Los Angeles residents’ responses to O₃ exposure. A group of 8 responders (3M, 5F) and 9 nonresponders (4M, 5F) were exposed to 180 ppb O₃ for 2 hours with heavy intermittent exercise (\( \dot{V}_E = 35 \) L/min per m² BSA) on four occasions (spring, fall, winter, and the following spring). In responders, relative to the first spring exposures, FEV₁ responses were attenuated in the fall and winter, but returned to similar decrements the following spring. By comparison, the nonresponders, on average, showed no FEV₁ decrements on any of the four occasions. In subjects recruited regardless of FEV₁ responsiveness to O₃ from the area around Chapel Hill, NC, no seasonal effect of ambient O₃ exposure on FEV₁ responses following chamber exposures to O₃ has been observed (Hazucha et al., 2003; McDonnell et al., 1985c).

Based on studies reviewed in previous O₃ AQCDs, several conclusions can be drawn about repeated 1- to 2-hour O₃ exposures. Repeated exposures to O₃ causes enhanced (i.e., greater decrements) FVC and FEV₁ responses on the second day of exposure. The enhanced response appears to depend to some extent on the magnitude of the initial response (Horvath et al., 1981). Small responses to the first O₃ exposure are less likely to result in an enhanced response on the second day of O₃ exposure.
(Folinsbee et al., 1994). With continued daily exposures (i.e., beyond the second day) there is a substantial (or even total) attenuation of pulmonary function responses, typically on the third to fifth days of repeated O₃ exposure. This attenuation of responses is lost in 1 week (Kulle et al., 1982; Linn et al., 1982b) or perhaps 2 weeks (Horvath et al., 1981) without O₃ exposure. In temporal conjunction with pulmonary function changes, symptoms induced by O₃ (e.g., cough, pain on deep inspiration, and chest discomfort), are also increased on the second exposure day but are attenuated with repeated O₃ exposure thereafter (Folinsbee et al., 1995; Foxcroft and Adams, 1986; Linn et al., 1982b; Folinsbee et al., 1980). In longer-duration (4-6.6 hours), lower-concentration studies that do not cause an enhanced second-day response, the attenuation of response to O₃ appears to proceed more rapidly (Folinsbee et al., 1994).

Consistent with other investigators, Frank et al. (2001) found FVC and FEV₁ decrements to be significantly attenuated following four consecutive days of exposure to O₃ (250 ppb, 2 hours). However, the effects of O₃ on the small airways (assessed by a combined index of isovolumetric forced expiratory flow between 25 and 75% of vital capacity [FEF₂₅₋₇₅] and flows at 50% and 75% of FVC) showed a persistent functional reduction from Day 2 through Day 4. Notably, in contrast to FVC and FEV₁ which exhibited a recovery of function between days, there was a persistent effect of O₃ on small airways function such that the baseline function on Day 2 through Day 4 was depressed relative to Day 1. Frank et al. (2001) also found neutrophil (PMN) numbers in BAL remained significantly higher following O₃ (24 hours after last O₃ exposure) compared to FA. Markers from bronchoalveolar lavage fluid (BALF) following 4 consecutive days of both 2-hour (Devlin et al., 1997) and 4-hour (Jorres et al., 2000; Christian et al., 1998) exposures have indicated ongoing cellular damage irrespective of the attenuation of some cellular inflammatory responses of the airways, lung function and symptoms response. These data suggest that the persistent small airways dysfunction assessed by Frank et al. (2001) is likely induced by both neurogenic and inflammatory mediators, since the density of bronchial C-fibers is much lower in the small than large airways.

Summary of Controlled Human Exposure Studies on Lung Function

Responses in humans exposed to O₃ concentrations found in the ambient environment include: decreased inspiratory capacity; mild bronchoconstriction; rapid, shallow breathing pattern during exercise; and symptoms of cough and pain on deep inspiration (U.S. EPA, 2006b, 1996a). Discussed in subsequent Section 6.2.2.1 and Section 6.2.3.1, controlled exposure to O₃ also results in airway hyperresponsiveness, pulmonary inflammation, immune system activation, and epithelial injury (Que et al., 2011; Mudway and Kelly, 2004a). Reflex inhibition of inspiration results in a decrease in forced vital capacity and, in combination with mild bronchoconstriction, contributes to a decrease in the FEV₁. Healthy young adults exposed to O₃ concentrations ≥ 60 ppb develop statistically significant reversible, transient decrements in lung function and symptoms of breathing discomfort if minute ventilation or duration of exposure is increased sufficiently.
With repeated O₃ exposures over several days, FEV₁ and symptom responses become attenuated in both healthy individuals and asthmatics, but this attenuation of responses is lost after about a week without exposure (Gong et al., 1997a; Folinsbee et al., 1994; Kulle et al., 1982). In contrast to the attenuation of FEV₁ responses, there appear to be persistent O₃ effects on small airways function as well as ongoing cellular damage during repeated exposures.

There is a large degree of intersubject variability in lung function decrements (McDonnell, 1996). However, these lung function responses tend to be reproducible within a given individual over a period of several months indicating differences in the intrinsic responsiveness of individuals (Hazucha et al., 2003; McDonnell et al., 1985c). In healthy young adults, O₃-induced decrements in FEV₁ do not appear to depend on sex (Hazucha et al., 2003), body surface area or height (McDonnell et al., 1997), lung size or baseline FVC (Messineo and Adams, 1990). There is limited evidence that blacks may experience greater O₃-induced decrements in FEV₁ than age-matched whites (Que et al., 2011; Seal et al., 1993). Healthy children experience similar spirometric responses but lesser symptoms from O₃ exposure relative to young adults (McDonnell et al., 1985b). On average, spirometric and symptom responses to O₃ exposure appear to decline with increasing age beyond about 18 years of age (McDonnell et al., 1999b; Seal et al., 1996). There is a tendency for slightly increased spirometric responses in individuals with mild asthma and allergic rhinitis relative to healthy young adults (Jorres et al., 1996). Spirometric responses in subjects with asthma appear to be affected by baseline lung function, i.e., responses increase with disease severity (Horstman et al., 1995).

Available information on recovery of lung function following O₃ exposure indicates that an initial phase of recovery in healthy individuals proceeds relatively rapidly, with acute spirometric and symptom responses resolving within about 2 to 4 hours (Folinsbee and Hazucha, 1989). Small residual lung function effects are almost completely resolved within 24 hours. One day following O₃ exposure, persistent effects on the small airways assessed by decrements in FEF₂₅₋₇₅ and altered ventilation distribution have been reported (Frank et al., 2001; Foster et al., 1997).

### 6.2.1.2 Epidemiology

The O₃-induced lung function decrements consistently demonstrated in controlled human exposure studies (Section 6.2.1.1) provide biological plausibility for the epidemiologic evidence consistently linking short-term increases in ambient O₃ concentration with lung function decrements in diverse populations. In the 1996 and 2006 O₃ AQCDs, coherence with controlled human exposure study results was found not only for epidemiologic associations observed in groups with expected higher ambient O₃ exposures and higher exertion levels, including children attending summer camps and adults exercising or working outdoors, but also for associations observed in children and individuals with asthma (U.S. EPA, 2006b, 1996a). Recent
epidemiologic studies focused more on children with asthma rather than groups with increased outdoor exposures or other healthy populations. Whereas recent studies contributed less consistent evidence, the cumulative body of evidence indicates decreases in lung function in association with increases in ambient O$_3$ concentration in children with asthma. Collectively, studies in adults with asthma and individuals without asthma found both O$_3$-associated decreases and increases in lung function. Recent studies did provide additional data to assess whether particular lags of O$_3$ exposure were more strongly associated with decrements in lung function; whether O$_3$ associations were confounded by copollutant exposures; and whether associations were modified by factors such as corticosteroid (CS) use, genetic polymorphisms, and elevated BMI.

**Populations with Increased Outdoor Exposures**

Epidemiologic studies primarily use ambient O$_3$ concentrations to represent exposure; however, few studies have accounted for time spent outdoors, which has been shown to influence the relationship between ambient concentrations and individual exposures to O$_3$ (Section 4.3.3). Epidemiologic studies of individuals engaged in outdoor recreation, exercise, or work are noteworthy for the likely greater extent to which ambient O$_3$ concentrations represent ambient O$_3$ exposures. Ambient O$_3$ concentrations, locations, and time periods for epidemiologic studies of populations with increased outdoor exposures are presented in Table 6-2. Most of these studies measured ambient O$_3$ at the site of subjects’ outdoor activity and related lung function changes to the O$_3$ concentrations measured during outdoor activity, which have contributed to higher O$_3$ personal exposure-ambient concentration correlations and ratios (Section 4.3.3). Because of improved O$_3$ exposure estimates, measurement of lung function before and after discrete periods of activity, and examination of O$_3$ effects during exertion when the dose of O$_3$ reaching the lungs may be higher due to higher ventilation and inhalation of larger volumes of air, epidemiologic studies of populations with increased outdoor exposures are more comparable to controlled human exposure studies. The collective body of epidemiologic evidence clearly demonstrates decrements in lung function in association with increases in ambient O$_3$ exposure during outdoor activity (Figure 6-4 [and Table 6-3], Figure 6-5 [and Table 6-4], Figure 6-6 [and Table 6-5]. Expanding upon findings from controlled human exposure studies, these epidemiologic studies provide strong evidence for respiratory effects in children and adults related to ambient O$_3$ exposure.
Table 6-2  Mean and upper percentile O₃ concentrations in epidemiologic studies of lung function in populations with increased outdoor exposures.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O₃ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Berry et al. (1991)</td>
<td>Mercer County, NJ</td>
<td>July 1988</td>
<td>1-h max(^a)</td>
<td>NR</td>
<td>Max: 204</td>
</tr>
<tr>
<td>Avol et al. (1990)</td>
<td>Idyllwild, CA</td>
<td>June-August 1988</td>
<td>1-h avg(^b)</td>
<td>94</td>
<td>Max: 161</td>
</tr>
<tr>
<td>Burnett et al. (1990)</td>
<td>Lake Couchiching, Ontario, Canada</td>
<td>June-July 1983</td>
<td>1-h avg(^b)</td>
<td>59</td>
<td>Max: 95</td>
</tr>
<tr>
<td>Higgins et al. (1990)</td>
<td>San Bernardino, CA</td>
<td>June-July 1987</td>
<td>1-h avg(^b)</td>
<td>123</td>
<td>Max: 245</td>
</tr>
<tr>
<td>Raizenne et al. (1989)</td>
<td>Lake Erie, Ontario, Canada</td>
<td>June-August 1986</td>
<td>1-h avg(^b)</td>
<td>71</td>
<td>Max: 143</td>
</tr>
<tr>
<td>Spektor et al. (1988a)</td>
<td>Fairview Lake, NJ</td>
<td>July-August 1984</td>
<td>1-h avg(^b)</td>
<td>53</td>
<td>Max: 113</td>
</tr>
<tr>
<td>Neas et al. (1999)</td>
<td>Philadelphia, PA</td>
<td>July-September 1993</td>
<td>12-h avg(^a) (9 a.m. - 9 p.m.)</td>
<td>57.5 (near Camp 1) 55.9 (near Camp 2)</td>
<td>Max (near Camp 1): 106</td>
</tr>
<tr>
<td>Nickmilder et al. (2007)</td>
<td>Southern Belgium</td>
<td>July-August 2002</td>
<td>1-h max</td>
<td>NR</td>
<td>Max (across 6 camps): 24.6-112.8(^c)</td>
</tr>
<tr>
<td>Hoppe et al. (2003)</td>
<td>Munich, Germany</td>
<td>Summers 1991, 1992</td>
<td>30-min max (1-4 p.m.)</td>
<td>High O₃ days: 62.1 Control O₃ days: 26.6</td>
<td>Max (overall): 82</td>
</tr>
<tr>
<td>Spektor et al. (1988b)</td>
<td>Tuxedo, NY</td>
<td>June-August 1985</td>
<td>Exercise-time avg (15 - 55 min)</td>
<td>NR</td>
<td>Max: 124</td>
</tr>
<tr>
<td>Selwyn et al. (1985)</td>
<td>Houston, TX</td>
<td>May-October 1981</td>
<td>Exercise-time 15-min max (4-7 p.m.)</td>
<td>47</td>
<td>Max: 135</td>
</tr>
<tr>
<td>Brunekreef et al. (1994)</td>
<td>Eastern Netherlands</td>
<td>June-August 1991</td>
<td>Exercise-time avg(^a) (10-145 min)</td>
<td>44.4(^c)</td>
<td>Max: 99.5(^c)</td>
</tr>
<tr>
<td>Braun-Fahrlander et al. (1994)</td>
<td>Southern Switzerland</td>
<td>May-October 1989</td>
<td>Exercise-time 30-min avg (1-4 p.m.)</td>
<td>NR</td>
<td>Max: 80(^c)</td>
</tr>
<tr>
<td>Castillejos et al. (1995)</td>
<td>Mexico City, Mexico</td>
<td>June 1990-October 1991</td>
<td>1-h max(^a)</td>
<td>112.3</td>
<td>Max: 365</td>
</tr>
<tr>
<td>Study*</td>
<td>Location</td>
<td>Study Period</td>
<td>O₃ Averaging Time</td>
<td>Mean/Median Concentration (ppb)</td>
<td>Upper Percentile Concentrations (ppb)</td>
</tr>
<tr>
<td>--------</td>
<td>----------</td>
<td>--------------</td>
<td>-------------------</td>
<td>-------------------------------</td>
<td>-------------------------------------</td>
</tr>
<tr>
<td>Hoek et al. (1993)</td>
<td>Wageningen, Netherlands</td>
<td>May-July 1989</td>
<td>1-h max&lt;sup&gt;a&lt;/sup&gt;</td>
<td>NR</td>
<td>Max: 105&lt;sup&gt;c&lt;/sup&gt;</td>
</tr>
<tr>
<td>Hoppe et al. (1995)</td>
<td>Munich, Germany</td>
<td>April-September 1993</td>
<td>30-min max (1-4 p.m.)</td>
<td>High O₃ days: 64 Control O₃ days: 32</td>
<td>Max (overall): 77</td>
</tr>
<tr>
<td>Chan and Wu (2005)</td>
<td>Taichung City, Taiwan</td>
<td>November-December 2001</td>
<td>8-h avg (9 a.m.-5 p.m.)</td>
<td>35.6</td>
<td>Max: 65.1</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-h max</td>
<td>52.6</td>
<td>95.5</td>
</tr>
<tr>
<td>Brauer et al. (1996)</td>
<td>British Columbia, Canada</td>
<td>June-August 1993</td>
<td>1-h max&lt;sup&gt;a&lt;/sup&gt;</td>
<td>40</td>
<td>Max: 84</td>
</tr>
<tr>
<td>Romieu et al. (1998b)</td>
<td>Mexico City, Mexico</td>
<td>March-August 1996</td>
<td>Work-shift avg (mean 9 h)&lt;sup&gt;b&lt;/sup&gt;</td>
<td>67.3</td>
<td>95th: 105.8</td>
</tr>
<tr>
<td>Thaller et al. (2008)</td>
<td>Galveston, TX</td>
<td>Summers 2002-2004</td>
<td>1-h max&lt;sup&gt;d&lt;/sup&gt;</td>
<td>35 (median)</td>
<td>Max: 118</td>
</tr>
</tbody>
</table>

* Note: Studies presented in order of first appearance in the text of this section.
NR = not reported.

<sup>a</sup>Some or all measurements obtained from monitors located off site of outdoor activity.

<sup>b</sup>1-h avg preceding lung function measurement, as reported in the pooled analysis by Kinney et al. (1996).

<sup>c</sup>Concentrations converted from µg/m³ to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).

<sup>d</sup>Individual-level estimates calculated from concentrations measured in different segments of hiking trail.

### Children Attending Summer Camps

Studies of children attending summer camps, most of which were discussed in the 1996 O₃ AQCD, have provided important evidence of the effect of ambient O₃ exposure on respiratory effects in young, healthy children. In addition to the improved exposure assessment as described above, these studies were noted for their daily assessment of lung function by trained staff over 1- to 2-week periods in the mornings and late afternoons before and after hours of outdoor activity (Thurston et al., 1997; Berry et al., 1991; Spektor and Lippmann, 1991; Avol et al., 1990; Burnett et al., 1990; Higgins et al., 1990; Raizenne et al., 1989; Spektor et al., 1988a).

In groups mostly comprising healthy children (ages 7-17 years), decrements in FEV₁ were associated consistently with increases in ambient O₃ concentration averaged over the 1-12 hours preceding lung function measurement (Figure 6-4 [and Table 6-3]). Kinney et al. (1996) corroborated this association in a re-analysis combining 5,367 lung function measurements collected from 616 healthy children from six studies (Spektor and Lippmann, 1991; Avol et al., 1990; Burnett et al., 1990; Higgins et al., 1990; Raizenne et al., 1989; Spektor et al., 1988a). Based on uniform statistical methods, a -20 ml (95% CI: -25, -14) change in afternoon FEV₁ was estimated for a 40-ppb increase in O₃ concentration averaged over the 1 hour before lung function measurement (Kinney et al., 1996) (all effect estimates are standardized to increments specific to the O₃ averaging time as detailed in Section 2.5). All of the studies in the pooled analysis were conducted during summer months but were diverse in locations examined (i.e., Northeast U.S., Canada,
California), range in ambient concentrations of \( \text{O}_3 \) (presented within Table 6-2) and other pollutants measured, and magnitudes of association observed. Study-specific effect estimates ranged between a 0.76 and 48 mL decrease or between a 0.3% and 2.2% decrease in study mean \( \text{FEV}_1 \) per 40-ppb increase in 1-h avg \( \text{O}_3 \).

Among camp studies (including the pooled analysis, plus others), associations for peak expiratory flow (PEF) were more variable than were those for \( \text{FEV}_1 \), as indicated by the wider range in effect estimates and wider 95% CIs (Figure 6-4 [and Table 6-3]). Nonetheless, in most cases, increases in ambient \( \text{O}_3 \) concentration were associated with decreases in PEF. The largest \( \text{O}_3 \)-associated decrease in PEF (mean 2.8% decline per 40-ppb increase in 1-h max \( \text{O}_3 \)) was found in a group of campers with asthma, in whom an increase in ambient \( \text{O}_3 \) concentration also was associated with increases in chest symptoms and bronchodilator use (Thurston et al., 1997).

For both \( \text{FEV}_1 \) and PEF, the magnitude of association was not related to the study mean ambient 1-h avg or max \( \text{O}_3 \) concentration. With exclusion of results from Spektor and Lippmann (1991), larger \( \text{O}_3 \)-associated \( \text{FEV}_1 \) decrements were found in populations with lower mean \( \text{FEV}_1 \). No such trend was found with mean PEF. Sufficient data were not available to assess whether the temporal variability in \( \text{O}_3 \) concentrations, activity levels of subjects, or associations with other pollutants contributed to between-study heterogeneity in \( \text{O}_3 \) effect estimates.
Note: Results generally are presented in order of increasing mean ambient O₃ concentration. Effect estimates are from single-pollutant models and are standardized to a 40-ppb increase for 1-h avg or 1-h max O₃ concentration and a 30-ppb increase for 12-h avg O₃ concentration.

**Figure 6-4** Changes in FEV₁ (mL) or PEF (mL/sec) in association with ambient O₃ concentrations among children attending summer camp.
Table 6-3  Changes in FEV\textsubscript{1} or PEF in association with ambient O\textsubscript{3} concentrations among children attending summer camp for studies presented in Figure 6-4.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Population, Mean FEV\textsubscript{1} or PEF (mL or mL/sec)</th>
<th>Standardized Percent Change (95% CI)a</th>
<th>Standardized Effect Estimate (95% CI)a</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>FEV\textsubscript{1} (mL)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Spektor et al. (1988a)</td>
<td>Fairview Lake, NJ</td>
<td>91 campers without asthma ages 8-15 yr, 2,140</td>
<td>-0.93 (-1.5, -0.35)b</td>
<td>-20.0 (-32.5, -7.5)b</td>
</tr>
<tr>
<td>Spektor and Lippmann (1991)</td>
<td>Fairview Lake, NJ</td>
<td>46 campers without asthma ages 8-14 yr, 2,390</td>
<td>-2.2 (-3.0, -1.3)b</td>
<td>-51.6 (-72.8, -30.4)b</td>
</tr>
<tr>
<td>Burnett et al. (1990)</td>
<td>Lake Couchiching, Ontario, Canada</td>
<td>29 campers without asthma ages 7-15 yr, 2,410</td>
<td>-0.32 (-1.7, 1.1)b</td>
<td>-7.6 (-42.1, 26.9)b</td>
</tr>
<tr>
<td>Raizenne et al. (1998)</td>
<td>Lake Erie, Ontario, Canada</td>
<td>112 campers without asthma mean age 11.6 yr, 2,340</td>
<td>-0.50 (-0.83, -0.16)b</td>
<td>-11.6 (-19.4, -3.8)b</td>
</tr>
<tr>
<td>Avol et al. (1999)</td>
<td>Pine Springs, CA</td>
<td>295 campers without asthma ages 8-17 yr, 2,190</td>
<td>-0.58 (-1.0, -0.12)b</td>
<td>-12.8 (-23.0, -2.6)b</td>
</tr>
<tr>
<td>Higgins et al. (1996)</td>
<td>San Bernardino, CA</td>
<td>43 campers without asthma ages 7-13 yr, 2,060</td>
<td>-1.6 (-2.4, -0.87)b</td>
<td>-33.6 (-49.3, -17.9)b</td>
</tr>
<tr>
<td>Kinney et al. (1996)</td>
<td>Pooled analysis of preceding 6 studies</td>
<td>616 campers without asthma ages 7-15 yr, 2,300</td>
<td>-0.87 (-1.1, -0.63)</td>
<td>-20.0 (-25.5, -14.5)b</td>
</tr>
<tr>
<td>Berry et al. (1991)</td>
<td>Hamilton, NJ</td>
<td>14 campers without asthma age &lt;14 yr, NA</td>
<td>NA</td>
<td>32.8 (6.9, 58.7)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>PEF (mL/sec)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Spektor et al. (1988a)</td>
<td>Fairview Lake, NJ</td>
<td>91 campers without asthma ages 8-15 yr, 4,360</td>
<td>-1.8 (-3.3, -0.40)b</td>
<td>-80.0 (-142.7, -17.3)b</td>
</tr>
<tr>
<td>Burnett et al. (1990)</td>
<td>Lake Couchiching, Ontario, Canada</td>
<td>29 campers without asthma ages 7-15 yr, 5,480</td>
<td>-1.9 (-3.8, -0.05)b</td>
<td>-106.4 (-209.9, -2.9)b</td>
</tr>
<tr>
<td>Raizenne et al. (1998)</td>
<td>Lake Erie, Ontario, Canada</td>
<td>112 campers without asthma mean age 11.6 yr, 5,510</td>
<td>-0.07 (-0.56, 0.41)b</td>
<td>-4.0 (-30.7, 22.7)b</td>
</tr>
<tr>
<td>Avol et al. (1999)</td>
<td>Pine Springs, CA</td>
<td>295 campers without asthma ages 8-17 yr, 4,520</td>
<td>1.9 (0.71, 3.1)b</td>
<td>86.8 (31.9, 142)b</td>
</tr>
<tr>
<td>Higgins et al. (1996)</td>
<td>San Bernardino, CA</td>
<td>43 campers without asthma ages 7-13 yr, 5,070</td>
<td>-0.87 (-2.1, 0.34)b</td>
<td>-44.0 (-105, 17.2)b</td>
</tr>
<tr>
<td>Kinney et al. (1996)</td>
<td>Pooled analysis of preceding 6 studies</td>
<td>616 campers without asthma ages 7-15 yr, 4,222</td>
<td>0.16 (-0.45, 0.77)b</td>
<td>6.8 (-19.1, 32.7)b</td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>CT River Valley, CT</td>
<td>166 campers with asthma ages 7-13 yr, 5,333</td>
<td>-2.8 (-4.9, -0.59)</td>
<td>-146.7 (-261.7, -31.7)</td>
</tr>
<tr>
<td>Neas et al. (1999)</td>
<td>Philadelphia, PA</td>
<td>156 campers without asthma ages 6-11 yr, 4,717</td>
<td>-0.58 (-1.5, 0.33)</td>
<td>-27.5 (-70.8, 15.8)</td>
</tr>
<tr>
<td>Berry et al. (1991)</td>
<td>Hamilton, NJ</td>
<td>14 campers without asthma age &lt;14 yr, NA</td>
<td>NA</td>
<td>-40.4 (-132.1, 51.3)</td>
</tr>
</tbody>
</table>

*Includes studies from Figure 6-4.
NA = Data not available.

aAll results are standardized to a 40-ppb increase in 1-h avg or 1-h max O\textsubscript{3}, except that from Neas et al. (1999), which is standardized to a 30-ppb increase in 12-h avg (9 a.m.-9 p.m.) O\textsubscript{3}.
bEffect estimates based on results reported in the pooled analysis by Kinney et al. (1996).
Similar to controlled human exposure studies, some camp studies found interindividual variability in the magnitude of O\textsubscript{3}-associated changes in lung function. Based on separate regression analyses of serial measurements from individual subjects, increases in ambient O\textsubscript{3} concentration were associated with a wide range of changes in lung function across subjects (Berry et al., 1991; Higgins et al., 1990; Spektor et al., 1988a). For example, among children attending camp in Fairview Lake, NJ, 36% of subjects had statistically significant O\textsubscript{3}-associated decreases in FEV\textsubscript{1}, and the 90th percentile of response was a 6.3% decrease in FEV\textsubscript{1} per a 40-pbb increase in 1-h avg O\textsubscript{3} (Spektor et al., 1988a).

In contrast with previous studies, a recent study of children attending six different summer camps in Belgium did not find an association between ambient O\textsubscript{3} concentration and lung function (Nickmilder et al., 2007). This study examined similar ambient O\textsubscript{3} concentrations as did previous studies (Table 6-2) but used a less rigorous methodology. Lung function was measured only once in each subject, and mean lung function was compared among camps. Children at camps with higher daily 1-h max or 8-h max O\textsubscript{3} concentrations did not consistently have larger decreases in mean intraday FEV\textsubscript{1} or FEV\textsubscript{1}/FVC (Nickmilder et al., 2007).

**Populations Exercising Outdoors**

As discussed in the 1996 and 2006 O\textsubscript{3} AQCDs, epidemiologic studies of adults exercising outdoors have provided evidence for lung function decrements in healthy adults associated with increases in ambient O\textsubscript{3} exposure during exercise with durations (10 min to 12 hours) and intensities (heart rates 121-190 beats per min) in the range of those examined in controlled human exposure studies (Table 6-1). Associations were found consistently in studies of adults exercising outdoor for up to 2 hours, which similar to the camp studies, measured lung function before and after exercise by trained staff on multiple occasions. Collectively, studies of exercising adults found FEV\textsubscript{1} decrements of 1.3 to 1.5% per unit increase in O\textsubscript{3} (Figure 6-5 [and Table 6-4]). The magnitude of association did not appear to be related to study mean ambient O\textsubscript{3} concentrations (Table 6-2), exercise duration, or the mean heart rate measured during exercise (Figure 6-5 [and Table 6-4]). Increases in ambient O\textsubscript{3} concentration generally were associated with decreases in lung function in the smaller body of studies of children exercising outdoors (Table 6-4).

\textsuperscript{1}Effect estimates were standardized to a 40-ppb increase in O\textsubscript{3} averaged over 15 min to 1 h and a 30-ppb increase for O\textsubscript{3} averaged over 2 to 12 hours.
Studie generally are presented in order of increasing duration of outdoor exercise. Data for mean heart rate refer to the maximum or mean measured during exercise or in different groups or conditions as described in Table 6-4.

a bpm = beats per minute. NR = Not reported. Effect estimates are from single-pollutant models and are standardized to a 40-ppb increase for O₃ concentrations averaged over 15 minutes to 1 hour and a 30-ppb increase for O₃ concentrations averaged over 2 to 12 hours.

**Figure 6-5** Percent change in FEV₁ in association with ambient O₃ concentrations among adults exercising outdoors.
Table 6-4 Percent change in FEV$_1$ in association with ambient O$_3$ concentrations among adults exercising outdoors for studies presented in Figure 6-5, and among children exercising outdoors.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Population</th>
<th>Exercise Duration, Mean Heart Rate</th>
<th>O$_3$ Averaging Time</th>
<th>Parameter</th>
<th>Standardized Percent Change (95% CI)$^a$</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Studies of adults</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Brunekreef et al. (1994)</td>
<td>Eastern Netherlands</td>
<td>29 adults exercising, ages 18-37 yr</td>
<td>10 min - 2.4 h, HR: 161 bpm (training), 176 bpm (races)</td>
<td>Exercise duration</td>
<td>FEV$_1$, PEF</td>
<td>-1.3 (-2.2, -0.37), -2.5 (-3.8, -1.2)</td>
</tr>
<tr>
<td>Spektor et al. (1988b)</td>
<td>Tuxedo, NY</td>
<td>30 adults exercising, ages 21-44 yr</td>
<td>15 - 55 min, HR:162 bpm if $\dot{V}_E$ &gt;100 L, 145 bpm if $\dot{V}_E$ 60-100 L</td>
<td>30-min avg</td>
<td>FEV$_1$</td>
<td>-1.3 (-2.0, -0.64)</td>
</tr>
<tr>
<td>Hoppe et al. (2003)</td>
<td>Munich, Germany</td>
<td>43 adults and children exercising, ages 13-38 yr</td>
<td>2 h, HR: NR</td>
<td>30-min max (1-4 p.m.)</td>
<td>FEV$_1$, PEF</td>
<td>-1.3 (-2.6, 0.10), -2.8 (-5.9, 0.31)</td>
</tr>
<tr>
<td>Girardot et al. (2006)</td>
<td>Great Smoky Mt, TN</td>
<td>354 adult day hikers, ages 18-82 yr</td>
<td>1.8-9 h, max HR:121 bpm</td>
<td>Hike duration</td>
<td>FEV$_1$, PEF</td>
<td>0.72 (-0.46, 1.90), 3.5 (-0.11, 7.2)</td>
</tr>
<tr>
<td>Korrick et al. (1998)</td>
<td>Mt. Washington, NH</td>
<td>530 adult day hikers, ages 18-64 yr</td>
<td>2-12 h, max HR: 122 bpm</td>
<td>Hike duration</td>
<td>FEV$_1$, PEF</td>
<td>-1.5 (-2.8, -0.24), -0.54 (-4.0, 2.9)</td>
</tr>
<tr>
<td>Selwyn et al. (1985)</td>
<td>Houston, TX</td>
<td>24 adults exercising, ages 29-47 yr</td>
<td>Duration: NR, max HR: 179 bpm in males, 183 bpm in females</td>
<td>15-min max</td>
<td>FEV$_1$, PEF</td>
<td>-16 mL (-28.8, -3.2)$^b$</td>
</tr>
<tr>
<td><strong>Studies of children not included in Figure 6-4.</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Braun-Fahrlander et al. (1994)</td>
<td>Southern Switzerland</td>
<td>128 children exercising, ages 9-11 yr</td>
<td>10 min, max HR: 180 bpm</td>
<td>30-min avg</td>
<td>PEF</td>
<td>-3.8 (-6.7, -0.96)</td>
</tr>
<tr>
<td>Castillejos et al. (1995)</td>
<td>Mexico City, Mexico</td>
<td>40 children exercising, ages 7-11 yr</td>
<td>2 periods, each with 15 min exercise and 15 min rest, max HR: &lt;190 bpm</td>
<td>1-h avg over combined exercise-rest period</td>
<td>FEV$_1$</td>
<td>-0.48 (-0.72, -0.24)</td>
</tr>
<tr>
<td>Hoek et al. (1993)</td>
<td>Wageningen, Netherlands</td>
<td>65 children exercising, ages 7-12 yr</td>
<td>25 min-1.5 h, HR: NR</td>
<td>1-h avg during exercise</td>
<td>PEF</td>
<td>1.9 (0.83, 3.0)</td>
</tr>
</tbody>
</table>

*Includes studies from Figure 6-5, plus others.
HR = heart rate, bpm = beats per minute, $\dot{V}_E$ = minute ventilation, NR = Not reported.

$^a$Effect estimates are standardized to a 40-ppb increase for O$_3$ concentrations averaged over 15 minutes to 1 hour and a 30-ppb increase for O$_3$ concentrations averaged over 2 to 12 hours.

$^b$Results not included in the figure because data were not available to calculate percent change in lung function.
Compared with the studies of individuals exercising outdoors described above, studies of day-hikers assessed lung function only on one day per subject but examined longer periods of outdoor activity and included much larger sample sizes. Studies of adult day-hikers had similar design but produced contrasting results (Girardot et al., 2006; Korrick et al., 1998). Among 530 hikers on Mt. Washington, NH, Korrick et al. (1998) reported posthike declines in FEV₁ and FVC of 1.5% and 1.3%, respectively, per a 30-ppb increase in 2- to 12-h avg O₃. Associations with FEV₁/FVC, FEF₂₅-₇₅%, and PEF were weaker. In contrast, among 354 hikers on Great Smoky Mt, TN, Girardot et al. (2006) found that higher O₃ concentrations were associated with posthike increases in many of the same lung function indices (Figure 6-5 [and Table 6-4]). These studies were similar in the examination of a mostly white, healthy population and of changes in lung function associated with ambient O₃ concentrations measured on site during multihour (2-12 hours) periods of outdoor exercise. Mean O₃ concentrations were similar as were the population mean and variability in lung function. However, Girardot et al. (2006) differed from Korrick et al. (1998) in several aspects, including a shorter hike time (mean: 5 versus 8 hours), older age of subjects (mean: 43 versus 35 yr), and measurement of lung function by a larger number of less well-trained technicians. The impact of these differences on the heterogeneity in results between the studies was not examined.

Similar to the camp studies, some studies of outdoor exercise examined and found interindividual variability in the magnitude of O₃-associated decreases in lung function. In Korrick et al. (1998), although a 30-ppb increase in 2- to 12-h avg ambient O₃ concentration was associated with a group mean change in FEF₂₅-₇₅% of -0.81% (95% CI: -4.9, 3.3), some individuals experienced a >10% decline. The odds of >10% decline in FEF₂₅-₇₅% increased with increasing ambient O₃ concentration (OR: 2.3 [95% CI: 1.2, 6.7] per 30-ppb increase in 2- to 12-h avg O₃). Likewise, Hoppe et al. (2003) found that compared with days with 30-min max (1-4 p.m.) ambient O₃ concentrations <40 ppb, on days with O₃ >50 ppb, 14% of athletes had at least a 10% decrease in lung function or 20% increase in airway resistance.

Outdoor Workers

Consistent findings in outdoor workers add to the evidence that short-term increases in ambient O₃ exposure can decrease lung function in healthy adults (Figure 6-6 [and Table 6-5]). Except for Hoppe et al. (1995), studies used central site ambient O₃ concentrations. However, in outdoor workers, ambient concentrations have been more highly correlated with and similar in magnitude to personal exposures (Section 4.3.3) likely because workers spend long periods of time outdoors (6-14 hours across studies) and the O₃ averaging times examined correspond to periods of outdoor work. For example, in a subset of berry pickers, the correlation and ratio of personal to ambient 24-h avg O₃ concentrations (15 km from work site) were 0.64 and 0.96, respectively (Brauer and Brook, 1997). The 6-h avg personal-ambient ratio in a population of shoe cleaners in Mexico City, Mexico, was 0.56 (O'Neill et al., 2003). Many studies of outdoor workers found that in addition to same-day concentrations, O₃ concentrations lagged 1 or 2 days (Chan and Wu, 2005; Brauer et
Similar to other populations with increased outdoor exposure, most of the magnitudes of O₃-associated lung function decrements in outdoor workers were small, i.e., <1% to 3.4% per unit increase in O₃ concentration. The magnitude of decrease was not found to depend strongly on duration of outdoor work or ambient O₃ concentration. The largest decrease (6.4% per 40-ppb increase in 1-h max O₃) was observed in berry pickers in British Columbia who were examined during a period of relatively low ambient O₃ concentrations (work shift mean: 26.0 ppb [SD: 11.8]) but had long daily periods of outdoor work (8-14 hours) (Brauer et al., 1996) (Figure 6-6 [and Table 6-5]). However, a much smaller O₃-associated decrease in FEV₁ was found in shoe cleaners in Mexico City who were examined during a period of higher O₃ concentrations (work shift mean: 67.3 ppb [SD: 24]) but had a period of outdoor work that was as long as that of the berry pickers. The smallest magnitude of decrease (-0.4% [95% CI: -0.8, 0] in afternoon FEV₁/FVC per 40-ppb increase in 1-h max O₃) was observed in lifeguards in Galveston, TX (Thaller et al., 2008) whose outdoor work periods were shorter than those of the berry pickers but characterized by a similar range of ambient O₃ concentrations. Not all studies provided information on ventilation rate or pulse rate, thus it was not possible to ascertain whether differences in the magnitude of O₃-associated lung function decrement across studies were related to differences in the level of exertion of among the various groups of workers.

---

₁Effect estimates were standardized to a 40-ppb increase for O₃ averaged over 30 minutes to 1 hour and a 30-ppb increase for O₃ averaged over 8 hours or 12 hours.
Study | Population | Parameter | $O_3$ Lag | Subgroup
--- | --- | --- | --- | ---
Thaller et al. (2008) | Lifeguards | FVC | 0 | 
 |  | FEV$_1$/FVC | 
Brauer et al. (1996) | Berry pickers | FEV$_1$ | 0 | 
 |  | 1 | 
Hoppe et al. (1995) | Forestry workers | FEV$_1$ | 0 | 
Romieu et al. (1998) | Shoe Cleaners | FEV$_1$ | 0 | Placebo
 |  |  | Antioxidant supplement
 |  | 0-1 avg | Placebo
 |  |  | Antioxidant supplement

Note: Studies generally are presented in order of increasing mean ambient $O_3$ concentration. Effect estimates are from single-pollutant models and are standardized to a 40-ppb increase for 30-min, 1-h avg, or 1-h max $O_3$ concentrations.

Figure 6-6  Percent change in lung function in association with ambient $O_3$ concentrations among outdoor workers.
Table 6-5  Percent change in FEV₁ or FEV₁/FVC in association with ambient O₃ concentrations among outdoor workers for studies presented in Figure 6-6.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Population</th>
<th>Parameter</th>
<th>FVIC/Max</th>
<th>1-h max</th>
<th>12-h avg (7 a.m.-7 p.m.)</th>
<th>O₃ Lag</th>
<th>Subgroup</th>
<th>Standardized Percent Change (95% CI)*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Thaller et al.</td>
<td>Galveston, TX</td>
<td>142 lifeguards, ages 16-27 yr</td>
<td>FVC</td>
<td>6-8 h</td>
<td>0</td>
<td>-0.24 (-0.28, 0.72)</td>
<td>0.15</td>
<td>(-0.60, 0.90)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>FEV₁/FVC</td>
<td></td>
<td>1-h max</td>
<td>12-h avg (7 a.m.-7 p.m.)</td>
<td>0.15</td>
<td>(-0.60, 0.90)</td>
<td></td>
</tr>
<tr>
<td>Brauer et al.</td>
<td>British Columbia, Canada</td>
<td>58 berry pickers, ages 10-69 yr</td>
<td>FEV₁</td>
<td>8-14 h</td>
<td>0</td>
<td>-5.4 (-6.5, -4.3)</td>
<td>-6.4</td>
<td>(-8.0, -4.7)</td>
<td></td>
</tr>
<tr>
<td>Hoppe et al.</td>
<td>Munich, Germany</td>
<td>41 forestry workers, ages 20-60 yr</td>
<td>FEV₁</td>
<td>Not reported</td>
<td>0-1</td>
<td>-1.4 (-3.0, 0.16)</td>
<td>0.03</td>
<td>(-1.8, 1.1)</td>
<td></td>
</tr>
<tr>
<td>Romieu et al.</td>
<td>Mexico City, Mexico</td>
<td>47 male shoe cleaners, mean (SD) age: 38.9 (10) yr</td>
<td>FEV₁ Mean (SD): 9 (1 h)</td>
<td>1-h avg before lung function measurement</td>
<td>0 Placebo Antioxidant</td>
<td>-2.1 (-3.3, -0.85)</td>
<td>-0.52 (-2.0, 0.97)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chan and Wu</td>
<td>Taichung City, Taiwan</td>
<td>43 mail carriers, Mean (SD) age: 39 (8) yr</td>
<td>Nighttime PEF</td>
<td>8 h</td>
<td>0</td>
<td>-1.6 (-2.2, -1.1)</td>
<td>-1.9</td>
<td>(-2.5, -1.3)</td>
<td></td>
</tr>
</tbody>
</table>

*Includes results from Figure 6-6, plus others.

**Effect estimates are standardized to a 40-ppb increase for 30-min, 1-h avg, or 1-h max O₃ and a 30-ppb increase for 8-h avg or 12-h avg O₃.

*PEF results not included in figure.

Associations at Lower Ozone Concentrations

In some studies of populations with increased outdoor exposures, O₃-associated lung function decrements were observed when maximum or average ambient O₃ concentrations over 30 minutes to 12 hours did not exceed 80 ppb (Chan and Wu, 2005; Korrick et al., 1998; Hoppe et al., 1995; Braun-Fahlander et al., 1994) (presented within Table 6-2). Korrick et al. (1998) found lung function decrements in association with higher hike-time average (2-12 hours) O₃ concentrations in the range 40-74 ppb but not <40 ppb. Several other studies that included higher maximum ambient O₃ concentrations restricted analyses to observations with 10-min to 1-h avg O₃ concentrations <80 ppb (Table 6-6). Higgins et al. (1990) found that O₃-associated lung function decrements in children attending camp were limited largely to 1-h avg ambient concentrations >120 ppb; however, many other studies found associations in the lower range of O₃ concentrations (Table 6-6). Among adults exercising outdoors, Spektor et al. (1988b) found that for most lung function
parameters, effect estimates in analyses restricted to 30-min max ambient O₃ concentrations <80 ppb were similar to those obtained for the full range of O₃ concentrations (Table 6-6). In a study of children attending summer camp, similar effects were estimated for the full range of 1-h avg O₃ concentrations and those <60 ppb (Spektor et al., 1988a). Brunekreef et al. (1994) found increases in ambient O₃ concentration (10-min to 1-hour) during outdoor exercise to be associated with decreases in FEV₁ in analyses restricted to concentrations <61 (Table 6-6) and <51 ppb (quantitative results not reported). Whereas Brunekreef et al. (1994) found that effect estimates were near zero with O₃ concentrations <41 ppb, Brauer et al. (1996) found that associations persisted with 1-h max O₃ concentrations <40 ppb (quantitative results not provided).

**Table 6-6** Associations between ambient O₃ concentration and FEV₁ decrements in different ranges of ambient O₃ concentrations.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Concentration Range</th>
<th>Standardized Percent Change (95% CI)ᵃ</th>
</tr>
</thead>
<tbody>
<tr>
<td>Brunekreef et al. (1994)</td>
<td>Eastern Netherlands</td>
<td>29 adults exercising, ages 18-37 yr</td>
<td>10-min to 2.4-h avg during exercise</td>
<td>Full range O₃ &lt;61 ppb</td>
<td>-1.3 (-2.2, -0.37)</td>
</tr>
<tr>
<td>Spektor et al. (1988a)</td>
<td>Fairview Lake, NJ</td>
<td>91 children without asthma at camp, ages 8-15 yr</td>
<td>1-h avg before afternoon FEV₁ measurement</td>
<td>Full range O₃ &lt;80 ppb</td>
<td>-2.7 (-3.3, -2.0)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ &lt;60 ppb</td>
<td>-1.4 (-2.5, -0.34)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-2.2 (-3.7, -0.80)</td>
</tr>
<tr>
<td>Spektor et al. (1988b)</td>
<td>Tuxedo, NY</td>
<td>30 adults exercising, ages 21-44 yr</td>
<td>30-min avg during exercise</td>
<td>Full range O₃ &lt;80 ppb</td>
<td>-1.3 (-2.0, -0.64)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-1.3 (-2.4, -0.08)</td>
</tr>
<tr>
<td>Korrick et al. (1998)</td>
<td>Mt. Washington, NH</td>
<td>530 adult day hikers, ages 18-64 yr</td>
<td>2-12 h avg during hike</td>
<td>Full range O₃ 40-74 ppb</td>
<td>-1.5 (-2.8, -0.24)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-2.6 (-4.9, -0.32)</td>
</tr>
<tr>
<td>Higgins et al. (1990)</td>
<td>San Bernardino, CA</td>
<td>43 children without asthma at camp, ages 7-13 yr</td>
<td>1-h avg around time of FEV₁ measurement</td>
<td>&gt;120 ppb</td>
<td>-1.4 (-2.8, 0.03)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>&lt;120 ppb</td>
<td>0.35 (-1.3, 2.0)</td>
</tr>
</tbody>
</table>

ᵃResults are presented in order of maximum O₃ concentration included in models. Effect estimates are standardized to a 40-ppb increase for O₃ concentrations averaged over 10 min to 1 h and a 30-ppb increase for O₃ concentrations averaged over 2 to 12 h.

**Children with Asthma**

Increases in ambient O₃ concentration have been associated with lung function decrements in children with asthma in epidemiologic studies conducted across diverse geographical locations and a range of ambient O₃ concentrations (Table 6-7). Whereas most studies of populations with increased outdoor exposures monitored O₃ concentrations at the site of subjects’ outdoor activities and used trained staff to measure lung function, studies of children with asthma relied more on O₃ measured at central monitoring sites and lung function measured by subjects. However, these methods for exposure and outcome assessment in studies of children with asthma likely are sources of nondifferential measurement error. Further, compared with the
camp studies, studies of children with asthma have provided an understanding of the changes in lung function associated with patterns of outdoor activity and ambient O₃ exposure that likely better represent those of children in the general population. These studies also have provided more information on potential at-risk populations for O₃-associated respiratory effects and on potential confounding by copollutant exposure or meteorology.

<p>| Table 6-7 Mean and upper percentile concentrations of O₃ in epidemiologic studies of lung function in children with asthma. |
|---|---|---|---|---|
| Study* | Location | Study Period | O₃ Averaging Time | Mean/Median Concentration (ppb) | Upper Percentile Concentrations (ppb) |
| Jalaludin et al. (2000) | Sydney, Australia | Feb-Dec 1994 | 15-h avg (6 a.m.-9 p.m.) | 12 | Max: 43 |
| | | | 1-h max | 26 | |
| Lewis et al. (2005) | Detroit, MI | Feb 2001-May 2002 | 24-h avg | 27.6, 26.5* | Overall max: 66.3* |
| | | | 8-h max | 40.4, 41.4* | Overall max: 92.0* |
| Just et al. (2002) | Paris, France | April-June 1996 | 24-h avg | 30.0b | Max: 61.7b |
| Hoppe et al. (2003) | Munich, Germany | Summers 1992-1995 | 30-min max (1 p.m.-4 p.m.) | High O₃ days: 66.9c | Max: 91 high O₃ daysc |
| | | | | Control O₃ days: 32.5c | Max: 39 Control O₃ daysc |
| Thurston et al. (1997) | CT River Valley, CT | June 1991-1993 | 1-h max | 83.6c | Max: 160c |
| | | | 1-h max | 102 | Max: 309 |
| O'Connor et al. (2008) | Boston, MA; Bronx, Manhattan NY; Chicago, IL; Dallas, TX, Seattle, WA; Tucson, AZ (ICAS) | Aug 1998-July 2001 | 24-h avg | NR | NR |
| Mortimer et al. (2002) | Bronx, East Harlem, NY; Baltimore, MD; Washington, DC; Detroit, MI; Cleveland, OH; Chicago, IL; St. Louis, MO (NCICAS) | June-Aug 1993 | 8-h avg (10 a.m.-6 p.m.) | 48 | NR |
| Mortimer et al. (2000) | | | | | |
| Gielens et al. (1997) | Amsterdam, Netherlands | Apr-July 1995 | 8-h max | 34.2c | Max: 56.5b |
| Liu et al. (2009a); Dales et al. (2009) | Windsor, ON, Canada | Oct-Dec 2005 | 24-h avg | 13.0 | 95th: 26.5 |
| | | | 1-h max | 27.2 | 75th: 32.8 |</p>
<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O₃ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rabinovitch et al. (2004)</td>
<td>Denver, CO</td>
<td>Nov-Mar 1999-2002</td>
<td>1-h max</td>
<td>28.2</td>
<td>75th: 36.0, Max 70.0</td>
</tr>
<tr>
<td>Barraza-Villarreal et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>June 2003-June 2005</td>
<td>8-h moving avg</td>
<td>31.6</td>
<td>Max: 66.3</td>
</tr>
<tr>
<td>Delfino et al. (2004)</td>
<td>Alpine, CA</td>
<td>September-October 1999; April-June 2000</td>
<td>8-h max</td>
<td>62.9</td>
<td>90th: 83.9, Max: 105.9</td>
</tr>
<tr>
<td>Hernández-Cadena et al. (2009)</td>
<td>Mexico City, Mexico</td>
<td>May-September 2005</td>
<td>24-h avg</td>
<td>26.3</td>
<td>75th: 35.3, Max: 62.8</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-h max</td>
<td>74.5</td>
<td>75th: 92.5, Max: 165</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.

ICAS = Inner City Asthma Study, NR = Not Reported, NCICAS = National Cooperative Inner-City Asthma Study.

Measurements at two sites established by investigators and located within 5 km of most subjects’ residences.

Concentrations converted from µg/m³ to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).

Measured where subjects spent daytime hours.

In a majority of studies, including a large U.S. multicity study and several smaller studies conducted in the United States, Mexico City, Mexico, and Europe, an increase in ambient O₃ concentration (various averaging times and lags) was associated with a decrement in FEV₁ ([Figure 6-7][and Table 6-8]) or PEF ([Figure 6-8][and Table 6-9]) in children with asthma. Results were more variable for FEV₁ than for PEF. In most studies, FEV₁ was measured by technicians whereas PEF was measured by study subjects or their parents. However, associations with O₃ also were found with PEF measured by trained technicians ([Romieu et al., 2004b; Thurston et al., 1997]), which are subject to less measurement error. Further, in some studies, associations with FEV₁ were limited to specific subgroups. Some studies found that increases in ambient O₃ concentration were associated with greater lung function variability, i.e., a deviation from a baseline level. These results pointed to associations of O₃ with poorer lung function, as indicated by a decrease from the individual’s mean lung function over the study period ([Jalaludin et al., 2000]), a decrease in lung function over the course of the day ([Lewis et al., 2005]), or a decrease in the lowest daily measurement ([Just et al., 2002]). Within many studies, increases in O₃ concentration were associated with decreases in lung function and increases in respiratory symptoms at the same or similar lag ([Just et al., 2002; Mortimer et al., 2002; Gielen et al., 1997; Romieu et al., 1997; Thurston et al., 1997; Romieu et al., 1996]) (see [Figure 6-12][and Table 6-20]) for symptom results).
Note: Results generally are presented in order of increasing mean ambient O₃ concentration. CS = Corticosteroid, URI = Upper respiratory infection. Effect estimates are from single-pollutant models and are standardized to a 40-ppb increase for 30-min or 1-h max O₃ concentrations, a 30-ppb increase for 8-h max or 8-h avg O₃ concentrations, and a 20-ppb increase for 24-h avg O₃ concentrations.

Figure 6-7 Percent change in FEV₁ in association with ambient O₃ concentrations among children with asthma.
Table 6-8 Percent change in FEV$_1$ in association with ambient O$_3$ concentrations among children with asthma for studies presented in Figure 6-7 plus others.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/Population</th>
<th>O$_3$ Averaging Time</th>
<th>O$_3$ Lag</th>
<th>Parameter</th>
<th>Subgroup</th>
<th>Standardized Percent Change (95% CI)$^a$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Liu et al.</td>
<td>Windsor, ON, Canada</td>
<td>24-h avg</td>
<td>0</td>
<td>FEV$_1$</td>
<td></td>
<td>-0.89 (-3.5, 1.8)</td>
</tr>
<tr>
<td>(2009a)</td>
<td>182 children with asthma, ages 9-14 yr</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.44 (-2.4, 1.6)</td>
</tr>
<tr>
<td>Lewis et al.</td>
<td>Detroit, MI</td>
<td>8-h max</td>
<td>1</td>
<td>Lowest daily FEV$_1$</td>
<td>CS user With URI</td>
<td>-1.9 (-10.4, 7.5)</td>
</tr>
<tr>
<td>(2005)</td>
<td>86 children with asthma, age 9.1 (1.4) yr</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-5.5 (-9.5, -1.5)</td>
</tr>
<tr>
<td>Hoppe et al.</td>
<td>Munich, Germany</td>
<td>30-min max</td>
<td>1</td>
<td>Afternoon FEV$_1$</td>
<td>Without asthma With asthma</td>
<td>0.93 (-0.80, 2.7)</td>
</tr>
<tr>
<td>(2003)</td>
<td>43 people with asthma, ages 12-23 yr</td>
<td>(1-4 p.m.)</td>
<td></td>
<td></td>
<td></td>
<td>-0.56 (-4.6, 3.7)</td>
</tr>
<tr>
<td></td>
<td>44 children without asthma, ages 6-8 yr</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.09 (-1.7, 1.6)</td>
</tr>
<tr>
<td>Barraza-Villarreal et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>8-h avg</td>
<td>0-4 avg</td>
<td>FEV$_1$</td>
<td>50 without asthma 158 with asthma</td>
<td>-1.5 (-4.7, 1.7)</td>
</tr>
<tr>
<td></td>
<td>208 children, ages 6-14 yr</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.12 (-2.0, 1.8)</td>
</tr>
<tr>
<td>Romieu et al.</td>
<td>Mexico City, Mexico</td>
<td>1-h max</td>
<td>1</td>
<td>FEV$_1$</td>
<td></td>
<td>Placebo -0.21 (-0.77, 0.36)</td>
</tr>
<tr>
<td>(2002)</td>
<td>158 children with asthma, ages 6-17 yr</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant supplement 0.05 (-0.60, 0.69)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Placebo, moderate/severe asthma -1.1 (-2.0, -0.19)</td>
</tr>
<tr>
<td>Romieu et al.</td>
<td>Mexico City, Mexico</td>
<td>1-h max</td>
<td>1</td>
<td>FEV$_1$</td>
<td>GSTP1 Ile/Ile or Ile/Val GSTP1 Val/Val</td>
<td>-0.51 (-1.1, 0.05)</td>
</tr>
<tr>
<td>(2006)</td>
<td>151 children with asthma, mean age 9 yr</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0.50 (-0.25, 1.3)</td>
</tr>
<tr>
<td>Study*</td>
<td>Location/Population</td>
<td>O₃ Averaging Time</td>
<td>O₃ Lag</td>
<td>Parameter</td>
<td>Subgroup</td>
<td>Standardized Percent Change (95% CI)a</td>
</tr>
<tr>
<td>--------</td>
<td>---------------------</td>
<td>------------------</td>
<td>--------</td>
<td>-----------</td>
<td>----------</td>
<td>-------------------------------------</td>
</tr>
<tr>
<td>Dales et al. (2009)</td>
<td>Windsor, ON, Canada</td>
<td>1-h max</td>
<td>0</td>
<td>Evening percent predicted FEV₁</td>
<td></td>
<td>-0.47 (-1.9, 0.95)</td>
</tr>
<tr>
<td>Rabinovitch et al. (2004)</td>
<td>Denver, CO</td>
<td>1-h max</td>
<td>0-2 avg</td>
<td>Morning FEV₁ (mL)</td>
<td></td>
<td>55 (-2.4, 108)</td>
</tr>
<tr>
<td>O'Connor et al. (2008)</td>
<td>Boston, MA; Bronx, Manhattan NY; Chicago, IL; Dallas, TX; Seattle, WA; Tucson, AZ</td>
<td>24-h avg</td>
<td>1-5 avg</td>
<td>Percent predicted FEV₁</td>
<td></td>
<td>-0.41 (-1.0, 0.21)</td>
</tr>
</tbody>
</table>

*Includes studies in Figure 6-7, plus others

CS = corticosteroid, URI = Upper respiratory infection.

**Effect estimates are standardized to a 40-ppb increase for 30-min or 1-h max O₃, a 30-ppb increase for 8-h max or 8-h avg O₃, and a 20-ppb increase for 24-h avg O₃.

bResults not presented in Figure 6-7 because a different form of FEV₁ with a different scale was examined or because sufficient data were not available to calculate percent change in FEV₁.
<table>
<thead>
<tr>
<th>Study</th>
<th>Parameter</th>
<th>O&lt;sub&gt;3&lt;/sub&gt; Lag</th>
<th>Subgroup</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gielen et al. (1997)</td>
<td>Evening PEF</td>
<td>0</td>
<td>All subjects</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2</td>
<td>Normal BW</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2</td>
<td>Low BW</td>
</tr>
<tr>
<td></td>
<td>Morning PEF</td>
<td>2</td>
<td>No medication</td>
</tr>
<tr>
<td>Mortimer et al. (2002)</td>
<td>Morning PEF</td>
<td>1</td>
<td>All subjects</td>
</tr>
<tr>
<td></td>
<td></td>
<td>3</td>
<td>Normal BW</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1-5 avg</td>
<td>Low BW</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>No medication</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>CS user</td>
</tr>
<tr>
<td>Mortimer et al. (2000)</td>
<td>Morning PEF</td>
<td>1</td>
<td>All subjects</td>
</tr>
<tr>
<td></td>
<td></td>
<td>3</td>
<td>Normal BW</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1-5 avg</td>
<td>Low BW</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>No medication</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>CS user</td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>PEF</td>
<td>0</td>
<td>All subjects</td>
</tr>
<tr>
<td>Romieu et al. (2004b)</td>
<td>FEF&lt;sub&gt;25-75%&lt;/sub&gt;</td>
<td>1</td>
<td>Placebo, GSTM1 null</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Placebo, GSTM1 positive</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Antioxidant, GSTM1 null</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Antioxidant, GSTM1 positive</td>
</tr>
<tr>
<td>Romieu et al. (1996)</td>
<td>Evening PEF</td>
<td>0</td>
<td>All subjects</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2</td>
<td>Normal BW</td>
</tr>
<tr>
<td>Romieu et al. (1997)</td>
<td>Evening PEF</td>
<td>0</td>
<td>All subjects</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2</td>
<td>Normal BW</td>
</tr>
</tbody>
</table>

Note: Results generally are presented in order of increasing mean ambient O<sub>3</sub> concentration. BW = birth weight, CS = Corticosteroid. Effect estimates are from single pollutant models and are standardized to a 40-ppb increase for 1-h max O<sub>3</sub> concentrations and a 30-ppb increase for 8-h max or 8-h avg O<sub>3</sub> concentrations.

Figure 6-8 Percent change in PEF or FEF<sub>25-75%</sub> in association with ambient O<sub>3</sub> concentrations among children with asthma.
### Table 6-9
Percent change in PEF or FEF<sub>25-75%</sub> in association with ambient O<sub>3</sub> concentrations among children with asthma for studies presented in Figure 6-8 plus others.

<table>
<thead>
<tr>
<th>Study&lt;sup&gt;a&lt;/sup&gt;</th>
<th>Location/Population</th>
<th>O&lt;sub&gt;3&lt;/sub&gt; Averaging Time</th>
<th>O&lt;sub&gt;3&lt;/sub&gt; Lag</th>
<th>Parameter</th>
<th>Subgroup</th>
<th>Standardized Percent Change (95% CI)&lt;sup&gt;a&lt;/sup&gt;</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gielen et al. (1997)</td>
<td>Amsterdam, Netherlands 61 children with asthma, ages 7-13 yr</td>
<td>8-h max</td>
<td>0</td>
<td>Evening PEF</td>
<td></td>
<td>1.3 (-0.25, 2.9)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>2</td>
<td>Evening PEF</td>
<td></td>
<td>-1.3 (-2.8, 0.16)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>2</td>
<td>Morning PEF</td>
<td></td>
<td>-1.3 (-2.6, -0.08)</td>
</tr>
<tr>
<td>Mortimer et al. (2002)</td>
<td>Bronx, East Harlem, NY; Baltimore, MD; Washington, DC; Detroit, MI, Cleveland, OH; Chicago, IL; St. Louis, MO 846 children with asthma, ages 4-9 yr</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>1</td>
<td>Evening PEF</td>
<td></td>
<td>-0.12 (-0.76, 0.52)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>3</td>
<td>Morning PEF</td>
<td>All subjects</td>
<td>-0.64 (-1.2, -0.10)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-5 avg</td>
<td></td>
<td></td>
<td>-1.2 (-2.1, -0.26)</td>
</tr>
<tr>
<td>Mortimer et al. (2000)</td>
<td>Bronx, East Harlem, NY; Baltimore, MD; Washington, DC; Detroit, MI, Cleveland, OH; Chicago, IL; St. Louis, MO 846 children with asthma, ages 4-9 yr</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>1-5 avg</td>
<td>Morning PEF</td>
<td>Normal BW Low BW (&lt;5.5 lbs.) No medication CS user</td>
<td>-0.60 (-1.6, 0.39)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Normal BW</td>
<td>-3.6 (-5.2, -2.0)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Low BW (&lt;5.5 lbs.)</td>
<td>-1.1 (-3.0, 0.84)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>No medication</td>
<td>-1.2 (-2.5, 0.11)</td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>CT River Valley, CT 166 children with asthma, ages 7-13 yr</td>
<td>1-h max</td>
<td>0</td>
<td>Intraday change PEF</td>
<td></td>
<td>-2.8 (-4.9, -0.59)</td>
</tr>
<tr>
<td>Romieu et al. (2004b)</td>
<td>Mexico City, Mexico 158 children with asthma, mean age 9 yr</td>
<td>1-h max</td>
<td></td>
<td></td>
<td>Placebo, GSTM1 null</td>
<td>-2.3 (-4.2, -0.44)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Placebo, GSTM1 positive</td>
<td>-0.48 (-1.7, 0.74)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant, GSTM1 null</td>
<td>-0.16 (-1.8, 1.6)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant, GSTM1 positive</td>
<td>0.24 (-1.3, 1.8)</td>
</tr>
<tr>
<td>Romieu et al. (1996)</td>
<td>Northern Mexico City, Mexico 71 children with asthma, ages 5-7 yr</td>
<td>1-h max</td>
<td>0</td>
<td></td>
<td></td>
<td>-0.17 (-0.79, 0.46)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>2</td>
<td>Evening PEF</td>
<td></td>
<td>-0.55 (-1.3, 0.19)</td>
</tr>
<tr>
<td>Romieu et al. (1997)</td>
<td>Southern Mexico City, Mexico 65 children with asthma, ages 5-13 yr</td>
<td>1-h max</td>
<td>0</td>
<td></td>
<td></td>
<td>-0.52 (-1.0, -0.01)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>2</td>
<td>Evening PEF</td>
<td></td>
<td>-0.06 (-0.70, 0.58)</td>
</tr>
</tbody>
</table>
### Studies not included in Figure 6-8

<table>
<thead>
<tr>
<th>Study</th>
<th>Location/Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag</th>
<th>Parameter</th>
<th>Subgroup</th>
<th>Standardized Percent Change (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jalaludin et al. (2000)</td>
<td>Sydney, Australia 45 children with asthma and AHR, mean (SD) age 9.6 (1) yr</td>
<td>24-h avg 1-h max 0</td>
<td>Daily deviation from mean PEF</td>
<td>-5.2 (-8.3, -2.2)</td>
<td>-1.1 (-2.4, 0.18)</td>
<td></td>
</tr>
<tr>
<td>Wiwatanadate and Trakultivakorn (2010)</td>
<td>Chiang Mai, Thailand 31 children with asthma, ages 4-11 yr</td>
<td>24-h avg 0 5</td>
<td>Daily avg PEF (L/min)</td>
<td>1.0 (-1.6, 3.6)</td>
<td>-2.6 (-5.2, 0)</td>
<td></td>
</tr>
<tr>
<td>O'Connor et al. (2008)</td>
<td>Boston, MA; Bronx, Manhattan NY; Chicago, IL; Dallas, TX; Seattle, WA; Tucson, AZ 861 Children with asthma, mean (SD) age 7.7 (2.0) yr</td>
<td>24-h avg 1-5 avg 5</td>
<td>Change in percent predicted PEF</td>
<td>-0.22 (-0.86, 0.43)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Just et al. (2002)</td>
<td>Paris, France 82 children with asthma, mean (SD) age 10.9 (2.5) yr</td>
<td>8-h avg 0-2 avg 2</td>
<td>Percent variability PEF</td>
<td>15.3 (0, 30.6)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

* Includes studies in Figure 6-8, plus others

BW = birth weight, CS = corticosteroid, AHR = Airway hyperresponsiveness.

**Effect estimates are standardized to a 40-ppb increase for 1-h max O₃, a 30-ppb increase for 8-h max or avg O₃, and a 20-ppb increase for 24-h avg O₃.

Results are not presented in Figure 6-8 because a different form of PEF with a different scale was examined or because sufficient data were not available to calculate percent change in PEF.

Outcome defined as the normalized percent deviation from individual mean PEF during the study period. Quantitative results from generalized estimating equations were provided only for models that included PM₁₀ and NO₂.

The most geographically representative data were provided by the large, multi-U.S. city National Cooperative Inner City Asthma Study (NCICAS) (Mortimer et al., 2002; Mortimer et al., 2000) and Inner-City Asthma Study (ICAS) (O'Connor et al., 2008). Although the two studies differed in the cities, seasons, racial distribution of subjects, and lung function indices examined, results were fairly similar. In ICAS, which included children with asthma and atopy (i.e., allergic sensitization) and year-round examinations of lung function, a 20-ppb increase in the lag 1-5 average of 24-h avg O₃ was associated with a 0.41-point decrease in percent predicted FEV₁ (95% CI: -1.0, 0.21) and a 0.22-point decrease in percent predicted PEF (95% CI: -0.86, 0.43) (O'Connor et al., 2008).

Increases in lag 1-5 avg O₃ (8-h avg, 10 a.m.-6 p.m.) also were associated with declines in PEF in NCICAS, which included different U.S. cities, summer-only measurements, larger proportions of Black and Hispanic children, and fewer subjects with atopy (79%) (Mortimer et al., 2002). Ozone concentrations lagged 3 to 5 days were associated with larger PEF decrements than were O₃ concentrations lagged 1 to 2 days (Figure 6-8 [and Table 6-9]). NCICAS additionally identified groups potentially at increased risk of O₃-associated PEF decrements, namely, males,
children of Hispanic ethnicity, children living in crowded housing, and as indicated in Figure 6-8 (and Table 6-9), children with birth weight <5.5 lbs (Mortimer et al., 2000). Somewhat paradoxically, O₃ was associated with a larger decrease in PEF among subjects taking cromolyn, medication typically used to treat asthma due to allergy, but a smaller decrease among subjects with allergic sensitization (as determined by skin prick test). NCICAS also indicated robust associations with consideration of other sources of heterogeneity. Except for Baltimore, MD, effect estimates were similar across the study cities (1.1 to 1.7% decrease in PEF per 30-ppb increase in lag 1-5 avg of 8-h avg O₃). Results were similar with O₃ averaged from all available city monitors and concentrations averaged from the three monitors closest to subjects’ ZIP code centroid (1.2% and 1.0% decrease in PEF, respectively, per 30-ppb increase in O₃). At concentrations <80 ppb, a 30-ppb increase in lag 1-5 of 8-h avg O₃ was associated with a 1.4% decrease (95% CI: -2.6, -0.21) in PEF, (Mortimer et al., 2002) which was similar to the effect estimated for the full range of O₃ concentrations (Figure 6-8 [and Table 6-9]). In a study of children with asthma in the Netherlands, Gielen et al. (1997) estimated similar effects on PEF for the full range of 8-h max O₃ concentrations and concentrations <51 ppb.

Several but not all controlled human exposure studies have reported slightly larger O₃-induced FEV₁ decrements in adults with asthma than adults without asthma (Section 6.2.1.1). However, in the few epidemiologic studies that compared children with and without asthma, evidence did not conclusively indicate that children with asthma were at increased risk of O₃-associated lung function decrements. Hoppe et al. (2003) and Jalaludin et al. (2000) generally found larger O₃-associated decrements in FVC and PEF, respectively, in children with asthma; whereas Raizenne et al. (1989) did not consistently demonstrate differences between campers with and without asthma. In their study of children in Mexico City, Mexico, Barraza-Villarreal et al. (2008) estimated larger O₃-associated decreases in children without asthma; however, 72% of these children had atopy. These findings indicate that children with atopy, who also have airway inflammation and similar respiratory symptoms, may experience respiratory effects from short-term ambient O₃ exposure.

As shown in Figure 6-7 (and Table 6-8) and Figure 6-8 (and Table 6-9), lung function decrements in children with asthma mostly ranged from <1% to 2% per unit increase in ambient O₃ concentration. Larger magnitudes of decrease, were found in children with asthma who were using CS, had a concurrent upper respiratory infection (URI), were GSTM1 null, had airway hyperresponsiveness, or had increased outdoor exposure (Romieu et al., 2006; Lewis et al., 2005; Romieu et al., 2004b; Jalaludin et al., 2000) than among children with asthma overall (Barraza-Villarreal et al., 2008; Lewis et al., 2005; Delfino et al., 2004; Romieu et al., 2002). For example, Jalaludin et al. (2000) estimated a -5.2% deviation from mean FEV₁ per 20-ppb increase in 24-h avg O₃ concentration among children with asthma and airway hyperresponsiveness and a much smaller -0.71% deviation among children with asthma without airway hyperresponsiveness. In a group of 86 children with asthma in Detroit, MI, Lewis et al. (2005) reported that associations between ambient

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1Effect estimates were standardized to a 40-, 30-, and 20-ppb increase for 1-h max, 8-h max or 8-h avg, and 24-h avg O₃.
O₃ concentration and FEV₁ were confined largely to children with asthma who used CS or had a concurrent URI, 7.3% and 4.9% decreases, respectively, in the mean of lowest daily FEV₁ per 30-ppb increase in 8-h max ambient O₃ concentration.

Heterogeneity in response to ambient O₃ exposure also was demonstrated by observations that some children with asthma experienced larger O₃-associated lung function decrements than the population mean effect estimate. Similar observations were made in controlled human exposure studies (Section 6.2.1.1). Mortimer et al. (2002) found that for a 30-ppb increase in lag 1-5 avg of 8-h avg O₃, there was a 30% ([95% CI: 4%, 61%]) higher incidence of >10% decline in PEF. Likewise, Hoppe et al. (2003) found that while the percentages of O₃-associated lung function decrements were variable and small, 47% of children with asthma experienced a >10% decline in FEV₁, FVC, or PEF or 20% increase in airway resistance on days with 30-min (1-4 p.m.) max ambient O₃ concentrations >50 ppb relative to days with <40 ppb O₃.

Effect Modification

Effect modification by corticosteroid use

In controlled human exposure studies, CS treatment of subjects with asthma generally has not prevented O₃-induced FEV₁ decrements (Section 6.2.1.1). Epidemiologic evidence is equivocal, with findings that use of inhaled CS attenuated (Hernández-Cadena et al., 2009), increased (Lewis et al., 2005), and did not affect (Mortimer et al., 2000) ambient O₃-associated lung function decrements. In winter-only studies, consideration of CS use largely did not influence associations between ambient O₃ and various lung function indices (Liu et al., 2009a; Rabinovitch et al., 2004). Similarly equivocal epidemiologic evidence was found for modification of associations with respiratory symptoms (Section 6.2.4.1). The assessment of effect modification by CS use has been hampered by differences in the severity of asthma among CS users and the definition of CS use. Additionally, investigators did not assess adherence to reported CS regimen, and misclassification of CS use may bias findings. For example, Mortimer et al. (2000) classified children by no or any CS use at baseline but did not measure daily use during the study period. Lewis et al. (2005) defined CS use as use for at least 50% of study days and estimated larger O₃-associated FEV₁ decrements among CS users (Figure 6-7 [and Table 6-8]) than among CS nonusers (quantitative results not reported). In this study, most children with moderate to severe asthma (91%) were classified as CS users. However, CS users had a higher percent predicted FEV₁. In contrast, Hernández-Cadena et al. (2009) observed larger O₃-related decrements in FEV₁ among the 60 CS nonusers than among the 25 CS users. A definition for CS use was not provided; however, children with persistent asthma were included among the group of CS nonusers. Thus, across studies, both CS use and nonuse have been used to indicate more severe, uncontrolled asthma.
Effect modification by antioxidant capacity

Ozone is a powerful oxidant whose secondary oxidation products have been described to initiate the key modes of action that mediate decreases in lung function, including the activation of neural reflexes (Section 5.3.2). Additionally, O₃ exposure of humans and animals has induced changes in the levels of antioxidants in the ELF (Section 5.3.3). These observations provide biological plausibility for diminished antioxidant capacity to increase the risk of O₃-associated respiratory effects and for augmented antioxidant capacity to decrease risk.

Antioxidant supplementation

Controlled human exposure studies have demonstrated the protective effects of α-tocopherol (vitamin E) and ascorbate (vitamin C) supplementation on O₃-induced lung function decrements (Section 6.2.1.1), and an epidemiologic study of children with asthma conducted in Mexico City, Mexico, produced similar findings. Particularly among children with moderate to severe asthma, an increase in ambient O₃ concentration was associated with a smaller decrease in FEV₁ in the group supplemented with vitamin C and E as compared with the placebo group (Romieu et al., 2002) (Figure 6-7 [and Table 6-8]). Romieu et al. (2009) also demonstrated an interaction between dietary antioxidant intake and ambient O₃ concentrations by finding that the main effect of diet was modified by ambient O₃ concentrations. Diets high in antioxidant vitamins and/or omega-3 fatty acids protected against FEV₁ decrements at 8-h max O₃ concentrations ≥ 38 ppb. Results for the main effect of O₃ on FEV₁ or effect modification by diet were not presented.

Genetic polymorphisms

Antioxidant capacity also can be characterized by variants in genes encoding oxidant metabolizing enzymes with altered enzymatic activity. A potential role for such genetic variants in modifying O₃-associated health effects is biologically plausible given the well-characterized evidence for the secondary oxidation products of O₃ mediating downstream effects and has been indicated in some epidemiologic studies. Specifically, ambient O₃-associated FEF₂₅₋₇₅% decrements were larger among children with asthma with the GSTM1 null genotype, which is associated with lack of oxidant metabolizing activity (Romieu et al., 2004b). The difference in association between GSTM1 null and positive subjects was minimal in children supplemented with antioxidant vitamins (Figure 6-8 [and Table 6-9]). Controlled human exposure studies have not consistently found larger O₃-induced lung function decrements in GSTM1 null subjects (Section 6.2.1.1). Effect modification by GSTP1 variants is less clear. Romieu et al. (2006) observed larger O₃-associated decreases in FEV₁ in children with asthma with the GSTP1 Ile/Ile or Ile/Val variant, which are associated with relatively higher oxidative metabolism activity (Figure 6-7 [and Table 6-8]). An increase in ambient O₃ concentration was associated with an increase in FEV₁ among children with the GSTP1 Val/Val variant, which is associated with reduced oxidative metabolism. Rather than reflecting effect modification by the GSTP1 variant, these results may reflect effect modification by asthma severity, as 77% of
subjects with the GSTP1 Ile/Ile genotype had moderate to severe asthma. In support of this alternate hypothesis, another analysis of the same cohort indicated a larger \( O_3 \)-associated decrement in \( FEV_1 \) among children with moderate to severe asthma than among all children with asthma (Romieu et al., 2002).

**Exposure Measurement Error**

Across the studies of children with asthma, lung function decrements were associated with ambient \( O_3 \) concentrations assigned to subjects using various exposure assessment methods. As described in Section 4.3.3, exposure measurement error due to use of ambient concentrations measured at central sites has varied, depending on the population and season examined. Because there are a limited number of studies of each method, it is difficult to conclude that a particular method of exposure assessment produced stronger results.

Seasonal differences have been observed in the personal-ambient \( O_3 \) relationship (Section 4.3.3); however, in children with asthma, \( O_3 \)-associated lung function decrements were found in studies conducted in summer months and over multiple seasons. Lung function was associated with \( O_3 \) measured on site of subjects’ daytime hours in summer months (Hoppe et al., 2003; Thurston et al., 1997), factors that have contributed to higher personal-ambient \( O_3 \) ratios and correlations. Many year-round studies in Mexico City, Mexico (Romieu et al., 2006; 2004b; 2002; 1997; 1996), and a study in Detroit, MI (Lewis et al., 2005) found associations with \( O_3 \) measured at sites within 5 km of children’s home or school. Children with asthma examined by Romieu et al. (2006); (2004b; 2002) had a personal-ambient ratio and correlation for 48- to 72-h avg \( O_3 \) concentrations of 0.17 and 0.35, respectively (Ramírez-Aguilar et al., 2008). These findings indicate that the effects of personal \( O_3 \) exposure on lung function decrements may have been underestimated in the children in Mexico City. Associations were found with \( O_3 \) concentrations averaged across multiple community monitoring sites (O’Connor et al., 2008; Just et al., 2002; Mortimer et al., 2002; Jalaludin et al., 2000) and measured at a single site (Gielen et al., 1997), which may be attributable to observations of high temporal correlation among \( O_3 \) concentrations measured at multiple sites within a region (Darrow et al., 2011a; Gent et al., 2003).

Studies of children with asthma restricted to winter months provided little evidence of an association between various single- and multi-day lags of ambient \( O_3 \) concentration and lung function decrements with several observations of \( O_3 \)-associated increases in lung function (Dales et al., 2009; Liu et al., 2009a; Rabinovitch et al., 2004). One explanation for these results may be lower indoor than outdoor \( O_3 \) concentrations, variable indoor to outdoor ratios, and lower correlations between personal and ambient \( O_3 \) concentrations in non-summer months (Sections 4.3.2 and 4.3.3). As noted for other respiratory endpoints such as respiratory hospital admissions, ED visits, and mortality, associations with \( O_3 \) generally are lower in colder seasons.
**Adults with Respiratory Disease**

Relative to studies in children with asthma, studies of adults with asthma or COPD have been limited in number. Details from these studies regarding location, time period, and ambient O$_3$ concentrations are presented in Table 6-10. Increases in ambient O$_3$ concentration were not consistently associated with lung function decrements in adults with respiratory disease. Several different exposure assessment methods were used, including monitoring personal exposures (Delfino et al., 1997), monitoring on site of outdoor activity (Girardot et al., 2006; Korrick et al., 1998), and using measurements from one (Peacock et al., 2011; Wiwatanadate and Liwsrisakun, 2011; Thaller et al., 2008; Ross et al., 2002) to several central monitors (Khatri et al., 2009; Lagorio et al., 2006; Park et al., 2005a). There was not a clear indication that differences in exposure assessment methodology contributed to inconsistencies in findings.
Table 6-10  Mean and upper percentile concentrations of O₃ in epidemiologic studies of lung function in adults with respiratory disease.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O₃ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Delfino et al.</td>
<td>Alpine, CA</td>
<td>May-July 1994</td>
<td>12-h avg personal (8 a.m.-8 p.m.)</td>
<td>18</td>
<td>90th: 38 Max: 80</td>
</tr>
<tr>
<td>Girardot et al.</td>
<td>Great Smoky Mountain NP, TN</td>
<td>August-October 2002 June-August 2003</td>
<td>Hike-time avg (2-9 h)</td>
<td>48.1a</td>
<td>Max: 74.2a</td>
</tr>
<tr>
<td>Korrick et al.</td>
<td>Mt. Washington, NH</td>
<td>Summer 1991, 1992</td>
<td>Hike-time avg (2-12 h)</td>
<td>40a</td>
<td>Max: 74a</td>
</tr>
<tr>
<td>Wiwatanadate and Liwsrisakun</td>
<td>Chiang Mai, Thailand</td>
<td>August 2005-June 2006</td>
<td>24-h avg</td>
<td>17.5</td>
<td>90th: 26.8 Max: 34.7</td>
</tr>
<tr>
<td>Thaller et al.</td>
<td>Galveston, TX</td>
<td>Summer 2002-2004</td>
<td>1-h max</td>
<td>35 (median)</td>
<td>Max: 118</td>
</tr>
<tr>
<td>Ross et al.</td>
<td>East Moline, IL</td>
<td>April-October 1994</td>
<td>8-h avg</td>
<td>41.5</td>
<td>Max: 78.3</td>
</tr>
<tr>
<td>Khatri et al.</td>
<td>Atlanta, GA</td>
<td>May-September 2003, 2005, 2006</td>
<td>8-h max</td>
<td>With asthma: 61 (median)b No asthma: 56 (median)c</td>
<td>75th (with asthma): 74b 75th (no asthma): 64b</td>
</tr>
<tr>
<td>Lagorio et al.</td>
<td>Rome, Italy</td>
<td>May-June, November-December 1999</td>
<td>24-h avg</td>
<td>Spring: 36.2c Winter: 8.2c</td>
<td>Overall max: 48.6c</td>
</tr>
<tr>
<td>Park et al.</td>
<td>Incheon, Korea</td>
<td>March-June 2002</td>
<td>24-h avg</td>
<td>Dust event days: 23.6 Control days: 25.1</td>
<td>NR</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.
NR = Not reported.

Individual-level estimates calculated from concentrations measured in different segments of hiking trail.

[1] Individual-level estimates calculated based on time spent in the vicinity of various O₃ monitors.

Comparisons of adults with asthma (8-18% of study population) and without asthma did not conclusively demonstrate that adults with asthma had larger ambient O₃-associated lung function decrements. Several studies examined on-site or central-site ambient O₃ concentrations measured while subjects were outdoors. Ambient O₃ measured during time spent outdoors has been closer in magnitude and more correlated with personal exposures (Section 4.3.3). In a panel study of lifeguards (ages 16-27 years) in Galveston, TX, a larger O₃-associated decrement in FEV₁/FVC was found among the 16 lifeguards with asthma (-1.6% [95% CI: -2.8, -0.4] per 40 ppb increase in 1-h max O₃) than among the 126 lifeguards without asthma (-0.40% [95% CI: -0.80, 0] per 40-ppb increase in 1-h max O₃) (Brooks, 2010). In Korrick et al. (1998), hikers with a history of asthma or wheeze had larger...
O₃-associated lung function decrements (e.g., -4.4% [95% CI: -7.5, -1.2] in FEV₁ per 30-ppb increase in 2-12 h avg O₃). In contrast, Girardot et al. (2006) generally did not find O₃-associated lung function decrements in hikers with or without respiratory disease history. In a cross-sectional study of 38 adults with asthma and 13 adults without asthma, Khatri et al. (2009) used central site O₃ measurements but aimed to account for spatial variability by calculating an average of concentrations measured at sites closest to each subject’s location during each hour. Investigators reported a larger O₃-associated decrease in percent predicted FEV₁/FVC in the 38 subjects with atopy (with or without asthma) (-12 points [95% CI: -21, -3] per 30-ppb increase in lag 2 of 8-h max O₃) than in subjects with asthma (-4.7 points [95% CI: -12, 2.3]). Among adults with asthma, O₃ was associated with an increase in FEV₁.

In panel studies that exclusively examined adults with asthma, increases in ambient O₃ concentrations, across the multiple lags examined, generally were associated with increases in lung function (Wiwatanadate and Liwsrisakun, 2011; Lagorio et al., 2006; Park et al., 2005a). These studies were conducted in Europe and Asia during periods of low ambient O₃ concentrations, including one conducted in Korea during a period of dust storms (Park et al., 2005a).

Some studies included children and adults with asthma. Among subjects ages 9-46 years (41% adults) in Alpine, CA with low personal 12-h avg O₃ exposures (55% samples below limit of detection) and a majority of sampling hours spent indoors (mean 71%), Delfino et al. (1997) reported that neither increases in 12-h avg personal exposure nor increases in ambient O₃ concentration were associated with decreases in PEF. Ross et al. (2002) examined subjects ages 5-49 years (proportion of adults not reported) in East Moline, IL and found that a 20-ppb increase in lag 0 (of 24-h avg O₃) was associated with a 2.6 L/min decrease (95% CI: -4.3, -0.90) in evening PEF. In this population with asthma, an increase in lag 0 ozone also was associated with an increase in symptom score.

Controlled human exposure studies have found diminished, statistically nonsignificant O₃-induced lung function responses in older adults with COPD (Section 6.2.1.1). Similarly, epidemiologic studies do not provide strong evidence that short-term increases in ambient O₃ exposure result in lung function decrements in adults with COPD. Inconsistent associations were reported for PEF, FEV₁, and FVC in a study that followed 94 adults with COPD (ages 40-83 years) in London, England daily over two years (Peacock et al., 2011). For example, an increase in lag 1 of 8-h max O₃ was associated with a decrease in PEF in an analysis of summer 1996 (-1.7 L/min [95% CI: -3.1, -0.39] per 30-ppb increase O₃), but the association was near null and imprecise in summer 1997 (-0.21 L/min [95% CI: -2.4, 2.0]). Further, in this study, an increase in ambient O₃ concentration was associated with lower odds of a large PEF decrement (OR for a >20% drop from an individual’s median value: 0.89 [95% CI: 0.72, 1.10] per 30-ppb increase in lag 1 of 8-h max O₃) and was not consistently associated with increases in respiratory symptoms (Peacock et al., 2011). Inconsistent associations also were reported in a small panel study of 11 adults with COPD (mean age 67 years) in Rome, Italy (Lagorio et al., 2006).
Populations Not Restricted to Individuals with Asthma

Several studies have examined associations between ambient O_3 concentrations and lung function in groups that included children with and without asthma; however, a limited number of studies have examined groups of children or adults restricted to healthy individuals. Details from studies not restricted to individuals with asthma regarding location, time period, and ambient O_3 concentrations are presented in Table 6-11.
Table 6-11  Mean and upper percentile concentrations of O₃ in epidemiologic studies of lung function in populations not restricted to individuals with asthma.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O₃ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Avol et al. (1998b)</td>
<td>6 southern CA communities</td>
<td>Spring and summer 1994</td>
<td>24-h avg personal</td>
<td>NR</td>
<td>NR</td>
</tr>
<tr>
<td>Hoppe et al. (2003)</td>
<td>Munich, Germany</td>
<td>Summers 1992-1995</td>
<td>30-min max (1-4 p.m.)</td>
<td>High O₃ days: 70.4* Control O₃ days: 29.8*</td>
<td>Max (high O₃ days): 99* Max (control O₃ days): 39*</td>
</tr>
<tr>
<td>Chen et al. (1999)</td>
<td>3 Taiwanese communities</td>
<td>May 1995-January 1996</td>
<td>1-h max (8 a.m.-6 p.m.)</td>
<td>NR</td>
<td>Max: 110.3*</td>
</tr>
<tr>
<td>Gold et al. (1999)</td>
<td>Mexico City, Mexico</td>
<td>January-November 1991</td>
<td>24-h avg</td>
<td>52.0*</td>
<td>Max: 103*</td>
</tr>
<tr>
<td>Ulmer et al. (1997)</td>
<td>Freudenstadt and Villingen, Germany</td>
<td>March-October 1994</td>
<td>30-min avg</td>
<td>Freudenstadt median: 50.6 Villingen median: 32.1</td>
<td>Freudenstadt 95th: 89.8 Villingen 95th: 70.1</td>
</tr>
<tr>
<td>Linn et al. (1996)</td>
<td>Rubidoux, Upland, Torrance, CA</td>
<td>September-June 1992-1994</td>
<td>24-h avg personal 24-h avg central site</td>
<td>5 23</td>
<td>Max: 16 Max: 53</td>
</tr>
<tr>
<td>Neuberger et al. (2004)</td>
<td>Vienna, Austria</td>
<td>June-October 1999, January-April 2000</td>
<td>8-h max</td>
<td>24.4*</td>
<td>NR</td>
</tr>
<tr>
<td>Alexeeff et al. (2008); (2007)</td>
<td>Greater Boston, MA; NAS</td>
<td>January 1995-June 2005</td>
<td>48-h avg</td>
<td>41.1</td>
<td>75th: 48.7 Max: 72.8</td>
</tr>
<tr>
<td>Steinvil et al. (2009)</td>
<td>Tel Aviv, Israel</td>
<td>September 2002-November 2007</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>34.9</td>
<td>Max: 56.6</td>
</tr>
<tr>
<td>Naeher et al. (1999)</td>
<td>Multiple communities, VA</td>
<td>May-September 1995-1996</td>
<td>24-h avg</td>
<td>35.9 (avg of 13 monitors)</td>
<td>Max: 59.5</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section. NR = Not Reported, NAS = Normative Aging Study.
*Measured at subjects’ schools where lung function was measured. Measured at central monitoring sites established by investigators. Concentrations were averaged across four monitors.

**Children**

Based on studies available at the time of the 2006 O₃ AQCD, evidence consistently links increases in ambient O₃ concentration with decrements in FEV₁ and PEF in children (U.S. EPA, 2006b) (Figure 6-9 [and Table 6-12]). These associations were
found with personal O₃ exposures (Avol et al., 1998b), ambient O₃ measured at children’s schools where lung function was measured (Hoppe et al., 2003; Chen et al., 1999; Gold et al., 1999), and ambient O₃ measured at sites within the community (Ward et al., 2002; Ulmer et al., 1997; Linn et al., 1996). Among children in California who spent a mean 2-3 hours per day outdoors and whose personal-ambient O₃ correlation was 0.28 across multiple seasons, Avol et al. (1998b) found slightly larger O₃-associated decrements in FEV₁ and FVC for 24-h avg personal exposures than for 1-h max ambient measurements (Figure 6-9 [and Table 6-12]). The effect estimates for personal exposures were similar in magnitude to those found in other studies of children for ambient O₃ measured at schools (Hoppe et al., 2003; Chen et al., 1999). In another study of children in California, Linn et al. (1996) did not present results for personal O₃ exposures but found FEV₁ decrements in association with increases in ambient O₃ concentration in children who spent 1-2 hours per day outdoors and whose personal-ambient correlation was 0.61. Because of between-study heterogeneity in populations and ambient O₃ concentrations examined, it is difficult to assess how the method of exposure assessment may have influenced findings.

### Table 6-12

<table>
<thead>
<tr>
<th>Study</th>
<th>Parameter</th>
<th>O₃ Lag</th>
</tr>
</thead>
<tbody>
<tr>
<td>Linn et al. (1996)</td>
<td>Intraday change FEV₁</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>Intraday change FVC</td>
<td></td>
</tr>
<tr>
<td>Hoppe et al. (2003)</td>
<td>Afternoon FEV₁</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>Afternoon FVC</td>
<td></td>
</tr>
<tr>
<td>Scarlett et al. (1996)</td>
<td>FEV₀.₇₅</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>FVC</td>
<td></td>
</tr>
<tr>
<td>Chen et al. (1999)</td>
<td>FEV₁</td>
<td>1</td>
</tr>
<tr>
<td></td>
<td>FVC</td>
<td></td>
</tr>
<tr>
<td>Avol et al. (1998)</td>
<td>Intraday change FEV₁</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>Personal Ambient</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Intraday change FVC</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Personal Ambient</td>
<td></td>
</tr>
</tbody>
</table>

Note: Results generally are presented in order of increasing mean ambient O₃ concentration. Effect estimates are from single-pollutant models and are standardized to a 40-, 30-, and 20-ppb increase for 1-h (or 30-min) max, 8-h max, and 24-h avg O₃ concentrations, respectively.

*The 95% CI was constructed using a standard error that was estimated from the p-value.

**Figure 6-9** Percent change in FEV₁ or FVC in association with ambient O₃ concentrations in studies of children in the general population.
### Table 6-12  Percent change in FEV₁ or FVC in association with ambient O₃ concentrations in studies of children in the general population presented in Figure 6-9 plus others.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/ Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag Parameter</th>
<th>Standardized Percent Change (95% CI)a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Linn et al. (1996)</td>
<td>3 southern CA communities, 269 children, 4th and 5th grades</td>
<td>24-h avg</td>
<td>Intraday change FEV₁, Intraday change FVC</td>
<td>-0.58 (-1.0, -0.13)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.21 (-0.62, 0.20)</td>
</tr>
<tr>
<td>Hoppe et al. (2003)</td>
<td>Munich, Germany, 44 children, ages 6-8 yr</td>
<td>30-min max (1 - 4 p.m.)</td>
<td>Afternoon FEV₁, Afternoon FVC</td>
<td>-0.14 (-2.7, 2.5)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-1.4 (-3.9, 1.2)</td>
</tr>
<tr>
<td>Scarlett et al. (1996)</td>
<td>Surrey, England, 154 children, ages 7-11 yr</td>
<td>8-h max</td>
<td>FEV₀.₇₅, FVC</td>
<td>-0.04 (-0.32, 0.23)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>0.06 (-0.21, 0.33)</td>
</tr>
<tr>
<td>Chen et al. (1999)</td>
<td>3 Taiwan communities, 941 children, ages 8-13 yr</td>
<td>1-h max (8 a.m.-8 p.m.)</td>
<td>FEV₁, FVC</td>
<td>-1.5 (-2.8, -0.12)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-1.6 (-2.9, -0.33)</td>
</tr>
<tr>
<td>Avol et al. (1998b)</td>
<td>3 southern CA communities, 195 children, ages 10-12 yr</td>
<td>24-h avg personal, 1-h max ambient, 24-h avg personal, 1-h max ambient</td>
<td>Intraday change FEV₁, Intraday change FEV₁, Intraday change FVC, Intraday change FVC</td>
<td>-0.85 (-2.1, 0.42)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.49 (-1.5, 0.57)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-1.0 (-2.0, 0)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.50 (-1.3, 0.35)</td>
</tr>
</tbody>
</table>

**Studies of children not included in Figure 6-9c**

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/ Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag Parameter</th>
<th>Standardized Percent Change (95% CI)b</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ulmer et al. (1997)</td>
<td>Freudenstadt and Villingen, Germany, 135 children, ages 8-10 yr</td>
<td>30-min max</td>
<td>FEV₁ (mL)</td>
<td>-57 (-102, 13)</td>
</tr>
<tr>
<td>Ward et al. (2002)</td>
<td>Birmingham and Sandwell, England, 162 children, age 9 yr</td>
<td>24-h avg</td>
<td>Daily deviation from mean morning PEF (L/min)</td>
<td>-3.2 (-8.3, 2.0)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-6.7 (-12, -1.4)</td>
</tr>
<tr>
<td>Gold et al. (1999)</td>
<td>Mexico City, Mexico, 40 children, ages 8-11 yr</td>
<td>24-h avg</td>
<td>Intraday change PEF (% change)</td>
<td>-0.47 (-1.1, 0.11)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-3.8 (-6.7, -0.94)</td>
</tr>
</tbody>
</table>

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*a Effect estimates are standardized to a 40-, 30-, and 20-ppb increase for 1-h (or 30-min) max, 8-h max, and 24-h avg O₃, respectively.

*b The 95% CI was constructed using a standard error that was estimated from the p-value.

*c Results are not presented in Figure 6-9 because sufficient data were not available to calculate percent change in FEV₁, or PEF was analyzed.

*d Effect estimates are from analyses restricted to summer months.

In the limited number of studies that examined only healthy children, increases in ambient O₃ concentration were associated with decreases (Hoppe et al., 2003) or no change in lung function (Neuberger et al., 2004). Several studies that included small proportions (4-10%) of children with history of respiratory disease or symptoms found associations between increases in ambient O₃ concentration and lung function decrements (Chen et al., 1999; Ulmer et al., 1997; Scarlett et al., 1996). Based on analysis of interaction terms for O₃ concentration and asthma/wheeze history, Avol et al. (1998b) and Ward et al. (2002) did not find differences in O₃-associated lung function decrements between children with history of asthma or wheeze and healthy children. Combined, these lines of evidence suggest that the ambient O₃-associated
lung function decrements found in children overall were not solely due to effects in children with asthma, and that increases in ambient O₃ exposure may decrease lung function in healthy children.

Among the studies of children, the magnitudes of decrease in lung function per unit increase in ambient O₃ concentration¹ ranged from <1 to 4%, a range similar to that estimated in children with asthma. Comparable data were not adequately available to assess whether mean lung function differed between groups of children with asthma and healthy children. In contrast with studies of children with asthma, studies of children in the general population did not consistently find both O₃-associated decreases in lung function and O₃-associated increases in respiratory symptoms. For example, Gold et al. (1999) found O₃-associated decreases in PEF and increases in phlegm; however, the increase in phlegm was associated with lag 1 O₃ concentrations whereas the PEF decrement was found with single-day lags 2 to 4 of O₃. Also, O₃ was weakly associated with cough and shortness of breath among children in England (Ward et al., 2002) and was associated with a decrease in respiratory symptom score among children in California (Linn et al., 1996).

**Adults**

Compared with children, in a smaller body of studies, O₃ was less consistently associated with lung function decrements in populations of adults not restricted to those with asthma (Table 6-13). In a study that included only healthy adults, increases in ambient O₃ concentration were associated with decreases and increases in lung function across the various lags of exposure examined (Steinvil et al., 2009). Contrasting results also were found in studies of older adults (Alexeeff et al., 2008; Alexeeff et al., 2007; Hoppe et al., 2003).

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¹Effect estimates were standardized to a 40-, 30-, and 20-ppb increase for 1-h max (or 30-min max), 8-h max, and 24-h avg O₃.
**Table 6-13**  Associations between ambient O₃ concentration and lung function in studies of adults.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location/Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag</th>
<th>Parameter</th>
<th>O₃ Assessment Method/Subgroup</th>
<th>Standardized Effect Estimate (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Son et al. (2010)</td>
<td>Ulsan, Korea 2,102 children and adults, ages 7-97 yr</td>
<td>8-h max 0-2 avg</td>
<td>Change in percent predicted FEV₁</td>
<td>All monitor avg</td>
<td>-1.4 (-2.7, -0.08)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Nearest monitor</td>
<td>-0.76 (-1.8, 0.25)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>IDW</td>
<td>-1.1 (-2.2, 0.05)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Kriging</td>
<td>-1.4 (-2.6, -0.11)</td>
</tr>
<tr>
<td>Steinvil et al. (2009)</td>
<td>Tel Aviv, Israel 2,380 healthy adults, mean age 43 yr, 75th percentile: 52 yr</td>
<td>8-h avg (10 a.m. - 6 p.m.) 0-6 avg</td>
<td>FEV₁ (mL)</td>
<td></td>
<td>60 (0, 120)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>141 (49, 234)</td>
</tr>
<tr>
<td>Naeher et al. (1999)</td>
<td>Multiple communities, VA 473 women, ages 19 - 43 yr</td>
<td>24-h avg 0-2 avg</td>
<td>Evening PEF (L/min)</td>
<td></td>
<td>-1.7 (-3.4, 0.03)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Munich, Germany 41 older adults, ages 69 - 95 yr</td>
<td>30-min max (1-4 p.m.) 0-1 avg</td>
<td>% change in afternoon FEV₁</td>
<td></td>
<td>0.75 (-2.1, 3.7)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Greater Boston, MA; NAS 1,015 older adults, mean (SD) age: 68.8 (7.2) yr at baseline</td>
<td>24-h avg 0-1 avg</td>
<td>% change in FEV₁ GSTP1 Ile/Ile GSTP1 Ile/Val or Val/Val</td>
<td>-1.0 (-2.2, 0.20)</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Greater Boston, MA; NAS 904 older adults, mean (SD) age: 68.8 (7.3) yr at baseline</td>
<td>24-h avg 0-1 avg</td>
<td>% change in FEV₁ BMI &lt;30 BMI ≥ 30 No AHR AHR BMI ≥ 30 and AHR</td>
<td>-5.3 (-8.2, -2.3)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

IDW = Inverse distance weighting, NAS = Normative Aging Study, BMI = Body mass index, AHR = airway hyperresponsiveness.

Results generally are presented in order of increasing mean ambient O₃ concentration.

Effect estimates are standardized to a 40-ppb increase for 30-min max O₃, 30-ppb increase for 8-h max or 8-h avg O₃, and 20-ppb increase for 24-h avg O₃.

Despite mixed results overall, studies that found ambient O₃-associated lung function decrements in adults used various exposure assessment methods with potentially varying degrees of measurement error. These methods included the average of multiple intra-city monitors, nearest monitor, estimates from spatial interpolation (Son et al., 2010), average of monitors in multiple towns (Alexeeff et al., 2008; Alexeeff et al., 2007), and one site for multiple towns (Naeher et al., 1999). In a large cross-sectional study, conducted in 2,102 children and adults (mean age: 45 years) living near a petrochemical plant in Ulsan, Korea, Son et al. (2010) did not find a consistent difference in the magnitude of association with lung function among ambient O₃ concentrations averaged across 13 city monitors, concentrations from the nearest monitor, inverse distance-weighted concentrations, and estimates from kriging across the various lags examined (Table 6-13). Ozone concentrations were similar (<10% difference) and highly correlated (r = 0.84 – 0.96) among the methods.
Although the health status of subjects was not reported, the study population mean percent predicted FEV₁ was 82.85%, indicating a large proportion of subjects with underlying airway obstruction. Results from this study were not adjusted for meteorological factors and thus, confounding cannot be ruled out. Importantly, the similarities among exposure assessment methods in Son et al. (2010) may apply mostly to populations living within the same region of a city. The majority of women examined by Naether et al. (1999) lived >60 miles from the single available central site monitor. However, in the nonurban (southwest Virginia) study area, O₃ concentrations may be more spatially homogeneous (Section 4.6.2.1), and the concentrations measured at the single site may capture temporal variability in ambient exposures.

The inconsistent epidemiologic findings for older adults parallel observations from controlled human exposure studies (Section 6.2.1.1). In a study that followed adults ages 69-95 years during several summers in Germany, Hoppe et al. (2003) did not find decreases in lung function in association with ambient O₃ measured at subjects’ retirement home. However, recently, the Normative Aging Study found decrements in FEV₁ and FVC in a group of older men (mean [SD] age = 68.9 [7.2] years at first lung function measurement) in association with ambient O₃ concentrations averaged from four town-specific monitors (Alexeeff et al., 2008), which may less well represent spatial heterogeneity in ambient O₃ exposures. Among all subjects, who were examined once every three years for ten years, associations were found with several lags of 24-h avg O₃ concentration, i.e., 1- to 7-day avg (Alexeeff et al., 2008). Additionally, larger effects were estimated in adults with airway hyperresponsiveness, higher BMI (≥ 30), and GSTP1 Ile/Val or Val/Val genetic variants (Val/Val variant produces enzyme with reduced oxidative metabolism activity) (Alexeeff et al., 2008; Alexeeff et al., 2007) (Table 6-13). Larger O₃-related decrements in FEV₁ and FVC also were observed in subjects with long GT dinucleotide repeats in the promoter region of the gene for the antioxidant enzyme heme oxygenase-1 (Alexeeff et al., 2008), which has been associated with reduced inducibility (Hiltermann et al., 1998). In this cohort, O₃ also was associated with decreases in lung function in adults without airway hyperresponsiveness and those with BMI <30, indicating effects of O₃ on lung function in healthy men within the cohort. However, the findings may be generalizable only to this study population of older, predominately white men.

Confounding in epidemiologic studies of lung function

The 1996 O₃ AQCD noted uncertainty regarding confounding by temperature and pollen (U.S. EPA, 1996a); however, collective evidence does not indicate that these factors fully account for the associations observed between increases in ambient O₃ concentration and lung function decrements. Across the populations examined, most studies that found ambient O₃-associated lung function decrements, whether conducted in multiple seasons or only in summer, included temperature in statistical analyses. Some summer camp studies conducted detailed analysis of temperature. In most of these studies, temperature and O₃ were measured at the camps. In two
Northeast U.S. studies, an increase in temperature was associated with an increase in lung function (Thurston et al., 1997; Berry et al., 1991). This positive association likely accounted for the nearly 2-fold greater decrease in O₃-associated PEF found by Thurston et al. (1997) with temperature in the model than with O₃ alone.

In another Northeast U.S. camp study, Spektor et al. (1988a) estimated similar effects for O₃ in a model with and without a temperature-humidity index. In the re-analysis of six camp studies, investigators did not include temperature in models because temperature within the normal ambient range had not been shown to affect O₃-induced lung function responses in controlled human exposure studies (Kinney et al., 1996).

Pollen was evaluated in far fewer studies. Camp studies that examined pollen found that pollen independently was not associated with lung function decrements (Thurston et al., 1997; Avol et al., 1990). Many studies of individuals with asthma with follow-up over multiple seasons found O₃-associated decrements in lung function in models that adjusted for pollen counts (Just et al., 2002; Ross et al., 2002; Jalaludin et al., 2000; Gielen et al., 1997). In these studies, large proportions of subjects had atopy (22-98%), with some studies examining large proportions of subjects specifically with pollen allergy who would be more responsive to pollen exposure (Ross et al., 2002; Gielen et al., 1997).

A relatively large number of studies provided information on potential confounding by copollutants such as PM₂.₅, PM₁₀, NO₂, or SO₂. While studies were varied in how they evaluated confounding, most indicated that O₃-associated lung function decrements were not solely due to copollutant confounding. Some studies of subjects exercising outdoors indicated that ambient concentrations of copollutants such as NO₂, SO₂, or acid aerosol were low and thus, not likely to confound associations observed for O₃ (Hoppe et al., 2003; Brunekreef et al., 1994; Hoek et al., 1993).

In other studies of children with increased outdoor exposures, O₃ was consistently associated with decreases in lung function, whereas other pollutants such as PM₂.₅, sulfate, and acid aerosol individually showed variable associations across studies (Thurston et al., 1997; Castillejos et al., 1995; Berry et al., 1991; Avol et al., 1990; Spektor et al., 1988a). Most of these studies measured ambient pollutants on site of subjects’ outdoor activity and related lung function changes to the pollutant concentrations measured during outdoor activity. Thus, the degree of exposure measurement error likely is comparable for O₃ and copollutants.

Studies that conducted copollutant modeling generally found O₃-associated lung function decrements to be robust; i.e., most copollutant-adjusted effect estimates fell within the 95% CI of the single-pollutant effect estimates (Figure 6-10 [and Table 6-14]). These studies used central site measurements for both O₃ and copollutants. There may be residual confounding because of differential exposure measurement error for O₃ and copollutants due to differing spatial heterogeneity and indoor-outdoor ratios; however, the limited available evidence indicates that personal O₃ exposures are weakly correlated with personal PM₂.₅ and NO₂ exposures (Section 4.3.4.1). Whereas a few studies used the same averaging time for copollutants (Lewis et al., 2005; Jalaludin et al., 2000), more examined 1-h max or
8-h max O₃ and 24-h avg copollutant concentrations (Son et al., 2010; Chen et al., 1999; Romieu et al., 1997; Romieu et al., 1996). In a Philadelphia-area summer camp study, Neas et al. (1999) was among the few studies to find that the effect estimate for O₃ was attenuated to near zero in a copollutant model (24-h avg sulfate in this study) (Figure 6-10 [and Table 6-14]).

Ambient O₃ concentrations showed a wide range of correlations with copollutant concentrations (r = -0.31 to 0.74). In Sydney, Australia, Jalaludin et al. (2000) found low correlations of O₃ with PM₁₀ (r = 0.13) and NO₂ (r = -0.31), all averaged over 24 hours. In two-pollutant models, PM₁₀ and NO₂ remained associated with increases in PEF, and O₃ remained associated with decreases in PEF in children with asthma. In Detroit, MI, O₃ was moderately correlated with PM₂.₅ (Pearson r = 0.57) and PM₁₀ (Pearson r = 0.59), all averaged over 24 hours (Lewis et al., 2005).

Adjustment for PM₁₀ resulted in a large (more negative) change in the O₃-associated FEV₁ decrement in children with asthma, but only in CS users and not in children with concurrent URI (Figure 6-10 [and Table 6-14]). Studies conducted in Mexico City, Mexico, found small changes in O₃-associated PEF decrements with copollutant adjustment although different averaging times were used for copollutants (Romieu et al., 1997; Romieu et al., 1996) (Figure 6-10 [Table 6-14]). In these studies, O₃ was moderately correlated with copollutants such as NO₂ and PM₁₀ (range of Pearson r = 0.38 - 0.58). Studies conducted in Asia also found that associations between O₃ and lung function were robust to adjustment for weakly- to moderately-correlated copollutants; effect estimates for copollutants generally were attenuated, indicating that O₃ may confound associations of copollutants (Son et al., 2010; Chen et al., 1999).

In a summer camp study conducted in Connecticut, Thurston et al. (1997) found ambient concentrations of 1-h max O₃ and 12-h avg sulfate to be highly correlated (r = 0.74), making it difficult to separate their independent effects. With sulfate in the model, a larger decrease in PEF was estimated for O₃; however, the 95% CI was much wider (Figure 6-10 [and Table 6-14]). Investigators found that the association for sulfate was due to one day when the ambient concentrations of both pollutants were at their peak. With the removal of this peak day, the sulfate effect was attenuated, whereas O₃ effects remained robust (Thurston et al., 1997). Among children with asthma in Thailand, the O₃-associated decrease in PEF was robust to adjustment of SO₂; however, different lags were examined for O₃ (lag 5) and SO₂ (lag 4) (Wiwatanadate and Trakultivakorn, 2010).
<table>
<thead>
<tr>
<th>Study</th>
<th>O₃ metrics</th>
<th>O₃ with copollutant</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lewis et al. (2005)</td>
<td>24-h avg, Lag 2</td>
<td>None</td>
</tr>
<tr>
<td>Children with asthma</td>
<td>24-h avg, Lag 2</td>
<td>24-h avg, Lag 2 PM₂.₅</td>
</tr>
<tr>
<td>using CS</td>
<td>24-h avg, Lag 2</td>
<td>24-h avg, Lag 2 PM₇.₅</td>
</tr>
<tr>
<td>Children with asthma</td>
<td>24-h avg, Lag 2</td>
<td>None</td>
</tr>
<tr>
<td>with URI</td>
<td>24-h avg, Lag 2</td>
<td>24-h avg, Lag 2 PM₇.₅</td>
</tr>
<tr>
<td>Chen et al. (1999)</td>
<td>1-h max, Lag 1</td>
<td>None</td>
</tr>
<tr>
<td>Children</td>
<td>1-h max, Lag 1</td>
<td>24-h avg, Lag 1 NO₂</td>
</tr>
</tbody>
</table>

**Note:** Results are presented first for FEV₁ then for PEF and within these categories, then in order of increasing mean ambient O₃ concentration. CS = corticosteroid, URI = Upper respiratory infection. Effect estimates are standardized to a 40-, 30-, and 20-ppb increase for 1-h max, 12-h avg, and 24-h avg O₃, respectively. Black circles represent O₃ effect estimates from single pollutant models, and open circles represent O₃ effect estimates from copollutant models.

*Information was not available to calculate 95% CI of the copollutant model.*

**Figure 6-10**  Comparison of O₃-associated changes in lung function in single- and copollutant models.
### Table 6-14: Comparison of O$_3$-associated changes in lung function in single- and copollutant models for studies presented in Figure 6-10 plus others.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/Population</th>
<th>Parameter</th>
<th>O$_3$-associated Percent Change in Single-Pollutant Model (95% CI)$^a$</th>
<th>O$_3$-associated Percent Change in Copollutant Model (95% CI)$^a$</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>FEV$_1$</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lewis et al.</td>
<td>Detroit, MI</td>
<td></td>
<td>For 24-h avg, Lag 2</td>
<td>With 24-h avg, Lag 2 PM$_{2.5}$</td>
</tr>
<tr>
<td>(2005)</td>
<td>Children with asthma using CS</td>
<td>393 person-days</td>
<td>0.29 (-4.2, 5.0)</td>
<td>-0.18 (-11.0, 11.9)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>With 24-h avg, Lag 2 PM$_{10}$</td>
<td>-13.4 (-17.8, -8.8)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>For 24-h avg, Lag 2</td>
<td>With 24-h avg, Lag 2 PM$_{2.5}$</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>-0.18 (-11.0, 11.9)</td>
<td>-13.4 (-17.8, -8.8)</td>
</tr>
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<tr>
<td></td>
<td></td>
<td></td>
<td>For 24-h avg, Lag 2</td>
<td>With 24-h avg, Lag 2 PM$_{10}$</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>-0.18 (-11.0, 11.9)</td>
<td>-13.4 (-17.8, -8.8)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chen et al.</td>
<td>3 Taiwan communities</td>
<td></td>
<td>For 1-h max, Lag 1</td>
<td>With 24-h avg, Lag 1 NO$_2$</td>
</tr>
<tr>
<td>(1999)</td>
<td>941 children, ages 8-13 yr</td>
<td>FEV$_1$</td>
<td>-1.5 (-2.8, -0.12)</td>
<td>-2.0 (-3.5, -0.43)</td>
</tr>
<tr>
<td><strong>PEF</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Neas et al.</td>
<td>Philadelphia, PA</td>
<td></td>
<td>For 12-h avg, Lag 1</td>
<td>With 24-h avg, Lag 1 sulfate</td>
</tr>
<tr>
<td>(1999)</td>
<td>156 Children at summer camp, ages 6 - 11 yr</td>
<td>Morning PEF</td>
<td>-0.94 (-2.0, 0.08)</td>
<td>-0.02$^b$</td>
</tr>
<tr>
<td>Thurston et al.</td>
<td>CT River Valley</td>
<td></td>
<td>For 1-h max, Lag 0</td>
<td>With 12-h avg, Lag 0 sulfate</td>
</tr>
<tr>
<td>(1997)</td>
<td>166 Children with asthma at summer camp, ages 7-13 yr</td>
<td>Intraday change PEF</td>
<td>-2.8 (-4.9, -0.59)</td>
<td>-11.8 (-31.6, 8.1)</td>
</tr>
<tr>
<td>Romieu et al.</td>
<td>Northern Mexico City, Mexico</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(1996)</td>
<td>71 children with asthma, ages 5-7 yr</td>
<td>Evening PEF</td>
<td>For 1-h max, Lag 2</td>
<td>With 24-h avg, Lag 2 PM$_{2.5}$</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>-0.55 (-1.3, 0.19)</td>
<td>-0.24 (-1.2, 0.68)</td>
</tr>
<tr>
<td>Romieu et al.</td>
<td>Southern Mexico City, Mexico</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(1997)</td>
<td>65 children with asthma, ages 5-13 yr</td>
<td>Evening PEF</td>
<td>For 1-h max, Lag 0</td>
<td>With 24-h avg, Lag 0 PM$_{10}$</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>-0.52 (-1.0, -0.01)</td>
<td>-0.79 (-1.4, -0.16)</td>
</tr>
<tr>
<td>Study*</td>
<td>Location/Population</td>
<td>Parameter</td>
<td>O₃-associated Percent Change in Single-Pollutant Model (95% CI)a</td>
<td>O₃-associated Percent Change in Copollutant Model (95% CI)a</td>
</tr>
<tr>
<td>--------</td>
<td>---------------------</td>
<td>-----------</td>
<td>---------------------------------------------------------------</td>
<td>---------------------------------------------------------------</td>
</tr>
<tr>
<td>Jalaludin et al. (2000)</td>
<td>Sydney, Australia</td>
<td>Daily deviation from mean PEF</td>
<td>For 15-h (6 a.m.-9 p.m.) avg, Lag 0 -1.8 (-3.5, -0.19)</td>
<td>With 15-h avg, Lag 0 PM₁₀, -1.8 (-3.5, -0.19)</td>
</tr>
<tr>
<td></td>
<td>125 children with asthma or wheeze, mean (SD) age 9.6 (1.0) yr</td>
<td></td>
<td></td>
<td>With 15-h avg, Lag 0 NO₂, -1.8 (-3.4, -0.11)</td>
</tr>
<tr>
<td>Wiwatanadate and Trakultivakorn (2010)</td>
<td>Chiang Mai, Thailand</td>
<td>Daily avg PEF (L/min)</td>
<td>For 24-h avg, Lag 5 -2.6 (-5.2, 0)</td>
<td>With 24-h avg, Lag 4 SO₂ -3.2 (-6.2, -0.2)</td>
</tr>
<tr>
<td></td>
<td>31 children with asthma, ages 4-11 yr</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Son et al. (2010)</td>
<td>Ulsan, Korea</td>
<td>Change in percent predicted FEV₁</td>
<td>For 8-h max, Lag 0-2 avg (kriging) -1.4 (-2.6, -0.11)</td>
<td>With 24-h avg, Lag 2 PM₁₀ (kriging) -1.8 (-3.4, -0.25)</td>
</tr>
<tr>
<td></td>
<td>2,102 children and adults, ages 7-97 yr</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

* Includes studies in Figure 6-10 plus others.

CS = Corticosteroid, URI = Upper respiratory infection.

**Effect estimates are standardized to a 40-ppb increase for 1-h max O₃, 30-ppb increase for 12-h avg or 8-h max O₃, and 20-ppb increase for 24-h avg or 15-h avg O₃.**

bInformation was not available to calculate 95% CI.

cResults are not presented in Figure 6-10 because sufficient data were not available to calculate percent change in lung function.

Some studies did not provide quantitative results but only reported that O₃-associated lung function decrements remained statistically significant in models that included copollutants such as PM₁₀, NO₂, sulfate, nitrate, or ammonium (Romieu et al., 1998b; Brauer et al., 1996; Linn et al., 1996; Spektor et al., 1988b).

Several studies estimated robust O₃-associated lung function decrements in multipollutant models that most often included O₃, NO₂, and either PM₂.₅ or PM₁₀ (O’Connor et al., 2008; Thaller et al., 2008; Chan and Wu, 2005; Romieu et al., 2002; Korrick et al., 1998; Higgins et al., 1990). However, the independent effects of O₃ are more difficult to assess in relation to incremental changes in more than one copollutant.

Summary of Epidemiologic Studies of Lung Function

The cumulative body of epidemiologic evidence indicates that short-term increases in ambient O₃ concentration are associated with decrements in lung function in children with asthma (Figure 6-7 [and Table 6-8]) and children in the general population. In contrast with results from controlled human exposure studies, within-study epidemiologic comparisons did not consistently indicate larger ambient O₃-associated lung function decrements in groups with asthma (children or adults) than in groups without asthma. Notably, most epidemiologic studies were not designed to assess between-group differences. Based on comparisons between studies, differences were noted between children with and without asthma in so far as in studies of children with asthma, an increase in ambient O₃ concentration was associated with both lung function decrements and increases in...
respiratory symptoms (Just et al., 2002; Mortimer et al., 2002; Ross et al., 2002; Gielen et al., 1997; Romieu et al., 1997; Thurston et al., 1997; Romieu et al., 1996). In studies of children in the general population, increases in ambient O₃ concentration were associated with decreases in lung function but not increases in respiratory symptoms (Ward et al., 2002; Gold et al., 1999; Linn et al., 1996).

Across studies of children, there was no clear indication that a particular exposure assessment method using central site measurements produced stronger findings, despite potential differences in exposure measurement error. In children, lung function was associated with ambient O₃ concentrations measured on site of children’s daytime hours (Hoppe et al., 2003; Thurston et al., 1997), at children’s schools (Chen et al., 1999; Gold et al., 1999), at the closest site (Romieu et al., 2006; Lewis et al., 2005; Romieu et al., 2004b; Romieu et al., 2002; Romieu et al., 1997; Romieu et al., 1996), at multiple community sites then averaged (O'Connor et al., 2008; Just et al., 2002; Mortimer et al., 2002; Jalaludin et al., 2000), and at a single site (Ward et al., 2002; Gielen et al., 1997; Ulmer et al., 1997; Linn et al., 1996). Among children in California, Avol et al. (1998b) found slightly larger O₃-associated lung function decrements for 24-h avg personal exposures than for 1-h max ambient concentrations.

As noted in the 1996 and 2006 O₃ AQCDs, evidence clearly demonstrates ambient O₃-associated lung function decrements in children and adults engaged in outdoor recreation, exercise, or work. Moreover, several results in these populations indicated associations with 10-min to 12-h avg O₃ concentrations <80 ppb. These studies are noteworthy for their measurement of ambient O₃ on site of and at the time of subjects’ outdoor activity, factors that have contributed to higher O₃ personal exposure-ambient concentration correlations and ratios (Section 4.3.3). These epidemiologic results are well supported by observations from controlled human exposure studies of lung function decrements induced by O₃ exposure during exercise (Section 6.2.1.1). Although epidemiologic investigation was relatively sparse, increases in ambient O₃ concentration were not consistently associated with lung function decrements in adults with respiratory disease, healthy adults, or older adults.

Across the diverse populations examined, most effect estimates ranged from a <1 to 2% decrease in lung function per unit increase in O₃ concentration¹. Heterogeneity in O₃-associated respiratory effects within populations was indicated by observations of larger decreases (3-8%) in children with asthma with CS use or concurrent URI (Lewis et al., 2005) and older adults with airway hyperresponsiveness and/or BMI >30 (Alexeeff et al., 2007). Among children in Mexico City, Mexico, higher dietary antioxidant intake attenuated O₃-associated lung function decrements (Romieu et al., 2004b; 2002), similar to results from controlled human exposure studies. Each of these potential effect modifiers was examined in one to two populations; thus, firm conclusions about their influences are not warranted. Adding to the evidence for heterogeneity in response, Hoppe et al. (2003) and Mortimer et al. (2002) found that

¹ Effect estimates were standardized to a 40-, 30-, and 20-ppb increase for 1-h max, 8-h max, and 24-h avg O₃.
increases in ambient O₃ concentration were associated with increased incidence of >10% decline in lung function in children with asthma.

Collectively, epidemiologic studies examined and found lung function decrements in association with single-day O₃ concentrations lagged from 0 to 7 days and concentrations averaged over 2-10 days. More studies found associations with O₃ concentrations lagged 0 or 1 day (Son et al., 2010; Alexeeff et al., 2008; Lewis et al., 2005; Ross et al., 2002; Jalaludin et al., 2000; Chen et al., 1999; Romieu et al., 1997; Brauer et al., 1996; Romieu et al., 1996; Spektor et al., 1988b) than those lagged 5-7 days (Wiwatanadate and Trakultivakorn, 2010; Hernández-Cadena et al., 2009; Steinvil et al., 2009). Associations with multiday average concentrations (Son et al., 2010; Liu et al., 2009a; Barraza-Villarreal et al., 2008; O'Connor et al., 2008; Alexeeff et al., 2007; Mortimer et al., 2002; Ward et al., 2002; Gold et al., 1999; Naheer et al., 1999; Neas et al., 1999) indicate that elevated exposures over several days may be important. Within studies, O₃ concentrations for multiple lag periods were associated with lung function decrements, possibly indicating that multiple modes of action may be involved in the responses. Activation of bronchial C-fibers (Section 5.3.2) may lead to decreases in lung function as an immediate response to O₃ exposure, and increased airway hyperresponsiveness to antigens resulting from sensitization of airways by O₃ (Section 5.3.5) may mediate lung function responses associated with lagged or multiday O₃ exposures (Peden, 2011).

For single- and multi-day average O₃ concentrations, lung function decrements were associated with 1-h max, 8-h max, and 24-h avg O₃, with no strong difference in the consistency or magnitude of association among the averaging times. For example, among studies that examined multiple averaging times, Spektor and Lippmann (1991) found a larger magnitude of association for 1-h max O₃ than for 24-h avg O₃. However, other studies found larger magnitudes of association for longer averaging times [8-h max in Chan and Wu (2005) and 12-h avg in Thaller et al. (2008)] than for 1-h max O₃. Other studies found no clear difference among O₃ averaging times (Jalaludin et al., 2000; Chen et al., 1999; Scarlett et al., 1996; Berry et al., 1991).

Several studies found that associations with lung function decrements persisted at lower ambient O₃ concentrations. For O₃ concentrations averaged up to 1 hour during outdoor recreation or exercise, associations were found in analyses restricted to O₃ concentrations <80 ppb (Spektor et al., 1988a; Spektor et al., 1988b), 60 ppb (Brunekreef et al., 1994; Spektor et al., 1988a), and 50 ppb (Brunekreef et al., 1994). Among outdoor workers, Brauer et al. (1996) found a robust association using daily 1-h max O₃ concentrations <40 ppb. Ulmer et al. (1997) found a robust association in schoolchildren using 30-min max O₃ concentrations <60 ppb. For 8-h avg O₃ concentrations, associations with lung function decrements in children with asthma were found to persist at concentrations <80 ppb in a U.S. multicity study (for lag 1-5 avg) (Mortimer et al., 2002) and <51 ppb in a study conducted in the Netherlands (for lag 2) (Gielen et al., 1997).

Evidence did not demonstrate strong confounding by meteorological factors or copollutant exposures. Most O₃ effect estimates for lung function were robust to adjustment for temperature, humidity, and copollutants such as PM₁₀, NO₂, or
SO\textsubscript{2}. Although examined in few epidemiologic studies, O\textsubscript{3} was associated with decreases in lung function with adjustment for pollen or acid aerosols. The consistency of association in the collective body of evidence with and without adjustment for ambient copollutant concentrations and meteorological factors combined with evidence from controlled human exposure studies for the direct effects of O\textsubscript{3} exposure provide strong support for the independent effects of short-term ambient O\textsubscript{3} exposure on lung function decrements.

### 6.2.1.3 Toxicology

The 2006 O\textsubscript{3} AQCD concluded that pulmonary function decrements occur in a number of species with acute exposures (≤ 1 week), ranging from 0.25 to 0.4 ppm O\textsubscript{3} (\textit{U.S. EPA, 2006b}). Early work has demonstrated that during acute exposure of ~0.2 ppm O\textsubscript{3} in rats, the most commonly observed alterations are increased frequency of breathing and decreased tidal volume (i.e., rapid, shallow breathing). Decreased lung volumes are observed in rats with acute exposures to 0.5 ppm O\textsubscript{3}. At concentrations of ≥ 1 ppm, breathing mechanics (compliance and resistance) are also affected. Exposures of 6 hours/day for 5 days create a pattern of attenuation of pulmonary function decrements in both rats and humans without concurrent attenuation of lung injury and morphological changes, indicating that the attenuation did not result in protection against all the effects of O\textsubscript{3} (\textit{Tepper et al., 1989}). A number of studies examining the effects of O\textsubscript{3} on pulmonary function in rats, mice, and dogs are described in Table 6-13 on page 6-91 (\textit{U.S. EPA, 1996m}) of the 1996 O\textsubscript{3} AQCD, and Annex Table AX5-11 on page AX5-34 (\textit{U.S. EPA, 2006g}) of the 2006 O\textsubscript{3} AQCD (\textit{U.S. EPA, 2006b, 1996a}). Lung imaging studies using hyperpolarized \textsuperscript{3}He provide evidence of ventilation abnormalities in rats following exposure to 0.5 ppm O\textsubscript{3} (\textit{Crémillieux et al., 2008}). Rats were exposed to 0.5 ppm O\textsubscript{3} for 2 or 6 days, either continuously (22 hours/day) or alternatingly (12 hours/day). Dynamic imaging of lung filling (2 mL/sec) revealed delayed and incomplete filling of lung segments and lobes. Abnormalities were mainly found in the upper regions of the lungs and proposed to be due to the spatial distribution of O\textsubscript{3} exposure within the lung. Although the small number of animals used in the study (n = 3 to 7/group) makes definitive conclusions difficult, the authors suggest that the delayed filling of lung lobes or segments is likely a result of an increase in airway resistance brought about by narrowing of the peripheral small airways.

### 6.2.2 Airway Hyperresponsiveness

Airway hyperresponsiveness refers to a condition in which the conducting airways undergo enhanced bronchoconstriction in response to a variety of stimuli. Airway responsiveness is typically quantified by measuring changes in pulmonary function (e.g., FEV\textsubscript{1} or specific airway resistance [sRaw]) following the inhalation of an aerosolized specific (allergen) or nonspecific (e.g., methacholine)
bronchoconstricting agent or administration of another stimulus such as exercise or cold air. People with asthma are generally more sensitive to bronchoconstricting agents than those without asthma, and the use of an airway challenge to inhaled bronchoconstricting agents is a diagnostic test in asthma. Standards for airway responsiveness testing have been developed for the clinical laboratory (ATS, 2000a), although variation in methodology for administering the bronchoconstricting agent may affect the results (Cockcroft et al., 2005). There is a wide range of airway responsiveness in people without asthma, and responsiveness is influenced by a wide range of factors, including cigarette smoke, pollutant exposures, respiratory infections, occupational exposures, and respiratory irritants. Airways hyperresponsiveness in response to O₃ exposure has not been examined widely in epidemiologic studies; such evidence is derived primarily from controlled human exposure and toxicological studies as described below.

### 6.2.2.1 Controlled Human Exposures

Beyond its direct effect on lung function, experimental O₃ exposure has been shown to cause an increase in airway responsiveness in human subjects. Increased airway responsiveness can be an important consequence of exposure to ambient O₃, because the airways are then predisposed to narrowing upon inhalation of a variety of ambient stimuli.

Increases in airway responsiveness have been reported for exposures to 80 ppb O₃ and above. Horstman et al. (1990) evaluated airway responsiveness to methacholine in young healthy adults (22 M) exposed to 80, 100, and 120 ppb O₃ (6.6 hours, quasi continuous moderate exercise, 39 L/min). Dose-dependent decreases of 33, 47, and 55% in the cumulative dose of methacholine required to produce a 100% increase in sRaw after exposure to O₃ at 80, 100, and 120 ppb, respectively, were reported. Molfino et al. (1991) reported increased allergen-specific airway responsiveness in adults with mild asthma exposed to 120 ppb O₃ (1 hour resting exposure). Due to safety concerns, however, the exposures in the Molfino et al. (1991) study were not randomized with FA conducted first and O₃ exposure second. Attempts to reproduce the findings of Molfino et al. (1991) using a randomized exposure design have not found statistically significant changes in airway responsiveness at such low levels of O₃ exposure. At a considerably higher exposure to 250 ppb O₃ (3 hours, light-to-moderate intermittent exercise, 30 L/min), Jorres et al. (1996) found significant increases in specific and non-specific airway responsiveness of adults with mild asthma 3 hours following O₃ exposure. Kehrl et al. (1999) found increased reactivity to house dust mite antigen in adults with mild asthma and atopy 16-18 hours after exposure to 160 ppb O₃ (7.6 hours, light quasi continuous exercise, 25 L/min). Holz et al. (2002) demonstrated that repeated daily exposure to lower concentrations of 125 ppb O₃ (3 hours for four consecutive days; intermittent exercise, 30 L/min) causes an increased response to allergen challenge at 20 hours postexposure in allergic airway disease.
Ozone exposure of subjects with asthma, who characteristically have increased airway responsiveness at baseline relative to healthy controls (by nearly two orders of magnitude), can cause further increases in responsiveness (Kreit et al., 1989). Similar relative changes in airway responsiveness are seen in subjects with asthma and healthy control subject exposed to O₃ despite their markedly different baseline airway responsiveness. Several studies (Kehrl et al., 1999; Jorres et al., 1996; Molfino et al., 1991) have suggested an increase in specific (i.e., allergen-induced) airway reactivity. An important aspect of increased airway responsiveness after O₃ exposure is that this may provide biological plausibility for associations observed between increases in ambient O₃ concentration and increased respiratory symptoms in children with asthma (Section 6.2.4.1) and increased hospital admissions and ED visits for asthma (Section 6.2.7).

Changes in airway responsiveness after O₃ exposure appear to resolve more slowly than changes in FEV₁ or respiratory symptoms (Folinsbee and Hazucha, 2000). Studies suggest that O₃-induced increases in airway responsiveness usually resolve 18 to 24 hours after exposure, but may persist in some individuals for longer periods (Folinsbee and Hazucha, 1989). Furthermore, in studies of repeated exposure to O₃, changes in airway responsiveness tend to be somewhat less susceptible to attenuation with consecutive exposures than changes in FEV₁ (Gong et al., 1997a; Folinsbee et al., 1994; Kulle et al., 1982; Dimeo et al., 1981). Increases in airway responsiveness do not appear to be strongly associated with decrements in lung function or increases in symptoms (Aris et al., 1995). Recently, Que et al. (2011) assessed methacholine responsiveness in healthy young adults (83M, 55 F) one day after exposure to 220 ppb O₃ and FA for 2.25 hours (alternating 15 min periods of rest and brisk treadmill walking). Increases in airways responsiveness at 1 day post-O₃ exposure were not correlated with FEV₁ responses immediately following the O₃ exposure or with changes in epithelial permeability assessed 1 day post-O₃ exposure.

### 6.2.2.2 Toxicology

In addition to human subjects, a number of species, including nonhuman primates, dogs, cats, rabbits, and rodents, have been used to examine the effect of O₃ exposure on airway hyperresponsiveness (see Table 6-14 on page 6-93 (U.S. EPA, 1996n) of the 1996 O₃ AQCD and Annex Table AX5-12 on page AX5-36 (U.S. EPA, 2006h) of the 2006 O₃ AQCD). With a few exceptions, commonly used animal models have been guinea pigs, rats, or mice acutely exposed to O₃ concentrations of 1 to 3 ppm to induce airway hyperresponsiveness. These animal models are helpful for determining underlying mechanisms of general airway hyperresponsiveness and are relevant for understanding airway responses in humans. Although 1-3 ppm may seem like a high exposure concentration, based on $^{18}$O₃ (oxygen-18-labeled O₃) in the BALF of humans and rats, an exposure of 0.4 ppm O₃ in exercising humans appears roughly equivalent to an exposure of 2 ppm in resting rats (Hatch et al., 1994).
A limited number of studies have observed airway hyperresponsiveness in rodents and guinea pigs after exposure to less than 0.3 ppm O₃. As previously reported in the 2006 O₃ AQCD, one study demonstrated that a very low concentration of O₃ (0.05 ppm for 4 hours) induced airway hyperresponsiveness in some of the nine strains of rats tested (Depuydt et al., 1999). This effect occurred at a concentration of O₃ that was much lower than has been reported to induce airway hyperresponsiveness in any other species. Similar to the effects of O₃ on other endpoints, these observations suggest that a genetic component plays an important role in O₃-induced airway hyperresponsiveness in this species and warrants verification in other species. More recently, Chhabra et al. (2010) demonstrated that exposure of ovalbumin (OVA)-sensitized guinea pigs to 0.12 ppm for 2 hours/day for 4 weeks produced specific airway hyperresponsiveness to an inhaled OVA challenge. Interestingly, in this study, dietary supplementation of the guinea pigs with vitamins C and E ameliorated a portion of the airway hyperresponsiveness as well as indices of inflammation and oxidative stress. Larsen and colleagues conducted an O₃ C-R study in mice sensitized by 10 daily inhalation treatments with an OVA aerosol (Larsen et al., 2010). Although airway responsiveness to methacholine was increased in non-sensitized animals exposed to a single 3-hour exposure to 0.5, but not 0.1 or 0.25 ppm O₃, airway hyperresponsiveness was observed after exposure to 0.1 and 0.25 ppm O₃ in OVA-sensitized mice.

In order to evaluate the ability of O₃ to enhance specific and non-specific airway responsiveness, it is important to take into account the phenomenon of attenuation in the effects of O₃. Several studies have clearly demonstrated that some effects caused by acute exposure are absent after repeated or prolonged exposures to O₃. The ability of the pulmonary system to adapt to repeated insults to O₃ is complex, however, and experimental findings for attenuation to O₃-induced airway hyperresponsiveness are inconsistent. Airway hyperresponsiveness was observed in mice after a 3-hour exposure but not in mice exposed continuously for 72 hours to 0.3 ppm (Johnston et al., 2005b). However, the Chhabra study demonstrated O₃-induced airway hyperresponsiveness in guinea pigs exposed for 2 hours/day for 10 days (Chhabra et al., 2010). Besides the obvious species disparity, these studies differ in that the mice were exposed continuously for 72 hours, whereas the guinea pigs were exposed intermittently over 10 days, suggesting that attenuation might be lost with periods of rest in between O₃ exposures.

Overall, numerous toxicological studies have demonstrated that O₃-induced airway hyperresponsiveness occurs in guinea pigs, rats, and mice after either acute or repeated exposure to relevant concentrations of O₃. The mechanisms by which O₃ enhances the airway responsiveness to either specific (e.g., OVA) or non-specific (e.g., methacholine) bronchoprovocation are not clear but appear to be associated with complex cellular and biochemical changes in the conducting airways. A number of potential mediators and cells may play a role in O₃-induced airway hyperresponsiveness; mechanistic studies are discussed in greater detail in Section 5.3.
6.2.3 Pulmonary Inflammation, Injury and Oxidative Stress

In addition to physiological pulmonary responses, respiratory symptoms, and airway hyperresponsiveness, O₃ exposure has been shown to result in increased epithelial permeability and respiratory tract inflammation. In general, inflammation can be considered as the host response to injury and the induction of inflammation as evidence that injury has occurred. Inflammation induced by exposure of humans to O₃ can have several potential outcomes: (1) inflammation induced by a single exposure (or several exposures over the course of a summer) can resolve entirely; (2) continued acute inflammation can evolve into a chronic inflammatory state; (3) continued inflammation can alter the structure and function of other pulmonary tissue, leading to diseases such as fibrosis; (4) inflammation can alter the body’s host defense response to inhaled microorganisms, particularly in potentially at-risk populations such as the very young and old; and (5) inflammation can alter the lung’s response to other agents such as allergens or toxins. Except for outcome (1), the possible chronic responses have only been directly observed in animals exposed to O₃. It is also possible that the profile of response can be altered in persons with pre-existing pulmonary disease (e.g., asthma, COPD) or smokers. Oxidative stress has been shown to play a key role in initiating and sustaining O₃-induced inflammation. Secondary oxidation products formed as a result of reactions between O₃ and components of the ELF can increase the expression of cytokines, chemokines, and adhesion molecules and enhance airway epithelium permeability (Section 5.3.3. and Section 5.3.4).

6.2.3.1 Controlled Human Exposures

As reported in studies reviewed in the 1996 and 2006 O₃ AQCDs, acute O₃ exposure initiates an acute inflammatory response throughout the respiratory tract that has been observed to persist for at least 18-24 hours postexposure. A meta-analysis of 21 studies (Mudway and Kelly, 2004a) for varied experimental protocols (80-600 ppb O₃; 1-6.6 hours duration; light to heavy exercise; bronchoscopy at 0-24 hours post-O₃ exposure) showed that neutrophils (PMN) influx in healthy subjects was linearly associated (p <0.01) with total O₃ dose (i.e., the product of O₃ concentration, exposure duration, and V̇E). As with FEV₁ responses to O₃, within-individual inflammatory responses to O₃ are generally reproducible and correlated between repeat exposures (Holz et al., 1999). Some individuals also appear to be intrinsically more susceptible to increased inflammatory responses to O₃ exposure (Holz et al., 2005).

The presence of PMNs in the lung has long been accepted as a hallmark of inflammation and is an important indicator that O₃ causes inflammation in the lungs. Neutrophilic inflammation of tissues indicates activation of the innate immune system and requires a complex series of events that are normally followed by processes that clear the evidence of acute inflammation. Inflammatory effects have been assessed in vivo by lavage (proximal airway and bronchoalveolar), bronchial...
biopsy, and more recently, induced sputum. A single acute exposure (1-4 hours) of humans to moderate concentrations of O₃ (200-600 ppb) while exercising at moderate to heavy intensities results in a number of cellular and biochemical changes in the lung, including an inflammatory response characterized by increased numbers of PMNs, increased permeability of the epithelial lining of the respiratory tract, cell damage, and production of proinflammatory cytokines and prostaglandins (U.S. EPA, 2006b). These changes also occur in humans exposed to 80 and 200 ppb O₃ for 6-8 hours (Alexis et al., 2010; Peden et al., 1997; Devlin et al., 1991). Significant (p = 0.002) increases in sputum PMN (16-18 hours postexposure) relative to FA responses have been recently reported for 60 ppb O₃ which is the lowest exposure concentration that has been investigated in young healthy adults (Kim et al., 2011). Soluble mediators of inflammation such as the cytokines (e.g., IL-6, IL-8) and arachidonic acid metabolites (e.g., prostaglandin [PG]E₂, PGF₂α, thromboxane, and leukotrienes [LTs] such as LTB₄) have been measured in the BALF of humans exposed to O₃. In addition to their role in inflammation, many of these compounds have bronchoconstrictive properties and may be involved in increased airway responsiveness following O₃ exposure. The possible relationship between repetitive bouts of acute inflammation in humans caused by O₃ and the development of chronic respiratory disease is unknown.

**Asthma**

Inflammatory responses to O₃ exposure have also been studied in subjects with asthma. Individuals with asthma exposed to 200 ppb O₃ for 4-6 hours with exercise show significantly more neutrophils in BALF (18 hours postexposure) than similarly exposed healthy individuals (Scannell et al., 1996; Basha et al., 1994). In subjects with allergic asthma who tested positive for *Dermatophagoides farinae* antigen, there was an eosinophilic inflammation (2-fold increase), as well as neutrophilic inflammation (3-fold increase) 18 hours after exposure to 160 ppb O₃ for 7.6 hours with exercise (Peden et al., 1997). In a study of subjects with intermittent asthma exposed to 400 ppb O₃ for 2 hours, increases in eosinophil cationic protein, neutrophil elastase and IL-8 were found to be significantly increased 16 hours postexposure and comparable in induced sputum and BALF (Hiltermann et al., 1999). At 18 hours post-O₃ exposure (200 ppb, 4 hours with exercise) and corrected for FA responses, Scannell et al. (1996) found significantly increased neutrophils in 18 adults with asthma (12%) compared to 20 healthy subjects (4.5%). This difference in inflammatory response was observed despite no group differences in spirometric responses to O₃. Scannell et al. (1996) also reported that IL-8 tends to be higher in the BALF of subjects with asthma compared to those without asthma following O₃ exposure, suggesting a possible mediator for the significantly increased neutrophilic inflammation in those subjects. Bosson et al. (2003) found significantly greater epithelial expression of IL-5, IL-8, granulocyte-macrophage colony-stimulating factor and epithelial cell-derived neutrophil-activating peptide-78 in subjects with asthma compared to healthy subjects following exposure to 200 ppb O₃ for 2 hours. In contrast, Stenfors et al. (2002) did not detect a difference in the O₃-induced increases in neutrophil numbers between 15 subjects with mild asthma and 15
healthy subjects by bronchial wash at the 6 hours postexposure time point. However, the subjects with asthma were on average 5 years older than the healthy subjects in this study, and it is not yet known how age affects inflammatory responses. It is also possible that the time course of neutrophil influx differs between healthy individuals and those with asthma. Differences between subjects with asthma and healthy subjects in O₃-mediated activation of innate and adaptive immune responses have been observed in two studies (Hernandez et al., 2010; Bosson et al., 2003), as discussed in Section 6.2.5.4 and Section 5.4.2.2.

Vagaggini et al. (2002) investigated the effect of prior allergen challenge on responses in subjects with mild asthma exposed for 2 hours to 270 ppb O₃ or filtered air. At 6 hours postexposure, eosinophil numbers in induced sputum were found to be significantly greater after O₃ than after air exposures. Studies such as this suggest that the time course of eosinophil and neutrophil influx following O₃ exposure can occur at levels detectable within the airway lumen by as early as 6 hours. They also suggest that the previous or concurrent activation of pro-inflammatory pathways within the airway epithelium may enhance the inflammatory effects of O₃. For example, in an in vitro study of primary human nasal epithelial cells and BEAS-2B cell line, cytokine production induced by rhinovirus infection was enhanced synergistically by concurrent exposure to O₃ at 200 ppb for 3 hours (Spannhake et al., 2002).

A few studies have evaluated the effects of corticosteroid usage on the response of subjects with asthma to O₃. Vagaggini et al. (2007) evaluated whether corticosteroid usage would prevent O₃-induced lung function decrements and inflammatory responses in a group of subjects with mild persistent asthma (n = 9; 25 ± 7 years). In this study, subjects with asthma were randomly exposed on four occasions to 270 ppb O₃ or FA for 2 hours with moderate exercise. Exposures were preceded by four days of treatment with prednisone or placebo. Pretreatment with corticosteroids prevented an inflammatory response in induced sputum at 6 hours postexposure. FEV₁ responses were, however, not prevented by corticosteroid treatment and were roughly equivalent to those observed following placebo. Vagaggini et al. (2001) also found budesonide to decrease airway neutrophil influx in subjects with asthma following O₃ exposure. In contrast, inhalation of corticosteroid budesonide failed to prevent or attenuate O₃-induced responses in healthy subjects as assessed by measurements of lung function, bronchial reactivity and airway inflammation (Nightingale et al., 2000). High doses of inhaled fluticasone and oral prednisolone have each been reported to reduce inflammatory responses to O₃ in healthy individuals (Holz et al., 2005).

Stenfors et al. (2010) exposed adults with persistent asthma (n = 13; aged 33 years) receiving chronic inhaled corticosteroid therapy to 200 ppb O₃ for 2 hours with moderate exercise. At 18 hours postexposure, there was a significant O₃-induced increase in bronchoalveolar lavage (BAL) neutrophils, but not eosinophils. Bronchial biopsy also showed a significant O₃-induced increase in mast cells. Results from this study suggest that the protective effect of acute corticosteroid
therapy against inflammatory responses to O₃ in subjects with asthma demonstrated by Vagaggini et al. (2007) may be lost with continued treatment regimes.

**Associations between Inflammation and FEV₁ responses**

Studies reviewed in the 2006 O₃ AQCD reported that inflammatory responses do not appear to be correlated with lung function responses in either subjects with asthma or healthy subjects. In healthy adults (14 M, 6 F) and volunteers with asthma (12 M, 6 F) exposed to 200 ppb O₃ (4 hours with moderate quasi continuous exercise, \( \dot{V}_E = 44 \) L/min), percent PMN and total protein in BAL fluids were significantly increased in the subjects with asthma relative to the healthy controls. Spirometric measures of lung function were significantly decreased following the O₃ exposure in both groups, but were not significantly different between the subjects with asthma and healthy subjects. Effects of O₃ on PMN and total protein were not correlated with changes in FEV₁ or FVC (Balmes et al., 1997; Balmes et al., 1996). Devlin et al. (1991) exposed healthy adults (18 M) to 80 and 100 ppb O₃ (6.6-hours with moderate quasi continuous exercise, 40 L/min). In BAL fluid collected 18 hours after exposure to 100 ppb O₃, significant increases in PMNs, protein, PGE₂, fibronectin, IL-6, lactate dehydrogenase, and α-1 antitrypsin were found compared to FA. Similar but smaller increases in all mediators were found after exposure to 80 ppb O₃ except for protein and fibronectin. Changes in BAL markers were not correlated with changes in FEV₁. Holz et al. (1999) examined inflammatory responses in healthy subjects (n = 21) and those with asthma (n = 15) exposed to 125 and 250 ppb O₃ (3 hours, light intermittent exercise, 26 L/min). Significantly increased percent PMN in sputum due to O₃ exposure was observed in both healthy subjects and those with asthma following the 250 ppb exposure. At the lower 125 ppb exposure, only the group with asthma experienced statistically significant increases in the percent PMN. Significant decrements in FEV₁ were only found following exposure to 250 ppb; these changes in FEV₁ did not differ significantly between the group with asthma and healthy group and were not correlated with changes in PMN levels. Peden et al. (1997) also found no correlation between PMN and FEV₁ responses in 8 individuals with asthma exposed to 160 ppb O₃ for 7.6 hours with light-to-moderate exercise (\( \dot{V}_E = 25 \) L/min). However, a marginally significant correlation (\( r = -0.69, \) two-tailed p = 0.08, n = 7) was observed between increases in the percentage of eosinophils and FEV₁ responses following O₃ exposure.

In contrast to these earlier findings, Vagaggini et al. (2010) recently reported a significant (\( r = 0.61, \) p = 0.015) correlation between changes in FEV₁ changes in sputum neutrophils in subjects with mild-to-moderate asthma (n = 23; 33 ± 11 years) exposed to 300 ppb O₃ for 2 hours with moderate exercise. Eight subjects were categorized as “responders” based on >10% FEV₁ decrements. There were no baseline differences between responders and nonresponders. However, at 6 hours post-O₃ exposure, sputum neutrophils were significantly increased by 15% relative to FA in responders. The neutrophil increase in responders was also significantly greater than the 0.2% increase in nonresponders. Interestingly, the nonresponders in the Vagaggini et al. (2010) study experienced a significant O₃-induced 11.3%
increase in sputum eosinophils, while responders had a nonsignificant 2.6% decrease.

**Time Course of the Inflammatory Response**

The time course of the inflammatory response to O₃ in humans has not been fully characterized. Different markers exhibit peak responses at different times. Studies in which lavages were performed 1 hour after O₃ exposure (1 hours at 400 ppb or 4 hours at 200 ppb) have demonstrated that the inflammatory responses are quickly initiated ([Torres et al., 1997; Devlin et al., 1996; Schelegle et al., 1991](#)).

Inflammatory mediators and cytokines such as IL-8, IL-6, and PGE₂ are greater at 1 hour than at 18 hours post-O₃ exposure ([Torres et al., 1997; Devlin et al., 1996](#)). However, IL-8 still remained elevated at 18 hours post-O₃ exposure (4 hours at 200 ppb O₃ versus FA) in healthy subjects ([Balmes et al., 1996](#)). [Schelegle et al. (1991](#)) found increased PMNs in the “proximal airway” lavage at 1, 6, and 24 hours after O₃ exposure (4 hours at 200 ppb O₃), with a peak response at 6 hours. However, at 18-24 hours after O₃ exposure, PMNs remain elevated relative to 1 hour postexposure ([Torres et al., 1997; Schelegle et al., 1991](#)).

**Genetic Polymorphisms**

[Alexis et al. (2010)](#) recently reported that a 6.6-hour exposure with moderate exercise to 80 ppb O₃ caused increased sputum neutrophil levels at 18 hours postexposure in young healthy adults (n = 15; 24 ± 1 years). In a prior study, [Alexis et al. (2009)](#) found genotype effects on inflammatory responses to O₃, but not lung function responses following a 2-hour exposure to 400 ppb O₃. At 4 hours post-O₃ exposure, groups of both GSTM1 genotypes had significant increases in sputum neutrophils with a tendency for a greater increase in GSTM1-sufficient than null individuals. At 24 hours postexposure, neutrophils had returned to baseline levels in the GSTM1-sufficient individuals. In the GSTM1-null subjects, however, neutrophil levels increased further from 4 hours to 24 hours and were significantly greater than both baseline levels and 24 hours levels in GSTM1-sufficient individuals. [Alexis et al. (2009)](#) found that GSTM1-sufficient individuals (n = 19; 24 ± 3 years) had a decrease in macrophage levels at 4-24 hours postexposure to 400 ppb O₃ for 2 hours with exercise. These studies also provide evidence for activation of innate immunity and antigen presentation, as discussed in Section 5.3.6. Effects of the exposure apart from O₃ cannot be ruled out in the [Alexis et al. (2010); (2009)](#) studies, however, since no FA exposure was conducted.

[Vagaggini et al. (2010)](#) examined FEV₁ and sputum neutrophils in subjects with mild-to-moderate asthma (n = 23; 33 ± 11 years) exposed to 300 ppb O₃ for 2 hours with moderate exercise. Six of the subjects were NQO1 wild type and GSTM1 null, but this genotype was not found to be associated with O₃-induced changes in lung function or inflammatory responses to O₃. [Kim et al. (2011)](#) showed a significant (p = 0.002) increase in sputum neutrophil levels following a 6.6-hour exposure to
60 ppb O₃ relative to FA in young healthy adults (13 F, 11 M; 25.0 ± 0.5 years). There was no significant effect of GSTM1 genotype (half GSTM1-null) on the inflammatory responses observed in these individuals. Previously, inflammatory responses had only been evaluated down to a level of 80 ppb O₃.

Repeated Exposures

Changes in markers from BALF following both 2-hour (Devlin et al., 1997) and 4-hour (Jorres et al., 2000; Christian et al., 1998) repeated O₃ exposures (up to 5 days) indicate that there is ongoing cellular damage irrespective of the attenuation of some cellular inflammatory responses of the airways, pulmonary function, and symptom responses. Devlin et al. (1997) found that several indicators of inflammation (e.g., PMN, IL-6, PGE₂, fibronectin) were attenuated after 5 days of exposure (i.e., values were not different from FA). However, other markers (LDH, IL-8, total protein, epithelial cells) did not show attenuation, suggesting that tissue damage probably continues to occur during repeated exposure. Some cellular responses did not return to baseline levels for more than 10-20 days following O₃ exposure. Christian et al. (1998) showed decreased numbers of neutrophils and a decrease in IL-6 levels in healthy adults after 4 days of exposure versus the single exposure to 200 ppb O₃ for 4 hours. Jorres et al. (2000) also found that both functional and BALF cellular responses to O₃ were abolished at 24 hours postexposure following the fourth exposure day. However, levels of total protein, IL-6, IL-8, reduced glutathione and ortho-tyrosine were still increased significantly. In addition, visual scores (bronchoscopy) for bronchitis and erythema and the numbers of neutrophils in bronchial mucosal biopsies were increased. Results indicate that, despite an attenuation of some markers of inflammation in BALF and pulmonary function decrements, inflammation within the airways persists following repeated exposure to O₃. The continued presence of cellular injury markers indicates a persistent effect that may not necessarily be recognized due to the attenuation of spirometric and symptom responses.

Epithelial Permeability

A number of studies show that O₃ exposures increase epithelial cell permeability through direct (technetium-99m labeled diethylene triamine pentaacetic acid, ⁹⁹ᵐTc-DTPA, clearance) and indirect (e.g., increased BALF albumin, protein) techniques. Kehrl et al. (1987) showed increased ⁹⁹ᵐTc-DTPA clearance in healthy young adults (age 20-30 yrs) at 75 minutes postexposure to 400 ppb O₃ for 2 hours. Also in healthy young adults (age 26 ± 2 yrs), Foster and Stetkiewicz (1996) have shown that increased ⁹⁹ᵐTc-DTPA clearance persists for at least 18-20 hours post-O₃ exposure (130 minutes to average O₃ concentration of 240 ppb), and the effect is greater at the lung apices than at the base. In a older group of healthy adults (mean age = 35 yrs), Morrison et al. (2006) observed ⁹⁹ᵐTc-DTPA clearance at 1 hours and 6 hours postexposure to O₃ (100 and 400 ppb; 1 hour; moderate intermittent exercise,
\( \dot{V}_E = 40 \text{ L/min} \) to be similar and not statistically different from \(^{99m}\text{Tc-DTPA} \) clearance at 1-hour postexposure to FA (1 hour; \( \dot{V}_E = 40 \text{ L/min} \)).

Increased BALF protein, suggesting \( O_3 \)-induced changes in epithelial permeability, have also been reported at 1 hour and 18 hours postexposure (Devlin et al., 1997; Balmes et al., 1996). Meta-analysis of results from 21 publications (Mudway and Kelly, 2004a) for varied experimental protocols (80-600 ppb \( O_3 \); 1-6.6 hours duration; light to heavy exercise; bronchoscopy at 0-24 hours post-\( O_3 \) exposure), showed that increased BALF protein is associated with total inhaled \( O_3 \) dose (i.e., the product of \( O_3 \) concentration, exposure duration, and \( \dot{V}_E \)).

It has been postulated that changes in permeability associated with acute inflammation may provide increased access of inhaled antigens, particles, and other inhaled substances deposited on lung surfaces to the smooth muscle, interstitial cells, and the blood. Hence, increases in epithelial permeability following \( O_3 \) exposure might lead to increases in airway responsiveness to specific and nonspecific agents. Que et al. (2011) investigated this hypothesis in healthy young adults (83M, 55 F) exposed to 220 ppb \( O_3 \) for 2.25 hours (alternating 15 min periods of rest and brisk treadmill walking). As has been observed by others for \( FEV_1 \) responses, within-individual changes in permeability were correlated between sequential \( O_3 \) exposures. This indicates intrinsic differences in susceptibility to epithelial damage from \( O_3 \) exposure among individuals. Increases in epithelial permeability at 1 day post-\( O_3 \) exposure were not correlated with \( FEV_1 \) responses immediately following \( O_3 \) exposure or with changes in airway responsiveness to methacholine assessed 1 day post-\( O_3 \) exposure. The authors concluded that changes in \( FEV_1 \), permeability, and airway responsiveness following \( O_3 \) exposure were relatively constant over time in young healthy adults, although these responses appear to be mediated by differing physiologic pathways.

### 6.2.3.2 Epidemiology

In the 2006 \( O_3 \) AQCD, epidemiologic evidence of associations between short-term increases in ambient \( O_3 \) concentration (30-min or 1-h max) and changes in pulmonary inflammation was limited to a few observations of increases in nasal lavage levels of inflammatory cell counts, eosinophilic cationic protein, and myeloperoxidases (U.S. EPA, 2006b). In recent years, there has been a large increase in the number of studies assessing ambient \( O_3 \)-related changes in pulmonary inflammation and oxidative stress, types of biological samples collected (i.e., lower airway), and types of indicators examined. Most studies collected samples every 1 to 3 weeks resulting in a total of 3 to 8 samples per subject. These recent studies form a larger base to establish coherence with findings from controlled human exposure and animal studies that have measured the same or related biological markers. Additionally, results from these studies provide further biological plausibility for the associations observed between ambient \( O_3 \) concentrations and respiratory symptoms and asthma exacerbations.
Despite the strengths of studies of inflammation, research in this field continues to develop, and several uncertainties are recognized that may limit inferences from results indicating the effects of ambient O₃ exposure. Current areas of development include examination of the clinical relevance of the observed magnitudes of changes in biological markers of pulmonary inflammation (Murugan et al., 2009; Duramad et al., 2007), characterization of the time course of changes between biomarker levels and other endpoints of respiratory morbidity, development of standardized methodologies for collection, improvement of the sensitivity and specificity of assay methods, and characterization of subject factors (e.g., asthma severity, recent medication use) that contribute to inter-individual variability. These sources of uncertainty may contribute to differences in findings among studies.

Although most of the biomarkers examined in epidemiologic studies were not specific to the lung, most studies collected exhaled breath, exhaled breath condensate (EBC), nasal lavage fluid, or induced sputum with the aim of monitoring inflammatory responses in airways, as opposed to monitoring systemic responses in blood. The biomarker most frequently measured was exhaled nitric oxide (eNO), likely related to its ease of collection in the field and automated measurement. Other biological markers were examined in EBC, induced sputum, and nasal lavage fluid, which are hypothesized to represent the fluid lining the lower or upper airways and contain cytokines, cells, and/or markers of oxidative stress that mediate inflammatory responses (Balbi et al., 2007; Howarth et al., 2005; Hunt, 2002). Table 6-15 presents the locations, time periods, and ambient O₃ concentrations for studies examining associations with biological markers of pulmonary inflammation and oxidative stress. Many studies found that short-term increases in ambient O₃ concentration were associated with increases in pulmonary inflammation and oxidative stress, in particular, studies of children with asthma conducted in Mexico City, Mexico (Figure 6-11 [and Table 6-16] and Table 6-17).
Table 6-15  Mean and upper percentile O₃ concentrations in epidemiologic studies of biological markers of pulmonary inflammation and oxidative stress.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O₃ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Barraza-Villarreal et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>June 2003-June 2005</td>
<td>8-h moving avg</td>
<td>31.6</td>
<td>Max: 86.3</td>
</tr>
<tr>
<td>Berhane et al. (2011)</td>
<td>13 Southern California Communities</td>
<td>Sept 2004-June 2005</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>NR</td>
<td>NR</td>
</tr>
<tr>
<td>Liu et al. (2009a)</td>
<td>Windsor, ON, Canada</td>
<td>Oct-Dec 2005</td>
<td>24-h avg</td>
<td>13.0</td>
<td>95th: 26.5</td>
</tr>
<tr>
<td>Qian et al. (2009)</td>
<td>Boston, MA; New York, NY; Denver, CO; Philadelphia, PA; San Francisco, CA; Madison, WI (SOCS)</td>
<td>Feb 1997 - Jan 1999</td>
<td>8-h max</td>
<td>33.6</td>
<td>75th: 44.4, Max: 91.5</td>
</tr>
<tr>
<td>Romieu et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>Jan-Oct 2004</td>
<td>8-h max</td>
<td>31.1</td>
<td>75th: 38.3, Max: 60.7</td>
</tr>
<tr>
<td>Sienra-Monge et al. (2004)</td>
<td>Mexico City, Mexico</td>
<td>All-year 1999-2000</td>
<td>8-h max</td>
<td>66.2</td>
<td>Max: 142.5</td>
</tr>
<tr>
<td>Ferdinands et al. (2008)</td>
<td>Suburb of Atlanta, GA</td>
<td>Aug 2004</td>
<td>1-h max</td>
<td>61 (median)</td>
<td>75th: 67</td>
</tr>
<tr>
<td>Chimenti et al. (2009)</td>
<td>Sicily, Italy</td>
<td>Nov, Feb, July, year NR</td>
<td>8-h avg (7 a.m.-3 p.m.)</td>
<td>November: 32.7 (pre-race), 35.1 (race)</td>
<td>February: 37.0 (pre-race), 30.8 (race)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>July: 51.2 (pre-race), 46.1 (race)</td>
</tr>
<tr>
<td>Nickmilder et al. (2007)</td>
<td>Southern Belgium</td>
<td>July-Aug 2002</td>
<td>1-h max</td>
<td>NR</td>
<td>Max (across 6 camps): 24.6-112.8ᵇ</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>8-h max</td>
<td>NR</td>
<td>Max (across 6 camps): 19.0-81.1ᵇ</td>
</tr>
<tr>
<td>Delfino et al. (2010a)</td>
<td>Los Angeles, CA</td>
<td>Warm and cold season 2005-2007</td>
<td>24-h avg</td>
<td>Warm season median: 32.1ᶜ</td>
<td>Max: 76.4ᶜ</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Cool season median: 19.1ᶜ</td>
<td>Max: 44.9ᶜ</td>
</tr>
<tr>
<td>Adamkiewicz et al. (2004)</td>
<td>Steubenville, OH</td>
<td>Sept-Dec 2000</td>
<td>24-h avg</td>
<td>15.3</td>
<td>75th: 20.2, Max: 32.2</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-h avgᵈ</td>
<td>19.8</td>
<td>75th: 27.5, Max: 61.6</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.
NR = Not Reported, SOCS = Salmeterol Off Corticosteroids Study.
⁺Individual-level estimates were calculated based on time spent in the vicinity of various O₃ monitors.
ᵇConcentrations converted from µg/m³ to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).
ᶜMeasurements outside subject's residence (retirement home).
ᵈAverage O₃ concentration in the 1 hour preceding eNO collection.
Note: Results are presented first for children with asthma and adults with asthma then for older adults. Effect estimates are from single-pollutant models and are standardized to a 40-ppb increase for 1-h avg O₃ concentrations, a 30-ppb increase for 8-h max or 8-h avg O₃ concentrations, and a 20-ppb increase for 24-h avg O₃ concentrations.

### Figure 6-11
Percent change in exhaled nitric oxide (eNO) in association with ambient O₃ concentrations in populations with and without asthma.
Table 6-16  Percent change in exhaled nitric oxide (eNO) in association with ambient O₃ concentrations in populations with and without asthma for studies presented in Figure 6-11.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag</th>
<th>Subgroup</th>
<th>Standardized % Change (95% CI)a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Studies in individuals with asthma</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Liu et al. (2009a)</td>
<td>Windsor, ON, Canada</td>
<td>24-h avg</td>
<td>0</td>
<td>Without asthma</td>
<td>-25.1 (-42.9, -1.7)</td>
</tr>
<tr>
<td></td>
<td>182 children with asthma, ages 9-14 yr</td>
<td></td>
<td>1</td>
<td>With asthma</td>
<td>-17.5 (-32.1, -0.24)</td>
</tr>
<tr>
<td></td>
<td>Barraza-Villarreal et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>8-h moving avg</td>
<td>0</td>
<td>Without asthma</td>
</tr>
<tr>
<td></td>
<td>208 children, ages 6-14 yr</td>
<td></td>
<td></td>
<td>With asthma</td>
<td>6.2 (6.0, 6.5)</td>
</tr>
<tr>
<td></td>
<td>Berhane et al. (2011)</td>
<td>13 Southern California communities</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>1-23 cumulative avg</td>
<td>Without asthma</td>
</tr>
<tr>
<td></td>
<td>2,240 children, ages 6-10 yr</td>
<td></td>
<td></td>
<td>With asthma</td>
<td>26.0 (-1.4, 60.9)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Without allergy</td>
<td>25.5 (5.3, 49.6)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>With allergy</td>
<td>32.1 (12.0, 55.9)</td>
</tr>
<tr>
<td></td>
<td>Qian et al. (2009)</td>
<td>Boston, MA; New York, NY; Denver, CO; Philadelphia, PA; San Francisco, CA; Madison, WI</td>
<td>8-h max</td>
<td>0</td>
<td>Without asthma</td>
</tr>
<tr>
<td></td>
<td>119 children and adults with asthma, ages 12-65 yr</td>
<td></td>
<td>0-3 avg</td>
<td>With asthma</td>
<td>-1.0 (-1.8, -0.26)</td>
</tr>
<tr>
<td></td>
<td>Khatri et al. (2009)</td>
<td>Atlanta, GA</td>
<td>8-h max</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>Studies in older adults</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adamkiewicz et al. (2004)</td>
<td>Steubenville, Ohio</td>
<td>24-h avg</td>
<td>0</td>
<td></td>
<td>-5.7 (-25.9, 14.5)</td>
</tr>
<tr>
<td></td>
<td>29 older adults, ages 53-90 yr</td>
<td></td>
<td>1-h avg</td>
<td></td>
<td>-19.7 (-41.3, 1.9)</td>
</tr>
<tr>
<td>Delfino et al. (2010a)</td>
<td>Los Angeles, CA</td>
<td>24-h avg</td>
<td>0-4 avg</td>
<td>Cool season</td>
<td>35.2 (10.9, 59.5)</td>
</tr>
<tr>
<td></td>
<td>60 older adults, ages ≥ 65 yr</td>
<td></td>
<td></td>
<td>Warm season</td>
<td>-0.60 (-14.0, 12.8)</td>
</tr>
</tbody>
</table>

*Includes studies in Figure 6-11.

aEffect estimates are standardized to a 40-ppb, 30-ppb, and 20-ppb increase for 1-h avg, 8-h max or 8-h avg, and 24-h avg O₃, respectively.

bAverage O₃ concentration in the 1 hour preceding eNO collection.
Table 6-17  Associations between short-term ambient O₃ exposure and biological markers of pulmonary inflammation and oxidative stress.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location/Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag</th>
<th>Biological Marker</th>
<th>Subgroup</th>
<th>Standardized Effect Estimate (95% CI)³</th>
</tr>
</thead>
<tbody>
<tr>
<td>Liu et al. (2009a)</td>
<td>Windsor, ON, Canada, 182 children with asthma, ages 9 - 14 yr</td>
<td>24-h avg</td>
<td>0</td>
<td>EBC 8-isoprostane (% change)</td>
<td></td>
<td>16.2 (-13.9, 56.8)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>EBC TBARS (% change)</td>
<td></td>
<td>11.5 (-27.0, 70.1)</td>
</tr>
<tr>
<td>Romieu et al. (2008)</td>
<td>Mexico City, Mexico, 107 children with asthma, mean (SD) age 9.5 (2.1) yr</td>
<td>8-h max</td>
<td>0</td>
<td>EBC Malondialdehyde b</td>
<td></td>
<td>1.9 (1.1, 3.5)</td>
</tr>
<tr>
<td>Barraza-Villarreal et al. (2008)</td>
<td>Mexico City, Mexico, 208 children, ages 6 - 14 yr</td>
<td>8-h moving avg</td>
<td>0</td>
<td>Nasal lavage IL-8 (pg/mL)</td>
<td>Without asthma</td>
<td>1.6 (1.4, 1.9)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>With asthma</td>
<td></td>
<td>1.6 (1.4, 1.8)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>EBC pH</td>
<td>Without asthma</td>
<td>-0.10 (-0.27, 0.08)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>With asthma</td>
<td></td>
<td>-0.10 (-0.20, 0.01)</td>
</tr>
<tr>
<td>Sienna-Monge et al. (2004)</td>
<td>Mexico City, Mexico, 117 children with asthma, mean age 9 yr</td>
<td>8-h max 0-2 avg</td>
<td></td>
<td>Nasal lavage IL-8 e Placebo</td>
<td>2.2 (1.1, 4.7)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant</td>
<td></td>
<td>1.0 (0.44, 2.3)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Nasal lavage IL-6 e Placebo</td>
<td>2.7 (1.4, 5.1)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant</td>
<td></td>
<td>1.1 (0.53, 2.2)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Nasal lavage Uric acid Placebo</td>
<td>0.75 (0.44, 1.3)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant</td>
<td></td>
<td>1.3 (0.68, 2.4)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Nasal lavage Glutathione Placebo</td>
<td>0.79 (0.63, 0.98)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Antioxidant</td>
<td></td>
<td>0.80 (0.66, 0.96)</td>
</tr>
<tr>
<td>Khatri et al. (2009)</td>
<td>Atlanta, GA, 38 adults with asthma, ages 31 - 50 yr</td>
<td>8-h max</td>
<td>2</td>
<td>Blood eosinophils (% change)</td>
<td></td>
<td>2.4 (0.62, 4.2)</td>
</tr>
<tr>
<td>Ferdinands et al. (2008)</td>
<td>Atlanta, GA, 16 children exercising outdoors, ages 14 - 17 yr</td>
<td>1-h max</td>
<td>0</td>
<td>EBC pH (normalized score)</td>
<td></td>
<td>0.80 (-0.20, 1.8)</td>
</tr>
</tbody>
</table>

Results generally are presented in order of increasing mean ambient O₃ concentration. EBC = exhaled breath condensate, TBARS = thiobarbituric acid reactive substances, IL-8 = interleukin 8, IL-6 = interleukin 6, Antioxidant = group supplemented with vitamins C and E.

³Effect estimates are standardized to a 40-, 30- and 20-ppb increase for 1-h max, 8-h max or 8-h avg, and 24-h avg O₃, respectively.

³Effect estimates represent the ratio of the geometric means of biological marker per unit increase in O₃ concentration. A ratio <1 indicates a decrease in marker, and a ratio >1 indicates an increase in marker for an increase in O₃.

³Model analyzed log-transformed O₃. Decreases and increases in pH indicate increases and decreases in pulmonary inflammation, respectively.
Populations with Asthma

Exhaled Nitric Oxide

Neither NO nor eNO has been examined in the controlled human exposure or toxicological studies of O₃ exposure reviewed in this ISA. However, several lines of evidence support its analysis as an indicator of pulmonary inflammation. Inducible NO synthase can be activated by pro-inflammatory cytokines, and NO can be produced by cells such as neutrophils, eosinophils, and epithelial cells in the lung during an inflammatory response (Barnes and Liew, 1995). Further, eNO commonly is higher in individuals with asthma and increases during acute exacerbations (Jones et al., 2001; Kharitonov and Barnes, 2000).

As indicated in Figure 6-10 [and Table 6-16]), short-term increases in ambient O₃ concentration (8-h max or avg) were associated with increases in eNO in children with asthma. These studies used different methods to assign exposures using central site O₃ measurements: the site closest (within 5 km) to home or school (Barraza-Villarreal et al., 2008) and a single site per community (Berhane et al., 2011). Because information on spatial homogeneity of ambient O₃ concentrations and time spent outdoors was not available in these studies, it is not possible to assess whether these two methods produced different personal-ambient O₃ ratios and correlations. Liu et al. (2009a) (described in Section 6.2.1.2) reported O₃-associated decreases in eNO; however, this study was restricted to winter. Results for EBC markers of oxidative stress and lung function collectively also provided weak evidence of O₃-associated respiratory effects in this study. As described in Section 4.3.3, in non-summer months, indoor to outdoor O₃ ratios are lower as are personal-ambient ratios, making it more difficult to detect associations with ambient O₃ concentrations.

In contrast with controlled human exposure studies (Section 6.2.3.1), epidemiologic studies did not find larger O₃-associated increases in pulmonary inflammation in groups with asthma than in groups without asthma (Figure 6-11 [and Table 6-16]). Among children in Southern California, Berhane et al. (2011) estimated similar associations for a 1-23 day cumulative average of 8-h avg (10 a.m.-6 p.m.) O₃ in children with and without asthma. Among children in Mexico City, Mexico, Barraza-Villarreal et al. (2008) found a larger association (for lag 0 [of 8-max O₃]) in children without asthma, most of whom had atopy.

Studies that included adults with asthma produced contrasting results (Khatri et al., 2009; Qian et al., 2009). The multicity salmeterol (β-2 agonist) trial (Boston, MA; New York, NY; Denver, CO; Philadelphia, PA; San Francisco, CA; and Madison, WI) involved serial collection of eNO from 119 subjects with asthma, 87% of whom were 20-65 years of age (Qian et al., 2009). Ambient O₃ concentrations were averaged from all sites within 20 miles of subjects’ zipcode centroids, which in a repeated measures study, may capture the temporal variation in O₃ reasonably well (Darrow et al., 2011a; Gent et al., 2003). Among all subjects, increases in 8-h max O₃ at multiple lags (0 to 3 single-day and 0-4 avg) were associated with decreases in eNO. Results did not vary among the salmeterol-, CS-, and placebo-treated groups, indicating that the counterintuitive findings for O₃ were not only due to the reduction
of inflammation by medication. Qian et al. (2009) suggested that at higher concentrations, O₃ may transform NO in airways to reactive nitrogen species. However, this hypothesis was not supported by results from Khatri et al. (2009), who in Atlanta, GA examined overall higher 8-h max O₃ ambient concentrations than did Qian et al. (2009) and found that an increase in O₃ was associated with an increase in eNO in adults with asthma (36.6% [95% CI: 1.2, 71.9] per 30-ppb increase in lag 2 of 8-h max O₃). Although Khatri et al. (2009) was cross-sectional and did not adjust for any meteorological factors, it may have better characterized O₃ exposures because subjects were examined during warm months, and an 8-h max O₃ concentration was calculated for each subject using measurements at the site closest to his/her location each hour.

**Other biological markers of pulmonary inflammation and oxidative stress**

Short-term increases in ambient O₃ concentration were associated with changes in the levels of pro-inflammatory cytokines and cells, indicators of oxidative stress, and antioxidants (Table 6-17). Importantly, any particular biomarker was examined in only one to two studies, and the evidence in individuals with asthma is derived primarily from studies conducted in Mexico City, Mexico (Barraza-Villarreal et al., 2008; Romieu et al., 2008; Sienra-Monge et al., 2004). These studies measured ambient O₃ concentrations at sites within 5 km of subjects’ schools or homes. In a Mexico City cohort of children with asthma, school ambient O₃ concentrations averaged over 48 to 72 hours had a ratio and correlation with personal exposures (48- to 72-h avg) of 0.17 and 0.35, respectively (Ramírez-Aguilar et al., 2008). These observations suggest that the effects of personal O₃ exposure on inflammation may have been underestimated in the Mexico City studies. Despite the limited evidence, the epidemiologic findings are well supported by controlled human exposure and toxicological studies that measured the same or related endpoints.

Several of the modes of action of O₃ are mediated by reactive oxygen species (ROS) produced in the airways by O₃ (Section 5.3.3). These ROS are important mediators of inflammation as they regulate cytokine expression and inflammatory cell activity in airways (Heidenfelder et al., 2009). Controlled human exposure and toxicological studies, frequently have found O₃-induced increases in oxidative stress as shown by increases in prostaglandins (Section 5.3.3 and Section 6.2.3.1), which are produced by the peroxidation of cell membrane phospholipids (Morrow et al., 1990). Romieu et al. (2008) analyzed EBC malondialdehyde (MDA), a thiobarbituric acid reactive substance, which like prostaglandins, is derived from lipid peroxidation (Janero, 1990). For a 30-ppb increase in lag 0 of 8-h max O₃, the ratio of the geometric means of MDA was 1.9 (95% CI: 1.1, 3.5). Similar results were reported for lags 1 and 0-1 avg O₃. A limitation of the study was that 25% of EBC samples had nondetectable levels of MDA, and the random assignment of concentrations between 0 and 4.1 nmol to these samples may have contributed random measurement error to the estimated O₃ effects. Because MDA represents less than 1% of lipid peroxides and is present at low concentrations, its biological relevance has been questioned. However, Romieu et al. (2008) pointed to their observations of statistically significant
associations of EBC MDA levels with nasal lavage IL-8 levels to demonstrate its relationship with pulmonary inflammation.

Uric acid and glutathione are ROS scavengers that are present in the airway ELF. While the roles of these markers in the inflammatory cascade of asthma are not well defined, they have been observed to be consumed in response to short-term O₃ exposure as part of an antioxidant response in controlled human exposure and animal studies (Section 5.3.3). Results from an epidemiologic study also indicate that a similar antioxidant response may be induced by increases in ambient O₃ exposure. Sienra-Monge et al. (2004) found O₃-associated decreases in nasal lavage levels of uric acid and glutathione in children with asthma not supplemented with antioxidant vitamins (Table 6-17). The magnitudes of decrease were similar for O₃ concentrations lagged 2 or 3 days and averaged over 3 days.

Both controlled human exposure and toxicological studies have found O₃-induced increases in the cytokines IL-6 and IL-8 (Section 5.3.3, Section 6.2.3.1, and Section 6.2.3.3), which are involved in initiating an influx of neutrophils, a hallmark of O₃-induced inflammation (Section 6.2.3.1). Epidemiologic studies conducted in Mexico City, Mexico, had similar findings. Sienra-Monge et al. (2004) found that an increase in 8-h max O₃ was associated with increases in nasal lavage levels of IL-6 and IL-8 (placebo group), with larger effects estimated for lag 0-2 avg than for lag 2 or 3 O₃ (Table 6-17). In another cohort of children with asthma, a 30-ppb increase in lag 0 of 8-h max O₃ was associated with a 1.6 pg/mL increase (95% CI: 1.4, 1.8) in nasal lavage levels of IL-8 (Barraza-Villarreal et al., 2008). This study also reported a small O₃-associated decrease in EBC pH (Table 6-17). EBC pH, which is thought to reflect the proton-buffering capacity of ammonium in airways, decreases upon asthma exacerbation, and is negatively correlated with airway levels of pro-inflammatory cytokines (Carpagnano et al., 2005; Kostikas et al., 2002; Hunt et al., 2000).

Albeit with limited investigation, controlled human exposure studies have found O₃-induced increases in eosinophils in adults with asthma (Section 6.2.3.1). Eosinophils are believed to be the main effector cells that initiate and sustain inflammation in asthma and allergy (Schmekel et al., 2001). Consistent with these findings, in a cross-sectional study of adults with asthma in Atlanta, GA, a 30-ppb increase in lag 2 of 8-h max O₃ was associated with a 2.4% increase (95% CI: 0.62, 4.2) in blood eosinophils (Khatri et al., 2009). Potential confounding by weather was not evaluated in models.

The prominent influences demonstrated for ROS and antioxidants in mediating the respiratory effects of O₃ provide biological plausibility for effect modification by antioxidant capacity. Effect modification by antioxidant capacity has been described for O₃-associated lung function in controlled human exposure and epidemiologic studies (Section 6.2.1.1 and Section 6.2.1.2). An epidemiologic study conducted in Mexico City, Mexico, also found that vitamin C and E supplementation, which potentially increase antioxidant capacity, attenuated O₃-associated inflammation and oxidative stress. Among children with asthma supplemented daily with vitamin C
and E, the ratios of the geometric means of nasal lavage IL-6 and IL-8 per 30-ppb increases in lag 0-2 avg of 8-h max O₃ were approximately 1, reflecting no change with increases in O₃ concentration (Table 6-17) (Sienra-Monge et al., 2004). The results did not clearly delineate interactions among ambient O₃ concentrations, endogenous antioxidants, and dietary antioxidants (Table 6-17). Ozone was associated with increases in uric acid in the antioxidant group but decreases in the placebo group across the O₃ lags examined. Associations with glutathione were similar in the two groups. In another cohort, 8-h max O₃ concentrations ≥ 38 ppb enhanced the effects of diets high in antioxidant vitamins and/or omega-3 fatty acids in protecting against O₃-related increases in nasal lavage IL-8 (Romieu et al., 2009). Information on the main effects of O₃ or effect modification by diet was not presented.

The levels of several biological markers such as eNO, EBC pH, and MDA consistently differ between groups with and without asthma and change during an asthma exacerbation (Corradi et al., 2003; Hunt et al., 2000); however, the magnitudes of change associated with these overt effects are not well defined. Ozone-associated increases in interleukins and indicators of oxidative stress were small: 1-2% increase for each 30-ppb increase in 8-h max O₃ concentration (Table 6-17). Ozone-associated increases in eNO were larger: 6-36% increase per 30-ppb increase in 8-h max ambient O₃ concentration (Berhane et al., 2011; Delfino et al., 2010a; Khatri et al., 2009; Barraza-Villarreal et al., 2008). Some studies in populations with asthma found that increases in ambient O₃ concentration (the same lag) were associated with increases in pulmonary inflammation concurrently and respiratory symptoms. For example, among adults with asthma in Atlanta, GA, an increase in lag 2 ambient O₃ concentration was associated with increases in eNO, blood eosinophils, and a decrease in quality of life score, which incorporates indices for symptoms and activity limitations (Khatri et al., 2009). Also, among children with asthma in Mexico City, Mexico, lag 0 O₃ was associated with increases in eNO, nasal lavage IL-8, and concurrently assessed cough but not wheeze (Barraza-Villarreal et al., 2008).

**Children without Asthma**

In the limited investigation, short-term increases in ambient O₃ concentration (8-h max or avg) were associated with increases in pulmonary inflammation in children without asthma (Berhane et al., 2011; Barraza-Villarreal et al., 2008) (Figure 6-11) and Table 6-16 and Table 6-17). The study of children in Mexico City found a larger O₃-associated increase in eNO in the children without asthma than with asthma (13.5% versus 6.2% increase per 30-ppb increase in lag 0 of 8-h max O₃) (Barraza-Villarreal et al., 2008). Ozone was associated with similar magnitudes of change in IL-8 and EBC pH in children with and without asthma. A distinguishing feature of this study was that 72% of children without asthma had allergies. A study conducted in 13 Southern California communities also found that increases in ambient O₃ concentration (8-h avg, 10 a.m.-6 p.m.) were associated with increases in eNO in children with respiratory allergy (Berhane et al., 2011). Coherence for these
epidemiologic findings is provided by observations of O$_3$-induced allergic inflammation in animal models of allergy (Section 6.2.3.3 and Section 6.2.6).

Berhane et al. (2011) found O$_3$-associated increases in eNO in children without asthma and children without respiratory allergy, providing evidence for effects on pulmonary inflammation in healthy children. This study provided detailed information on differences in association among various lags of 8-h avg (10 a.m.-6 p.m.) O$_3$. Ozone concentrations averaged over the several hours preceding eNO collection were not significantly associated with eNO. Consistent with other studies examining pulmonary inflammation and oxidative stress, Berhane et al. (2011) found that relatively short lags of O$_3$, i.e., 1 to 5 days, were associated with increases in eNO. However, among several types of lag-based models, including unconstrained lag models, polynomial distributed lag models, spline-based distributed lag models, and cumulative lag models, a 23-day cumulative lag of O$_3$ best fit the data. Among the studies evaluated in this ISA, Berhane et al. (2011) was unique in evaluating and finding larger respiratory effects for multi-week (e.g., 13-30 days) average O$_3$ concentrations. A mechanism for the effects of O$_3$ peaking with a 23-day cumulative lag of exposure has not been delineated. Further, with examination of such long lag periods, there is greater potential for residual confounding by weather.

**Populations with Increased Outdoor Exposures**

With limited investigation, increases in ambient O$_3$ concentration were not consistently associated with pulmonary inflammation in populations engaged in outdoor activity or exercise. Common limitations of these studies were the small numbers of subjects and lack of consideration for potential confounding factors. A study in 16 adolescent long-distance runners near Atlanta, GA was noteworthy for the daily collection of EBC and the likely greater extent to which ambient O$_3$ concentrations represented ambient exposures because O$_3$ concentrations were measured during outdoor running at a site less than 1 mile from the exercise track (Ferdinands et al., 2008). Increases in 1-h max O$_3$ (lags 0 to 2) were associated with increases in EBC pH, indicating O$_3$-associated decreases in pulmonary inflammation. Among 9 adult male runners in Sicily, Italy examined 3 days before and 20 hours after 3 races in fall, winter, and summer, weekly average O$_3$ concentrations (8-h avg, 7 a.m.-3 p.m.) were positively correlated with apoptosis of neutrophils (Spearman’s r = 0.70, p <0.005) and bronchial epithelial cell differential counts (Spearman’s r = 0.47, p <0.05) but not with neutrophil or macrophage cell counts or levels of the pro-inflammatory cytokines TNF-α and IL-8 (Chimenti et al., 2009). Associations with O$_3$ concentrations measured during the races (mean 35 to 89 minutes) were not examined. This study provides evidence for new endpoints; however, the implications of findings are limited due to the lack of a rigorous statistical analysis.

In a cross-sectional study of children at camps in south Belgium, although lung function was not associated with O$_3$ measured at camps during outdoor activity, an association was found for eNO (Nickmilder et al., 2007). Children at camps with lag 0 1-h max O$_3$ concentrations >85.2 ppb had greater intraday increases in eNO.
compared with children at camps with O₃ concentrations <51 ppb. A benchmark dose analysis indicated that the threshold for an O₃-associated increase of 4.3 ppb eNO (their definition of increased pulmonary inflammation) was 68.6 ppb for 1-h max O₃ and 56.3 ppb for 8-h max O₃. While these results provide additional evidence for O₃-associated increases in pulmonary inflammation in healthy children, they should be interpreted with caution since they were not adjusted for any potential confounding factors and based on camp-level comparisons.

### Older Adults

The panel studies examining O₃-associated changes in eNO in older adults produced contrasting findings (Figure 6-1[1 and Table 6-16]). The studies differed with respect to geographic location, inclusion of healthy subjects, exposure assessment method, and lags of O₃ examined. Delfino et al. (2010a) followed 60 older adults with coronary artery disease in the Los Angeles, CA area for 6 weeks each during a warm and cool season; the specific months were not specified. Ambient O₃ was measured at subjects’ retirement homes, possibly reducing some exposure measurement error due to spatial variability. Multiday averages of O₃ (3- to 9-day) were associated with increases in eNO, with effect estimates increasing with increasing number of averaging days. In contrast with most other studies, an association was found in the cool season but not warm season (increase in eNO per 20-ppb increase in lag 0-4 avg of 24-h avg O₃: 35.2% [95% CI: 10.9, 59.5] in cool season, -0.06% [95% CI: -14.0, 12.8] in warm season). Despite these unusual findings for the cool season, they were similar to findings from another study of Los Angeles area adults with asthma, which indicated an O₃-associated decrease in indoor activity during the fall season (Eiswerth et al., 2005).

In a cool season (September-December) study conducted in older adults (ages 54-91 years) in Steubenville, OH, Adamkiewicz et al. (2004) found that increases in O₃ (1-h avg and 24-h avg before eNO collection) were associated with decreases in eNO, reflecting decreases in pulmonary inflammation (Figure 6-11 and Table 6-16). The study included healthy adults and those with asthma or COPD. A study in a subset of these adults illustrated why it is difficult to detect effects with central site O₃ concentrations in the cool season by showing that subjects spent ≥ 90% of time indoors and >77% at home and had a mean 24-h avg O₃ personal-ambient ratio of 0.27 (Sarnat et al., 2006a).

### Confounding in Epidemiologic Studies of Pulmonary Inflammation and Oxidative Stress

Except where noted in the preceding text; epidemiologic studies of pulmonary inflammation and oxidative stress accounted for potential confounding by meteorological factors. Increases in ambient O₃ concentration were associated with pulmonary inflammation or oxidative stress in models that adjusted for temperature and/or humidity (Delfino et al., 2010a; Barraza-Villarreal et al., 2008; Romieu et al.,
Final results from Sienra-Monge et al. (2004) and Berhane et al. (2011) were not adjusted for temperature because associations were not altered by adjustment for temperature. Most studies conducted over multiple seasons adjusted for season or time trend.

In evidence limited to a small number of studies conducted in Mexico City, Mexico, O₃-associated pulmonary inflammation and oxidative stress were not found to be confounded by PM₂.₅ or PM₁₀. These studies, which analyzed 8-hour averages for both O₃ and PM, found robust associations for O₃ (Barraza-Villarreal et al., 2008; Romieu et al., 2008; Sienra-Monge et al., 2004). Ozone and PM, both measured at central sites located within 5 km of subjects’ schools or homes, were moderately correlated (r = 0.46 – 0.54). Weak correlations have been found between personal exposures of O₃ and PM₂.₅ (Section 4.3.4.1). Only Romieu et al. (2008) provided quantitative results. Lag 0 of 8-h max O₃ was associated with a similar magnitude of increase in MDA without and with adjustment for lag 0 of 8-h max PM₂.₅ (ratio of geometric means for a 30-ppb increase: 1.3 [95% CI: 1.0, 1.7]). In comparison, the O₃-adjusted effect estimate for PM₂.₅ was cut in half.

Summary of Epidemiologic Studies of Pulmonary Inflammation and Oxidative Stress

Many epidemiologic studies provided evidence that short-term increases in ambient O₃ exposure increase pulmonary inflammation and oxidative stress in children with asthma, with evidence primarily provided by studies conducted in Mexico City. By also finding that associations were attenuated with higher antioxidant intake, these studies indicated that inhaled O₃ may be an important source of ROS in airways and/or may increase pulmonary inflammation via oxidative stress-mediated mechanisms. Studies also found O₃-associated increases in pulmonary inflammation in children with allergy (Berhane et al., 2011; Barraza-Villarreal et al., 2008). The limited available evidence in children and adults with increased outdoor exposures and older adults was inconclusive. Results did not indicate confounding of O₃ associations by temperature or humidity. Copollutant models were analyzed in a few studies conducted in Mexico City; O₃ effect estimates were robust to adjustment for moderately correlated (r = 0.46 – 0.54) PM₂.₅ or PM₁₀ (Barraza-Villarreal et al., 2008; Romieu et al., 2008; Sienra-Monge et al., 2004).

Ozone-associated increases in pulmonary inflammation and oxidative stress were found in studies that used varied exposure assessment methods: measurement on site of subjects’ outdoor activity (Nickmilder et al., 2007), average of concentrations measured at the closest site each hour Khatri et al. (2009), measurement at a site within 5 km of subjects’ schools or homes (Barraza-Villarreal et al., 2008; Romieu et al., 2008; Sienra-Monge et al., 2004), and measurement at single site per town (Berhane et al., 2011). While these methods may differ in the degree of exposure measurement error, in the limited body of evidence, there was not a clear indication that the method of exposure assessment influenced the strength or magnitude of associations.
Most studies examined and found associations with 8-h max or daytime 8-h avg O₃ concentrations, although associations also were found for 1-h max (Nickmilder et al., 2007) and 24-h avg O₃ (Delfino et al., 2010a). Collectively, studies examined single-day O₃ concentrations lagged from 0 to 5 days and concentrations averaged over 2 to 9 days. Lag 0 of 8-h max O₃ was most frequently examined and consistently associated with pulmonary inflammation and oxidative stress. However, in the few studies that examined multiple O₃ lags, multiday average 8-h max or 8-h avg concentrations were associated with larger increases in pulmonary inflammation and oxidative stress (Berhane et al., 2011; Delfino et al., 2010a; Sienra-Monge et al., 2004). These findings for multiday average O₃ concentrations are supported by controlled human exposure (Section 6.2.3.1) and animal studies (Section 6.2.3.3) that similarly have found that some markers of pulmonary inflammation remain elevated with O₃ exposures repeated over multiple days.

Several epidemiologic studies concurrently examined associations of ambient O₃ concentrations with biological markers of pulmonary inflammation and lung function or respiratory symptoms. Whether evaluated at the same or different lags of O₃, associations generally were stronger for biological markers of airway inflammation than for lung function within populations (Khatri et al., 2009; Barraza-Villarreal et al., 2008; Nickmilder et al., 2007). Controlled human exposure studies have demonstrated a lack of correlation between inflammatory and spirometric responses induced by O₃ exposure within subjects (Section 6.2.3.1). Evidence has suggested that O₃-related respiratory morbidity may occur via multiple mechanisms with varying time courses of action, and the examination of a limited number of O₃ lags in these aforementioned studies may explain some of the inconsistencies in associations of O₃ with measures of pulmonary inflammation and lung function. In contrast, based on examination in a few studies, increases in ambient O₃ concentration (at the same lag) were associated with increases in pulmonary inflammation and increases in respiratory symptoms or activity limitations in the same population of individuals with asthma (Khatri et al., 2009; Barraza-Villarreal et al., 2008).

6.2.3.3 Toxicology

The 2006 O₃ AQCD states that the “extensive human clinical and animal toxicological evidence, together with the limited available epidemiologic evidence, is clearly indicative of a causal role for O₃ in inflammatory responses in the airways” (U.S. EPA, 2006b). Airway ciliated epithelial cells and Type 1 cells are the most O₃-sensitive cells and are initial targets of O₃. These cells are damaged by O₃ and produce a number of pro-inflammatory mediators (e.g., interleukins [IL-6, IL-8], PGE₂) capable of initiating a cascade of events leading to PMN influx into the lung, activation of alveolar macrophages, inflammation, and increased permeability across the epithelial barrier. One critical aspect of inflammation is the potential for metaplasia and alterations in pulmonary morphology. Studies have observed increased thickness of the alveolar septa, presumably due to increased cellularity after acute exposure to O₃. Epithelial hyperplasia starts early in exposure and
increases in magnitude for several weeks, after which it plateaus until exposure ceases. When exposure persists for a month and longer, excess collagen and interstitial fibrosis are observed. This response, discussed in Chapter 7, continues to increase in magnitude throughout exposure and can even continue to increase after exposure ends (Last et al., 1984). Previously reviewed toxicological studies of the ability of O₃ to cause inflammation, injury, and morphological changes are described in Table 6-5 on page 6-25 (U.S. EPA, 1996f), Table 6-10 (U.S. EPA, 1996k) and Table 6-11 (U.S. EPA, 1996l) beginning on page 6-61 of the 1996 O₃ AQCD, and Annex Tables AX5-8 (U.S. EPA, 2006e) and AX5-9 (U.S. EPA, 2006f), beginning on page AX5-17 of the 2006 O₃ AQCD. Numerous recent in vitro and in vivo studies add to this very large body of evidence for O₃-induced inflammation and injury, and provide new information regarding the underlying mechanisms (see Section 5.3).

A number of species, including dogs, rabbits, guinea pigs, rats, and mice have been used as models to study the pulmonary effects of O₃, but the similarity of non-human primates to humans makes them an attractive model in which to study the pulmonary response to O₃. As reviewed in the 1996 and 2006 O₃ AQCDs, several pulmonary effects, including inflammation, changes in morphometry, and airway hyperresponsiveness, have been observed in macaque and rhesus monkeys after acute exposure to O₃ (Table 6-18 presents a highlight of these studies). Increases in inflammatory cells were observed after a single 8-hour exposure of adult rhesus monkeys to 1 ppm O₃ (Hyde et al., 1992). Inflammation was linked to morphometric changes, such as increases in necrotic cells, smooth muscle, fibroblasts, and nonciliated bronchiolar cells, which were observed in the trachea, bronchi, or respiratory bronchioles. Effects have also been observed after short-term repeated exposure to O₃ at concentrations that are more relevant to ambient O₃ concentrations. Morphometry changes in the lung, nose, and vocal cords were observed after exposure to 0.15 ppm O₃ for 8 hours/day for 6 days (Harkema et al., 1993; Dimitriadis, 1992; Harkema et al., 1987a).

Since 2006, however, only one study has been published regarding acute exposure of non-human primates to O₃ (a number of recent chronic studies in non-human primates are described in Chapter 7). In this study, a single 6-hour exposure of adult male cynomolgus monkeys to 1 ppm O₃ induced significant increases in inflammatory and injury markers, including BAL neutrophils, total protein, alkaline phosphatase, IL-6, IL-8, and G-CSF (Hicks et al., 2010a). Gene expression analysis confirmed the increases in the pro-inflammatory cytokine IL-8, which had been previously described in O₃ exposed rhesus monkeys (Chang et al., 1998). The anti-inflammatory cytokine IL-10 was also elevated, but the fold changes in IL-10 and G-CSF were relatively low and highly variable. The single exposure also caused necrosis and sloughing of the epithelial lining of the most distal portions of the terminal bronchioles and the respiratory bronchioles. Bronchiolitis, alveolitis, parenchymal and centriacinar inflammation were also observed. A second exposure protocol (two exposures with a 2-week inter-exposure period) resulted in similar inflammatory responses, with the exception of total protein and alkaline phosphatase levels which were attenuated, indicating that attenuation of some but not all lavage parameters occurred upon repeated exposure of non-human primates to O₃ (Hicks et
This variability in attenuation is similar to the findings of earlier reports in rodents (Wiester et al., 1996c) and non-human primates (Tyler et al., 1988).

Table 6-18 describes key morphometric studies conducted in non-human primates exposed to O3. Morphologic observations made by Dungworth (1976) and Dungworth et al. (1975) indicate that the rat and Bonnet monkey (Macaca radiata) are approximately equal in susceptibility to short-term effects of O3. Mild but discernible lesions were caused in both species by exposure to 0.2 ppm O3 for 8 hours/day for 7 days. The authors stated that detectable morphological effects in the rat occurred at levels as low as 0.1 ppm O3. In both species, the lesion occurred at the junction of the small airways and the gaseous exchange region. In rats, the prominent features were accumulation of macrophages, replacement of necrotic Type 1 epithelial cells with Type 2 cells, and damage to ciliated and nonciliated Clara cells. The principal site of damage was the alveolar duct. In monkeys, the prominent O3-induced injury was limited to the small airways. At 0.2 ppm O3, the lesion was observed at the proximal portion of the respiratory bronchioles. As concentrations of O3 were increased up to 0.8 ppm, the severity of the lesion increased, and the damage extended distally to involve the proximal portions of the alveolar duct.

Mellick et al. (1977) found similar but more pronounced effects when rhesus monkeys (3 to 5 years of age) were exposed to 0.5 and 0.8 ppm O3, 8 hours/day for 7 days. In these experiments, the respiratory bronchioles were the most severely damaged, whereas more distal parenchymal regions were unaffected. Major effects were hyperplasia and hypertrophy of the nonciliated bronchiolar epithelial cells and the accumulation of macrophages intraluminally. In mice, continuous exposure to 0.5 ppm O3 caused nodular hyperplasia of Clara cells after 7 days of exposure. Similar findings were reported by Schwartz (1976) and Schwartz et al. (1976), who exposed rats to 0.2, 0.5 or 0.8 ppm O3 for 8 or 24 hours/day for 1 week. Changes observed within the proximal alveoli included infiltration of inflammatory cells and swelling and necrosis of Type 1 cells. In the terminal bronchiole, the changes reported were shortened cilia, clustering of basal bodies in ciliated cells suggesting ciliogenesis, and reduction in height or loss of cytoplasmic luminal projection of the Clara cells. Effects were seen at O3 concentrations as low as 0.2 ppm. A dose-dependent pulmonary response to the three levels of O3 was evident. No differences were observed in morphologic characteristics of the lesions between rats exposed continuously and those exposed intermittently for 8 hours/day.
Exposure of adult BALB/c mice to 0.1 ppm O₃ for 4 hours increased BAL levels of keratinocyte chemoattractant (KC; IL-8 homologue) (~ 6-fold), IL-6 (~12-fold), and TNF-α (~ 2-fold) (Damera et al., 2010). Additionally, O₃ increased BAL neutrophils by 21% without changes in other cell types. A trend of increased neutrophils with increased O₃ concentration (0.12-2 ppm) was observed in BALB/c mice exposed for 3 hours (Jang et al., 2005). Although alterations in the epithelium of the airways were not evident in 129J mice after 4 hours of exposure to 0.2 ppm O₃ (Plopper et al., 2006), detachment of the bronchiolar epithelium was observed in SD rats after 5 days or 60 days of exposure to 0.25 ppm O₃ (Oyarzún et al., 2005). Subacute (65 hours) exposure to 0.3 ppm O₃ induced pulmonary inflammation, cytokine induction, and enhanced vascular permeability in wild type mice of a mixed background (129/Ola and C57BL/6) and these effects were exacerbated in metallothionein I/II knockout mice.
mice (Inoue et al., 2008). Three hours or 72 hours of exposure to 0.3 ppm O₃ resulted in similar levels of IL-6 expression in the lungs of C57BL/6 mice (Johnston et al., 2005b), along with increases in BAL protein, sTNFR1, and sTNFR2. Increased neutrophils were observed only after the 72-hour exposure, and neither exposure resulted in detectable levels of IL-6 or KC protein. Levels of BAL protein, sTNFR1, and sTNFR2 were higher in the 72-hour exposure group than in the 3-hour exposure group. In another study, the same subacute (72 hours) exposure protocol elicited increases in BALF protein, IP-10, sTNFR1, macrophages, neutrophils, and IL-6, IL-1α, and IL-1β expression (Johnston et al., 2007). Yoon et al. (2007) exposed C57BL/6J mice continuously to 0.3 ppm O₃ for 6, 24, 48, or 72 hours, and observed elevated levels of KC, MIP-2, metalloproteinases, and inflammatory cells in the lungs at various time points. A similar exposure protocol using C3H/HeJ and C3H/OuJ mice demonstrated elevations in protein, PMNs, and KC, which were predominantly TLR 4 pathway dependent based on their prominence in the TLR 4 sufficient C3H/OuJ strain Bauer et al. (2011). C3H/OuJ mice also had elevated levels of the heat-shock protein HSP70, and further experiments in HSP70 deficient mice indicated a role for this particular pathway in O₃-related injury, discussed in more detail in Chapter 5.

As reviewed in the 2006 O₃ AQCD, the time course for changes in BAL depends on the parameters being studied. Similarly, after exposing adult C57BL mice to 0.5 ppm O₃ for 3 hours, Han et al. (2008) observed early (5 hours postexposure) increases in BAL TNF-α and IL-1β, which diminished by 24 hours postexposure. Total BAL protein was elevated at 24 hours, but there were only minimal or negligible changes in LDH, total cells, or PMNs. Ozone increased BAL mucin levels (with statistical significance by 24 hours postexposure), and significantly elevated surfactant protein D at both time points. Prior intratracheal (IT) exposure to multiwalled carbon nanotubes enhanced most of these effects, but the majority of responses to the combined exposure were not greater than those to nanotubes alone. Ozone exposure did not induce markers of oxidative stress in lung tissue, BAL, or serum. Consistent with this study, Aibo et al. (2010) did not detect changes in BAL inflammatory cell numbers in the same mouse strain after a 6-hour exposure to 0.25 or 0.5 ppm. The majority of inflammatory cytokines (pulmonary or circulating) were not significantly changed (as assessed 9 hours post-O₃ exposure). Exposure of C57BL/6 mice to 1 ppm for 3 hours increased BAL total cells, neutrophils, and KC; these responses were greatest at 24 hours postexposure. F2-isoprostane (8-isoprostane), a marker of oxidative stress, was also elevated by O₃, peaking at 48 hours postexposure (Voynow et al., 2009).

Atopic asthma appears to be a risk factor for more severe airway inflammation induced by experimental O₃ exposure in humans (Balmes et al., 1997; Scannell et al., 1996), and allergic animal models are often used to investigate the effects of O₃ on this potentially at-risk population. Farraj et al. (2010) exposed allergen-sensitized adult male BALB/c mice to 0.5 ppm O₃ for 5 hours once per week for 4 weeks. Ovalbumin-sensitized mice exposed to O₃ had significantly increased BAL eosinophils by 85% and neutrophils by 103% relative to OVA sensitized mice exposed to air, but these changes were not evident upon histopathological evaluation.
of the lung, and no O₃ induced lesions were evident in the nasal passages. Ozone increased BAL levels of N-acetyl-glucosaminidase (NAG; a marker of injury) and protein. DEP co-exposure (2.0 mg/m³, nose only) inhibited these responses. These pro-inflammatory effects in an allergic mouse model have also been observed in rats. Wagner et al. (2007) exposed the relatively O₃-resistant Brown Norway rat strain to 1 ppm O₃ after sensitizing and challenging with OVA. Rats were exposed for 2 days, and airway inflammation was assessed one day later. Filtered air for controls contained less than 0.02 ppm O₃. Histopathology indicated that O₃ induced site-specific lung lesions in the centriacinar regions, characterized by wall thickening partly due to inflammatory cells influx. BAL neutrophils were elevated by O₃ in allergic rats, and modestly increased in non-allergic animals (not significant). A slight (but not significant) increase in macrophages was observed, but eosinophil numbers were not affected by O₃. Soluble mediators of inflammation (Cys-LT, MCP-1, and IL-6) were elevated by O₃ in allergic animals but not non-allergic rats. Treatment with γT, which neutralizes oxidized lipid radicals and protects lipids and proteins from nitrosative damage, did not alter the morphologic character or severity of the centriacinar lesions caused by O₃, nor did it reduce neutrophil influx. It did, however, significantly reduce O₃-induced soluble inflammatory mediators in allergic rats. The effects of O₃ in animal models of allergic asthma are discussed in Section 6.2.6.

In summary, a large number of toxicology studies have demonstrated that acute exposure to O₃ produces injury and inflammation in the mammalian lung, supporting the observations in controlled human exposure studies (Section 6.2.3.1) and epidemiologic studies (Section 6.2.3.2). These acute changes, both in inflammation and morphology, provide a limited amount of evidence for long term sequelae of exposure to O₃. Related alterations resulting from long term exposure, such as fibrotic changes, are discussed in Chapter 7.

### 6.2.4 Respiratory Symptoms and Medication Use

Controlled human exposure and toxicological studies have described modes of action through which short-term O₃ exposure may increase respiratory symptoms by demonstrating O₃-induced airway hyperresponsiveness (Section 6.2.2) and pulmonary inflammation (Section 6.2.3.1 and Section 6.2.3.3). Epidemiologic studies have not widely examined associations between ambient O₃ concentrations and airway hyperresponsiveness but have found O₃-associated increases in pulmonary inflammation and oxidative stress (Section 6.2.3.2). In addition to lung function decrements, controlled human exposure studies clearly indicate O₃-induced increases in respiratory symptoms including pain on deep inspiration, shortness of breath, and cough. This evidence is detailed in Section 6.2.1.1; however, salient observations include an increase in respiratory symptoms with increasing concentration and duration of O₃ exposure and activity level of exposed subjects (McDonnell et al., 1999b). Further, increases in total subjective respiratory symptoms have been reported following 5.6 and 6.6 hours of exposure to 60 ppb O₃ relative to baseline.
At 70 ppb, Schelegle et al. (2009) observed a statistically significant O$_3$-induced FEV$_1$ decrement of 6.1% at 6.6 hours and a significant increase in total subjective symptoms at 5.6 and 6.6 hours. The findings for O$_3$-induced respiratory symptoms in controlled human exposure studies and the evidence integrated across disciplines describing underlying modes of action provide biological plausibility for epidemiologic associations observed between short-term increases in ambient O$_3$ concentration and increases in respiratory symptoms.

In epidemiologic studies, respiratory symptom data typically are collected by having subjects (or their parents) record symptoms and medication use in a diary without direct supervision by study staff. Several limitations of symptom reports are well recognized: recall error if not recorded daily, differences among subjects in the interpretation of symptoms, differential reporting by subjects with and without asthma, and occurrence in a smaller percentage of the population compared with changes in lung function and biological markers of pulmonary inflammation. Nonetheless, symptom diaries remain a convenient tool to collect individual-level data from a large number of subjects and allow modeling of associations between daily changes in O$_3$ concentration and daily changes in respiratory morbidity over multiple weeks or months. Importantly, most of the limitations described above are sources of random measurement error that can bias effect estimates to the null or increase the uncertainty around effect estimates. Furthermore, because respiratory symptoms are associated with limitations in activity and function and are the primary reason for using medication and seeking medical care, the evidence is directly coherent with the associations consistently observed between increases in ambient O$_3$ concentration and increases in asthma ED visits (Section 6.2.7.3).

Most studies of respiratory symptoms were conducted in individuals with asthma, and as was concluded in the 2006 O$_3$ AQCD (U.S. EPA, 2006b, 1996a), the collective body of epidemiologic evidence indicates that short-term increases in ambient O$_3$ concentration are associated with increases in respiratory symptoms in children with asthma. Studies also found O$_3$-associated increases in the use of asthma medication by children. In a smaller body of studies, increases in ambient O$_3$ concentration were associated with increases in respiratory symptoms in adults with asthma. Ozone-associated increases in respiratory symptoms in healthy populations were not as clearly indicated.

### 6.2.4.1 Children with Asthma

#### Respiratory Symptoms

Table 6-19 presents the locations, time periods, and ambient O$_3$ concentrations for studies examining respiratory symptoms and medication use in children with asthma. The evidence supporting associations between short-term increases in ambient O$_3$ concentration and increases in respiratory symptoms in children with asthma is derived mostly from examination of 1-h max, 8-h max, or 8-h avg O$_3$ concentrations.
and strong findings from a large body of single-region or single-city studies (Figure 6-12 and Table 6-20). The few available U.S. multicity studies produced less consistent associations, but the overall body of epidemiologic evidence remains compelling. As detailed below, because of specific methodological distinctions, results from some multicity studies were not given greater consideration than results from single city studies in weighing the evidence for ambient O₃ exposure and respiratory symptoms.

Similar to lung function, associations with respiratory symptoms in children with asthma were found with ambient O₃ concentrations assigned to subjects using various methods with potentially different degrees of exposure measurement error. As was discussed for lung function, methods included measurement of O₃ on site of and at the time of outdoor activity (Thurston et al., 1997), which is associated with higher ambient-personal O₃ correlations and ratios (Section 4.3.3); O₃ concentrations measured at sites within 5 km of subjects’ home or school (Escamilla-Núñez et al., 2008; Romieu et al., 2006; Romieu et al., 1997; Romieu et al., 1996); O₃ measured at a single city site (Gielen et al., 1997); and O₃ concentrations averaged across multiple sites (Gent et al., 2003; Mortimer et al., 2002). In analyses with O₃ averaged across multiple sites, which were restricted to warm seasons, O₃ concentrations within the region were temporally correlated as indicated by high statewide correlations [median r = 0.83 in Gent et al. (2003)] or similar odds ratios for O₃ averaged across all within-city monitors and that averaged from the three closest sites (Mortimer et al., 2002). In these panel studies, the ambient concentrations averaged across sites may have well represented the temporal variability in subjects’ ambient O₃ exposures.
Table 6-19  Mean and upper percentile O$_3$ concentrations in epidemiologic studies of respiratory symptoms, medication use, and activity levels in children with asthma.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O$_3$ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Thurston et al. (1997)</td>
<td>CT River Valley, CT</td>
<td>June 1991-1993</td>
<td>1-h max</td>
<td>83.6&lt;sup&gt;a&lt;/sup&gt;</td>
<td>Max: 160&lt;sup&gt;a&lt;/sup&gt;</td>
</tr>
<tr>
<td>Escamilla-Nuñez et al. (2008)</td>
<td>Mexico City, Mexico</td>
<td>July-Mar 2003-2005</td>
<td>1-h max</td>
<td>86.5</td>
<td>NR</td>
</tr>
<tr>
<td>Gent et al. (2003)</td>
<td>CT, Southern MA</td>
<td>Apr-Sept 2001</td>
<td>8-h rolling avg, 1-h max</td>
<td>51.3, median 50.0, 58.6, median 55.5</td>
<td>Max: 99.6, Max: 125.5</td>
</tr>
<tr>
<td>Mortimer et al. (2002); Mortimer et al. (2000)</td>
<td>Bronx, East Harlem, NY; Baltimore, MD; Washington, DC; Detroit, MI; Cleveland, OH; Chicago, IL; St. Louis, MO (NCICAS)</td>
<td>June-Aug 1993</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>48</td>
<td>NR</td>
</tr>
<tr>
<td>Gielen et al. (1997)</td>
<td>Amsterdam, Netherlands</td>
<td>Apr-July 1995</td>
<td>8-h max</td>
<td>34.2&lt;sup&gt;b&lt;/sup&gt;</td>
<td>Max: 56.5&lt;sup&gt;b&lt;/sup&gt;</td>
</tr>
<tr>
<td>Delfino et al. (2003)</td>
<td>Los Angeles, CA</td>
<td>Nov 1999-Jan 2000</td>
<td>8-h max, 1-h max</td>
<td>17.1, 25.4</td>
<td>90th: 26.1, Max: 37, 90th: 38.0, Max: 52</td>
</tr>
<tr>
<td>Rabinovitch et al. (2004)</td>
<td>Denver, CO</td>
<td>Nov-Mar 1999-2002</td>
<td>1-h max</td>
<td>28.2</td>
<td>75th: 60, Max: 70.0</td>
</tr>
<tr>
<td>Schildcrout et al. (2006)</td>
<td>Albuquerque, NM; Baltimore, MD; Boston, MA; Denver, CO; San Diego, CA; Seattle, WA; St. Louis, MO; Toronto, ON, Canada (CAMP)</td>
<td>May-Sept 1994-1995</td>
<td>1-h max</td>
<td>Range in medians across cities: 43.0-65.8</td>
<td>Range in 90th across cities: 61.5-94.7</td>
</tr>
<tr>
<td>Jalaludin et al. (2004)</td>
<td>Sydney, Australia</td>
<td>Feb-Dec 1994</td>
<td>15-h avg (6 a.m.-9 p.m.)</td>
<td>12</td>
<td>Max: 43</td>
</tr>
<tr>
<td>Study*</td>
<td>Location</td>
<td>Study Period</td>
<td>O₃ Averaging Time</td>
<td>Mean/Median Concentration (ppb)</td>
<td>Upper Percentile Concentrations (ppb)</td>
</tr>
<tr>
<td>--------</td>
<td>----------</td>
<td>--------------</td>
<td>-------------------</td>
<td>----------------------------------</td>
<td>--------------------------------------</td>
</tr>
<tr>
<td>O'Connor et al. (2008)</td>
<td>Boston, MA; Bronx, Manhattan NY; Chicago, IL; Dallas, TX, Seattle, WA; Tucson, AZ (ICAS)</td>
<td>Aug 1998-July 2001</td>
<td>24-h avg</td>
<td>NR</td>
<td>NR</td>
</tr>
<tr>
<td>Mann et al. (2010)</td>
<td>Fresno/Clovis, California</td>
<td>Winter-Summer 2000-2005</td>
<td>8-h max</td>
<td>49.4 (median)</td>
<td>75th: 69.5, Max: 120.0</td>
</tr>
<tr>
<td>Just et al. (2002)</td>
<td>Paris, France</td>
<td>Apr-June 1996</td>
<td>24-h avg</td>
<td>30.0&lt;sup&gt;b&lt;/sup&gt;</td>
<td>Max: 61.7&lt;sup&gt;b&lt;/sup&gt;</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.
NR = Not Reported, NCICAS = National Cooperative Inner-City Asthma Study, CAMP = Childhood Asthma Management Program, ICAS = Inner City Asthma Study.
<sup>a</sup>Measured on site of subjects’ outdoor activity.
<sup>b</sup>Concentrations converted from µg/m³ to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).
Note: Results are presented first for aggregate indices of symptoms then for individual symptoms. Within each category, results generally are organized in order of increasing mean ambient O₃ concentration. LRS = lower respiratory symptoms, URS = upper respiratory symptoms. Odds ratios are from single-pollutant models and are standardized to a 40-, 30-, and 20-ppb increase for 1-h max, 8-h max (or 8-h avg or 15-h avg), and 24-h avg O₃ concentrations, respectively.

Figure 6-12  Associations between ambient O₃ concentrations and respiratory symptoms in children with asthma.
Table 6-20  Associations between ambient $O_3$ concentrations and respiratory symptoms in children with asthma for studies presented in Figure 6-12.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/Population</th>
<th>O$_3$ Averaging Time</th>
<th>O$_3$ Lag</th>
<th>Symptom</th>
<th>Subgroup</th>
<th>Standardized OR (95% CI)$^a$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Studies examining aggregates of symptoms</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
| Delfino et al. (2003) | Los Angeles, CA  
22 children with asthma, ages 10-16 yr | 8-h max  
1-h max | 0 | Bothersome symptoms |          | 0.75 (0.24, 2.30)  
1.09 (0.39, 3.03) |
| Rabinovitch et al. (2004)  | Denver, CO  
86 children with asthma, ages 6-12 yr | 1-h max | 0-2 avg | Daytime symptoms |          | 1.32 (1.01, 1.74) |
| Schildcrout et al. (2006)  | Albuquerque, NM; Baltimore, MD; Boston, MA; Denver, CO; San Diego, CA; Seattle, WA; St. Louis, MO; Toronto, ON, Canada  
990 children with asthma, ages 5-12 yr | 1-h max | 0-2 sum | Asthma symptoms |          | 1.08 (0.89, 1.31)  
1.01 (0.92, 1.12) |
| Gielen et al. (1997)  | Amsterdam, Netherlands  
61 children with asthma, ages 7-13 yr | 8-h max | 0 | LRS  
URS | All subjects | 1.04 (0.75, 1.45)  
1.16 (1.02, 1.32) |
| Mortimer et al. (2002); Mortimer et al. (2000)  | Bronx, East Harlem, NY; Baltimore, MD; Washington, DC; Detroit, MI, Cleveland, OH; Chicago, IL; St. Louis, MO  
846 children with asthma, ages 4-9 yr | 8-h avg  
(10 a.m.- 6 p.m.) | 1-4 avg | Morning symptoms | All subjects  
Centers | 1.06 (0.88, 1.27)  
1.21 (1.04, 1.41)  
1.02 (0.88, 1.18)  
1.19 (1.02, 1.38)  
1.35 (1.04, 1.74)  
1.08 (0.62, 1.87)  
2.13 (1.12, 4.04)  
1.39 (0.98, 1.98)  
1.17 (0.79, 1.72)  
1.59 (1.00, 2.52)  
1.35 (0.92, 1.96) |
| Romieu et al. (1996)  | Northern Mexico City, Mexico  
71 children with asthma, ages 5-7 yr | 1-h max | 0 | LRS | All subjects | 1.07 (1.02, 1.12) |
| Romieu et al. (1997)  | Southern Mexico City, Mexico  
65 children with asthma, ages 5-13 yr | 1-h max | 0 | LRS | All subjects | 1.09 (1.04, 1.14) |
|        | Studies examining individual symptoms                                                  |                       |           |                          |          |                             |
| Jalaludin et al. (2004)  | Sydney, Australia  
125 children with asthma, mean age 9.6 yr | 15-h avg  
(6 a.m.- 9 p.m.) | 0 | Wheeze | All subjects | 0.93 (0.63, 1.37)  
1.15 (0.94, 1.41) |
| O’Connor et al. (2008)  | Boston, MA; Bronx, Manhattan NY; Chicago, IL; Dallas, TX; Seattle, WA; Tucson, AZ  
861 children with asthma, mean (SD) age 7.7 (2.0) yr | 24-h avg  
1-19 avg | 0 | Wheeze/ cough | All subjects | 1.02 (0.86, 1.21) |
<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag</th>
<th>Symptom</th>
<th>Subgroup</th>
<th>Standardized OR (95% CI)a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Just et al. (2002)</td>
<td>Paris, France 82 children with asthma, mean (SD) age 10.9 (2.5) yr</td>
<td>24-h avg</td>
<td>0</td>
<td>Nocturnal cough</td>
<td></td>
<td>1.17 (0.72, 1.91)</td>
</tr>
<tr>
<td>Ostro et al. (2001)</td>
<td>Los Angeles, CA 138 children with asthma, ages 6-13 yr</td>
<td>1-h max</td>
<td>3</td>
<td>Wheeze</td>
<td></td>
<td>0.94 (0.88, 1.00)</td>
</tr>
<tr>
<td>Escamilla- Núñez et al. (2008)</td>
<td>Mexico City, Mexico 147 children with asthma, mean (SD) age 9.6 (2.1) yr</td>
<td>1-h max</td>
<td>1</td>
<td>Wheeze</td>
<td></td>
<td>1.08 (1.03, 1.14)</td>
</tr>
<tr>
<td>Mann et al. (2010)</td>
<td>Fresno/Clovia, California 280 children with asthma, ages 6-11 yr</td>
<td>8-h max</td>
<td>0</td>
<td>Wheeze All Fungi allergic</td>
<td></td>
<td>1.00 (0.84, 1.19)</td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>CT River Valley, CT 166 children with asthma, ages 7-13 yr</td>
<td>1-h max</td>
<td>0</td>
<td>Chest symptoms</td>
<td></td>
<td>1.28 (1.09, 1.51)</td>
</tr>
<tr>
<td>Romieu et al. (2006)</td>
<td>Mexico City, Mexico 151 children with asthma, mean age 9 yr</td>
<td>1-h max</td>
<td>0-5 avg</td>
<td>Difficulty breathing</td>
<td>GSTM1 positive</td>
<td>1.10 (0.98, 1.24)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>GSTM1 null</td>
<td>1.17 (1.02, 1.33)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>GSTP1 Ile/Ile or Ile/Val</td>
<td>1.06 (0.94, 1.20)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>GSTP1 Val/Val</td>
<td>1.30 (1.10, 1.53)</td>
</tr>
<tr>
<td>Gent et al. (2003)b</td>
<td>CT, Southern MA 130 children with asthma on maintenance medication</td>
<td>1-h max</td>
<td>0</td>
<td>Wheeze</td>
<td></td>
<td>1.00 (reference)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ &lt;43.2 ppb</td>
<td>1.04 (0.89, 1.21)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ 43.2-51.5 ppb</td>
<td>1.16 (1.00, 1.35)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ 51.6-58.8 ppb</td>
<td>1.16 (1.00, 1.35)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ 58.9-72.6 ppb</td>
<td>1.22 (0.97, 1.53)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ ≥ 72.7 ppb</td>
<td>1.00 (reference)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ &lt;43.2 ppb</td>
<td>1.11 (0.91, 1.36)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ 43.2-51.5 ppb</td>
<td>1.01 (0.83, 1.23)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ 51.6-58.8 ppb</td>
<td>1.16 (0.97, 1.39)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>O₃ 58.9-72.6 ppb</td>
<td>1.31 (0.97, 1.77)</td>
</tr>
</tbody>
</table>

*Includes studies for Figure 6-12, plus others.
LRS = Lower respiratory symptoms, URS = Upper respiratory symptoms.

Effect estimates are standardized to a 40, 30, and 20 ppb increase for 1-h max, 8-h max (or 8-h avg or 15-h avg), and 24-h avg O₃, respectively.

Results not included in Figure 6-12 because results presented per quintile of ambient O₃ concentration.

Among U.S. multicity studies of children with asthma, each of which examined a different O₃ averaging time, O₃ was not consistently associated with increases in respiratory symptoms (O'Connor et al., 2008; Schildcrout et al., 2006; Mortimer et al., 2002). In the NCICAS (described in Section 6.2.1.2), which analyzed greater than 11,000 person-days of data during one warm season, increases in most evaluated lags of O₃ (1 to 4 and 1-4 avg) were associated with increases in asthma symptoms. A 30-ppb increase in lag 1-4 avg, of 8-h avg (10 a.m.-6 p.m.), O₃ was associated with an increase in morning asthma symptoms with an OR of 1.35 (95% CI: 1.04, 1.69) (Mortimer et al., 2002). The OR was similar in an analysis restricted to O₃ concentrations <80 ppb. Associations were similarly strong for lags 2 and 4 of O₃ but weaker for lags 1 and 3 (Figure 6-12 [and Table 6-20]).
Like NCICAS, the U.S. multicity Childhood Asthma Management Program (CAMP, with two cities in common with NCICAS, Table 6-19) collected daily symptom data, analyzed data collected between May and September, and evaluated multiple lags of O3 (Schildcrout et al., 2006). However, associations in CAMP were weaker for all evaluated lags of O3. In meta-analyses that combined city-specific estimates, a 40-ppb increase in lag 0 of 1-h max O3 was associated with asthma symptoms with an OR of 1.08 (95% CI: 0.89, 1.31). Odds ratios for lags 1 and 2 and the lag 0-2 sum of O3 were between 1.0 and 1.03. In this study, data available from an average of 12 subjects per day per city were used to produce city-specific ORs. The person-days of data contributing to each city-specific model were likely less than those of the other multicity studies. These city-specific ORs then were combined in meta-analyses to produce study-wide ORs. Because of these methodological details of CAMP, power to detect associations with O3 likely was less than that for other pollutants (which were analyzed using year-round data), other multicity studies, and several available single-city studies.

Inconsistent associations between wheeze and nighttime asthma were reported in the ICAS cohort (described in Section 6.2.1.2) (O'Connor et al., 2008); however, the results are considered separately from the other available evidence because symptom incidence was examined in association with 19-day avg (of 24-h avg) concentrations of O3. Most evidence, whether from multi- or single-city studies, indicates associations of respiratory symptoms with shorter lags of O3 up to a few days. The implications of ICAS results are more limited because of a lack of a well-characterized mode of action for respiratory symptoms resulting from longer lag periods of O3 exposure. ICAS was precluded from examining shorter lag periods because data were collected every 2 months on the number of days with symptoms during the previous 2 weeks.

Several longitudinal studies conducted in different cohorts of children with asthma in Mexico City, Mexico examined and found increases in respiratory symptoms in association with 1-h max O3 concentrations (Escamilla-Nuñez et al., 2008; Romieu et al., 2006; Romieu et al., 1997; Romieu et al., 1996). Romieu et al. (1997); (1996) found larger increases in symptoms in association with increases in 1-h max O3 at lag 0 than at lag 1 or 2. Recent studies in Mexico City expanded on earlier evidence by indicating associations with multiday averages of O3 concentrations. Romieu et al. (2006) and Escamilla-Nuñez et al. (2008) found that ORs for associations of ambient 1-h max O3 concentrations with respiratory symptoms and medication use increased as the number of averaging days increased (up to lag 0-5 avg).

Studies of children with asthma examined factors that may modify symptom responses to ambient O3 exposure but did not produce conclusive evidence. Larger O3-associated (8-h avg [10 a.m.-6 p.m.] or 8-h max) increases in symptoms were found in children taking asthma medication, although the specific medications examined differed between studies. As with results for PEF, in the NCICAS multicity cohort, O3-associated increases in morning symptoms were larger in children taking cromolyn (used to treat asthma with allergy) or beta-agonists/xanthines than in children taking no medication. Odds ratios were similar in
children taking steroids and children taking no medication (Figure 6-12 [and Table 6-20]) (Mortimer et al., 2000). Among children with asthma in Southern New England, O₃-associated increases in symptoms were limited mostly to children taking steroids, cromolyn, or leukotriene inhibitors for maintenance (Gent et al., 2003).

In most studies of children with asthma, a majority of subjects (52 to 100%) had atopy as determined by sensitization to any examined allergen. While other studies found O₃-associated increases in pulmonary inflammation in children with atopy (Section 6.2.3.2) and in animal models of allergy (Section 6.2.3.3), evidence did not indicate that the risk of O₃-associated respiratory symptoms differed in children with asthma with and without atopy. In the NCICAS, Mortimer et al. (2000) found that an increase in lag 1-4 avg (8-h avg, 10 a.m.-6 p.m.) O₃ was associated with a similar increased incidence of asthma symptoms among the 79% of subjects with atopy and the 21% of subjects without atopy (Figure 6-12 [and Table 6-20]). Odds ratios for O₃ did not differ by residential allergen levels. Among children with asthma in Fresno, CA, most ORs for associations of single- and multi-day lags of 8-h max O₃ concentrations (0-14 days) with wheeze were near or below 1.0 among all subjects. Among the various O₃ lags examined, increases in O₃ were not consistently associated with increases in wheeze in subjects with cat or fungi allergy either (Mann et al., 2010).

Romieu et al. (2006) found differences in O₃-associated respiratory symptoms by genetic variants in GST enzymes, particularly, GSTP1 but less so for GSTM1. Compared with GSTP1 Ile/Ile or Ile/Val subjects, larger effects were estimated for GSTP1 Val/Val subjects (Figure 6-12 [and Table 6-20]). The largest OR was found for difficulty breathing in children with asthma who had both GSTM1 null and GSTP1 Val/Val genotypes (OR: 1.49 [95% CI: 1.14, 1.93] per 40-ppb increase in lag 0-5 avg of 1-h max O₃). These results are consistent with those described for antioxidant capacity modifying O₃-associated changes in lung function (Section 6.2.1.2) and pulmonary inflammation (Section 6.2.3.2 for results in the same cohort (Sienra-Monge et al., 2004)); however, effect modification by GSTP1 variants has not been consistent. (Romieu et al., 2006) found an O₃-associated decrease in FEV₁ only in children with GSTP1 Ile/Ile or Ile/Val genotype. Among children in southern California, GSTP1 Ile/Ile was associated with greater risk of asthma onset (Section 7.2.1). Asthma prevalence has not been consistently associated with a particular GSTP1 genotype either (Tamer et al., 2004; Mapp et al., 2002; Hemmingsen et al., 2001).

**Asthma Medication Use**

Although recent studies contributed mixed evidence, the collective body of evidence supports associations between increases in ambient O₃ concentration and increased asthma medication use in children (Figure 6-13 [and Table 6-21]). Most studies examined and found associations with 1-h max O₃ concentrations lagged 0 or 1 day; however, associations also were found for multiday average O₃ concentrations (lag
0-5 avg in Romieu et al. (2006) and lags 0-2 avg and 0-4 avg in Just et al. (2002). Within several studies, associations of \( \text{O}_3 \) were similar for respiratory symptoms and asthma medication use (Escamilla-Nuñez et al., 2008; Romieu et al., 2006; Schildcrout et al., 2006; Jalaludin et al., 2004; Romieu et al., 1997; Thurston et al., 1997). As an exception, Romieu et al. (1996) found that \( \text{O}_3 \) was associated with an increase in respiratory symptoms but not bronchodilator use, and Rabinovitch et al. (2004) indicated statistically significant associations with symptoms but not bronchodilator use (OR not reported). A few studies found higher odds of \( \text{O}_3 \)-associated increases in asthma medication use than in respiratory symptoms (Just et al., 2002; Ostro et al., 2001).

<table>
<thead>
<tr>
<th>Study</th>
<th>Medication</th>
<th>( \text{O}_3 ) Lag</th>
<th>Subgroup</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jalaludin et al. (2004)</td>
<td>Beta-agonist, no steroid</td>
<td>1</td>
<td></td>
</tr>
<tr>
<td>Gielen et al. (1997)</td>
<td>Bronchodilator</td>
<td>0</td>
<td></td>
</tr>
<tr>
<td>Schildcrout et al. (2006)</td>
<td>Rescue inhaler</td>
<td>0</td>
<td></td>
</tr>
<tr>
<td>Ostro et al. (2001)</td>
<td>Extra medication</td>
<td>1 0.7</td>
<td>moderate/severe asthma Los Angeles</td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>Beta-agonist</td>
<td>0</td>
<td></td>
</tr>
<tr>
<td>Romieu et al. (2006)</td>
<td>Bronchodilator</td>
<td>0-5 avg 1.1</td>
<td>GSTM1 positive GSTM1 null GSTP1 Ile/Ile/Ile/Val GSTP1 Val/Val</td>
</tr>
<tr>
<td>Romieu et al. (1996)</td>
<td>Bronchodilator</td>
<td>0</td>
<td></td>
</tr>
<tr>
<td>Romieu et al. (1997)</td>
<td>Bronchodilator</td>
<td>0</td>
<td></td>
</tr>
</tbody>
</table>

Note: Results generally are presented in order of increasing mean ambient \( \text{O}_3 \) concentration. Odds ratios are from single-pollutant models and are standardized to a 40-ppb increase for 1-h max \( \text{O}_3 \) and a 30-ppb increase for 8-h max or 15-h avg \( \text{O}_3 \).

Figure 6-13  Associations between ambient \( \text{O}_3 \) concentrations and asthma medication use.
Table 6-21  Associations between ambient O₃ concentrations and asthma medication use for studies presented in Figure 6-13.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/Population</th>
<th>O₃ Averaging Time</th>
<th>O₃ Lag</th>
<th>Medication</th>
<th>Subgroup</th>
<th>Standardized OR (95% CI)a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jalaludin et al. (2004)</td>
<td>Sydney, Australia 125 children with asthma, mean age 9.6 yr</td>
<td>15-h avg (6 a.m.-9 p.m.)</td>
<td>1</td>
<td>Beta-agonist, no CS</td>
<td></td>
<td>1.06 (0.91, 1.23)</td>
</tr>
<tr>
<td>Gielen et al. (1997)</td>
<td>Amsterdam, Netherlands 61 children with asthma, ages 7-13 yr</td>
<td>8-h max</td>
<td>0</td>
<td>Bronchodilator</td>
<td></td>
<td>1.10 (0.78, 1.55)</td>
</tr>
<tr>
<td>Schildcrout et al. (2006)</td>
<td>Albuquerque, NM; Baltimore, MD; Boston, MA; Denver, CO; San Diego, CA; Seattle, WA; St. Louis, MO; Toronto, ON, Canada 990 children with asthma, ages 5-12 yr</td>
<td>1-h max</td>
<td>0</td>
<td>Rescue inhaler</td>
<td></td>
<td>1.01 (0.89, 1.15)</td>
</tr>
<tr>
<td>Ostro et al. (2001)</td>
<td>Los Angeles, CA 138 children with asthma, ages 6-13 yr</td>
<td>1-h max</td>
<td>1</td>
<td>Any extra medication</td>
<td>Moderate/severe asthma</td>
<td>1.15 (1.12, 1.19)</td>
</tr>
<tr>
<td></td>
<td>Los Angeles</td>
<td></td>
<td></td>
<td></td>
<td>Los Angeles</td>
<td>1.10 (1.03, 1.19)</td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>CT River Valley, CT 166 children with asthma, ages 7-13 yr</td>
<td>1-h max</td>
<td>0</td>
<td>Beta-agonist</td>
<td></td>
<td>1.17 (0.96, 1.44)</td>
</tr>
<tr>
<td>Romieu et al. (2006)</td>
<td>Mexico City, Mexico 151 children with asthma, mean age 9 yr</td>
<td>1-h max 0-5 avg</td>
<td>0</td>
<td>Bronchodilator</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Mexico City, Mexico 71 children with asthma, ages 5-7 yr</td>
<td>1-h max</td>
<td>0</td>
<td>Bronchodilator</td>
<td></td>
<td>0.97 (0.93, 1.01)</td>
</tr>
<tr>
<td>Romieu et al. (1996)</td>
<td>Northern Mexico City, Mexico 65 children with asthma, ages 5-13 yr</td>
<td>1-h max</td>
<td>0</td>
<td>Bronchodilator</td>
<td></td>
<td>1.02 (1.00, 1.05)</td>
</tr>
<tr>
<td>Just et al. (2002)</td>
<td>Paris, France 82 children with asthma, mean (SD) age 10.9 (2.5) yr</td>
<td>24-h avg</td>
<td>0</td>
<td>Beta-agonist, no steroid</td>
<td></td>
<td>3.95 (1.22, 12.9)</td>
</tr>
<tr>
<td>Gent et al. (2003)</td>
<td>CT, southern MA 130 children with asthma on maintenance medication</td>
<td>1-h max</td>
<td>0</td>
<td>Bronchodilator</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

*Includes studies in Figure 6-13, plus others.
CS = Corticosteroid

Effect estimates are standardized to a 40-ppb increase for 1-h max O₃, a 30-ppb increase for 8-h max or 15-h avg O₃, and a 20-ppb increase for 24-h avg O₃.

Results not included in Figure 6-13. Results from Just et al. (2002) were out of range of other estimates, and results from Gent et al. (2003) were presented per quintile of ambient O₃ concentration.
Changes in Activity

While investigation has been limited, evidence does not consistently demonstrate O₃-associated diminished activity in children with asthma (O'Connor et al., 2008; Delfino et al., 2003). These studies examined different O₃ averaging times and lags. In the multicity ICAS cohort, O'Connor et al. (2008) found that a 20-ppb increase in lag 1-19 avg (of 24-h avg O₃) was associated with a 10% lower odds (95% CI: -26, 10) of slow play. In a small (n = 22) panel study conducted in children with asthma in Los Angeles CA, Delfino et al. (2003) found that a 40-ppb increase in lag 0 of 1-h max O₃ was associated with an increase in symptoms that interfered with daily activity with an OR of 7.14 (95% CI: 1.18, 43.2). Several studies reported increases in school absenteeism in children with asthma in association with increases in ambient O₃ concentration with long lag periods (14-day and 30-day distributed lags, 19-day avg) (O'Connor et al., 2008; Gilliland et al., 2001; Chen et al., 2000). Whereas Chen et al. (2000) and O'Connor et al. (2008) examined absences for any reason, Gilliland et al. (2001) found associations with absences for respiratory illnesses. Despite this evidence, several limitations are notable, including the lack of a well-characterized mode of action for respiratory effects occurring with longer lag periods of O₃ exposure and the potential for residual seasonal confounding with examination of long lag periods. In analyses of single-day lags, Gilliland et al. (2001) found associations with O₃ lagged 1 to 5 days, indicating respiratory absences may be affected by O₃ exposures with shorter lag periods.

6.2.4.2 Adults with Respiratory Disease

Within a small body of studies, several found that increases in ambient O₃ concentration (8-hour or 1-h max) were associated with increases in respiratory symptoms in adults with asthma (Khatri et al., 2009; Feo Brito et al., 2007; Ross et al., 2002). Details from studies of respiratory symptoms in adults with respiratory disease regarding location, time period, and ambient O₃ concentrations are presented in Table 6-22. These studies used different exposure assessment methods: concentrations averaged from sites closest to subjects' location each hour (Khatri et al., 2009) or concentrations measured at one (Ross et al., 2002) or multiple (Feo Brito et al., 2007) city sites. Park et al. (2005a) found inconsistent associations for 24-h avg O₃ measured at 10 city sites among the various symptoms and medication use examined in adults with asthma in Korea during a period of dust storms. In a study of adults with COPD in London, England, increases in lag 1 of 8-h max O₃ (at a single city site) were associated with higher odds of dyspnea and sputum changes but lower odds of nasal discharge, wheeze, or upper respiratory symptoms (Peacock et al., 2011).
Table 6-22  Mean and upper percentile O₃ concentrations in epidemiologic studies of respiratory symptoms and medication use in adults with respiratory disease.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O₃ Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Khatri et al. (2009)</td>
<td>Atlanta, GA</td>
<td>May-Sept 2003, 2005, 2006</td>
<td>8-h max</td>
<td>61 (median)a</td>
<td>75th: 74a</td>
</tr>
<tr>
<td>Feo Brito et al. (2007)</td>
<td>Ciudad Real and Puertollano, Spain</td>
<td>May-June 2000-2001</td>
<td>1-h max</td>
<td>65.9 (Ciudad Real)b</td>
<td>Max: 101.5b (Ciudad Real); 138.2b (Puertollano)</td>
</tr>
<tr>
<td>Eiswerth et al. (2005)</td>
<td>Glendora, CA</td>
<td>Oct-Nov 1983</td>
<td>1-h max</td>
<td>NR</td>
<td>NR</td>
</tr>
<tr>
<td>Ross et al. (2002)</td>
<td>East Moline, IL</td>
<td>Apr-Oct 1994</td>
<td>8-h avg</td>
<td>41.5</td>
<td>Max: 78.3</td>
</tr>
<tr>
<td>Park et al. (2005a)</td>
<td>Incheon, Korea</td>
<td>Mar-June 2002</td>
<td>24-h avg</td>
<td>Dust event days: 23.6</td>
<td>NR</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.
NR = Not Reported
aIndividual-level estimates were derived based on time spent in the vicinity of various O₃ monitors.
bConcentrations converted from µg/m³ to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).

Some studies that included adults with asthma examined populations with a high prevalence of atopy. In a study of children and adults with asthma (at least 53% with atopy), Ross et al. (2002) found that an increase in lag 1-3 avg of 8-h max O₃ was associated with an increase in symptom score and asthma medication use. Feo Brito et al. (2007) followed 137 adults with asthma in two central Spain cities. All subjects had pollen allergy and were examined during pollen season. In Puertollano, O₃ concentrations were obtained from four city monitors, and a 40-ppb increase in lag 3 of 1-h max O₃ was associated with a 14.3% increase (95% CI: 3.6, 26.0) in the number of subjects reporting respiratory symptoms, adjusting only for time trend. The association was imprecise in Ciudad Real (2.3% increase [95% CI: -14, 21%] per 40-ppb increase in lag 4 of 1-h max O₃), a city characterized by lower ambient air pollution levels and a narrower range of ambient O₃ concentrations as measured at a single site established by investigators.

Cross-sectional studies reported ambient O₃-associated decreases in activity in adults with asthma; however, due to various limitations in the collective body of evidence, firm conclusions are not warranted. Although conducted over single seasons, the studies did not consider potential confounding by meteorological factors. In a warm season study in Atlanta, GA (described in Section 6.2.1.2), Khatri et al. (2009) found
that a 30-ppb increase in lag 2 of 8-h max O₃ was associated with a 0.69-point decrease (95% CI: -1.3, -0.11) in the Juniper quality of life score, which incorporates indices for symptoms, mood, and activity limitations (7-point scale). In a fall study conducted in the Los Angeles, CA area in individuals with asthma (age 16 years and older), Eiswerth et al. (2005) found that a 40-ppb increase in 1-h max O₃ was associated with a 2.4% (95% CI: 0.83, 4) lower probability of indoor activity but higher probability of outdoor activity. The authors acknowledged that their findings were unexpected and may have been influenced by lack of control for potential confounders, but they interpreted the decrease in indoor activities as rest replacing chores. In contrast with the aforementioned studies, a panel study of individuals with asthma (ages 13-78 years) in Thailand found that a 20-ppb increase in lag 4 of 24-h avg O₃ was associated with a 26% (95% CI: 4, 43) lower odds of symptoms that interfered with activities (Wiwatanadate and Liwsrisakun, 2011).

6.2.4.3 Populations not Restricted to Individuals with Asthma

Locations, time periods, and ambient O₃ concentrations for studies of respiratory symptoms in populations not restricted to individuals with asthma are presented in Table 6-23. Most studies examined children, and in contrast with lung function results (Section 6.2.1.2), short-term increases in ambient O₃ concentration were not consistently associated with increases in respiratory symptoms in children in the general population (Figure 6-14 [and Table 6-24]). Because examination was limited, conclusions about the effects of ambient O₃ exposure on respiratory symptoms in adults are not warranted.
Table 6-23  Mean and upper percentile O\textsubscript{3} concentrations in epidemiologic studies of respiratory symptoms in populations not restricted to individuals with asthma.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Study Period</th>
<th>O\textsubscript{3} Averaging Time</th>
<th>Mean/Median Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Neas et al.</td>
<td>Uniontown, PA</td>
<td>June-August 1990</td>
<td>12-h avg (8 a.m.-8 p.m.)</td>
<td>50</td>
<td>NR</td>
</tr>
<tr>
<td>(1995)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Linn et al.</td>
<td>Rubidoux, Upland, Torrance, CA</td>
<td>September-June 1992-1994</td>
<td>24-h avg personal, 24-h avg ambient</td>
<td>5</td>
<td>Max: 16, Max: 53</td>
</tr>
<tr>
<td>(1996)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Hoek and Brunekreef</td>
<td>Deurne and Enkhuizen, Netherlands</td>
<td>March-July 1989</td>
<td>1-h max</td>
<td>Deurne: 57, Enkhuizen: 59</td>
<td>Max: 107, Max: 114</td>
</tr>
<tr>
<td>(1995)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rodriguez et al.</td>
<td>Perth, Australia</td>
<td>All-year, 1996-2003</td>
<td>24-h avg, 1-h max</td>
<td>28</td>
<td>Max: 74, Max: 95</td>
</tr>
<tr>
<td>(2007)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Moon et al.</td>
<td>4 cities, South Korea</td>
<td>April-May 2003</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>NR</td>
<td>NR</td>
</tr>
<tr>
<td>(2009)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(2002)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Triche et al.</td>
<td>Southwestern VA</td>
<td>June-August 1995-1996</td>
<td>24-h avg, 8-h max, 1-h max</td>
<td>35.2, 54.5, 60.8</td>
<td>75th: 40.6, 56.6, 75th: 64.1, 87.6, 75th: 70.0, 95.0</td>
</tr>
<tr>
<td>(2006)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Gold et al.</td>
<td>Mexico City, Mexico</td>
<td>January-November 1991</td>
<td>24-h avg</td>
<td>52.0\textsuperscript{a}</td>
<td>Max: 103\textsuperscript{a}</td>
</tr>
<tr>
<td>(1998)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Apte et al.</td>
<td>Multiple U.S. cities (NR)</td>
<td>Winter or summer 1994-1998</td>
<td>24-h avg</td>
<td>Workday avg (8 a.m.-5 p.m.)</td>
<td>34.2\textsuperscript{b}, Max: 86.2\textsuperscript{b}</td>
</tr>
<tr>
<td>(2008)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.
NR = Not Reported.
\textsuperscript{a}Measured at subject’s schools.
\textsuperscript{b}Concentrations converted from µg/m\textsuperscript{3} to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).

Children

Evidence of O\textsubscript{3}-associated increases in respiratory symptoms in children was inconsistent, which did not appear to be attributable to the differences in exposure assessment method among studies [e.g., O\textsubscript{3} measured at a single site (Linn et al., 1996; Hoek and Brunekreef, 1995), O\textsubscript{3} averaged across multiple city sites (Rodriguez et al., 2007), O\textsubscript{3} measured at sites near schools (Moon et al., 2009; Ward et al., 2002)]. Some studies that found weak or inconsistent associations between ambient O\textsubscript{3} concentrations and respiratory symptoms in children found O\textsubscript{3}-associated decrements in lung function (Ward et al., 2002; Linn et al., 1996). In their study of healthy children in Uniontown, Pennsylvania, Neas et al. (1995) found differences in association with respiratory symptoms between two estimates of O\textsubscript{3} exposure using ambient O\textsubscript{3} measurements from one central site in town. Subjects...
spent a mean 5.4 hours outdoors during the 12-hour period (8 a.m.-8 p.m.) over which O₃ concentrations were averaged and symptoms were reported. Evening cough was more strongly associated with O₃ concentrations weighted by time spent outdoors (OR: 2.20 [95% CI: 1.02, 4.75] per 30-ppb increase in lag 0 of 12-h avg O₃) than with unweighted O₃ concentrations (OR: 1.36 [95% CI: 0.86, 2.13]). Time spent outdoors has been shown to influence O₃ personal-ambient ratios and correlations (Section 4.3.3); thus, the weighted O₃ concentrations may have represented personal O₃ exposures better.

<table>
<thead>
<tr>
<th>Study</th>
<th>Symptom</th>
<th>Lag</th>
<th>Subgroup</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ward et al. (2002)</td>
<td>Wheeze, summer</td>
<td>0</td>
<td>All subjects</td>
</tr>
<tr>
<td>Hoek and Brunekreef</td>
<td>Cough</td>
<td>0</td>
<td>Jeju Island</td>
</tr>
<tr>
<td>(1995)</td>
<td>Any symptom</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Moon et al. (2009)</td>
<td>URS</td>
<td>0</td>
<td>All subjects</td>
</tr>
<tr>
<td>Neas et al. (1995)</td>
<td>Evening cough</td>
<td>0</td>
<td>Jeju Island</td>
</tr>
<tr>
<td>Triche et al. (2006)</td>
<td>Wheeze</td>
<td>0, 24-h avg</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>0, 8-h max</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>0, 1-h max</td>
<td></td>
</tr>
<tr>
<td>Gold et al. (1999)</td>
<td>Phlegm</td>
<td>0</td>
<td></td>
</tr>
</tbody>
</table>

Note: Results generally are presented in increasing order of mean ambient O₃ concentration. URS = Upper respiratory symptoms. Odds ratios are from single-pollutant models and are standardized to a 40-, 30-, and 20-ppb increase for 1-h max, 8-h max (or 8-h avg or 12-h avg), and 24-h avg O₃ concentrations, respectively.

Figure 6-14  Associations between ambient O₃ concentrations and respiratory symptoms in children in the general population.
### Table 6-24 Associations between ambient O₃ concentrations and respiratory symptoms in children in the general population for studies represented in Figure 6-14.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location/ Population</th>
<th>O₃ Lag</th>
<th>O₃ Averaging Time</th>
<th>Symptom</th>
<th>Subgroup</th>
<th>Standardized OR (95% CI)*a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ward et al. (2002)</td>
<td>Birmingham and Sandwell, England, 162 children, age 9 yr</td>
<td>0</td>
<td>24-h avg</td>
<td>Wheeze, summer</td>
<td></td>
<td>0.69 (0.51, 0.94)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Cough, summer</td>
<td></td>
<td>0.98 (0.80, 1.21)</td>
</tr>
<tr>
<td>Hoek and Brunekreef (1995)</td>
<td>Deurne and Enkhuizen, Netherlands, 300 children, ages 7-11 yr</td>
<td>0</td>
<td>1-h max</td>
<td>Cough</td>
<td></td>
<td>0.86 (0.61, 1.22)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Any symptom</td>
<td></td>
<td>0.94 (0.76, 1.16)</td>
</tr>
<tr>
<td>Moon et al. (2009)</td>
<td>4 cities, South Korea, 696 children, ages &lt;13 yr</td>
<td>0</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>URS All subjects</td>
<td>Jeju Island</td>
<td>0.96 (0.90, 1.03)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1.11 (0.95, 1.30)</td>
</tr>
<tr>
<td>Neas et al. (1995)</td>
<td>Uniontown, PA, 83 healthy children, 4th and 5th grades</td>
<td>0</td>
<td>12-h avg (8 a.m.-8 p.m.)</td>
<td>Evening symptom</td>
<td></td>
<td>2.20 (1.02, 4.75)b</td>
</tr>
<tr>
<td>Triche et al. (2006)</td>
<td>Southwestern VA, 61 infants of mothers with asthma, age &lt;1 yr</td>
<td>0</td>
<td>24-h avg</td>
<td>Wheeze</td>
<td></td>
<td>2.34 (1.02, 5.37)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>8-h max</td>
<td></td>
<td></td>
<td>1.48 (0.49, 4.41)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-h max</td>
<td></td>
<td></td>
<td>1.73 (0.48, 6.22)</td>
</tr>
<tr>
<td>Gold et al. (1999)</td>
<td>Mexico City, Mexico, 40 children, ages 8-11 yr</td>
<td>1</td>
<td>24-h avg</td>
<td>Phlegm</td>
<td></td>
<td>1.02 (1.00, 1.04)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1.02 (1.00, 1.04)</td>
</tr>
<tr>
<td>Linn et al. (1996)c</td>
<td>Rubidoux, Upland, Torrance, CA, 269 children, 4th and 5th grades</td>
<td>0</td>
<td>24-h avg</td>
<td>Evening symptom score</td>
<td></td>
<td>-0.96 (-2.2, 0.26)</td>
</tr>
</tbody>
</table>

*Includes studies in Figure 6-14, plus others.

URS = Upper respiratory symptoms

*a Effect estimates are standardized to a 40-, 30-, and 20-ppb increase for 1-h max, 8-h max (or 8-h avg or 12-h avg), and 24-h avg O₃, respectively.

b O₃ concentrations were weighted by the proportion of time spent outdoors.

c Results not presented in Figure 6-14 because outcome is a continuous variable indicating intensity of symptoms (negative indicates improvement in symptoms).

Several other panel studies of children, in which asthma prevalence ranged from 0 to 50%, reported null or negative associations between various averaging times and lags of ambient O₃ concentration and respiratory symptoms (Moon et al., 2009; Rodriguez et al., 2007; Ward et al., 2002; Linn et al., 1996; Hoek and Brunekreef, 1995). Among children in Mexico City, Gold et al. (1999) reported an increase in phlegm in association with an increase in lag 1 of 24-h avg O₃ concentration measured at schools; however, investigators acknowledged being unable to distinguish between the effects of O₃ and PM₁₀ due to their high correlation (r = 0.75).

Unlike other studies that examined ambient O₃ concentrations from a single monitoring site, Triche et al. (2006) found respiratory symptoms to be associated with O₃ measured at a site that for some subjects was located >100 miles away from home (Figure 6-14 [and Table 6-24]). Subjects included infants in Southwestern VA. Odds ratios were 46-73% larger in the group who had mothers with asthma than
among all infants (Triche et al., 2006). Larger ORs were found for 24-h avg than 1-h or 8-h max O₃ concentrations, particularly for wheeze but less so for difficulty breathing. While these results suggested that children with mothers with asthma may be at increased risk of O₃-related respiratory morbidity, the authors acknowledged that mothers with asthma may be more likely to report symptoms in their children. Additionally, transient wheeze, which is common in infants, may not predict respiratory morbidity later in life. In another cohort of children with parental history of asthma that was followed to an older age (5 years), increases in ambient O₃ concentration (increment of effect estimate not reported) were not associated with increases in respiratory symptoms (Rodriguez et al., 2007).

**Adults**

A cross-sectional study of 4,200 adult workers from 100 office buildings across the U.S. found that multiple ambient O₃ metrics, including the 24-hour average, the workday average (8 a.m.-5 p.m.), and the late workday (3-6 p.m.) average, were associated with similar magnitudes of increase in building-related symptoms (Apte et al., 2008). Office workers likely have a low personal-ambient O₃ correlation and ratio, thus the implications of these findings compared to those of the other respiratory symptom studies are limited.

### 6.2.4.4 Confounding in Epidemiologic Studies of Respiratory Symptoms and Medication Use

Epidemiologic evidence does not indicate that confounding by meteorological factors or copollutant exposures fully accounts for associations observed between short-term increases in ambient O₃ concentration and respiratory symptoms and medication use. Except where specified in the text, studies found O₃-associated increases in respiratory symptoms or medication use in statistical models that adjusted for temperature. Thurston et al. (1997) found no independent association between temperature and respiratory symptoms among children with asthma at summer camps. A few studies additionally included humidity in models (Triche et al., 2006; Ross et al., 2002).

Several studies that examined populations with a high prevalence of atopy found O₃-associated increases in respiratory symptoms and asthma medication use with adjustment for daily pollen counts (Just et al., 2002; Ross et al., 2002; Gielen et al., 1997). Gielen et al. (1997) and Ross et al. (2002) examined populations with a high prevalence of grass pollen allergy (52% and 38%, respectively). In a study conducted over multiple seasons, Ross et al. (2002) found a similar magnitude of association between O₃ and morning symptoms and medication use with adjustment for pollen counts. Feo Brito et al. (2007) followed adults in central Spain specifically with asthma and pollen allergy. In one city, O₃ was associated with an increase in the number of subjects reporting symptoms. A smaller increase was estimated for pollen. Conversely, in another city, pollen was associated with an increased reporting of
respiratory symptoms, whereas O₃ was not. The results suggested that O₃ and pollen may have independent effects that vary by location, depending on the mix of ambient pollutants.

Results from copollutant models did not indicate strong confounding by copollutants such as PM₂.₅, PM₁₀, sulfate, SO₂, or NO₂ (Table 6-25). Notably, studies examined different averaging times for O₃ (1-h max or 8-h avg) and copollutants (3-h avg to 24-h avg) and reported a range of correlations between O₃ and copollutants, which may complicate interpretation of copollutant model results. Information on potential copollutant confounding of asthma medication use results was limited. The association between O₃ and bronchodilator use did not change with adjustment for PM₂.₅ in Gent et al. (2003) but decreased in magnitude with adjustment for 12-h avg sulfate in Thurston et al. (1997). In Thurston et al. (1997) and Gent et al. (2003), 1-h max O₃ was highly correlated with 12-h avg sulfate (r = 0.74) and 24-h avg PM₂.₅ (r = 0.77), respectively, making it difficult to distinguish the independent effects of O₃. Studies conducted concurrently in two areas of Mexico City, Mexico, examined 1-h max O₃ and 24-h avg PM₁₀ or PM₂.₅ and found robust ORs for respiratory symptoms for both O₃ and PM (Romieu et al., 1997; Romieu et al., 1996). Romieu et al. (1997) reported a moderate correlation between 1-h max O₃ and 24-h avg PM₁₀ (r = 0.47). Associations between O₃ and respiratory symptoms were observed in NCICAS in copollutant models with SO₂, NO₂, or PM₁₀, which were examined with different averaging times and lags than was O₃ (Mortimer et al., 2002) (Table 6-25). Also difficult are interpretations of the O₃-associated increases in respiratory symptoms found with adjustment for two copollutants in the same model (i.e., PM₂.₅ plus NO₂ or PM₁₀₂.₅) (Escamilla-Nuñez et al., 2008; Triche et al., 2006).
Table 6-25  Associations between ambient O₃ concentrations and respiratory symptoms in single- and copollutant models.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location/Population</th>
<th>O₃ Metrics</th>
<th>Symptom</th>
<th>OR for O₃ in Single-Pollutant Model (95% CI)a</th>
<th>OR for O₃ in Copollutant Model (95% CI)a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mortimer et al. (2002)</td>
<td>Bronx, East Harlem, NY; Baltimore, MD; Washington, DC; Detroit, MI; Cleveland, OH; Chicago, IL; St. Louis, MO</td>
<td>8 cities with SO₂ data</td>
<td>8-h avg (10 a.m.-6 p.m.)</td>
<td>Morning symptoms</td>
<td>With lag 1-2 avg, 3-h avg SO₂</td>
</tr>
<tr>
<td></td>
<td>846 children with asthma, ages 4-9 yr</td>
<td>Lag 1-4 avg</td>
<td></td>
<td>1.35 (1.04, 1.74)</td>
<td>1.23 (0.94, 1.61)</td>
</tr>
<tr>
<td></td>
<td>7 cities with NO₂ data</td>
<td>1.25 (0.94, 1.67)</td>
<td>With lag 1-6 avg, 24-h avg NO₂</td>
<td>1.14 (0.85, 1.55)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>3 cities with PM₁₀ data</td>
<td>1.21 (0.61, 2.41)</td>
<td>With lag 1-2 avg, 24-h avg PM₁₀</td>
<td>1.08 (0.49, 2.39)</td>
<td></td>
</tr>
<tr>
<td>Gent et al. (2003)b</td>
<td>CT, Southern MA</td>
<td>1-h max, Lag 0</td>
<td>&lt;43.2 ppb</td>
<td>Wheeze</td>
<td>with lag 0, 24-h avg PM₂.₅</td>
</tr>
<tr>
<td></td>
<td>130 children with asthma on maintenance medication</td>
<td>43.2-51.5 ppb</td>
<td></td>
<td>1.00 (reference)</td>
<td>1.00 (reference)</td>
</tr>
<tr>
<td></td>
<td>51.6-58.8 ppb</td>
<td>1.04 (0.89, 1.21)</td>
<td>With lag 0, 12-h avg sulfate</td>
<td>1.05 (0.90, 1.23)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>58.9-72.6 ppb</td>
<td>1.16 (1.00, 1.35)</td>
<td>With lag 0, 12-h avg sulfate</td>
<td>1.18 (1.00, 1.38)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>≥ 72.7 ppb</td>
<td>1.16 (1.00, 1.35)</td>
<td>With lag 0, 12-h avg sulfate</td>
<td>1.25 (1.05, 1.50)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>1.22 (0.97, 1.53)</td>
<td></td>
<td>1.47 (1.13, 1.90)</td>
<td></td>
</tr>
<tr>
<td>Thurston et al. (1997)</td>
<td>CT River Valley</td>
<td>1-h max, Lag 0</td>
<td></td>
<td>Chest symptoms</td>
<td>With lag 0, 12-h avg sulfate</td>
</tr>
<tr>
<td></td>
<td>166 children with asthma, ages 7-13 yr</td>
<td>1.21 (1.12, 1.31)</td>
<td></td>
<td>1.19 (1.06, 1.35)</td>
<td>1.07 (0.92, 1.24)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1.20 (1.09, 1.32)</td>
<td></td>
<td>With lag 0, 12-h avg sulfate</td>
<td>1.06 (1.02, 1.10)</td>
</tr>
<tr>
<td>Romieu et al. (1996)</td>
<td>Northern Mexico City, Mexico</td>
<td>1-h max, Lag 0</td>
<td></td>
<td>Lower respiratory symptoms</td>
<td>With lag 0, 24-h avg PM₂.₅</td>
</tr>
<tr>
<td></td>
<td>71 children with asthma, ages 5-7 yr</td>
<td>1.07 (1.02, 1.12)</td>
<td></td>
<td>1.06 (1.02, 1.10)</td>
<td></td>
</tr>
<tr>
<td>Romieu et al. (1997)</td>
<td>Southern Mexico City, Mexico</td>
<td>1-h max, Lag 0</td>
<td></td>
<td>Lower respiratory symptoms</td>
<td>With lag 0, 24-h avg PM₁₀</td>
</tr>
<tr>
<td></td>
<td>65 children with asthma, ages 5-13 yr</td>
<td>1.09 (1.04, 1.14)</td>
<td></td>
<td>1.09 (1.01, 1.19)</td>
<td></td>
</tr>
</tbody>
</table>

Results generally are presented in order of increasing mean ambient O₃ concentration.
aORs are standardized to a 40- and 30-ppb increase for 1-h max and 8-h avg O₃, respectively.
bTemperature not included in models.

6.2.4.5 Summary of Epidemiologic Studies of Respiratory Symptoms and Asthma Medication Use

Comprising a majority of available evidence, single-city and -region epidemiologic studies provide consistent evidence for the effects of short-term increases in ambient O₃ exposure on increasing respiratory symptoms and asthma medication use in children with asthma (Figure 6-12 and Table 6-20 and Figure 6-13 and Table 6-21). Evidence from the few available U.S. multiplicity studies is less consistent (O'Connor et al., 2008; Schildcrout et al., 2006; Mortimer et al., 2002). However, methodological differences among studies make comparisons across the
multicity studies difficult. Because of fewer person-days of data (Schildcrout et al., 2006) or examination of 19-day averages of ambient O₃ concentrations (O'Connor et al., 2008), results from recent multicity studies were not given greater consideration than results from single-city studies in weighing the evidence for ambient O₃ exposure and respiratory symptoms in children with asthma. Findings from a small body of studies indicate O₃-associated increases in respiratory symptoms in adults with asthma. Associations between short-term increases in ambient O₃ concentration and reduced activity in children or adults with asthma are not clearly demonstrated. While O₃-associated increases in school absenteeism were found in children with asthma, evidence for respiratory-related absences and for O₃ exposure lags shorter than 14 days is sparse. The implications of results for multi-week averages of ambient O₃ concentrations are limited because of the lack of a well-characterized mode of action for such lags of O₃ exposure and the greater potential for residual seasonal confounding with examination of long lag periods. Short-term increases in ambient O₃ concentration were not consistently associated with increases in respiratory symptoms in groups comprising children with and without asthma.

Increases in respiratory symptoms and asthma medication use were associated with increases in ambient O₃ concentration assigned to subjects using various methods. Associations were found with methods likely to represent ambient exposures better, including O₃ measured on site and at the time of children’s outdoor activity (Thurston et al., 1997) and concentrations weighted by time spent outdoors (Neas et al., 1995). However, associations also were found with methods that varied in their representation of ambient exposures and spatial variability in ambient concentrations, i.e., concentrations averaged among subjects’ locations each hour (Khatri et al., 2009), measured within 5 km of schools or homes (Escamilla-Nuñez et al., 2008; Romieu et al., 2006; Romieu et al., 1997; Romieu et al., 1996), averaged across multiple sites (Feo Brito et al., 2007; Gent et al., 2003; Mortimer et al., 2002), and measured at a single site (Ross et al., 2002; Gielen et al., 1997).

Associations with respiratory symptoms were demonstrated most frequently for 1-h max and 8-h max or avg O₃, and within-study comparisons indicated similar ORs for 1-h max and 8-h max O₃ (Delfino et al., 2003; Gent et al., 2003). Respiratory symptoms also were associated with 12-h avg and 24-h avg O₃ (Jalaludin et al., 2004; Gold et al., 1999; Neas et al., 1995). Epidemiologic studies examined respiratory symptoms associated with O₃ concentrations lagged 0 to 5 days and those averaged over 2 to 19 days. While O₃ at lags 0 or 1 were consistently associated with respiratory symptoms, several studies found larger ORs for multiday averages (3- to 6-day) of O₃ (Escamilla-Nuñez et al., 2008; Romieu et al., 2006; Just et al., 2002; Mortimer et al., 2002; Ross et al., 2002). Epidemiologic findings for lagged or multiday average O₃ are supported by evidence that O₃ sensitizes bronchial smooth muscle to hyperreactivity and thus acts as a primer for subsequent exposure to antigens such as allergens (Section 5.3.5). Many studies examined populations with asthma with a high prevalence of atopy (52-100%). In these populations, sensitization of airways provides a biologically plausible mode of action by which increases in respiratory symptoms result from increases in O₃ exposure after a lag or accumulated over several days. Further support is provided by findings in controlled
human exposure studies that airway hyperresponsiveness (Section 6.2.2.1) and some indicators of inflammation (Section 6.2.3.1) remained elevated following repeated O₃ exposures and by observations from epidemiologic studies that increases in pulmonary inflammation were associated with multiday average O₃ concentrations (Section 6.2.3.2).

Epidemiologic study results did not indicate that O₃-associated increases in respiratory symptoms are confounded by temperature, pollen, or copollutants. In limited analysis, ambient O₃ was associated with respiratory symptoms with adjustment for copollutants, primarily PM. However, identifying the independent effects of O₃ in some studies was complicated due to the high correlations observed between O₃ and PM or different lags and averaging times examined for copollutants. Nonetheless, the robustness of associations in some studies of individuals with asthma with and without adjustment for ambient copollutant concentrations combined with findings from controlled human exposure studies for the direct effect of O₃ exposure provide substantial evidence for the independent effects of short-term ambient O₃ exposure on increasing respiratory symptoms.

6.2.5 Lung Host Defenses

The mammalian respiratory tract has a number of closely integrated defense mechanisms that, when functioning normally, provide protection from the potential health effects attributed to exposure to a wide variety of inhaled particles and microbes. For simplicity, these interrelated defenses can be divided into two major parts: (1) nonspecific (transport, phagocytosis, and bactericidal activity) and (2) specific (immunologic) defense mechanisms. A variety of sensitive and reliable methods have been used to assess the effects of O₃ on these components of the lung’s defense system to provide a better understanding of the health effects associated with the inhalation of this pollutant. The 2006 O₃ AQCD stated that animal toxicological studies provide extensive evidence that acute O₃ exposures as low as 0.08 to 0.5 ppm can cause increases in susceptibility to infectious diseases due to modulation of lung host defenses. Table 6-6 through Table 6-9 (U.S. EPA, 1996g, h, i, j) beginning on page 6-41 of the 1996 O₃ AQCD (U.S. EPA, 1996a), and Annex Table AX5-7 (U.S. EPA, 2006d), beginning on page AX5-8 of the 2006 O₃ AQCD (U.S. EPA, 2006b), present studies on the effects of O₃ on host defense mechanisms. This section discusses the various components of host defenses, such as the mucociliary escalator, the phagocytic, bactericidal, and regulatory role of the alveolar macrophages (AMs), the adaptive immune system, and integrated mechanisms that are studied by investigating the host’s response to experimental pulmonary infections.

6.2.5.1 Mucociliary Clearance

The mucociliary system is one of the lung’s primary defense mechanisms. It protects the conducting airways by trapping and quickly removing material that has been
deposited or is being cleared from the alveolar region by migrating alveolar macrophages. Ciliary movement directs particles trapped on the overlying mucous layer toward the pharynx, where the mucus is swallowed or expectorated.

The effectiveness of mucociliary clearance can be determined by measuring such biological activities as the rate of transport of deposited particles; the frequency of ciliary beating; structural integrity of the ciliated cells; and the size, number, and distribution of mucus-secreting cells. Once this defense mechanism has been altered, a buildup of both viable and nonviable inhaled substances can occur on the epithelium which may jeopardize the health of the host, depending on the nature of the uncleared substance. Impaired mucociliary clearance can result in an unwanted accumulation of cellular secretions, increased infections, chronic bronchitis, and complications associated with COPD. A number of previous studies with various animal species have examined the effect of O3 exposure on mucociliary clearance and reported morphological damage to the cells of the tracheobronchial tree from acute and sub-chronic exposure to O3 0.2 ppm and higher. The cilia were either completely absent or had become noticeably shorter or blunt. Once these animals were placed in a clean-air environment, the structurally damaged cilia regenerated and appeared normal (U.S. EPA, 1986). Based on such morphological observations, related effects such as ciliostasis, increased mucus secretions, and a slowing of mucociliary transport rates might be expected. However, no measurable changes in ciliary beating activity have been reported due to O3 exposure alone. Essentially no data are available on the effects of prolonged exposure to O3 on ciliary functional activity or on mucociliary transport rates measured in the intact animal. In general, functional studies of mucociliary transport have observed a delay in particle clearance soon after acute exposure. Decreased clearance is more evident at higher doses (1 ppm), and there is some evidence of attenuation of these effects (U.S. EPA, 1986). However, no recent studies have evaluated the effects of O3 on mucociliary clearance.

6.2.5.2 Alveolobronchiolar Transport Mechanism

In addition to the transport of particles deposited on the mucous surface layer of the conducting airways, particles deposited in the deep lung may be removed either up the respiratory tract or through interstitial pathways to the lymphatic system. The pivotal mechanism of alveolobronchiolar transport involves the movement of AMs with phagocytized particles to the bottom of the mucociliary escalator. Failure of the AMs to phagocytize and sequester the deposited particles from the vulnerable respiratory membrane can lead to particle entry into the interstitial spaces. Once lodged in the interstitium, particle removal is more difficult and, depending on the toxic or infectious nature of the particle, its interstitial location may allow the particle to set up a focus for pathologic processes. Although some studies show reduced early (tracheobronchial) clearance after O3 exposure, late (alveolar) clearance of deposited material is accelerated, presumably due to macrophage influx (which in itself can be damaging due to proteases and oxidative reactions in these cells).
6.2.5.3 Alveolar Macrophages

Within the gaseous exchange region of the lung, the first line of defense against microorganisms and nonviable particles that reach the alveolar surface is the AM. This resident phagocyte is responsible for a variety of activities, including the detoxification and removal of inhaled particles, maintenance of pulmonary sterility via destruction of microorganisms, and interaction with lymphocytes for immunologic protection. Under normal conditions, AMs seek out particles deposited on the alveolar surface and ingest them, thereby sequestering the particles from the vulnerable respiratory membrane. To adequately fulfill their defense function, the AMs must maintain active mobility, a high degree of phagocytic activity, and an optimally functioning biochemical and enzyme system for bactericidal activity and degradation of ingested material. As discussed in previous AQCDs, short periods of O₃ exposure can cause a reduction in the number of free AMs available for pulmonary defense, and these AMs are more fragile, less phagocytic, and have decreased lysosomal enzyme activities required for killing pathogens. For example, in results from earlier work in rabbits, a 2-hour exposure to 0.1 ppm O₃ inhibited phagocytosis and a 3-hour exposure to 0.25 ppm decreased lysosomal enzyme activities (Driscoll et al., 1987; Hurst et al., 1970). Similarly, AMs from rats exposed to 0.1 ppm O₃ for 1 or 3 weeks exhibited reduced hydrogen peroxide production (Cohen et al., 2002). A controlled human exposure study reported decrements in the ability of alveolar macrophages to phagocytize yeast following exposure of healthy volunteers to 80 to 100 ppb O₃ for 6.6-hour during moderate exercise (Devlin et al., 1991). Although the percentage of phagocytosis-capable macrophages was unchanged by O₃ exposure, the number of yeast engulfed was reduced when phagocytosis was complement-dependent. However, there was no difference in the ability of macrophages to produce superoxide anion after O₃ exposure. These results are consistent with those from another controlled human exposure study in which no changes in the level of lysosomal enzymes or superoxide anion production were observed in macrophages lavaged from healthy human subjects exposed to 400 ppb O₃ for 2 hours with heavy intermittent exercise (Koren et al., 1989). More recently, Lay et al. (2007) observed no difference in phagocytic activity or oxidative burst capacity in macrophages or monocytes from sputum or blood collected from healthy volunteers after a 2-hour exposure to 400 ppb O₃ with moderate intermittent exercise. However, another study (Alexis et al., 2009) found that oxidative burst and phagocytic activity in macrophages increased in GSTM1 null subjects compared to GSTM1 positive subjects, who had relatively unchanged macrophage function parameters after an O₃ exposure identical to that of Lay et al. (2007). Collectively, these studies demonstrate that O₃ can affect multiple steps or aspects required for proper macrophage function, but any C-R relationship appears complex and genotype may be a consideration. A few other recent studies have evaluated the effects of O₃ on macrophage function, but these are of questionable relevance due to the use of in vitro exposure systems and amphibian animal models (Mikerov et al., 2008c; Dohm et al., 2005; Klestadt et al., 2005).
6.2.5.4 Infection and Adaptive Immunity

General Effects on the Immune System

The effects of O₃ on the immune system are complex and dependent on the exposure regimen and the observation period. According to toxicological studies it appears that the T-cell-dependent functions of the immune system are more affected than B-cell-dependent functions (U.S. EPA, 2006b). Generally, there is an early immunosuppressive effect that subsides with continued O₃ exposure, resulting in either a return to normal responses or an enhancement of immune responses. However, this is not always the case as Aranyi et al. (1983) showed decreased T-cell mitogen reactions in mice after subchronic (90-day) exposure to 0.1 ppm O₃. Earlier studies report changes in cell populations in lymphatic tissues (U.S. EPA, 2006b). A more recent study in mice demonstrated that numbers of certain T-cell subsets in the spleen were reduced after exposure to 0.6 ppm O₃ (10h/day x 15d) (Feng et al., 2006).

The inflammatory effects of O₃ involve the innate immune system, and as such, O₃ can affect adaptive (or acquired) immunity via alterations in antigen presentation and costimulation by innate immune cells such as macrophages and dendritic cells.

Several recent controlled human exposure studies demonstrate increased expression of molecules involved in antigen presentation or costimulation. Lay et al. (2007) collected sputum monocytes from healthy volunteers exposed to 400 ppb O₃ for 2 hours with moderate intermittent exercise and detected increases in HLA-DR, used to present antigen to T-cells, and CD86, a costimulatory marker necessary for T-cell activation. Upregulation of HLA-DR was also observed by Alexis et al. (2009) in sputum dendritic cells and macrophages from GSTM1 null subjects exposed to 400 ppb O₃ for 2 hours with moderate intermittent exercise. On airway monocytes from healthy volunteers 24 hours after exposure to 80 ppb O₃ for 6.6 hours with moderate intermittent exercise, HLA-DR, CD86, and CD14 (a molecule involved in bacterial endotoxin reactivity) were increased, whereas CD80, a costimulatory molecule of more heterogeneous function, was decreased (Alexis et al., 2010). Patterns of expression on macrophages were similar, except that HLA-DR was found to be significantly decreased after O₃ exposure and CD86 was not significantly altered. An increase in IL-12p70, a macrophage and dendritic cell product that activates T-cells, was correlated with increased numbers of dendritic cells. It should be noted that these results are reported as comparisons to baseline as there was no clean air control (Alexis et al., 2010; Alexis et al., 2009). Another controlled human exposure study reported no increase in IL-12p70 in sputum from healthy subjects or those with atopy or atopy and asthma following a 2-hour exposure to 400 ppb O₃ with intermittent moderate exercise (Hernandez et al., 2010). Levels of HLA-DR, CD14 and CD86 were not increased on macrophages collected from any of these subjects. It is difficult to compare these results to those of Lay et al. (2007) and Alexis et al. (2010) due to differences in O₃ concentration, cell type examined, and timing of postexposure analysis.
Although no controlled human exposure studies have examined the effects of O₃ on the ability to mount antigen-specific responses, upregulation of markers associated with innate immune activation and antigen presentation could potentially enhance adaptive immunity and increase immunologic responses to antigens. While enhanced adaptive immunity may bolster defenses against infection, it also may enhance allergic responses (Section 6.2.6).

In animal models, O₃ has been found to alter responses to antigenic stimulation. For example, antibody responses to a T-cell-dependent antigen were suppressed after a 56-day exposure of mice to 0.8 ppm O₃, and a 14-day exposure to 0.5 ppm O₃ decreased the antiviral antibody response following influenza virus infection (Jakab and Hmieleski, 1988); the latter impairment may lead to lowered resistance to re-infection. The immune response is highly influenced by the temporal relationship between O₃ exposure and antigenic stimulation. When O₃ exposure preceded Listeria infection, there were no effects on delayed-type hypersensitivity or splenic lymphoproliferative responses; however, when O₃ exposure occurred during or after Listeria infection was initiated, these immune responses were suppressed (Van Loveren et al., 1988). In another study, a reduction in mitogen activated T-cell proliferation was observed after exposure to 0.6 ppm O₃ for 15 days that could be ameliorated by antioxidant supplementation. Antigen-specific proliferation decreased by 60%, indicating attenuation of the acquired immunity needed for subsequent memory responses (Feng et al., 2006). Ozone exposure also skewed the ex-vivo cytokine responses elicited by non-specific stimulation toward inflammation, decreasing IL-2 and increasing IFN-γ. Modest decreases in immune function assessed in the offspring of O₃-exposed dams (mice) were observed by Sharkhuu et al. (2011). The ability to mount delayed-type hypersensitivity responses was significantly suppressed in 42 day-old offspring when dams were exposed to 0.8 or 1.2 ppm O₃, but not 0.4 ppm, from gestational day 9-18. Humoral responses to immunization with sheep red blood cells were unaffected, as were other immune parameters such as splenic populations of CD45+ T-cells, iNKT-cells, and levels of IFN-γ, IL-4, and IL-17 in the BALF. Generally, continuous exposure to O₃ impairs immune responses for the first several days of exposure, followed by an adaptation to O₃ that allows a return of normal immune responses. Most species show little effect of O₃ exposures prior to immunization, but show a suppression of responses to antigen in O₃ exposures post-immunization.

Microbial Infection

Bacterial infection

A relatively large body of evidence shows that O₃ increases susceptibility to bacterial infections. The majority of studies in this area were conducted before the 1996 O₃ AQCD was published and many are included in Table 6-9 (U.S. EPA, 1996j) on page 6-53 of that document (U.S. EPA, 1996a). Known contributing factors are impaired mucociliary streaming, altered chemotaxis/motility, defective phagocytosis of bacteria, decreased production of lysosomal enzymes or superoxide radicals by
alveolar macrophages, and decreased IFN-γ levels. In animal models of bacterial infection, exposure to 0.08 ppm O₃ increases streptococcus-induced mortality, regardless of whether O₃ exposure precedes or follows infection (Miller et al., 1978; Coffin and Gardner, 1972; Coffin et al., 1967). Increases in mortality are due to the infectious agent, thereby reflecting functional impairment of host defenses. Exercise and copollutants can enhance the effects of O₃ in infectivity models. Although both mice and rats exhibit impaired bactericidal macrophage activity after O₃ exposure, mortality due to infection is only observed in mice. Additionally, although mice and humans share many host defense mechanisms, there is little compelling evidence from epidemiologic studies to suggest an association between O₃ exposure and decreased resistance to bacterial infection, and the etiology of respiratory infections is not easily identified via ICD codes (Section 6.2.7.3).

**Viral infection**

Only a few studies, described in previous AQCDs, have examined the effects of O₃ exposure on the outcome of viral respiratory infection [see Table 6-9 on page 6-53 of the 1996 O₃ AQCD (U.S. EPA, 1996)]. Some studies show increased mortality in animals, while others show diminished severity and increased survival time. There is little to no evidence from studies of animals or humans to suggest that O₃ increases the incidence of respiratory viral infection in humans. In human volunteers infected with rhinovirus prior to O₃ exposure (0.3 ppm for 5 consecutive days), no effect was observed on viral titers, IFN-γ production, or blood lymphocyte proliferative responses to viral antigen (Henderson et al., 1988). In vitro cell culture studies of human bronchial epithelial cells indicate O₃-induced exacerbation of human rhinovirus infection (Spannhake et al., 2002), but this is of limited relevance. More recent studies on the interactions of O₃ and viral infections have not been published. Natural killer (NK) cells, which destroy virally infected cells and tumors in the lung, appear to be inhibited by higher concentrations of O₃ and either unaffected or stimulated at lower concentrations. Several studies show decreases in NK cell activity following acute exposures ranging from 0.8 to 1 ppm (Gilmour and Jakab, 1991; Van Loveren et al., 1990; Burleson et al., 1989). However, Van Loveren et al. (1990) showed that a 1-week exposure to 0.2 or 0.4 ppm O₃ increased NK cell activity, and an urban pattern of exposure (base of 0.06 ppm with peaks of 0.25 ppm) had no effect on NK cell activity after 1, 3, 13, 52, or 78 weeks of exposure (Selgrade et al., 1990). A more recent study demonstrated a 35% reduction in NK cell activity after exposure of mice to 0.6 ppm O₃ (10h/day x 15d) (Feng et al., 2006). The defective IL-2 production demonstrated in this study may impair NK cell activation. Alternatively, NK cell surface charge may be altered by ROS, decreasing their adherence to target cells (Nakamura and Matsunaga, 1998).

### 6.2.5.5 Summary of Lung Host Defenses

Taken as a whole, the data clearly indicate that an acute O₃ exposure impairs the host defense capability of animals, primarily by depressing AM function and perhaps also
by decreasing mucociliary clearance of inhaled particles and microorganisms. Coupled with limited evidence from controlled human exposure studies, this suggests that humans exposed to O₃ could be predisposed to bacterial infections in the lower respiratory tract. The seriousness of such infections may depend on how quickly bacteria develop virulence factors and how rapidly PMNs are mobilized to compensate for the deficit in AM function. It remains unclear how O₃ might affect antigen presentation and the costimulation required for T-cell activation, given the mixed results from controlled human exposure studies, but there is toxicological evidence for suppression of T-cell-dependent functions by O₃, including reductions in antigen-specific proliferation and antibody production, indicating the potential for impaired acquired immunity and memory responses. To date, a limited number of epidemiologic studies have examined associations between O₃ exposure and hospital admissions or ED visits for respiratory infection, pneumonia, or influenza. Results have been mixed, and in some cases conflicting (see Section 6.2.7.2 and Section 6.2.7.3). With the exception of influenza, it is difficult to ascertain whether cases of respiratory infection or pneumonia are of viral or bacterial etiology. A study that examined the association between O₃ exposure and respiratory hospital admissions in response to an increase in influenza intensity did observe an increase in respiratory hospital admissions (Wong et al., 2009), but information from toxicological studies of O₃ and viral infections is ambiguous.

### 6.2.6 Allergic and Asthma-Related Responses

Effects resulting from combined exposures to O₃ and allergens have been studied in a variety of animal species, generally as models of experimental asthma. Pulmonary function and airways hyperresponsiveness in animal models of asthma are discussed in Section 6.2.1.3 and Section 6.2.2.2. Previous evidence indicates that O₃ exposure skews immune responses toward an allergic phenotype. For example, Gershwin et al. (1981) reported that O₃ (0.8 and 0.5 ppm for 4 days) exposure caused a 34-fold increase in the number of IgE (allergic antibody)-containing cells in the lungs of mice. In general, the number of IgE-containing cells correlated positively with levels of anaphylactic sensitivity. In humans, allergic rhinoconjunctivitis symptoms are associated with increases in ambient O₃ concentrations (Riediker et al., 2001). Recent controlled human exposure studies have observed O₃-induced changes indicating allergic skewing. Airway eosinophils, which participate in allergic disease and inflammation, were observed to increase in volunteers with atopy and mild asthma 18 hours following a 7.6-hour exposure to 160 ppb O₃ with light intermittent exercise (Peden et al., 1997). No increase in airway eosinophils was observed 4 hours after exposure of healthy subjects or those with atopic or atopy and asthma to 400 ppb O₃ for 2 hours with moderate intermittent exercise (Hernandez et al., 2010). However, subjects with atopy did exhibit increased IL-5, a cytokine involved in eosinophil recruitment and activation, suggesting that perhaps these two studies observed the same effect at different time points. Epidemiologic studies describe associations between eosinophils and short- (Section 6.2.3.2) or long-term (Section 7.2.5) O₃ exposure, as do chronic exposure studies in non-human primates.
Hernandez et al. (2010) also observed increased expression of high and low affinity IgE receptors on sputum macrophages from atopic asthmatics, which may enhance IgE-dependent inflammation. Sputum levels of IL-4 and IL-13, both pro-allergic cytokines that aid in the production of IgE, were unaltered in all groups. The lack of increase in IL-4 levels in sputum reported by Hernandez et al. (2010), along with increased IL-5, is consistent with results from Bosson et al. (2003), in which IL-5 (but not IL-4 levels) increased in bronchial epithelial biopsy specimens following exposure of subjects with atopy and mild asthma to 200 ppb O₃ for 2 hours with moderate intermittent exercise. IL-5 was not elevated in specimens obtained from healthy (no asthma) O₃-exposed subjects. Collectively, findings from these studies suggest that O₃ can induce or enhance certain components of allergic inflammation in individuals with atopy or atopic asthma.

Ozone enhances inflammatory and allergic responses to allergen challenge in sensitized animals. Short-term exposure (2 days) to 1 ppm O₃ exacerbated allergic rhinitis and lower airway allergic inflammation in Brown Norway rats, a rat strain that is comparatively less sensitive to O₃ than other rats or humans (Wagner et al., 2009; 2007). OVA-sensitized rats were intranasally challenged with OVA on days 1 and 2, and exposed to 0 or 1 ppm O₃ (8 hours/day) on days 4 and 5. Analysis at day 6 indicated that O₃ exposure enhanced intraepithelial mucosubstances in the nose and airways, induced cys-LTs, MCP-1, and IL-6 production in BALF, and upregulated expression of the pro-allergic cytokines IL-5 and IL-13. These changes were not evident in non-allergic controls. All of these responses were blunted by gamma-tocopherol (γT; vitamin E) therapy. γT neutralizes oxidized lipid radicals, and protects lipids and proteins from nitrosative damage from NO-derived metabolites. Farraj et al. (2010) exposed allergen-sensitized adult male BALB/c mice to 0.5 ppm O₃ for 5 hours once per week for 4 weeks. Ozone exposure and O₃/DEP (2.0 mg/m³) co-exposure of OVA-sensitized mice elicited significantly greater serum IgE levels than in DEP-exposed OVA-sensitized mice (98% and 89% increases, respectively). Ozone slightly enhanced levels of BAL IL-5, but despite increases in IgE, caused a significant decrease in BAL IL-4 levels. IL-10, IL-13, and IFN-γ levels were unaffected. Lung resistance and elastance were unaffected in allergen sensitized mice exposed solely to 0.5 ppm O₃ once a week for 4 weeks (Farraj et al., 2010). However, co-exposure to O₃ and diesel exhaust particles increased lung resistance.

In addition to exacerbating existing allergic responses, O₃ can also act as an adjuvant to produce sensitization in the respiratory tract. In a model of murine asthma, using OVA free of detectable endotoxin, inclusion of 1 ppm O₃ during the initial exposures to OVA (2 hours, days 1 and 6) enhanced the inflammatory and allergic responses to subsequent allergen challenge (Hollingsworth et al., 2010). Compared to air exposed animals, O₃-exposed mice exhibited significantly higher levels of total cells, macrophages, eosinophils, and PMNs in BALF, and increased total serum IgE. Pro-allergic cytokines IL-4, and IL-5 were also significantly elevated, along with pleiotropic Th2 cytokine IL-9 (associated with bronchial hyperresponsiveness) and pro-inflammatory IL-17, produced by activated T-cells. Based on lower inflammatory, IgE, and cytokine responses in Toll-like receptor 4 deficient mice, the effects of O₃ seem to be dependent on TLR 4 signaling, as are a number of other
biological responses to O₃ according to studies by Hollingsworth et al. (2004), Kleeberger et al. (2000) and Garantziotis et al. (2010). The involvement of TLR 4, along with its endogenous ligand, hyaluronan, in O₃-induced responses described in these studies has been corroborated by a controlled human exposure study by Hernandez et al. (2010), who found increased TLR 4 expression and elevated levels of hyaluronic acid in volunteers with atopy or atopic asthma exposed to 400 ppb O₃. This pathway is discussed in more detail in Chapter 5. Examination of dendritic cells (DCs) from the draining thoracic lymph nodes indicated that O₃ did not enhance the migration of DCs from the lungs to the lymph nodes, nor did it alter the expression of functional DC markers such as CD40, MHC class II, or CD83. However, O₃ did increase expression of CD86, which is generally associated with Th2 responses and was detected at higher levels on DCs from donors with allergic asthma compared to those from healthy donors Chen et al. (2006b). Increased CD86 has also been observed on airway cells collected from human subjects following exposure to O₃ in studies by Lay et al. (2007) and Alexis et al. (2009), but not Hernandez et al. (2010) (study details described in Section 6.2.5.4).

Ozone exposure during gestation has modest effects on allergy and asthma related endpoints in adult offspring. When dams were exposed to 1.2 ppm O₃ (but not 0.8 ppm) from gestational day 9-18, some allergic and inflammatory responses to OVA sensitization and challenge were reduced compared to air exposed controls. Such responses included IgE levels and eosinophils, and were observed only in mice that were immunized early in life (PND 3) as opposed to later (PND 42), perhaps due to the proximity of O₃ and antigen exposure. The effects of gestational O₃ exposure on immune function have not been widely studied, and although reductions in allergic endpoints are not generally observed in association with O₃, other parameters of immune function were found to be reduced, so a more global immunosuppression may underlie these effects.

In addition to pro-allergic effects, O₃ could also make airborne allergens more allergenic. When combined with NO₂, O₃ has been shown to enhance nitration of common protein allergens, which may increase their allergenicity Franze et al. (2005).

6.2.7 Hospital Admissions, Emergency Department Visits, and Physicians Visits

6.2.7.1 Summary of Findings from 2006 O₃ AQCD

The 2006 O₃ AQCD evaluated numerous respiratory ED visits and hospital admissions studies, which consisted primarily of time-series studies conducted in the U.S., Canada, Europe, South America, Australia and Asia. Upon collectively evaluating the scientific evidence, the 2006 O₃ AQCD concluded that “the overall evidence supports a causal relationship between acute ambient O₃ exposures and
increased respiratory morbidity resulting in increased ED visits [and hospital admissions] during the warm season” (U.S. EPA, 2006b). This conclusion was “strongly supported by the human clinical, animal toxicologic[al], and epidemiologic evidence for [O₃-induced] lung function decrements, increased respiratory symptoms, airway inflammation, and airway hyperreactivity” (U.S. EPA, 2006b).

Since the completion of the 2006 O₃ AQCD, relatively fewer studies conducted in the U.S., Canada, and Europe have examined the association between short-term exposure to ambient O₃ and respiratory hospital admissions and ED visits with a growing number of studies having been conducted in Asia. This section focuses primarily on multicity studies because they examine the effect of O₃ on respiratory-related hospital admissions and ED visits over a large geographic area using a consistent statistical methodology. Single-city studies that encompass a large number of hospital admissions or ED visits, or included a long study-duration were also evaluated because these studies have more power to detect whether an association exists between short-term O₃ exposure and respiratory hospital admissions and ED visits compared to smaller single-city studies. Additional single-city studies were also evaluated within this section, if they were conducted in locations not represented by the larger single-city and multicity studies, or examined population-specific characteristics not included in the larger studies that may modify the association between short-term O₃ exposure and respiratory-related hospital admissions or ED visits. The remaining single-city studies identified were not evaluated in this section due to factors such as inadequate study design or insufficient sample size.

It should be mentioned that when examining the association between short-term O₃ exposure and respiratory health effects that require medical attention, it is important to distinguish between hospital admissions and ED visits. This is because it is likely that a small percentage of respiratory ED visits will be admitted to the hospital; therefore, respiratory ED visits may represent potentially less serious, but more common outcomes. As a result, in the following sections respiratory hospital admission and ED visit studies are evaluated individually. Additionally, within each section, results are presented as either a collection of respiratory diagnoses or as individual diseases (e.g., asthma, COPD, pneumonia and other respiratory infections) in order to evaluate the potential effect of short-term O₃ exposure on each respiratory-related outcome. The ICD codes (i.e., ICD-9 or ICD-10) that encompass each of these endpoints are presented in Table 6-26 along with the air quality characteristics of the city, or across all cities, included in each study evaluated in this section.
Table 6-26  Mean and upper percentile concentrations of respiratory-related hospital admission and emergency department (ED) visit studies evaluated.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Type of Visit (ICD9/10)</th>
<th>Averaging Time</th>
<th>Mean Concentration (ppb)</th>
<th>Upper Percentile Concentrations (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>90 U.S. cities (NMMAPS)[b]</td>
<td>Hospital Admissions: All respiratory (460-519)</td>
<td>1-h max</td>
<td>34.9-60.0</td>
<td>75th: 46.8-68.8</td>
</tr>
<tr>
<td></td>
<td>32 European cities (APHEA)[c]</td>
<td></td>
<td></td>
<td></td>
<td>75th: 15.3-49.4</td>
</tr>
<tr>
<td></td>
<td>12 Canadian cities</td>
<td></td>
<td></td>
<td></td>
<td>75th: 8.4-12.4</td>
</tr>
<tr>
<td>Cakmak et al. (2006)</td>
<td>10 Canadian cities</td>
<td>Hospital Admissions: All respiratory (466, 480-486, 490, 491, 492, 493, 494, 496)</td>
<td>24-h avg</td>
<td>17.4</td>
<td>Max: 38.0-79.0</td>
</tr>
<tr>
<td>Biggeri et al. (2005)</td>
<td>4 Italian cities</td>
<td>Hospital Admissions: All respiratory (460-519)</td>
<td>8-h max</td>
<td>50th: 11.0-38.1</td>
<td>75th: 107.5-115.1</td>
</tr>
<tr>
<td>Dales et al. (2006)</td>
<td>11 Canadian cities</td>
<td>Hospital Admissions: Respiratory disorders (486, 768.9, 769, 770.8, 786, 799.0, 799.1)</td>
<td>24-h avg</td>
<td>17.0</td>
<td>95th: 24.9-46.0</td>
</tr>
<tr>
<td>Lin et al. (2008a)</td>
<td>11 New York regions</td>
<td>Hospital Admissions: Respiratory diseases (466, 490-493, 496)</td>
<td>8-h max[d]</td>
<td>44.1</td>
<td>Max: 217.0</td>
</tr>
<tr>
<td>Wong et al. (2009)</td>
<td>Hong Kong</td>
<td>Hospital Admissions: All respiratory (460-519) COPD (490-496)</td>
<td>8-h max[d]</td>
<td>18.8</td>
<td>Max: 100.3</td>
</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>36 U.S. cities</td>
<td>Hospital Admissions: COPD (490-496, excluding 493) Pneumonia (480-487)</td>
<td>8-h max</td>
<td>Warm (May-September): 45.8 Cool (October-April): 27.6</td>
<td>NR</td>
</tr>
<tr>
<td>Yang et al. (2005b)</td>
<td>Vancouver, Canada</td>
<td>Hospital Admissions: COPD (490-492, 494, 496)</td>
<td>24-h avg</td>
<td>14.1</td>
<td>95th: 31.0</td>
</tr>
<tr>
<td>Zanobetti and Schwartz (2006)</td>
<td>Boston, MA</td>
<td>Hospital Admissions: Pneumonia (480-487)</td>
<td>24-h avg</td>
<td>22.4</td>
<td>95th: 47.6</td>
</tr>
<tr>
<td>Silverman and Ito (2010)</td>
<td>New York, NY</td>
<td>Hospital Admissions: Asthma (493)</td>
<td>8-h max</td>
<td>Warm (April-August): 41.0</td>
<td>75th: 53</td>
</tr>
<tr>
<td>Study</td>
<td>Location</td>
<td>Type of Visit (ICD9/10)</td>
<td>Averaging Time</td>
<td>Mean Concentration (ppb)(^a)</td>
<td>Upper Percentile Concentrations (ppb)(^a)</td>
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<tr>
<td>Stieb et al. (2009)</td>
<td>7 Canadian cities</td>
<td>ED Visits: Asthma (493) COPD (490-492, 494-496) Respiratory infection (464, 466, 480-487)</td>
<td>24-h avg</td>
<td>18.4</td>
<td>75th: 19.3-28.6</td>
</tr>
<tr>
<td>Tolbert et al. (2007)</td>
<td>Atlanta, GA</td>
<td>ED Visits: All respiratory (460-465, 460.0, 466.1, 466.11, 466.19, 477, 480-486, 491, 492, 493, 496, 786.07, 786.09)</td>
<td>8-h max</td>
<td>Warm: 53.0</td>
<td>75th: 67.0 90th: 82.1 Max: 147.5</td>
</tr>
<tr>
<td>Darrow et al. (2011a)</td>
<td>Atlanta, GA</td>
<td>ED Visits: All respiratory (460-466, 477, 480-486, 491, 492, 493, 496, 786.09)</td>
<td>8-h max</td>
<td>Warm (March-October): 8-h max: 53</td>
<td>8-h max:75th: 67 8-h max:Max: 148</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Day-time</td>
<td>Warm (March-October): Day-time: 45</td>
<td>1-h max</td>
<td>Warm (March-October): 1-h max: 62</td>
</tr>
<tr>
<td></td>
<td></td>
<td>24-h avg</td>
<td>Warm (March-October): 24-h avg: 30</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>Commute</td>
<td>Warm (March-October): Commute: 35</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>Night-time</td>
<td>Warm (March-October): Night-time: 14</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Villeneuve et al. (2007)</td>
<td>Alberta, CAN</td>
<td>ED Visits: Asthma (493)</td>
<td>8-h max</td>
<td>Summer (April-September): 38.0</td>
<td>Summer: 75th: 46.0 Winter: 75th: 31.5</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Winter (October-March): 24.3</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ito et al. (2007b)</td>
<td>New York, NY</td>
<td>ED Visits: Asthma (493)</td>
<td>8-h max</td>
<td>All year: 30.4</td>
<td>All year: 95th: 68.0 Warm months: 95th: 77.0 Cold months: 95th: 33.0</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Warm (April-September): 42.7</td>
<td></td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>Cold (October-March): 18.0</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Strickland et al. (2010)</td>
<td>Atlanta, GA</td>
<td>ED Visits: Asthma (493) Wheeze (786.07 after 10/1/98, 786.09 before 10/1/98)</td>
<td>8-h max</td>
<td>All year: 45.4</td>
<td>NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>Warm (May-October): 55.2</td>
<td></td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>Cold (November-April): 34.5</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mar and Koenig (2009)</td>
<td>Seattle, WA</td>
<td>ED Visits: Asthma (493-493.9)</td>
<td>1-h max</td>
<td>Warm (May-October): 1-h max: 38.6</td>
<td>75th: 1-h max: 45.5 8-h max: 39.2</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>8-h max</td>
<td>8-h max: 32.2</td>
<td></td>
</tr>
<tr>
<td>Arbex et al. (2009)</td>
<td>Sao Paulo, Brazil</td>
<td>ED Visits: COPD (J40-44)</td>
<td>1-h max</td>
<td>48.8</td>
<td>75th: 61.0 Max: 143.8</td>
</tr>
</tbody>
</table>

\(^a\) Concentrations in ppb.
<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Type of Visit (ICD9/10)</th>
<th>Averaging Time</th>
<th>Mean Concentration (ppb)(^a)</th>
<th>Upper Percentile Concentrations (ppb)(^a)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Orazzo et al. (2009)*</td>
<td>6 Italian cities</td>
<td>ED Visits: Wheezing</td>
<td>8-h max</td>
<td>Summer (April-September): 21.1-44.3</td>
<td>NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Winter (October-March): 11.5-27.9</td>
<td></td>
</tr>
<tr>
<td>Burra et al. (2009)</td>
<td>Toronto, Canada</td>
<td>Physician Visits: ED Asthma (493)</td>
<td>1-h max</td>
<td>33.3</td>
<td>95th: 66</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Max: 121</td>
<td></td>
</tr>
<tr>
<td>Villeneuve et al. (2006b)</td>
<td>Toronto, Canada</td>
<td>Physician Visits: Allergic rhinitis (177)</td>
<td>8-h max</td>
<td>30.0</td>
<td>Max: 98.7</td>
</tr>
<tr>
<td>Sinclair et al. (2010)</td>
<td>Atlanta, GA</td>
<td>Physician Visits: Asthma</td>
<td>8-h max</td>
<td>Total Study Period: All-year: 44.0</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>Upper respiratory infection</td>
<td></td>
<td>25 mo Period: All-year: 47.9</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>Lower respiratory infection</td>
<td></td>
<td>Warm: 61.2</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Cold: 27.8</td>
<td></td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td>28 mo Period: All-year: 40.7</td>
<td></td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td>Warm: 51.8</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Cold: 26.0</td>
<td></td>
</tr>
</tbody>
</table>

\(^a\)Some studies did not present an overall value for the mean, middle and/or upper percentiles of the O\(_3\) distribution; as a result, the range of the mean, middle, and/or upper percentiles across all of the cities included in the study are presented.

\(^b\)Study only presented median concentrations.

\(^c\)Study presented concentrations as µg/m\(^3\) Concentration was converted to ppb using the conversion factor of 0.51 assuming standard temperature (25°C) and pressure (1 atm).

\(^d\)A subset of the European and U.S. cities included in the mortality analyses were used in the hospital admissions analyses: 8 of the 32 European cities and 14 of 90 U.S. cities.

\(^e\)Hospital admission data was coded using three classifications (ICD-10-CA, ICD-9, and ICD-9-CM). Attempts were made by the original investigators to convert diagnosis from ICD-10-CA back to ICD-9.

\(^f\)Only 4 of the 8 cities included in the study collected O\(_3\) data.

\(^g\)O\(_3\) measured from 10:00 a.m. to 6:00 p.m.

\(^h\)Only 35 of the 36 cities included in the analysis had O\(_3\) data.

\(^i\)Commute (7:00 a.m. to 10:00 a.m., 4:00 p.m. to 7:00 p.m.); day-time (8:00 a.m. to 7:00 p.m.); Night-time (12:00 a.m. to 6:00 a.m.).

\(^j\)Means represent population-weighted O\(_3\) concentrations.

\(^k\)O\(_3\) measured from 8:00 a.m. to 4:00 p.m.

\(^l\)This study did not report the ICD codes used for the conditions examined. The 25-month period represents August 1998-August 2000, and the 28-month period represents September 2000-December 2002. This study defined the warm months as April – October and the cold months as November-March.

### 6.2.7.2 Hospital Admission Studies

**Respiratory Diseases**

The association between exposure to an air pollutant, such as O\(_3\), and daily respiratory-related hospital admissions has primarily been examined using all respiratory-related hospital admissions within the range of ICD-9 codes 460-519. Recent studies published since the 2006 O\(_3\) AQCD (U.S. EPA, 2006b) attempt to further examine the effect of O\(_3\) exposure on respiratory-related hospital admissions through a multicity design that examines O\(_3\) effects across countries using a standardized methodology; multicity studies that examine effects within one country;
and multi- and single-city studies that attempt to examine potential modifiers of the \( \text{O}_3 \)-respiratory-related hospital admission relationship.

The Air Pollution and Health: A European and North American Approach (APHENA) study combined data from existing multicity study databases from Canada, Europe (APHEA2) (Katsouyanni et al., 2001), and the U.S. (NMMAPS) (Samet et al., 2000) in order to “develop more reliable estimates of the potential acute effects of air pollution on human health [and] provide a common basis for [the] comparison of risks across geographic areas” (Katsouyanni et al., 2009). In an attempt to address both of these issues, the investigators conducted extensive sensitivity analyses to evaluate the robustness of the results to different model specifications (e.g., penalized splines [PS] versus natural splines [NS]) and the extent of smoothing to control for seasonal and temporal trends. The trend analyses consisted of subjecting the models to varying extent of smoothing selected either a priori (i.e., 3 df/year, 8 df/year, and 12 df/year), which was selected through exploratory analyses using between 2 and 20 df, or by using the absolute sum of the residuals of the partial autocorrelation function (PACF). Although the investigators did not identify the model they deemed to be the most appropriate for comparing the results across study locations, they did specify that “overall effect estimates (i.e., estimates pooled over several cities) tended to stabilize at high degrees of freedom” (Katsouyanni et al., 2009). Therefore, in discussion of the results across the three study locations below, the 8 df/year results are presented for both the PS and NS models because: (1) 8 df/year is most consistent with the extent of temporal adjustment used in previous and recent large multicity studies in the U.S. (e.g., NMMAPS); (2) the risk estimates for 8 df/year and 12 df/year are comparable for all three locations; (3) the models that used the PACF method did not report the actual degrees of freedom chosen; and (4) the 3 df/year and the PACF method resulted in negative \( \text{O}_3 \) risk estimates, which is inconsistent with the results obtained using more aggressive seasonal adjustments and suggests inadequate control for seasonality. Additionally, in comparisons of results across studies in figures, only the results from one of the spline models (i.e., NS) are presented because it has been previously demonstrated that alternative spline models result in relatively similar effect estimates (HEI, 2003). This observation is consistent with the results of the APHENA analysis that was conducted with a higher number of degrees of freedom (e.g., \( \geq 8 \) df/year) to account for temporal trends.

Katsouyanni et al. (2009) examined respiratory hospital admissions for people aged 65 years and older using 1-h max \( \text{O}_3 \) data. The extent of hospital admission and \( \text{O}_3 \) data varied across the 3 datasets: Canadian dataset included 12 cities with data for 3 years (1993-1996) per city; European dataset included 8 cities with each city having data for between 2 and 8 years from 1988-1997; and the U.S. dataset included 14 cities with each city having data for 4 to 10 years from 1985-1994 and 7 cities having only summer \( \text{O}_3 \) data. The investigators used a three-stage hierarchical model to account for within-city, within region, and between region variability. Results were presented individually for each region (Figure 6-15 [and Table 6-27]). Ozone and PM\(_{10}\) concentrations were weakly correlated in all locations in the summer (ranging from \( r = 0.27 \) – 0.40), but not in the winter.
In the Canadian cities, using all-year data, a 40 ppb increase in 1-h max \(O_3\) concentrations at lag 0-1 was associated with an increase in respiratory hospital admissions of 8.9% (95% CI: 0.79, 16.8%) in a PS model and 8.1% (95% CI: 0.24, 16.8%) in a NS model (Katsouyanni et al., 2009). The results were somewhat sensitive to the lag day selected, reduced when using a single-day lag (e.g., lag 1) (PS: 6.0%; NS: 5.5%) and increased when using a distributed lag model (PS: 18.6%; NS: 20.4%). When adjusted for PM\(_{10}\), the magnitude of the effect estimate was attenuated, but remained positive with it being slightly larger in the NS model (5.1% [95% CI: -6.6, 18.6%]) compared to the PS model (3.1% [95% CI: -8.3, 15.9%]). However, in the Canadian dataset the copollutant analysis was only conducted using a 1-day lag. The large confidence intervals for both models could be attributed to the reduction in days included in the copollutant analyses as a result of the every-6th-day PM sampling schedule. When the analysis was restricted to the summer months, stronger associations were observed between \(O_3\) and respiratory hospital admissions across the lags examined, ranging from ~22 to 37% (the study does not specify whether these effect estimates are from a NS or PS model). Because \(O_3\) concentrations across the cities included in the Canadian dataset are low (median concentrations ranging from 6.7-8.3 ppb [Table 6-26]), the standardized increment of 40 ppb for a 1-h max increase in \(O_3\) concentrations represents an unrealistic increase in \(O_3\) concentrations in Canada and increases the magnitude, not direction, of the observed risk estimate. As a result, calculating the \(O_3\) risk estimate using the 40 ppb increment does not accurately reflect the observed risk of \(O_3\)-related respiratory hospital admissions. Although this increment adequately characterizes the distribution of 1-h max \(O_3\) concentrations across the U.S. and European datasets, it misrepresents the observed \(O_3\) concentrations in the Canadian dataset. As a result in summary figures, for comparability, effect estimates from the Canadian dataset are presented for both a 5.1 ppb increase in 1-h max \(O_3\) concentrations (i.e., an approximate interquartile range [IQR] increase in \(O_3\) concentrations across the Canadian cities) as well as the 40 ppb increment used throughout the ISA.

In Europe, weaker but positive associations were also observed in year round analyses; 2.9% (95% CI: 0.63, 5.0%) in the PS model and 1.6% (95% CI: -1.7, 4.2%) in the NS model at lag 0-1 for a 40 ppb increase in 1-h max \(O_3\) concentrations (Katsouyanni et al., 2009). Additionally, at lag 1, associations between \(O_3\) and respiratory hospital admissions were also reduced, but in contrast to the lag 0-1 analysis, greater effects were observed in the NS model (2.9% [95% CI: 1.0, 4.9%]) compared to the PS model (1.5% [95% CI: -2.2, 5.4%]). Unlike the Canadian analysis, a distributed lag model provided limited evidence of an association between \(O_3\) and respiratory hospital admissions. To compare with the Canadian results, focused on adjustment for PM\(_{10}\) at lag 1, \(O_3\) effect estimates using the European dataset were increased in the PS model (2.5% [95% CI: 0.39-4.8%]) and remained robust in the NS model (2.4% [95% CI: 0.08, 4.6%]). However, the European analysis also examined the effect of adjusting for PM\(_{10}\) at lag 0-1 and found results were attenuated, but remained positive in both models (PS: 0.8% [95% CI: -2.3, 4.0%]; NS: 0.8% [95% CI: -1.8, 3.6%]). Unlike the Canadian and U.S. datasets, the European dataset consisted of daily PM data. The investigators did not observe stronger associations in the summer-only analyses for the European cities at lag 0-1.
(PS: 0.4% [95% CI: -3.2, 4.0%]; NS: 0.2% [95% CI: -3.3, 3.9%]), but did observe some evidence for larger effects during the summer, an \(~2.5\%\) increase, at lag 1 in both models (the study does not present the extent of temporal smoothing used for these models).

Note: Black circles = all-year results; open circles = all-year results in copollutant model with PM\textsubscript{10}; and red circles = summer only results. For Canada, lag days with an “a” next to them represent the risk estimates standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O\textsubscript{3} concentrations.

**Figure 6-15** Percent increase in respiratory hospital admissions from natural spline models with 8 df/yr for a 40 ppb increase in 1-h max O\textsubscript{3} concentrations for each location of the APHENA study.
Table 6-27  Corresponding effect estimates for Figure 6-15.

<table>
<thead>
<tr>
<th>Location*</th>
<th>Season</th>
<th>Lag*</th>
<th>Copollutant</th>
<th>% Increase (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>U.S.</strong></td>
<td>All-year</td>
<td>1</td>
<td></td>
<td>2.62 (0.63, 4.64)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1</td>
<td>PM$_{10}$</td>
<td>2.14 (-0.08, 4.40)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td></td>
<td>2.38 (0.00, 4.89)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td>PM$_{10}$</td>
<td>1.42 (-1.33, 4.23)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>DL(0-2)</td>
<td></td>
<td>3.34 (0.02-6.78)</td>
</tr>
<tr>
<td></td>
<td>Summer</td>
<td>0-1</td>
<td></td>
<td>2.14 (-0.63, 4.97)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1</td>
<td></td>
<td>2.78 (-0.02, 5.71)</td>
</tr>
<tr>
<td><strong>Canada</strong></td>
<td>All-year</td>
<td>1</td>
<td></td>
<td>5.54 (-0.94, 12.4)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1a</td>
<td></td>
<td>0.69 (-0.12, 1.50)$^a$</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1</td>
<td>PM$_{10}$</td>
<td>5.13 (-6.62, 18.6)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1a</td>
<td>PM$_{10}$</td>
<td>0.64 (-0.87, 2.20)$^a$</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td></td>
<td>8.12 (0.24, 16.8)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1a</td>
<td></td>
<td>1.00 (0.03, 2.00)$^a$</td>
</tr>
<tr>
<td></td>
<td></td>
<td>DL(0-2)</td>
<td></td>
<td>20.4 (4.07, 40.2)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>DL(0-2)$^a$</td>
<td></td>
<td>2.4 (0.51, 4.40)$^a$</td>
</tr>
<tr>
<td></td>
<td>Summer</td>
<td>1</td>
<td></td>
<td>21.4 (15.0, 29.0)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1a</td>
<td></td>
<td>2.50 (1.80, 3.30)$^a$</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td></td>
<td>32.0 (18.6, 47.7)</td>
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<td></td>
<td>0-1a</td>
<td></td>
<td>3.60 (2.20, 5.10)$^a$</td>
</tr>
<tr>
<td></td>
<td></td>
<td>DL(0-2)</td>
<td></td>
<td>37.1 (11.5, 67.5)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>DL(0-2)$^a$</td>
<td></td>
<td>4.1 (1.40, 6.80)$^a$</td>
</tr>
<tr>
<td><strong>Europe</strong></td>
<td>All-year</td>
<td>1</td>
<td></td>
<td>2.94 (1.02, 4.89)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1</td>
<td>PM$_{10}$</td>
<td>2.38 (0.08, 4.64)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td></td>
<td>1.58 (-1.71, 4.15)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td>PM$_{10}$</td>
<td>0.87 (-1.79, 3.58)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>DL(0-2)</td>
<td></td>
<td>0.79 (-4.46, 6.37)</td>
</tr>
<tr>
<td></td>
<td>Summer</td>
<td>1</td>
<td></td>
<td>2.46 (-0.63, 5.54)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>0-1</td>
<td></td>
<td>0.24 (-3.32, 3.91)</td>
</tr>
</tbody>
</table>

*For effect estimates in Figure 6-15.

$^a$For Canada, lag days with an “a” next to them represent the risk estimates standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O$_3$ concentrations.

$^b$Unless noted, risk estimates standardized to 40 ppb for a 1-h max increase in O$_3$ concentrations.

For the U.S. in year round analyses, the investigators reported a 1.4% (95% CI: -0.9, 3.9%) increase in the PS model and 2.4% (95% CI: 0.0, 4.9%) increase in the NS model in respiratory hospital admissions at lag 0-1 for a 40 ppb increase in 1-h max O$_3$ concentrations with similar results for both models at lag 1 (Katsouyanni et al., 2009). The distributed lag model provided results similar to those observed in the European dataset with the PS model (1.1% [95% CI: -3.0, 5.3%]), but larger
effects in the NS model (3.3% [95% CI: 0.02, 6.8%]), which is consistent with the Canadian results. With adjustment for PM$_{10}$ using the U.S. data (i.e., every-6th-day PM data), results were attenuated, but remained positive at lag 0-1 (PS: 0.6% [95% CI: -2.0, 3.3%]; NS: 1.4% [95% CI: -1.3, 4.2%]) which is consistent with the results presented for the European dataset. However, at lag 1, U.S. risk estimates remained robust to the inclusion of PM$_{10}$ in copollutant models as was observed in the Canadian and European datasets. Compared to the all-year analyses, the investigators did not observe stronger associations in the summer-only analysis at either lag 0-1 (~2.2%) or lag 1 (~2.8%) in both the PS and NS models (the study does not present the extent of temporal smoothing used for these models).

Several additional multicity studies examined respiratory disease hospital admissions in Canada and Europe. Cakmak et al. (2006b) evaluated the association between ambient O$_3$ concentrations and respiratory hospital admissions for all ages in 10 Canadian cities from April 1993 to March 2000. The primary objective of this study was to examine the potential modification of the effect of ambient air pollution on daily respiratory hospital admissions by education and income using a time-series analysis conducted at the city-level. The authors calculated a pooled estimate across cities for each pollutant using a random effects model by first selecting the lag day with the strongest association from the city-specific models. For O$_3$, the mean lag day across cities that provided the strongest association and for which the pooled effect estimate was calculated was 1.2 days. In this study, all-year O$_3$ concentrations were used in the analysis, and additional seasonal analyses were not conducted. Cakmak et al. (2006b) reported a 4.4% increase (95% CI: 2.2, 6.5%) in respiratory hospital admissions for a 20 ppb increase in 24-h avg O$_3$ concentrations.

The investigators only examined the potential effect of confounding by other pollutants through the use of a multipollutant model (i.e., two or more additional pollutants included in the model), which is difficult to interpret due to the potential multicollinearity between pollutants. Cakmak et al. (2006b) also conducted an extensive analysis of potential modifiers, specifically sex, educational attainment, and family income, of the association between air pollution and respiratory hospital admissions. When stratified by sex, the increase in respiratory hospital admissions due to short-term O$_3$ exposure were similar in males (5.2% [95% CI: 3.0, 7.3%]) and females (4.2% [95% CI: 1.8, 6.6%]). In addition, the examination of effect modification by income found no consistent trend across the quartiles of family income. However, there was evidence that individuals with an education level less than the 9th grade were disproportionately affected by O$_3$ exposure (4.6% [95% CI: 1.8, 7.5%]) compared to individuals that completed grades 9-13 (1.7% [95% CI: -1.9, 5.3%]), some university or trade school (1.4% [95% CI: -2.0, 5.1%]), or have a university diploma (0.66% [95% CI: -3.3, 4.7%]). The association between O$_3$ and respiratory hospital admissions in individuals with an education level less than the 9th grade was the strongest association across all of the pollutants examined.

A multicity study conducted in Europe by Biggeri et al. (2005) examined the association between short-term O$_3$ exposure and respiratory hospital admissions for all ages in four Italian cities from 1990 to 1999. In this study, O$_3$ was only measured during the warm season (May-September). The authors examined associations
between daily respiratory hospital admissions and short-term O₃ exposure at the city-level using a time-series analysis. Pooled estimates were calculated by combining city-specific estimates using fixed and random effects models. The investigators found no evidence of an association between O₃ exposure and respiratory hospital admissions in the warm season in both the random (0.1% [95% CI: -5.2, 5.7%]; distributed lag 0-3) and fixed effects (0.1% [95% CI: -5.2, 5.7%]; distributed lag 0-3) models for a 30 ppb increase in 8-h max O₃ concentrations.

Additional studies examined associations between short-term O₃ exposure and respiratory hospital admissions specifically in children. In a multicity study conducted in Canada, Dales et al. (2006) examined the association between all-year ambient O₃ concentrations and neonatal (ages 0-27 days) respiratory hospital admissions in 11 Canadian cities from 1986 to 2000. The investigators used a statistical analysis approach similar to Cakmak et al. (2006b) (i.e., time-series analysis to examine city-specific associations, and then a random effects model to pool estimates across cities). The authors reported that for O₃, the mean lag day across cities that provided the strongest association was 2 days. The authors reported a 5.4% (95% CI: 2.9, 8.0%) increase in neonatal respiratory hospital admissions for a 20 ppb increase in 24-h avg O₃ concentrations at lag-2 days. The results from Dales et al. (2006) provide support for the associations observed in a smaller scale study that examined O₃ exposure and pediatric respiratory hospital admissions in New York state (Lin et al., 2008a). Lin et al. (2008a), when examining single-day lags of 0 to 3 days, observed a positive association between O₃ and pediatric (i.e., <18 years) respiratory admissions at lag 2 (results not presented quantitatively) in a two-stage Bayesian hierarchical model analysis of 11 geographic regions of New York state from 1991 to 2001. Additionally, in copollutant models with PM10 collected every-6th day, the authors found region-specific O₃ associations with respiratory hospital admissions remained relatively robust.

Overall, the evidence from epidemiologic studies continues to support an association between short-term O₃ exposure and respiratory-related hospital admissions, but it remains unclear whether certain factors (individual- or population-level) modify this association. Wong et al. (2009) examined the potential modification of the relationship between ambient O₃ (along with NO₂, SO₂, and PM10) and respiratory hospital admissions by influenza intensity in Hong Kong for the period 1996 – 2002. In this study air pollution concentrations were estimated by centering non-missing daily air pollution data on the annual mean for each monitor and then an overall daily concentration was calculated by taking the average of the daily centered mean across all monitors. Influenza intensity was defined as a continuous variable using the proportion of weekly specimens positive for influenza A or B instead of defining influenza epidemics. This approach was used to avoid any potential bias associated with the unpredictable seasonality of influenza in Hong Kong where there are traditionally two seasonal peaks, which is in contrast to the single peaking influenza season in the U.S. (Wong et al., 2009). In models that examined the baseline effect (i.e., without taking into consideration influenza intensity) of short-term O₃ exposure, the authors found a 3.6% (95% CI: 1.9, 5.3%) and 3.2% (95% CI: 1.0, 5.4%) increase in respiratory hospital admissions at lag 0-1 for a 30 ppb increase in
8-h max O₃ concentrations for the all age and ≥ 65 age groups, respectively. When examining influenza intensity, Wong et al. (2009) reported that the association between short-term exposure to O₃ and respiratory hospital admissions was stronger with higher levels of influenza intensity: additional increase in respiratory hospital admissions above baseline of 1.4% (95% CI: 0.24, 2.6%) for all age groups and 2.4% (95% CI: 0.94, 3.8%) for those 65 and older when influenza activity increased from 0% to 10%. No difference in effects was observed when stratifying by sex.

**Cause-Specific Respiratory Outcomes**

In the 2006 O₃ AQCD a limited number of studies were identified that examined the effect of short-term O₃ exposure on cause-specific respiratory hospital admissions. The limited evidence “reported positive O₃ associations with... asthma and COPD, especially... during the summer or warm season” (U.S. EPA, 2006b). Of the studies evaluated since the completion of the 2006 O₃ AQCD, more have focused on identifying whether O₃ exposure is associated with specific respiratory-related hospital admissions, including COPD, pneumonia, and asthma, but the overall body of evidence remains small.

**Chronic Obstructive Pulmonary Disease**

Medina-Ramon et al. (2006) examined the association between short-term exposure to ambient O₃ and PM₁₀ concentrations and Medicare hospital admissions among individuals ≥ 65 years of age for COPD in 35 cities in the U.S. for the years 1986-1999. The cities included in this analysis were selected because they monitored PM₁₀ on a daily basis. In this study, city-specific results were obtained using a monthly time-stratified case-crossover analysis. A meta-analysis was then conducted using random effects models to combine the city-specific results. All cities measured O₃ from May through September, while only 16 of the cities had year-round measurements. The authors reported a 1.6% increase (95% CI: 0.48, 2.9%) in COPD admissions for lag 0-1 in the warm season for a 30 ppb increase in 8-h max O₃ concentrations. In examination of single-day lags, stronger associations were observed for lag 1 (2.9% [95% CI: 1.8, 4.0%]) compared to lag 0 (-1.5% [95% CI: -2.7, -0.24%]). The authors found no evidence of increased associations in cool season (-1.9% [95% CI: -3.6, -0.06%]; lag 0-1) or year round (0.24% [95% CI: -0.78, 1.2%]; lag 0-1) analyses. In a copollutant model restricted to days in which PM₁₀ was available, the association between O₃ and COPD hospital admissions remained robust. Of note, the frequency of PM₁₀ measurements varied across cities with measurements collected either every 2, 3, or 6 days. The authors conducted additional analyses to examine potential modification of the warm season estimates for O₃ and COPD admissions by several city-level characteristics: percentage living in poverty, emphysema mortality rate (as an indication of smoking), daily summer apparent temperature, and percentage of households using central air conditioning. Of the city-level characteristics examined, stronger associations were only reported for cities with a smaller variability in daily apparent summer temperature.
In a single-city study conducted in Vancouver from 1994-1998, a location with low ambient O₃ concentrations (Table 6-26), Yang et al. (2005b) examined the association between O₃ and COPD. Ozone was moderately inversely correlated with CO (r = -0.56), NO₂ (r = -0.32), and SO₂ (r = -0.34), and weakly inversely correlated with PM₁₀ (r = -0.09), suggesting that the observed O₃ effect is likely not only due to a positive correlation with other pollutants. Yang et al. (2005b) examined 1- to 7-day (e.g., 0-6 days) lagged moving averages and observed an 8.8% (95% CI: -12.5, 32.6%) increase in COPD admissions for lag 0-3 per 20 ppb increase in 24-h avg O₃ concentrations. In two-pollutant models with every-day data for NO₂, SO₂, or PM₁₀ at lag 0-3, O₃ risk estimates remained robust, but were increased slightly when CO was added to the model (Figure 6-20 [and Table 6-29]).

In the study discussed above, Wong et al. (2009) also examined the potential modification of the relationship between ambient O₃ and COPD hospital admissions by influenza intensity. The authors also found evidence of an additional increase in COPD admissions above baseline when influenza activity increased from 0% to 10% of 1.0% (95% CI: -0.82, 2.9%) for all age groups and 2.4% (95% CI: 0.41, 4.4%) for those 65 and older. The baseline increase in COPD hospital admissions at lag 0-1 for a 30 ppb increase in 8-h max O₃ concentrations was 8.5% (95% CI: 5.6, 11.4%) for the all age and 4.2% (95% CI: 1.1, 7.3%) ≥ 65 age groups.

Pneumonia

In addition to COPD, Medina-Ramon et al. (2006) examined the association between short-term exposure to ambient O₃ and PM₁₀ concentrations and Medicare hospital admissions among individuals ≥ 65 years of age for pneumonia (ICD-9: 480-487). The authors reported an increase in pneumonia-hospital admissions in the warm season (2.5% [95% CI: 1.6, 3.5%] for a 30 ppb increase in 8-h max O₃ concentrations; lag 0-1). Similar to the results observed for COPD hospital admissions, pneumonia-hospital admissions associations were stronger at lag 1 (2.6% [95% CI: 1.8, 3.4%]) compared to lag 0 (0.06% [95% CI: -0.72, 0.78%]), and no evidence of an association was observed in the cool season or year round. In two-pollutant models restricted to days for which PM₁₀ data was available, as discussed above, the association between O₃ exposure and pneumonia-hospital admissions remained robust (results not presented quantitatively). The authors also examined potential effect modification of the warm season estimates for O₃-related pneumonia-hospital admissions, as was done for COPD, by several city-level characteristics. Stronger associations were reported in cities with a lower percentage of central air conditioning use. Across the cities examined, the percentage of households having central air conditioning ranged from 6 to 93%. The authors found no evidence of effect modification of the O₃-pneumonia-hospital admission relationship when examining the other city-level characteristics.

Results from a single-city study conducted in Boston, MA, did not support the results presented by Medina-Ramon et al. (2006). Zanobetti and Schwartz (2006) examined the association of O₃ and pneumonia Medicare hospital admissions for the period 1995-1999. Ozone was weakly positively correlated with PM₂.₅ (r = 0.20) and
weakly inversely correlated with black carbon, NO$_2$, and CO (-0.25, -0.14, and -0.30, respectively). In an all-year analysis, the investigators reported a 3.8% (95% CI: -7.9, -0.1%) decrease in pneumonia admissions for a 20 ppb increase in 24-h avg O$_3$ concentrations at lag 0 and a 6.0% (95% CI: -11.1, -1.4%) decrease for the average of lags 0 and 1. It should be noted that the mean daily counts of pneumonia admissions was low for this study, ~14 admissions per day compared to ~271 admissions per day for Medina-Ramon et al. (2006). However, in analyses with other pollutants Zanobetti and Schwartz (2006) did observe positive associations with pneumonia-hospital admissions, indicating that the low number of daily hospital admission counts probably did not influence the O$_3$ pneumonia-hospital admissions association in this study.

Asthma

There are relatively fewer studies that examined the association between short-term exposure to O$_3$ and asthma hospital admissions, presumably due to the limited power given the relative rarity of asthma hospital admissions compared to ED or physician visits. A study from New York City examined the association of 8-h max O$_3$ concentrations with severe acute asthma admissions (i.e., those admitted to the Intensive Care Unit [ICU]) during the warm season in the years 1999 through 2006 (Silverman and Ito, 2010). In this study, O$_3$ was moderately correlated with PM$_{10}$ ($r = 0.59$). When stratifying by age, the investigators reported positive associations with ICU asthma admissions for the 6- to 18-year age group (26.8% [95% CI: 1.4, 58.2%] for a 30 ppb increase in 8-h max O$_3$ concentrations at lag 0-1), but little evidence of associations for the other age groups examined (<6 years, 19-49, 50+, and all ages). However, positive associations were observed for each age-stratified group and all ages for non-ICU asthma admissions, but again the strongest association was reported for the 6- to 18-years age group (28.2% [95% CI: 15.3, 41.5%]; lag 0-1). In two-pollutant models, O$_3$ effect estimates for both non-ICU and ICU hospital admissions remained robust to adjustment for PM$_{2.5}$. In an additional analysis, using a smooth function, the authors examined whether the shape of the C-R curve for O$_3$ and asthma hospital admissions (i.e., both general and ICU for all ages) is linear. To account for the potential confounding effects of PM$_{2.5}$, Silverman and Ito (2010) also included a smooth function of PM$_{2.5}$ lag 0-1. When comparing the curve to a linear fit line the authors found that the linear fit is a reasonable approximation of the C-R relationship between O$_3$ and asthma hospital admissions around and below the level of the 1997 O$_3$ NAAQS (Figure 6-16).
Note: The average of 0-day and 1-day lagged 8-hour O₃ was used in a two-pollutant model with PM_{2.5} lag 0-1, adjusting for temporal trends, day of the week, and immediate and delayed weather effects. The solid lines are smoothed fit data, with long broken lines indicating 95% confidence bands. The density of lines at the bottom of the figure indicates sample size.

Source: Reprinted with permission of the American Academy of Allergy, Asthma & Immunology (Silverman and Ito, 2010).

**Figure 6-16** Estimated relative risks (RRs) of asthma hospital admissions for 8-h max O₃ concentrations at lag 0-1 allowing for possible nonlinear relationships using natural splines.

**Averting Behavior**

The studies discussed above have found consistent positive associations between short-term O₃ exposure and respiratory-related hospital admissions, however, the strength of these associations may be underestimated due to the studies not accounting for averting behavior. As discussed in Section 4.6.6, a recent study (Neidell and Kinney, 2010; Neidell, 2009) conducted in Southern California demonstrated that controlling for avoidance behavior increases O₃ effect estimates for respiratory hospital admissions, specifically for children and older adults. These analyses show that on days where no public alert was issued warning of high O₃ concentrations there was an increase in asthma hospital admissions. Although only one study has examined averting behavior and this study is limited to the outcome of asthma hospital admissions in one location (i.e., Los Angeles, CA) for the years...
1989-1997, it does provide preliminary evidence indicating that epidemiologic studies may underestimate associations between O$_3$ exposure and health effects by not accounting for behavioral modification when public health alerts are issued.

### 6.2.7.3 Emergency Department Visit Studies

Overall, relatively fewer studies have examined the association between short-term O$_3$ exposure and respiratory-related ED visits, compared to hospital admissions. In the 2006 O$_3$ AQCD, positive, but inconsistent, associations were observed between O$_3$ and respiratory-related ED visits with effects generally occurring during the warm season. Since the completion of the previous AQCD, larger studies have been conducted, in terms of sample size, study duration, and in some cases multiple cities, to examine the association between O$_3$ and ED visits for all respiratory diseases, COPD, and asthma.

### Respiratory Disease

A large single-city study conducted in Atlanta, GA, by Tolbert et al. (2007), and subsequently re-analyzed by Darrow et al. (2011a) using different air quality data, provides evidence for an association between short-term exposures to ambient O$_3$ concentrations and respiratory ED visits. Tolbert et al. (2007) examined the association between air pollution, both gaseous pollutants and PM and its components, and respiratory disease ED visits in all ages from 1993 to 2004. The correlations between O$_3$ and the other pollutants examined ranged from 0.2 for CO and SO$_2$ to 0.5-0.6 for the PM measures. Using an a priori average of lags 0-2 for each air pollutant examined, the authors reported a 3.9% (95% CI: 2.7, 5.2%) increase in respiratory ED visits for a 30 ppb increase in 8-h max O$_3$ concentrations during the warm season [defined as March-October in Darrow et al. (2011a)]. In copollutant models, limited to days in which data for all pollutants were available, O$_3$ respiratory ED visits associations with CO, NO$_2$, and PM$_{10}$, were attenuated, but remained positive (results not presented quantitatively).

Darrow et al. (2011a) examined the same health data as Tolbert et al. (2007), but used air quality data from one centrally located monitor instead of the average of multiple monitors. This study primarily focused on exploring whether differences exist in the association between O$_3$ exposure and respiratory-related ED visits depending on the exposure metric used (i.e., 8-h max, 1-h max, 24-h avg, commuting period [7:00 a.m. to 10:00 a.m.; 4:00 p.m. to 7:00 p.m.], day-time [8:00 a.m. to 7:00 p.m.] and night-time [12:00 a.m. to 6:00 a.m.]). An ancillary analysis of the spatial variability of each exposure metric conducted by Darrow et al. (2011a) found a rather homogenous spatial distribution of O$_3$ concentrations (r > ~0.8) as the distance from the central monitor increased from 10 km to 60 km for all exposure durations, except the night-time metric. The relatively high spatial correlation gives confidence in the use of a single monitor and the resulting risk estimates. To examine the association between the various O$_3$ exposure metrics and respiratory ED visits,
the authors conceptually used a time-stratified case-crossover framework where control days were selected as those days within the same calendar month and maximum temperature as the case day. However, instead of conducting a traditional case-crossover analysis, the authors used a Poisson model with indicator variables for each of the strata (i.e., parameters of the control days). Darrow et al. (2011a) found using an a priori lag of 1 day, the results were somewhat variable across exposure metrics. The strongest associations with respiratory ED visits were found when using the 8-h max, 1-h max, and day-time exposure metrics with weaker associations using the 24-h avg and commuting period exposure metrics; a negative association was observed when using the night-time exposure metric (Figure 6-17). These results indicate that using the 24-h avg exposure metric may lead to smaller O₃-respiratory ED visits risk estimates due to: (1) the dilution of relevant O₃ concentrations by averaging over hours (i.e., nighttime hours) during which O₃ concentrations are known to be low and (2) potential negative confounding by other pollutants (e.g., CO, NO₂) during the nighttime hours (Darrow et al., 2011a).

Figure 6-17  Risk ratio for respiratory ED visits and different O₃ exposure metrics in Atlanta, GA, from 1993-2004.
In an additional study conducted in 6 Italian cities, Orazzo et al. (2009) examined respiratory ED visits for ages 0-2 years in 6 Italian cities from 1996 to 2000. However, instead of identifying respiratory ED visits using the traditional approach of selecting ICD codes as was done by Tolbert et al. (2007) and Darrow et al. (2011a), Orazzo et al. (2009) used data on wheeze extracted from medical records as an indicator of lower respiratory disease. This study examined daily counts of wheeze in relation to air pollution using a time-stratified case-crossover approach in which control days were matched on day of week in the same month and year as the case day. The authors found no evidence of an association between 8-h max O3 concentrations and respiratory ED visits in children aged 0-2 years in models that examined both single-day lags and moving averages of lags from 0-6 days in year-round and seasonal analyses (i.e., warm and cool seasons). In all-year analyses, the percent increase in total wheeze ranged from -1.4% to -3.3% for a 0-1 to 0-6 day lag, respectively.

**COPD**

Stieb et al. (2009) also examined the association between short-term O3 exposure and COPD ED visits in 7 Canadian cities. Across cities, in an all-year analysis, O3 was found to be positively associated with COPD ED visits (2.4% [95% CI: -1.9, 6.9%] at lag 1 and 4.0% [95% CI: -0.54, 8.6%] at lag 2 for a 20 ppb increase in 24-h avg O3 concentrations). In seasonal analyses, larger effects were observed between O3 and ED visits for COPD during the warm season (i.e., April-September) 6.8% [95% CI: 0.11, 13.9%] (lag day not specified); with no associations observed in the winter season. Stieb et al. (2009) also examined associations between respiratory-related ED visits, including COPD, and air pollution at sub-daily time scales (i.e., 3-h avg of ED visits versus 3-h avg pollutant concentrations) and found no evidence of consistent associations between any pollutant and any respiratory outcome.

In a single-city study, Arbex et al. (2009) examined the association between COPD and several ambient air pollutants, including O3, in Sao Paulo, Brazil for the years 2001-2003 for individuals over the age of 40. Associations between O3 exposure and COPD ED visits were examined in both single-day lag (0-6 days) and polynomial distributed lag models (0-6 days). In all-year analyses, O3 was not found to be associated with an increase in COPD ED visits (results not presented quantitatively). The authors also conducted stratified analyses to examine the potential modification of the air pollutant-COPD ED visits relationship by age (e.g., 40-64, >64) and sex. In these analyses O3 was found to have an increase in COPD ED visits for women, but not for men or either of the age groups examined.

**Asthma**

In a study of 7 Canadian cities, Stieb et al. (2009) also examined the association between exposure to air pollution (i.e., CO, NO2, O3, SO2, PM10, PM2.5, and O3) and asthma ED visits. Associations between short-term O3 exposure and asthma ED visits were examined at the city level and then pooled using either fixed or random effects models depending on whether heterogeneity among effect estimates was
found to be statistically significant. Across cities, in an all-year analysis, the authors found that short-term O3 exposure was associated with an increase (4.7% [95% CI: -1.4, 11.1%] at lag 1 and 3.5% [95% CI: 0.33, 6.8%] at lag 2 for a 20 ppb increase in 24-h avg O3 concentrations) in asthma ED visits. The authors did not present the results from seasonal analyses for asthma but stated that no associations were observed between any pollutant and respiratory ED visits in the winter season.

As stated previously, in analyses of 3-h avg O3 concentrations, the authors observed no evidence of consistent associations between any pollutant and any respiratory outcome, including asthma. A single-city study conducted in Alberta, Canada Villeneuve et al. (2007) from 1992-2002 among individuals two years of age and older provides additional support for the findings from Stieb et al. (2009), but also attempts to identify those lifestages (i.e., 2-4, 5-14, 15-44, 45-64, 65-74, or 75+) at greatest risk to O3-induced asthma ED visits. In a time-referent case-crossover analysis, Villeneuve et al. (2007) found an increase in asthma ED visits in an all-year analysis across all ages (12.0% [95% CI: 6.8, 17.2] for a 30 ppb increase in max 8-h avg O3 concentrations at lag 0-2) with associations being stronger during the warmer months (19.0% [95% CI: 11.9, 28.1]). When stratified by age, the strongest associations were observed in the warm season for individuals 5-14 (28.1% [95% CI: 11.9, 45.1]; lag 0-2) and 15-44 (19.0% [95% CI: 8.5, 31.8]; lag 0-2). These associations were not found to be confounded by the inclusion of aeroallergens in age-specific models.

Several additional single-city studies have also provided evidence of an association between asthma ED visits and ambient O3 concentrations. Ito et al. (2007b) examined the association between short-term exposure to air pollution and asthma ED visits for all ages in New York City from 1999 to 2002. Similar to Darrow et al. (2011a), when examining the spatial distribution of O3 concentrations, Ito et al. (2007b) found a rather homogenous distribution (r ≥ ~0.80) when examining monitor-to-monitor correlations at distances up to 20 miles. Ito et al. (2007b) used three different weather models with varying extent of smoothing to account for temporal relationships and multicollinearity among pollutants and meteorological variables (i.e., temperature and dew point) to examine the effect of model selection on the air pollutant-asthma ED visit relationship. When examining O3, the authors reported a positive association with asthma ED visits, during the warm season across the models (ranging from 8.6 to 16.9%) and an inverse association in the cool season (ranging from -23.4 to -25.1%), at lag 0-1 for a 30 ppb increase in 8-h max O3 concentrations. Ito et al. (2007b) conducted copollutant models using a simplified version of the weather model used in NMMAPS analyses (i.e., terms for same-day temperature and 1-3 day average temperature). The authors found that O3 risk estimates were not substantially changed in copollutant models that used every-day data for PM2.5, NO2, SO2, and CO during the warm season (Figure 6-20 [and Table 6-29]).
Strickland et al. (2010) examined the association between O₃ exposure and pediatric asthma ED visits (ages 5-17 years) in Atlanta, GA, between 1993 and 2004 using air quality data over the same years as Darrow et al. (2011a) and Tolbert et al. (2007). However, unlike Darrow et al. (2011a) and Tolbert et al. (2007), which used single centrally located monitors or an average of monitors, respectively, Strickland et al. (2010) used population-weighting to combine daily pollutant concentrations across monitors. In this study, the authors developed a statistical model using hospital-specific time-series data that is essentially equivalent to a time-stratified case-crossover analysis (i.e., using interaction terms between year, month, and day-of-week to mimic the approach of selecting referent days within the same month and year as the case day). The authors observed a 6.4% (95% CI: 3.2, 9.6%) increase in ED visits for a 30 ppb increase in 8-h max O₃ concentrations at lag 0-2 in an all-year analysis. In seasonal analyses, stronger associations were observed during the warm season (i.e., May-October) (8.4% [95% CI: 4.4, 12.7%]; lag 0-2) than the cold season (4.5% [95% CI: -0.82, 10.0%]; lag 0-2). Strickland et al. (2011) confirmed these findings in an additional analysis using the same dataset, and found that the exposure assignment approach used (i.e., centrally located monitor, unweighted average across monitors, and population-weighted average across monitors) did not influence
pediatric asthma ED visit risk estimates for spatially homogeneous pollutants such as O₃.

In copollutant analyses conducted over the entire dataset for the gaseous pollutants (i.e., CO, NO₂), and limited to a subset of years (i.e., 1998-2004) for which daily PM data (i.e., PM₂.₅ elemental carbon, PM₂.₅ sulfate) were available, Strickland et al. (2010) found that O₃ risk estimates were not substantially changed when controlling for other pollutants (results not presented quantitatively). The authors also examined the C-R relationship between O₃ exposure and pediatric asthma ED visits and found that both quintile and loess C-R analyses (Figure 6-18) suggest that there are elevated associations with O₃ at 8-h max concentrations as low as 30 ppb. These C-R analyses do not provide evidence of a threshold level.

In a single-city study conducted on the West coast, Mar and Koenig (2009) examined the association between O₃ exposure and asthma ED visits (ICD-9 codes: 493-493.9) for children (<18) and adults (≥ 18) in Seattle, WA from 1998 to 2002. Of the total number of visits over the study duration, 64% of visits in the age group <18 comprised boys, and 70% of visits in the ≥ 18 age group comprised females. Mar and Koenig (2009) conducted a time-series analysis using both 1-h max and max 8-h avg O₃ concentrations. A similar magnitude and pattern of associations was observed at each lag examined using both metrics. Mar and Koenig (2009) presented results for single day lags of 0 to 5 days, but found consistent positive associations across individual lag days which supports the findings from the studies discussed above that examined multi-day exposures. For children, consistent positive associations were observed across all lags, ranging from a 19.1-36.8% increase in asthma ED visits for a 30 ppb increase in 8-h max O₃ concentrations with the strongest associations observed at lag 0 (33.1% [95% CI: 3.0, 68.5]) and lag 3 (36.8% [95% CI: 6.1, 77.2]). Ozone was also found to be positively associated with asthma ED visits for adults at all lags, ranging from 9.3-26.0%, except at lag 0. The slightly different lag times for children and adults suggest that children may be more immediately responsive to O₃ exposures than adults Mar and Koenig (2009).

**Respiratory Infection**

Although an increasing number of studies have examined the association between O₃ exposure and cause-specific respiratory ED visits this trend has not included an extensive examination of the association between O₃ exposure and respiratory infection ED visits. Stieb et al. (2009) also examined the association between short-term O₃ exposure and respiratory infection ED visits in 7 Canadian cities. In an all-year analysis, there was no evidence of an association between O₃ exposure and respiratory infection ED visits at any lag examined (i.e., 0, 1, and 2). Across cities, respiratory infections comprised the single largest diagnostic category, approximately 32% of all the ED visits examined, which also included myocardial infarction, heart failure, dysrhythmia, asthma, and COPD.
6.2.7.4 Outpatient and Physician Visit Studies

Several studies have examined the association between ambient O\textsubscript{3} concentrations and physician or outpatient (non-hospital, non-ED) visits for acute conditions in various geographic locations. Burra et al. (2009) examined asthma physician visits among patients aged 1-17 and 18-64 years in Toronto, Canada from 1992 to 2001. The authors found little or no evidence of an association between asthma physician visits and O\textsubscript{3}; however, seasonal analyses were not conducted. It should be noted that in this study, most of the relative risks for O\textsubscript{3} were less than one and statistically significant, perhaps due to an inverse correlation with another pollutant or an artifact of the strong seasonality of asthma visits. Villeneuve et al. (2006b) also focused on physician visits to examine the effect of short-term O\textsubscript{3} exposure on allergic rhinitis among individuals aged 65 or older in Toronto from 1995 to 2000. The authors did not observe any evidence of an association between allergic rhinitis physician visits and ambient O\textsubscript{3} concentrations in single-day lag models in an all-year analysis (results not presented quantitatively).

In a study conducted in Atlanta, GA, Sinclair et al. (2010) examined the association of acute asthma and respiratory infection (e.g., upper respiratory infections and lower respiratory infections) outpatient visits from a managed care organization with ambient O\textsubscript{3} concentrations as well as multiple PM size fractions and species from August 1998 through December 2002. The authors separated the analysis into two time periods (the first 25 months of the study period and the second 28 months of the study period), in order to compare the air pollutant concentrations and relationships between air pollutants and acute respiratory visits for the 25-month time-period examined in Sinclair and Tolsma (2004) to an additional 28-month time-period of available data from the Atlanta Aerosol Research Inhalation Epidemiology Study (ARIES). The authors found little evidence of an association between O\textsubscript{3} and asthma visits, for either children or adults, or respiratory infection visits in all-year analyses and seasonal analyses. For example, a slightly elevated RR for childhood asthma visits was observed during the 25-month period in the cold season (RR: 1.12 [95% CI: 0.86, 1.41]; lag 0-2 for a 30 ppb increase in 8-h max O\textsubscript{3}), but not in the warm season (RR: 0.97 [95% CI: 0.86, 1.10]; lag 0-2). During the 28-month period at lag 0-2, a slightly larger positive effect was observed during the warm season (RR: 1.06 [95% CI: 0.97, 1.17]), compared to the cold season (RR: 1.03 [95% CI: 0.87, 1.21]). Overall, these results contradict those from Strickland et al. (2010) discussed above. Although the mean number of asthma visits and O\textsubscript{3} concentrations in Sinclair et al. (2010) and Strickland et al. (2010) are similar the difference in results between the two studies could potentially be attributed to the severity of O\textsubscript{3}-induced asthma exacerbations (i.e., more severe symptoms requiring a visit to a hospital) and behavior, such as delaying a visit to the doctor for less severe symptoms.
6.2.7.5 Summary

The results of the recent studies evaluated largely support the conclusion of the 2006 O₃ AQCD. While fewer studies were published overall since the previous review, several multicity studies (e.g., Cakmak et al., 2006b; Dales et al., 2006) and a multi-continent study (Katsouyanni et al., 2009) provide supporting evidence for an association between short-term O₃ exposure and an increase in respiratory-related hospital admissions and ED visits. Across studies, different ICD-9 codes were used to define total respiratory causes, which may contribute to some heterogeneity in the magnitude of association. These findings are supported by single-city studies that used different exposure assignment approaches (i.e., average of multiple monitors, single monitor, population-weighted average) and averaging times (i.e., 1-h max and 8-h max).

Collectively, in both single-city and multicity studies there is continued evidence for increases in both hospital admissions and ED visits when examining all respiratory outcomes combined. Additionally, recent studies published since the 2006 O₃ AQCD support an association between short-term O₃ exposure and asthma (Strickland et al., 2010; Stieb et al., 2009) and COPD (Stieb et al., 2009; Medina-Ramon et al., 2006) hospital admissions and ED visits, with more limited evidence for pneumonia-hospital admissions and ED visits (Medina-Ramon et al., 2006; Zanobetti and Schwartz, 2006). As with total respiratory causes, studies used slightly different ICD-9 codes to define specific conditions. In seasonal analyses, stronger associations were observed in the warm season or summer months compared to the cold season, particularly for asthma (Strickland et al., 2010; Ito et al., 2007b) and COPD (Medina-Ramon et al., 2006) (Figure 6-19 [and Table 6-28]), which is consistent with the conclusions of the 2006 O₃ AQCD. There is also continued evidence that children are particularly at greatest risk to O₃-induced respiratory effects (Silverman and Ito, 2010; Strickland et al., 2010; Mar and Koenig, 2009; Villeneuve et al., 2007; Dales et al., 2006). Of note, the consistent associations observed across studies for short-term O₃ exposure and respiratory-related hospital admissions and ED visits was not supported by studies that focused on respiratory-related outpatient or physician visits. These differences could potentially be attributed to the severity of O₃-induced respiratory effects requiring more immediate treatment or behavioral factors that result in delayed visits to a physician. Although the collective evidence across studies indicates a consistent positive association between O₃ exposure and respiratory-related hospital admissions and ED visits, the magnitude of these associations may be underestimated due to behavioral modification in response to forecasted air quality (Neidell and Kinney, 2010; Neidell, 2009) (Section 4.6.6).

The studies that examined the potential confounding effects of copollutants found that O₃ effect estimates remained relatively robust upon the inclusion of PM (measured using different sampling strategies ranging from every-day to every-6th day) and gaseous pollutants in two-pollutant models (Figure 6-20 [and Table 6-29]). Additional studies that conducted copollutant analyses, but did not present quantitative results, also support these conclusions (Strickland et al., 2010; Tolbert et al., 2007; Medina-Ramon et al., 2006). Overall, recent studies provide
copollutant results that are consistent with those from the studies evaluated in the 2006 O₃ AQCD ([U.S. EPA, 2006b], Figure 7-12, page 7-80 of the 2006 O₃ AQCD), which found that O₃ respiratory hospital admissions risk estimates remained robust to the inclusion of PM in copollutant models.

Note: Effect estimates are for a 20 ppb increase in 24-h; 30 ppb increase in 8-h max; and 40 ppb increase in 1-h max O₃ concentrations. HA=hospital admission; ED=emergency department. Black=All-year analysis; Red=Summer only analysis; Blue=Winter only analysis.

a) Wheeze used as indicator of lower respiratory disease.
b) APHENA-Canada results standardized to approximate IQR of 5.1 ppb for 1-h max O₃ concentrations.
c) Study included 8 cities; but of those 8, only 4 had O₃ data.
d) non-ICU effect estimates.
e) The study did not specify the lag day of the summer season estimate.

Figure 6-19 Percent increase in respiratory-related hospital admission and ED visits in studies that presented all-year and/or seasonal results.
<table>
<thead>
<tr>
<th>Study*</th>
<th>ED Visit or Hospital Admission</th>
<th>Location</th>
<th>Age</th>
<th>Lag</th>
<th>Avg Time</th>
<th>% Increase (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Respiratory</strong></td>
<td></td>
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<td><strong>All-year</strong></td>
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<td>Hospital Admission</td>
<td>Hong Kong, China</td>
<td>All</td>
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<td>8-h max</td>
<td>3.58 (1.90, 5.29)</td>
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<td>Hospital Admission</td>
<td>10 Canadian cities</td>
<td>All</td>
<td>1.2</td>
<td>24-h avg</td>
<td>4.38 (2.19, 6.46)</td>
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<td>Dales et al. (2006)</td>
<td>Hospital Admission</td>
<td>11 Canadian cities</td>
<td>0-27 days</td>
<td>2</td>
<td>24-h avg</td>
<td>5.41 (2.88, 7.96)</td>
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<td>ED Visit</td>
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<td>0-6</td>
<td>8-h max</td>
<td>-3.34 (-11.2, 5.28)</td>
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<td>APHENA-europe</td>
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<td>0-1</td>
<td>1-h max</td>
<td>1.58 (-1.71, 4.15)</td>
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<td>20.4 (4.07, 40.2)</td>
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<td>1-h max</td>
<td>2.4 (0.51, 4.40)</td>
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<td>All</td>
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<td>8-h max</td>
<td>2.08 (1.25, 2.91)</td>
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<td>0-2</td>
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<td>1-h max</td>
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<tr>
<td><strong>All-year</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Stieb et al. (2009)</td>
<td>ED Visit</td>
<td>7 Canadian cities</td>
<td>All</td>
<td>2</td>
<td>24-h avg</td>
<td>3.48 (0.33, 6.76)</td>
</tr>
<tr>
<td>Villeneuve et al. (2007)</td>
<td>ED Visit</td>
<td>Alberta, CAN</td>
<td>&gt;2</td>
<td>0-2</td>
<td>8-h max</td>
<td>11.9 (6.8, 17.2)</td>
</tr>
<tr>
<td>Strickland et al. (2010)</td>
<td>ED Visit</td>
<td>Atlanta, GA</td>
<td>Children</td>
<td>0-2</td>
<td>8-h max</td>
<td>6.38 (3.19, 9.57)</td>
</tr>
<tr>
<td><strong>Warm</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Silverman and Ito (2010)d</td>
<td>Hospital Admission</td>
<td>New York, NY</td>
<td>All</td>
<td>0-1</td>
<td>8-h max</td>
<td>12.5 (8.27, 16.7)</td>
</tr>
<tr>
<td>Ito et al. (2007b)</td>
<td>ED Visit</td>
<td>New York, NY</td>
<td>All</td>
<td>0-1</td>
<td>8-h max</td>
<td>16.9 (10.9, 23.4)</td>
</tr>
<tr>
<td>Villeneuve et al. (2007)</td>
<td>ED Visit</td>
<td>Alberta, Canada</td>
<td>&gt;2</td>
<td>0-2</td>
<td>8-h max</td>
<td>19.0 (11.9, 28.1)</td>
</tr>
<tr>
<td>Mar and Koenig (2009)</td>
<td>ED Visit</td>
<td>Seattle, WA</td>
<td>18+</td>
<td>2</td>
<td>8-h max</td>
<td>19.1 (3.00, 40.5)</td>
</tr>
<tr>
<td>Strickland et al. (2010)</td>
<td>ED Visit</td>
<td>Atlanta, GA</td>
<td>Children</td>
<td>0-2</td>
<td>8-h max</td>
<td>8.43 (4.42, 12.7)</td>
</tr>
<tr>
<td>Silverman and Ito (2010)d</td>
<td>Hospital Admission</td>
<td>New York, NY</td>
<td>6-18</td>
<td>0-1</td>
<td>8-h max</td>
<td>28.2 (15.3, 41.5)</td>
</tr>
<tr>
<td>Mar and Koenig (2009)</td>
<td>ED Visit</td>
<td>Seattle, WA</td>
<td>&lt;18</td>
<td>0</td>
<td>8-h max</td>
<td>33.1 (3.00, 68.5)</td>
</tr>
<tr>
<td>Study*</td>
<td>ED Visit or Hospital Admission</td>
<td>Location</td>
<td>Age</td>
<td>Lag</td>
<td>Avg Time</td>
<td>% Increase (95% CI)</td>
</tr>
<tr>
<td>----------------</td>
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</tr>
<tr>
<td><strong>Cold</strong></td>
<td></td>
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<td></td>
</tr>
<tr>
<td>Ito et al. (2007b)</td>
<td>ED Visit</td>
<td>New York, NY</td>
<td>All</td>
<td>0-1</td>
<td>8-h max</td>
<td>-23.4 (-27.3, -19.3)</td>
</tr>
<tr>
<td>Villeneuve et al. (2007)</td>
<td>ED Visit</td>
<td>Alberta, Canada</td>
<td>&gt;2</td>
<td>0-2</td>
<td>8-h max</td>
<td>8.50 (0.00, 17.2)</td>
</tr>
<tr>
<td>Strickland et al. (2010)</td>
<td>ED Visit</td>
<td>Atlanta, GA</td>
<td>Children</td>
<td>0-2</td>
<td>8-h max</td>
<td>4.52 (-0.82, 10.1)</td>
</tr>
<tr>
<td><strong>COPD</strong></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>All-year</td>
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</tr>
<tr>
<td>Stieb et al. (2009)</td>
<td>ED Visit</td>
<td>7 Canadian cities</td>
<td>All</td>
<td>2</td>
<td>24-h avg</td>
<td>4.03 (-0.54, 8.62)</td>
</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>Hospital Admission</td>
<td>36 U.S. cities</td>
<td>65+</td>
<td>0-1</td>
<td>8-h max</td>
<td>0.24 (-0.78, 1.21)</td>
</tr>
<tr>
<td>Yang et al. (2005b)</td>
<td>Hospital Admission</td>
<td>Vancouver, Canada</td>
<td>65+</td>
<td>0-3</td>
<td>24-h avg</td>
<td>8.80 (-12.5, 32.6)</td>
</tr>
<tr>
<td><strong>Warm</strong></td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Stieb et al. (2009)*</td>
<td>ED Visit</td>
<td>7 Canadian cities</td>
<td>All</td>
<td>NR</td>
<td>24-h avg</td>
<td>6.76 (0.11, 13.9)</td>
</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>Hospital Admission</td>
<td>36 U.S. cities</td>
<td>65+</td>
<td>0-1</td>
<td>8-h max</td>
<td>1.63 (0.48, 2.85)</td>
</tr>
<tr>
<td><strong>Cold</strong></td>
<td></td>
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</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>Hospital Admission</td>
<td>36 U.S. cities</td>
<td>65+</td>
<td>0-1</td>
<td>8-h max</td>
<td>-1.85 (-3.60, -0.06)</td>
</tr>
<tr>
<td><strong>Pneumonia</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>All-year</td>
<td></td>
<td></td>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Zanobetti and Schwartz (2006)</td>
<td>Hospital Admission</td>
<td>Boston, MA</td>
<td>65+</td>
<td>0-1</td>
<td>24-h avg</td>
<td>-5.96 (-11.1, -1.36)</td>
</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>Hospital Admission</td>
<td>36 U.S. cities</td>
<td>65+</td>
<td>0-1</td>
<td>8-h max</td>
<td>1.81 (-0.72, 4.52)</td>
</tr>
<tr>
<td><strong>Warm</strong></td>
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<td></td>
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</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>Hospital Admission</td>
<td>36 U.S. cities</td>
<td>65+</td>
<td>0-1</td>
<td>8-h max</td>
<td>2.49 (1.57, 3.47)</td>
</tr>
<tr>
<td><strong>Cold</strong></td>
<td></td>
<td></td>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Medina-Ramon et al. (2006)</td>
<td>Hospital Admission</td>
<td>36 U.S. cities</td>
<td>65+</td>
<td>0-1</td>
<td>8-h max</td>
<td>-4.88 (-6.59, -3.14)</td>
</tr>
</tbody>
</table>

*Includes studies in Figure 6-19.

*Wheeze used as indicator of lower respiratory disease.

tAPHEA-Canada results standardized to approximate IQR of 5.1 ppb for 1-h max O₃ concentrations.

cStudy included 8 cities, but of those 8 only 4 had O₃ data.

dNon-ICU effect estimates.

eThe study did not specify the lag day of the summer season estimate.
Note: Effect estimates are for a 20 ppb increase in 24-h; 30 ppb increase in 8-h max; and 40 ppb increase in 1-h max O₃ concentrations.

aStudies that examined hospital admissions.
bA study that examined ED visits.
cRisk estimates from APHENA -Canada standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O₃ concentrations. Black = results from single-pollutant models; Red = results from copollutant models with PM10 or PM2.5; Yellow = results from copollutant models with CO; Blue = results from copollutant models with NO2; Green = results from copollutant models with SO2.

Figure 6-20 Percent increase in respiratory-related hospital admissions and ED visits for studies that presented single and copollutant model results.
To date only a few studies have examined the C-R relationship between short-term O₃ exposure and respiratory-related hospital admissions and ED visits. A preliminary examination of the C-R relationship found no evidence of a deviation from linearity when examining the association between short-term O₃ exposure and asthma hospital admissions (Silverman and Ito, 2010). Additionally, an examination of the C-R relationship for O₃ exposure and pediatric asthma ED visits found no evidence of a threshold with elevated associations with O₃ at concentrations as low as 30 ppb (Silverman and Ito, 2010; Strickland et al., 2010). However, in both studies there is uncertainty in the shape of the C-R curve at the lower end of the distribution of O₃ concentrations due to the low density of data in this range.

### Table 6-29  Corresponding effect estimates for Figure 6-20.

<table>
<thead>
<tr>
<th>Study*,a</th>
<th>Location</th>
<th>Visit Type</th>
<th>Age</th>
<th>Lag</th>
<th>Copollutant</th>
<th>% Increase (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>All-year: Respiratory</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-U.S. Hospital Admission</td>
<td>65+</td>
<td>1</td>
<td></td>
<td>PM₁₀</td>
<td>2.62 (0.63, 4.64)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>2.14 (-0.08, 4.40)</td>
</tr>
<tr>
<td></td>
<td>APHENA-Europe</td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>2.94 (1.02, 4.89)</td>
</tr>
<tr>
<td></td>
<td>APHENA-Canada</td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>2.38 (0.08, 4.64)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>5.54 (-0.94, 12.4)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>0.69 (-0.12, 1.50)b</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>5.13 (-6.62, 18.6)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>0.64 (-0.87, 2.20)b</td>
</tr>
<tr>
<td><strong>COPD</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Yang et al. (2005b)</td>
<td>Vancouver Hospital Admission</td>
<td>65+</td>
<td>0-3</td>
<td></td>
<td>CO</td>
<td>8.80 (-12.5, 32.6)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>NO₂</td>
<td>22.8 (-2.14, 50.7)</td>
</tr>
<tr>
<td></td>
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<td></td>
<td></td>
<td></td>
<td>SO₂</td>
<td>11.1 (-10.4, 37.6)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>13.4 (-8.40, 40.2)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>11.1 (-8.40, 37.6)</td>
</tr>
<tr>
<td><strong>Summer: Asthma</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ito et al. (2007b)</td>
<td>New York ED All</td>
<td>0-1</td>
<td></td>
<td></td>
<td>CO</td>
<td>16.9 (10.9, 23.4)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>NO₂</td>
<td>18.1 (12.1, 24.5)</td>
</tr>
<tr>
<td></td>
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<td></td>
<td></td>
<td>SO₂</td>
<td>10.2 (4.29, 16.4)</td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₂.₅</td>
<td>13.1 (7.16, 19.5)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₂.₅</td>
<td>12.7 (6.37, 19.3)</td>
</tr>
</tbody>
</table>

*aStudies included in Figure 6-20.

Averaging times: Katsouyanni et al. (2009) = 1-h max; Yang et al. (2005b) = 24-h avg; and Ito et al. (2007b) = 8-h max.

bRisk estimates standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O₃ concentrations.
In totality, building upon the conclusions of the 2006 O₃ AQCD, the evidence from recent studies continues to support an association between short-term O₃ exposure and respiratory-related hospital admissions and ED visits. Additional evidence also supports stronger associations during the warm season for specific respiratory outcomes such as asthma and COPD.

### 6.2.8 Respiratory Mortality

The epidemiologic, controlled human exposure, and toxicological studies discussed within this section (Section 6.2) provide evidence for multiple respiratory effects in response to short-term O₃ exposure. Additionally, the evidence from experimental studies indicates multiple potential pathways of O₃-induced respiratory effects, which support the continuum of respiratory effects that could potentially result in respiratory-related mortality. The 2006 O₃ AQCD found inconsistent evidence for an association between short-term O₃ exposure and respiratory mortality (U.S. EPA, 2006b). Although some studies reported a strong positive association between O₃ exposure and respiratory mortality, additional studies reported a small association or no association. The majority of recent multicity studies found consistent positive associations between short-term O₃ exposure and respiratory mortality, specifically during the summer months.

The APHENA study, described earlier in Section 6.2.7.2, (Katsouyanni et al., 2009) also examined associations between short-term O₃ exposure and mortality and found consistent positive associations for respiratory mortality in all-year analyses, except in the Canadian data set for ages ≥ 75, with an increase in the magnitude of associations in analyses restricted to the summer season across data sets and age ranges. Additional multicity studies from the U.S. (Zanobetti and Schwartz, 2008b), Europe (Samoli et al., 2009), Italy (Stafoggia et al., 2010), and Asia (Wong et al., 2010) that conducted summer season and/or all-year analyses provide additional support for an association between short-term O₃ exposure and respiratory mortality (Figure 6-37).

Of the studies evaluated, only the APHENA study (Katsouyanni et al., 2009) and the Italian multicity study (Stafoggia et al., 2010) conducted an analysis of the potential for copollutant confounding of the O₃-respiratory mortality relationship. In the APHENA study, specifically the European dataset, focused on the natural spline model with 8 df/year (as discussed in Section 6.2.7.2) and lag 1 results (as discussed in Section 6.6.2.1), respiratory mortality risk estimates were robust to the inclusion of PM₁₀ in copollutant models in all-year analyses with O₃ respiratory mortality risk estimates increasing in the Canadian and U.S. datasets compared to single-pollutant model results. In summer season analyses, respiratory O₃ mortality risk estimates were robust in the U.S. dataset and attenuated in the European dataset. Similarly, in the Italian multicity study (Stafoggia et al., 2010), which was limited to the summer season, respiratory mortality risk estimates were attenuated in copollutant models with PM₁₀. Based on the APHENA and Italian multicity results, O₃ respiratory mortality risk estimates were robust to the inclusion of PM₁₀.
mortality risk estimates appear to be moderately to substantially sensitive (e.g., increased or attenuated) to inclusion of PM$_{10}$. However, in the APHENA study, the mostly every-6th-day sampling schedule for PM$_{10}$ in the Canadian and U.S. datasets greatly reduced their sample size and limits the interpretation of these results.

### 6.2.9 Summary and Causal Determination

The 2006 O$_3$ AQCD concluded that there was clear, consistent evidence of a causal relationship between short-term O$_3$ exposure and respiratory effects (U.S. EPA, 2006b). This conclusion was substantiated by evidence from controlled human exposure and toxicological studies indicating a range of respiratory effects in response to short-term O$_3$ exposure, including pulmonary function decrements and increases in respiratory symptoms, lung inflammation, lung permeability, and airway hyperresponsiveness. Toxicological studies provided additional evidence for O$_3$-induced impairment of host defenses. Combined, these findings from experimental studies provided support for epidemiologic evidence, in which short-term increases in ambient O$_3$ concentration were consistently associated with decreases in lung function in populations with increased outdoor exposures, children with asthma, and healthy children; increases in respiratory symptoms and asthma medication use in children with asthma; and increases in respiratory-related hospital admissions and asthma-related ED visits. Short-term increases in ambient O$_3$ concentration also were consistently associated with increases in all-cause and cardiopulmonary mortality; however, the contribution of respiratory causes to these findings was uncertain.

Building on the large body of evidence presented in the 2006 O$_3$ AQCD, recent studies support associations between short-term O$_3$ exposure and respiratory effects. Controlled human exposure studies continue to provide the strongest evidence for lung function decrements in young healthy adults over a range of O$_3$ concentrations. Studies previously reported mean O$_3$-induced FEV$_1$ decrements of 6-8% at 80 ppb O$_3$ (Adams, 2006a, 2003a; McDonnell et al., 1991; Horstman et al., 1990), and recent evidence additionally indicates mean FEV$_1$ decrements of 6% at 70 ppb O$_3$ (Schelegle et al., 2009) and 2-3% at 60 ppb O$_3$ (Kim et al., 2011; Brown et al., 2008; Adams, 2006a) (Section 6.2.1.1). In healthy young adults, O$_3$-induced decrements in FEV$_1$ do not appear to depend on sex (Hazucha et al., 2003), body surface area or height (McDonnell et al., 1997), lung size or baseline FVC (Messineo and Adams, 1990). There is limited evidence that blacks may experience greater O$_3$-induced decrements in FEV$_1$ than do age-matched whites (Que et al., 2011; Seal et al., 1993). Healthy children experience similar spirometric responses but lesser symptoms from O$_3$ exposure relative to young adults (McDonnell et al., 1985b). On average, spirometric and symptom responses to O$_3$ exposure appear to decline with increasing age beyond about 18 years of age (McDonnell et al., 1999b; Seal et al., 1996). There is also a tendency for slightly increased spirometric responses in subjects with mild asthma and subjects with allergic rhinitis relative to healthy young adults (Jorres et
al., 1996). Spirometric responses in subjects with asthma appear to be affected by baseline lung function, i.e., responses increase with disease severity (Horstman et al., 1995).

Available information from controlled human exposure studies on recovery from O₃ exposure indicates that an initial phase of recovery in healthy individuals proceeds relatively rapidly, with acute spirometric and symptom responses resolving within about 2 to 4 hours (Folinsbee and Hazucha, 1989). Small residual lung function effects are almost completely resolved within 24 hours. Effects of O₃ on the small airways persisting a day following exposure, assessed by persistent decrement in FEF_{25-75\%} and altered ventilation distribution, may be due in part to inflammation (Frank et al., 2001; Foster et al., 1997). In more responsive individuals, this recovery in lung function takes longer (as much as 48 hours) to return to baseline. Some cellular responses may not return to baseline levels in humans for more than 10-20 days following O₃ exposure (Devlin et al., 1997). Airway hyperresponsiveness and increased epithelial permeability are also observed as late as 24 hours post-exposure (Que et al., 2011).

With repeated O₃ exposures over several days, spirometric and symptom responses become attenuated in both healthy individuals and individuals with asthma, but this attenuation is lost after about a week without exposure (Gong et al., 1997a; Folinsbee et al., 1994; Kulle et al., 1982). Airway responsiveness also appears to be somewhat attenuated with repeated O₃ exposures in healthy individuals, but becomes increased in individuals with pre-existing allergic airway disease (Gong et al., 1997a; Folinsbee et al., 1994). Some indicators of pulmonary inflammation are attenuated with repeated O₃ exposures. However, other markers such as epithelial integrity and damage do not show attenuation, suggesting continued tissue damage during repeated O₃ exposure (Devlin et al., 1997).

Consistent with controlled human exposure study findings, epidemiologic evidence indicates that lung function decrements are related to short-term increases in ambient O₃ concentration (Section 6.2.1.2). As described in the 1996 and 2006 O₃ AQCDs, the most consistent observations were those in populations engaged in outdoor recreation, exercise, or work. Epidemiologic evidence also demonstrates that increases in ambient O₃ concentration are associated with decreases in lung function in children with asthma (Figure 6-7 [and Table 6-8] and Figure 6-8 [and Table 6-9]) and children in the general population (Figure 6-9 [and Table 6-12]). Evidence in adults with respiratory disease and healthy adults is inconsistent. In children with asthma, lung function mostly was found to decrease by <1-2% per unit increase in O₃ concentration. However, within studies of children with asthma, increases in ambient O₃ concentration (at the same or similar lag) were associated both with decrements in lung function and increases in respiratory symptoms (Just et al., 2002; Mortimer et al., 2002; Ross et al., 2002; Gielen et al., 1997; Romieu et al., 1997; Thurston et al., 1997; Romieu et al., 1996). Biological plausibility for O₃-associated decrements in lung function found in controlled human exposure, epidemiologic, and

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1 Effect estimates were standardized to a 40-ppb increase for 1-h max O₃, a 30-ppb increase for 8-h max O₃, and a 20-ppb increase for 24-h avg O₃.
animal studies is provided by the well-documented effects of O₃ on activation of bronchial C-fibers (Section 5.3.2).

Across disciplines, studies have examined factors that may potentially increase the risk of O₃-induced decrements in lung function. In the controlled human exposure studies, there is a large degree of intersubject variability in lung function decrements, symptomatic responses, pulmonary inflammation, airway hyperresponsiveness, and altered epithelial permeability in healthy adults exposed to O₃ (Que et al., 2011; Holz et al., 2005; McDonnell, 1996). The magnitudes of pulmonary inflammation, airway hyperresponsiveness, and increases in epithelial permeability do not appear to be correlated, nor are these responses to O₃ correlated with changes in lung function, suggesting that different mechanisms may be responsible for these processes (Que et al., 2011; Balmes et al., 1997; Balmes et al., 1996; Aris et al., 1995). However, these responses tend to be reproducible within a given individual over a period of several months indicating differences in the intrinsic responsiveness of individuals (Holz et al., 2005; Hazucha et al., 2003; Holz et al., 1999; McDonnell et al., 1985c).

Numerous reasons for differences in the risk of individuals to O₃ exposure have been reported in the literature. These include dosimetric and mechanistic differences (Section 5.4). Further, evidence in all three disciplines suggests a role for antioxidant defenses (i.e., vitamin supplementation, genetic variants in oxidative metabolizing enzymes) in modulating respiratory responses to O₃. The biological plausibility of these findings is provided by the well-characterized evidence for O₃ exposure leading to the formation of secondary oxidation products that subsequently activate neural reflexes that mediate lung function decrements (Section 5.2.3) and that initiate pulmonary inflammation (Section 5.3.3).

Recent controlled human exposure studies (Section 6.2.3.1) and toxicological studies (Section 6.2.3.3) also continue to demonstrate lung injury and inflammatory responses upon O₃ exposure. Evidence from more than a hundred toxicological studies clearly indicates that O₃ induces damage and inflammation in the lung, and studies continue to elucidate the mechanistic pathways involved (Section 5.3). Though inflammation may resolve, continued cellular damage may alter the structure and function of pulmonary tissues. Recent controlled human exposure studies support previous findings for pulmonary inflammation but demonstrate effects at 60 ppb O₃, the lowest concentration evaluated. Building on the extensive experimental evidence, recent epidemiologic studies, most of which were conducted in Mexico City, indicate ambient O₃-associated increases in pulmonary inflammation in children with asthma. Multiple studies examined and found increases in eNO (Berhane et al., 2011; Khatri et al., 2009; Barraza-Villarreal et al., 2008). In some studies of subjects with asthma, increases in ambient O₃ concentration at the same lag were associated with both increases in pulmonary inflammation and respiratory symptoms (Khatri et al., 2009; Barraza-Villarreal et al., 2008). Although more limited in number, epidemiologic studies also found associations with cytokines such as IL-6 or IL-8 (Barraza-Villarreal et al., 2008; Sienra-Monge et al., 2004), eosinophils (Khatri et al., 2009), antioxidants (Sienra-Monge et al., 2004), and indicators of oxidative stress (Romieu et al., 2008) (Section 6.2.3.2). This epidemiologic evidence is coherent with results from controlled human exposure and
toxicological studies that demonstrated an induction or reduction of these same endpoints after O₃ exposure.

The evidence for O₃-induced pulmonary inflammation and airway hyperresponsiveness, largely demonstrated in controlled human exposure and toxicological studies, provides mechanistic support for O₃-associated increases in respiratory symptoms observed in both controlled human exposure and epidemiologic studies. Controlled human exposure studies of healthy, young adults demonstrate increases in respiratory symptoms induced by O₃ exposures <80 ppb (Schelegle et al., 2009; Adams, 2006a) (Section 6.2.1.1). Adding to this evidence, epidemiologic studies find effects in children with asthma. Evidence from the previous large multiplicity NCICAS and a large body of single-city and -region studies indicates that short-term increases in ambient O₃ concentration are associated with increases in respiratory symptoms and asthma medication use in children with asthma (Section 6.2.4.1). Weak evidence is available from the few recent U.S. multiplicity studies; however, they examined either fewer person-days of data (Schildcrout et al., 2006) or longer lags of ambient O₃ exposure (19-day avg versus exposures lagged or averaged over a few days) than are supported by controlled human exposure, toxicological, and other epidemiologic studies (O'Connor et al., 2008). Several epidemiologic studies found associations between ambient O₃ concentrations and respiratory symptoms in populations with asthma that also had a high prevalence of allergy (52-100%) (Escamilla-Nuñez et al., 2008; Feo Brito et al., 2007; Romieu et al., 2006; Just et al., 2002; Mortimer et al., 2002; Ross et al., 2002; Gielen et al., 1997). The strong evidence in populations with asthma and allergy is supported by observations of O₃-induced inflammation in animal models of allergy (Section 6.2.3.3), and may be explained mechanistically by the action of O₃ to sensitize bronchial smooth muscle to hyperreactivity and thus, potentially act as a primer for subsequent exposure to antigens such as allergens (Section 5.3.5).

Modification of innate and adaptive immunity is emerging as a mechanistic pathway contributing to the effects of O₃ on asthma and allergic airways disease (Section 5.3.6). While the majority of evidence comes from animal studies, controlled human exposure studies have found differences between subjects with asthma and healthy controls in O₃-mediated innate and adaptive immune responses (Section 5.4.2.2), suggesting that these pathways may be relevant to humans and may lead to the induction and exacerbation of asthma (Alexis et al., 2010; Hernandez et al., 2010; Alexis et al., 2009; Bosson et al., 2003).

The subclinical and overt respiratory effects observed across disciplines, as described above, collectively provide support for epidemiologic studies that demonstrate consistently positive associations between short-term O₃ exposure and respiratory-related hospital admissions and ED visits (Section 6.2.7). Consistent with evidence presented in the 2006 O₃ AQCD, recent multiplicity studies and a multicontinent study (i.e., APHENA) (Katsouyanni et al., 2009) found risk estimates ranging from an approximate 1.6 to 5.4% increase in all respiratory-related hospital admissions and ED visits in all-year analyses per unit increase in ambient O₃ concentration (as described in Section 2.5). Positive associations persisted in analyses restricted to the
summer season, but the magnitude varied depending on the study location
(Figure 6-19). Compared with studies reviewed in the 2006 O\textsubscript{3} AQCD, a larger
number of recent studies examined hospital admissions and ED visits for specific
respiratory outcomes. Although limited in number, both single- and multi-city studies
consistently found positive associations between short-term O\textsubscript{3} exposures and asthma
and COPD hospital admissions and ED visits, with more limited evidence for
pneumonia. Consistent with the conclusions of the 2006 O\textsubscript{3} AQCD, in studies that
conducted seasonal analyses, risk estimates were elevated in the warm season
compared to cold season or all-season analyses, specifically for asthma and COPD.
Although recent studies did not include detailed age-stratified results, the increased
risk of asthma hospital admissions (Silverman and Ito, 2010; Strickland et al., 2010;
Dales et al., 2006) observed for children strengthens the conclusion from the 2006 O\textsubscript{3}
AQCD that children are potentially at increased risk of O\textsubscript{3}-induced respiratory
effects (U.S. EPA, 2006b). Although the C-R relationship has not been extensively
examined, preliminary examinations found no evidence of a threshold between short-
term O\textsubscript{3} exposure and asthma hospital admissions and pediatric asthma ED visits,
with uncertainty in the shape of the C-R curve at the lower limit of ambient
concentrations in the U.S. (Silverman and Ito, 2010; Strickland et al., 2010).

Recent evidence extends the potential range of well-established O\textsubscript{3}-associated
respiratory effects by demonstrating associations between short-term ambient O\textsubscript{3}
 exposure and respiratory-related mortality. In all-year analyses, a multicontinent
(APHENA) and multicity (PAPA) study consistently found positive associations
with respiratory mortality with evidence of an increase in the magnitude of
associations in analyses restricted to the summer months. Further, additional
multicity studies conducted in the U.S. and Europe provide evidence supporting
stronger O\textsubscript{3}-respiratory mortality associations during the summer season
(Section 6.2.8).

Several epidemiologic studies of respiratory morbidity and mortality evaluated the
potential confounding effects of copollutants, in particular, PM\textsubscript{10}, PM\textsubscript{2.5}, or NO\textsubscript{2}. In
most cases, effect estimates remained robust to the inclusion of copollutants.
In some studies of lung function and respiratory symptoms, larger effects were
estimated for O\textsubscript{3} when copollutants were added to models. Ozone effect estimates for
respiratory-related hospital admissions and ED visits remained relatively robust upon
the inclusion of PM and gaseous pollutants in two-pollutant models (Strickland et al.,
2010; Tolbert et al., 2007; Medina-Ramon et al., 2006). Although copollutant
confounding was not extensively examined in studies of cause-specific mortality, O\textsubscript{3}-
respiratory mortality risk estimates remained positive but were moderately to
substantially sensitive (e.g., increased or attenuated) to the inclusion of PM\textsubscript{10} in
copollutant models (Stafoggia et al., 2010; Katsouyanni et al., 2009). However,
interpretation of these results for respiratory mortality requires caution due to the
limited PM datasets used in these studies as a result of the every 3rd- or 6th-day PM
sampling schedule employed in most cities. Together, these copollutant-adjusted
findings across respiratory endpoints provide support for the independent effects of
short-term exposures to ambient O\textsubscript{3}. 
Across the respiratory endpoints examined in epidemiologic studies, associations were found using several different exposure assessment methods that likely vary in how well ambient O\textsubscript{3} concentrations represent ambient exposures and between-subject variability in exposures. Evidence clearly demonstrated O\textsubscript{3}-associated lung function decrements in populations with increased outdoor exposures for whom ambient O\textsubscript{3} concentrations measured on site of outdoor activity and/or at the time of outdoor activity have been more highly correlated and similar in magnitude to personal O\textsubscript{3} exposures (Section 4.3.3). However, associations with respiratory effects also were found with ambient O\textsubscript{3} concentrations expected to have weaker personal-ambient relationships, including those measured at home or school, measured at the closest site, averaged from multiple community sites, and measured at a single site. Overall, there was no clear indication that a particular method of exposure assessment produced stronger findings.

An additional consideration in the evaluation of the epidemiologic evidence is the impact of behavioral modifications on observed associations. A study demonstrated that the magnitude of O\textsubscript{3}-associated asthma hospitalizations in Los Angeles, CA was underestimated due to behavioral modification in response to forecasted air quality (Section 4.6.5). It is important to note that the study was limited to one metropolitan area and used air quality data for the years 1989-1997, when the O\textsubscript{3} concentration that determines the designation of an O\textsubscript{3} action day, was much higher than it is currently.

Both panel and time-series epidemiologic studies found increases in respiratory effects in association with increases in O\textsubscript{3} concentrations using various exposure metrics (i.e., 24-h avg, 1-h max, and 8-h max O\textsubscript{3} concentrations). A majority of studies examined and found associations of respiratory symptoms with 1-h max or 8-h max O\textsubscript{3} concentrations and associations of pulmonary inflammation with 8-h max or daytime avg O\textsubscript{3}. Within study comparisons of associations of lung function and respiratory symptoms among various exposure metrics yielded mixed evidence. Within some studies, larger effects were estimated for shorter O\textsubscript{3} averaging times whereas in other studies, larger effects were estimated for longer averaging times or no difference was found among averaging times. Comparisons in a limited number of time-series studies indicate rather comparable risk estimates across exposure metrics with some evidence indicating that 24-h avg O\textsubscript{3} was associated with a smaller increase in risk of respiratory ED visits (Section 6.2.7.3). Overall, there was no indication that the consistency or magnitude of the observed association was stronger for a particular O\textsubscript{3} exposure metric. In examination of the lag structure of associations, epidemiologic evidence for the range of respiratory endpoints clearly supports associations with ambient O\textsubscript{3} concentrations lagged 0 to 1 day, which is consistent with the O\textsubscript{3}-induced respiratory effects observed in controlled human exposure studies. Several epidemiologic studies also found increased respiratory morbidity in association with O\textsubscript{3} concentrations averaged over multiple days (2 to 5 days). Across respiratory endpoints examined in epidemiologic studies, there was not strong evidence that the magnitude of association was larger for any particular lag.
In summary, recent studies evaluated since the completion of the 2006 O₃ AQCD support and expand upon the strong body of evidence that indicated a causal relationship between short-term O₃ exposure and respiratory health effects. Controlled human exposure studies continue to demonstrate O₃-induced decreases in FEV₁ and pulmonary inflammation at concentrations as low as 60 ppb. Epidemiologic studies provide evidence that increases in ambient O₃ exposure can result in lung function decrements; increases in respiratory symptoms and pulmonary inflammation in children with asthma; increases in respiratory-related hospital admissions and ED visits; and increases in respiratory mortality. Recent toxicological studies demonstrating O₃-induced inflammation, airway hyperresponsiveness, and impaired lung host defense have continued to support the biological plausibility and modes of action for the O₃-induced respiratory effects observed in the controlled human exposure and epidemiologic studies. Additionally, recent epidemiologic studies affirm that respiratory morbidity and mortality associations are stronger during the warm/summer months and remain relatively robust after adjustment for copollutants. The recent evidence integrated across toxicological, controlled human exposure, and epidemiologic studies, along with the total body of evidence evaluated in previous AQCDs, is sufficient to conclude that there is a causal relationship between short-term O₃ exposure and respiratory health effects.

6.3 Cardiovascular Effects

Overall, there have been a relatively small number of studies that have examined the potential effect of short-term O₃ exposure on the cardiovascular system. This was reflected in the 1996 O₃ AQCD by the limited discussion on possible O₃-related cardiovascular effects. The 2006 O₃ AQCD (U.S. EPA, 2006b) built upon the limited evidence described in the 1996 O₃ AQCD (U.S. EPA, 1996a) and further explored the potential relationship between short-term O₃ exposure and cardiovascular outcomes. The 2006 O₃ AQCD concluded that “O₃ directly and/or indirectly contributes to cardiovascular-related morbidity” but added that the body of evidence was limited. This conclusion was based on a controlled human exposure study that included hypertensive adult males, a few epidemiologic studies of physiologic effects, heart rate variability, arrhythmias, myocardial infarctions, and hospital admissions, and toxicological studies of heart rate, heart rhythm, and blood pressure.

6.3.1 Controlled Human Exposure

Ozone reacts rapidly on contact with respiratory tract lining fluids and is not absorbed or transported to extrapulmonary sites to any significant degree as such. Controlled human exposure studies discussed in the previous AQCDs failed to demonstrate any consistent extrapulmonary effects. Some controlled human exposure studies have attempted to identify specific markers of exposure to O₃ in blood. Buckley et al. (1975) reported a 28% increase in serum α-tocopherol and a 26%
increase in erythrocyte fragility in healthy males immediately following exposure to 500 ppb O₃ for 2.75 hours with exercise (unspecified activity level). However, in healthy adult males exposed during exercise ($\dot{V}_E = 44$ L/min) to 323 ppb O₃ (on average) for 130 min on 3 consecutive days, Foster et al. (1996) found a 12% reduction in serum $\alpha$-tocopherol 20 hours after the third day of O₃ exposure. Liu et al. (1999); (1997) used a salicylate metabolite, 2,3, dehydroxybenzoic acid (DHBA), to indicate increased levels of hydroxyl radical which hydroxylates salicylate to DHBA. Increased DHBA levels after exposure to 120 and 400 ppb suggest that O₃ increases production of hydroxyl radical. The levels of DHBA were correlated with changes in spirometry. Interestingly, simultaneous exposure of healthy adults to O₃ (120 ppb for 2 hours at rest) and concentrated ambient particles (CAPs) resulted in a diminished systemic IL-6 response compared with exposure to CAPs alone (Urch et al., 2010).

Devlin et al. (2012) recently evaluated systemic and cardiovascular responses in a group of young healthy adults (20 M, 3 F; median age 28.8 yrs) exposed to O₃ (300 ppb; 2 hours with alternating 15 min periods of rest and moderate-to-heavy exercise [$\dot{V}_E = 25$ L/min per BSA]). Relative to FA responses, immediately following the O₃ exposure there was an 85% increase in blood IL-8 (p< 0.025). There were also trends (p<0.10) for O₃-induced increases in blood IL-1β (56%) and blood TNF-α (10%). At 24 hrs postexposure, there were significant (p=0.025) increases blood IL-1β (65%) and CRP (104%). Beyond these markers of systemic inflammation, there were also changes in biomarkers of vascular effects following O₃ exposure. There were significant (p< 0.025) O₃-induced decreases in plasminogen activator inhibitor-1 (PAI-1) by 33% immediately following exposure and by 43% at 24 hrs postexposure. Plasminogen levels were also decreased by42% at 24 hrs post exposure (p<0.05). Finally, there was a tendency (p=0.065) for a 44% increase in tissue-type plasminogen activator (tPA). Based on the combination of an increase in tPA and a decrease in PAI-1, the authors suggested that O₃ exposure may activate the fibronolyisis system. Until replicated at an O₃ concentration more typical of ambient exposures, the results of this study and other high O₃ concentration exposure studies should be interpreted with caution.

Gong et al. (1998) exposed hypertensive (n = 10) and healthy (n = 6) adult males, 41 to 78 years of age, to FA and on the subsequent day to 300 ppb O₃ for 3 hours with intermittent exercise ($\dot{V}_E = 30$ L/min). The overall results did not indicate any major acute cardiovascular effects of O₃ in either the hypertensive individuals or healthy controls. Statistically significant O₃ effects for both groups combined were increases in heart rate, rate-pressure product, and the alveolar-to-arterial PO₂ gradient, suggesting that impaired gas exchange was being compensated for by increased myocardial work. The mechanism for the decrease in arterial oxygen tension in the Gong et al. (1998) study could be due to an O₃-induced ventilation-perfusion mismatch. Gong et al. (1998) suggested that by impairing alveolar-arterial oxygen transfer, the O₃ exposure could potentially lead to adverse cardiac events by decreasing oxygen supply to the myocardium. The subjects in the Gong et al. (1998) study had sufficient functional reserve so as to not experience significant ECG changes or myocardial ischemia and/or injury. In studies evaluating the exercise
performance of healthy adults, no significant effect of O₃ on arterial O₂ saturation has been observed (Schelegle and Adams, 1986).

Fakhri et al. (2009) evaluated changes in HRV among adult volunteers (n = 50; 27 ± 7 years) during 2-hour resting exposures to PM₂.₅ CAPs (127 ± 62 μg/m³) and O₃ (114 ± 7 ppb), alone and in combination. High frequency HRV was increased following CAPs-only (p = 0.046) and O₃-only (p = 0.051) exposures, but not in combination. The standard deviation of NN intervals and the square root of the mean squared differences of successive NN intervals also showed marginally significant (0.05 < p < 0.10) increase due to O₃ but not CAPS. Ten of the subjects in this study were characterized as “mildly” asthmatic, however, asthmatic status was not found to modify these effects. Power et al. (2008) also investigated HRV in a small group of mild-to-moderate allergic asthmatics (n = 5; mean age = 37 years) exposed for 4 hours during moderate intermittent exercise to FA, carbon and ammonium nitrate particles (313 ± 20 μg/m³), and carbon and ammonium nitrate particles (255 ± 37 μg/m³) + O₃ (200 ppb). Changes in frequency-domain variables for the particle and particle + O₃ exposures were not statistically significant compared with FA. Seemingly in contrast to Fakhri et al. (2009), the standard deviation of NN intervals and the square root of the mean squared differences of successive NN intervals also showed a significant (p = 0.01) decrease for both the particle and particle + O₃ exposures relative to FA responses. Using a similar protocol, Sivagangabalan et al. (2011) concluded that spatial dispersion of cardiac repolarization was most affected by the combined pollutant exposure of CAP + O₃ compared to FA in healthy adults.

In healthy young adults (20 M, 3 F; median age 28.8 yrs), Devlin et al. (2012) recently reported an O₃-induced reduction in high frequency HRV by 51% (p<0.025) immediately following O₃ exposure (300 ppb for 2 hr with intermittent exercise) that appeared to persist to 24 hrs postexposure (38% decrease, p<0.10). A small, 1.2% increase in the QT interval immediately after O₃ exposure relative to FA exposures was also observed. The authors suggested that changes in HRV and repolarization were likely mediated by nerve fibers that terminate in the lung. Changes in FEV₁ due to O₃ exposure are also mediated by C-fibers in the lung. There was an O₃-induced FEV₁ decrement of 11% in the Devlin et al. (2012) study, whereas the resting O₃ exposure used by Fakhri et al. (2009) is predicted to cause very small (~0.3%) decrements in FEV₁ (McDonnell et al., 2007). The induction of nerve fiber mediated responses may, in part, explain the reduction in high frequency HRV following a high level of exposure (300 ppb for 2 hr with intermittent exercise) in the Devlin et al. (2012) versus the increase in high frequency HRV observed following a lower level of exposure (120 ppb for 2hr during rest) by Fakhri et al. (2009).

Diastolic blood pressure increased by 2 mmHg following the combined O₃ + CAPs exposure, but was not altered by either O₃ or CAPs alone in the Fakhri et al. (2009) study. For a subset of the subjects without asthma in the Fakhri et al. (2009) study, Urch et al. (2005) previously reported a 6 mmHg increase in diastolic blood pressure following a 2-hour resting exposure to O₃ (120 ppb) + PM₂.₅ CAPs (150 μg/m³) in healthy adults (n = 23; 32 ± 10 years), which was statistically different from the 1 mmHg increase seen following FA exposure. Brook et al. (2002) found O₃
(120 ppb) + PM$_{2.5}$ CAPs (150 μg/m$^3$) in healthy adults (n = 25; 35 ± 10 years) caused brachial artery vasoconstriction. However, minimal change in diastolic blood pressure (0.9 mmHg increase) relative to FA (0.4 mmHg decrease) was observed. More recently, Sivagangabalan et al. (2011) observed a 4.2 mmHg increase in diastolic blood pressure following a 2-hour resting exposure to O$_3$ (110 ppb) + PM$_{2.5}$ CAPs (150 μg/m$^3$) in healthy adults (n = 25; 27 ± 8 years), which was statistically different from the 1.7 mmHg increase seen following the FA exposure. The CAP exposure alone also caused a 3 mmHg increase in diastolic blood pressure which was significantly more than following FA. However, similar to FA, the O$_3$ exposure alone caused a 1.8 mmHg increase in diastolic blood pressure. Overall, these studies indicate an effect of CAPs and CAP + O$_3$, but not O$_3$ alone, on diastolic blood pressure.

### 6.3.2 Epidemiology

The 2006 O$_3$ AQCD concluded that the “generally limited body of evidence is highly suggestive that O$_3$ directly and/or indirectly contributes to cardiovascular-related morbidity,” including physiologic effects (e.g., release of platelet activating factor [PAF]), HRV, arrhythmias, and myocardial infarctions, although the available body of evidence reviewed during the 2006 O$_3$ AQCD does not “fully substantiate links between ambient O$_3$ exposure and adverse cardiovascular outcomes” (U.S. EPA, 2006b). Since the completion of the 2006 O$_3$ AQCD an increasing number of studies have examined the relationship between short-term O$_3$ exposure and cardiovascular morbidity and mortality. These recent studies, as well as evidence from the previous AQCDs, are presented within this section.

#### 6.3.2.1 Arrhythmia

In the 2006 O$_3$ AQCD, conflicting results were observed when examining the effect of O$_3$ on arrhythmias (Dockery et al., 2005; Rich et al., 2005). A study by Dockery et al. (2005) reported no association between O$_3$ concentration and ventricular arrhythmias among patients with implantable cardioverter defibrillators (ICD) living in Boston, MA, although when O$_3$ concentration was categorized into quintiles, there was weak evidence of an association with increasing O$_3$ concentration (median O$_3$ concentration: 22.9 ppb). Rich et al. (2005) performed a re-analysis of this cohort using a case-crossover design and detected a positive association between O$_3$ concentration and ventricular arrhythmias. Recent studies were conducted in various locations and each used a different cardiac episode to define an arrhythmic event and a different time period of exposure, which may help explain observed differences across studies. Study-specific characteristics and air quality data for recent studies are reported in Table 6-30.
Multiple studies examined O$_3$-related effects on individuals with ICDs. A study of 518 ICD patients who had at least 1 tachyarrhythmia within a 10-year period (totaling 6,287 tachyarrhythmic event-days; 1993-2002) was conducted in Atlanta, Georgia (Metzger et al., 2007). Tachyarrhythmic events were defined as any ventricular tachyarrhythmic event, any ventricular tachyarrhythmic event that resulted in electrical therapy, and any ventricular tachyarrhythmic event that resulted in defibrillation. In the primary analysis, no evidence of an association was observed for a 30 ppb increase in 8-h max O$_3$ concentrations and tachyarrhythmic events (OR: 1.00 [95% CI: 0.92, 1.08]; lag 0). Season-specific as well as several sensitivity analyses (including the use of an unconstrained distributed lag model [lags 0-6]) were conducted resulting in similar null associations.

In a case-crossover analysis, a population of ICD patients in Boston, MA, previously examined by (Rich et al., 2005) was used to assess the association between air pollution and paroxysmal atrial fibrillation (PAF) episodes (Rich et al., 2006b). In addition to ventricular arrhythmias, ICD devices may also detect supraventricular arrhythmias, of which atrial fibrillation is the most common. Although atrial fibrillation is generally not considered lethal, it has been associated with increased premature mortality as well as hospitalization and stroke. Ninety-one electrophysiologist-confirmed episodes of PAF were ascertained among 29 patients. An association (OR: 3.86 [95% CI: 1.44, 10.28] per 40 ppb increase in 1-h max O$_3$ concentrations) was observed between increases in O$_3$ concentration during the concurrent hour (lag 0-hour) and PAF episodes. The estimated OR for the 24-hour moving average concentration was elevated (OR: 1.81 [95% CI: 0.86, 3.83] per 20 ppb), but weaker than the estimate for the shorter exposure window. The association between PAF and O$_3$ concentration in the concurrent hour during the

### Table 6-30 Characterization of O$_3$ concentrations (in ppb) from studies of arrhythmias.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Averaging Time</th>
<th>Mean Concentration (Standard Deviation)</th>
<th>Upper Range of Concentration</th>
</tr>
</thead>
<tbody>
<tr>
<td>Metzger et al. (2007)</td>
<td>Atlanta, GA</td>
<td>8-h max Summer only</td>
<td>53.9 (23)</td>
<td>Max: 148</td>
</tr>
<tr>
<td>Rich et al. (2006b)</td>
<td>Boston, MA</td>
<td>1-h</td>
<td>22.2*</td>
<td>75th: 33 Max: 119.5</td>
</tr>
<tr>
<td></td>
<td></td>
<td>24-h</td>
<td>22.6*</td>
<td>75th: 30.9 Max: 77.5</td>
</tr>
<tr>
<td>Rich et al. (2006a)</td>
<td>St. Louis, MO</td>
<td>24-h</td>
<td>21*</td>
<td>75th: 31</td>
</tr>
<tr>
<td>Anderson et al. (2010)</td>
<td>London, England</td>
<td>8-h max</td>
<td>8.08</td>
<td>75th: 11.5</td>
</tr>
<tr>
<td>Sarnat et al. (2006b)</td>
<td>Steubenville, OH</td>
<td>24-h Summer and Fall only</td>
<td>21.8 (12.6)</td>
<td>75th: 28.5 Max: 74.8</td>
</tr>
<tr>
<td></td>
<td></td>
<td>5 days</td>
<td>22.2 (9.1)</td>
<td>75th: 29.1 Max: 44</td>
</tr>
</tbody>
</table>

Note: Median presented (information on mean not given); studies presented in order of first appearance in the text of this section.
cold months was comparable to that during the warm months. In addition, no
evidence of a deviation from linearity between $O_3$ concentration and the log odds of
PAF was observed. Authors report that the difference between $O_3$ concentration and
observed effect between this study ($PAF$ and 1-hour $O_3$) and their previous study
(ventricular arrhythmias and 24-hour moving average $O_3$) (Rich et al., 2005) suggest
a more rapid response to air pollution for $PAF$ (Rich et al., 2006b).

In an additional study, Rich et al. (2006a) employed a case-crossover design to
examine the association between air pollution and 139 confirmed ventricular
arrhythmias among 56 ICD patients in St Louis, Missouri. The authors observed a
positive association with $O_3$ concentration (OR: 1.17 [95% CI: 0.58, 2.38] per 20 ppb
increase in 24-hour moving avg $O_3$ concentrations [lags 0-23 hours]). Although the
authors concluded these results were similar to their results from Boston, MA (Rich
et al., 2005), they postulated that the pollutants responsible for the increased risk in
ventricular arrhythmias are different ($O_3$ and $PM_{2.5}$ in Boston and sulfur dioxide in
St Louis).

Anderson et al. (2010) used a case-crossover framework to assess air pollution and
activation of ICDs among patients from all 9 ICD clinics in the London National
Health Service hospitals. “Activation” was defined as tachycardias for which the
defibrillator delivered treatment. Investigators modeled associations using
unconstrained distributed lags from 0 to 5 days. The sample consisted of 705 patients
with 5,462 activation days ($O_3$ concentration information was for 543 patients and
4,092 activation days). Estimates for the association with $O_3$ concentration were
consistently positive, although weak (OR: 1.09 [95% CI: 0.76, 1.55] per 30 ppb
increase in 8-h max $O_3$ concentrations at 0-1 day lag; OR: 1.04 [95% CI: 0.60, 1.81]
per 30 ppb increase in 8-h max $O_3$ concentrations at 0-5 day lag) (Anderson et al.,
2010).

In contrast to arrhythmia studies conducted among ICD patients, Sarnat et al. (2006b)
recruited non-smoking adults (age range: 54-90 years) to participate in a study of air
pollution and arrhythmias conducted over two 12-week periods during summer and
fall of 2000 in a region characterized by industrial pollution (Steubenville, Ohio).
Continuous ECG data acquired on a weekly basis over a 30-minute sampling period
were used to assess ectopy, defined as extra cardiac depolarizations within the atria
(supraventricular ectopy, SVE) or the ventricles (ventricular ectopy, VE). Increases
in the 5-day moving average (days 1-5) of $O_3$ concentration were associated with an
increased odds of SVE (OR: 2.17 [95% CI: 0.93, 5.07] per 20 ppb increase in
24-h avg $O_3$ concentrations). A weaker association was observed for VE (OR: 1.62
[95% CI: 0.54, 4.90] per 20 ppb increase in 24-h avg $O_3$ concentrations). The results
of the effect of 5-day $O_3$ concentration on SVE were robust to the inclusion of $SO_4^{2-}$
in the model [OR: 1.62 (95% CI: 0.54, 4.90)]. The authors indicate that the strong
associations observed at the 5-day moving averages, as compared to shorter time
periods, suggests a relatively long-acting mechanistic pathways, such as
inflammation, may have promoted the ectopic beats in this population (Sarnat et al.,
2006b).
Although many studies report positive associations, collectively, studies of arrhythmias report inconsistent results. This may be due to variation in study populations, length and season of averaging time, and outcome under study.

### 6.3.2.2 Heart Rate/Heart Rate Variability

In the 2006 O₃ AQCD, two large population-based studies of air pollution and HRV were summarized (Park et al., 2005b; Liao et al., 2004a). In addition, the biological mechanisms and potential importance of HRV were discussed. Briefly, the study of acute effects of air pollution on cardiac autonomic control is based on the hypothesis that increased air pollution levels may stimulate the autonomic nervous system and lead to an imbalance of cardiac autonomic control characterized by sympathetic activation unopposed by parasympathetic control (U.S. EPA, 2006b). Examples of HRV indices include the standard deviation of normal-to-normal intervals (SDNN), the square root of the mean of the sum of the squares of differences between adjacent NN intervals (r-MSSD), high-frequency power (HF), low-frequency power (LF), and the LF/HF ratio. Liao et al. (2004a) examined the association between air pollution and cardiac autonomic control in the fourth cohort examination (1996-1998) of the U.S.-based Atherosclerosis Risk in Communities Study. A decrease in log-transformed HF was associated with an increase in O₃ concentration among white study participants. Park et al. (2005b) examined the effects of air pollution on indices of HRV in a population-based study among men from the Normative Aging Study in Boston, Massachusetts. Several associations were observed with O₃ concentration and HRV outcomes. A reduction in LF was associated with increased O₃ concentration, which was robust to inclusion of PM₂.₅. The associations with all HRV indices and O₃ concentration were stronger among those with ischemic heart disease and hypertension. In addition to the population-based studies included in the 2006 O₃ AQCD was a study by Schwartz et al. (2005), who conducted a panel study to assess the relationship between exposure to summertime air pollution and HRV. A weak association of O₃ concentration during the hour immediately preceding the health measures was observed with r-MSSD among a study population that consisted of mostly older female participants. In summary, these studies suggest that short-term exposures to ambient O₃ concentrations are predictors of decreased HRV and that the relationship may be stronger among certain subgroups. More recent studies that examined the association between O₃ concentration and HRV are described below. Study-specific characteristics and O₃ concentrations for these studies are presented in Table 6-31.
Several follow-up examinations of HRV were conducted among the participants of the Normative Aging Study in Boston, Massachusetts. A trajectory cluster analysis was used to assess whether pollution originating from different locations had varying relationships with HRV (Park et al., 2007). Subjects who were examined on days when air parcels originated in the west had the strongest associations with O₃; however, the O₃ concentration in this cluster was low (24-h avg, 17.0 ppb) compared to the other clusters (24-h avg of 21.3-29.1 ppb). LF and SDNN decreased with increases in the 4-hour moving average of O₃ concentration from the west (LF decreased by 51.2% [95% CI: 1.6, 75.9%] and SDNN decreased by 28.2% [95% CI: -0.5, 48.7%] per 30 ppb increase in 4-h avg O₃ concentrations) (Park et al., 2007). The Boston air mass originating in the west traveled over Illinois, Indiana, and Ohio; states typically characterized by coal-burning power plants. Due to the low O₃ concentrations observed in the west cluster, the authors hypothesize that O₃ concentration on those days could be capturing the effects of other, secondary and/or transported pollutants from the coal belt or that the relationship between ambient O₃ concentration and personal exposure to O₃ is stronger during that period (supported by a comparatively low apparent temperature which could indicate a likelihood to keep windows open and reduced air conditioning use) (Park et al., 2007).

Table 6-31  Characterization of O₃ concentrations (in ppb) from studies of heart rate variability.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Averaging Time</th>
<th>Mean Concentration (Standard Deviation)</th>
<th>Upper Range of Concentration</th>
</tr>
</thead>
<tbody>
<tr>
<td>Park et al. (2007)</td>
<td>Boston, MA</td>
<td>24-h</td>
<td>Range of 17.0-29.1</td>
<td></td>
</tr>
<tr>
<td>Park et al. (2008)</td>
<td>Boston, MA</td>
<td>24-h</td>
<td>23.4 (13)</td>
<td></td>
</tr>
<tr>
<td>Baja et al. (2010)</td>
<td>Boston, MA</td>
<td>0 lag</td>
<td>23 (16)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>10-h lag</td>
<td>21 (15)</td>
<td></td>
</tr>
<tr>
<td>Wheeler et al. (2006)</td>
<td>Atlanta, GA</td>
<td>4-h</td>
<td>18.5</td>
<td>75th: 22.5</td>
</tr>
<tr>
<td></td>
<td></td>
<td>24-h</td>
<td>29.4</td>
<td></td>
</tr>
<tr>
<td>Zanobetti et al. (2010)</td>
<td>Boston, MA</td>
<td>0.5-h</td>
<td>20.7*</td>
<td>75th: 30.33</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2-h</td>
<td>20.5*</td>
<td>75th: 30.08</td>
</tr>
<tr>
<td></td>
<td></td>
<td>3-D</td>
<td>21.9*</td>
<td>75th: 28.33</td>
</tr>
<tr>
<td></td>
<td></td>
<td>5-D</td>
<td>22.8*</td>
<td>75th: 29.28</td>
</tr>
<tr>
<td>Ruidavets et al. (2005a)</td>
<td>Toulouse, France</td>
<td>8-h max</td>
<td>38.3 (14.8)</td>
<td>75th: 46.9</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Max: 80.3</td>
</tr>
<tr>
<td>Hampel et al. (2012)</td>
<td>Augsburg, Germany</td>
<td>1-h avg</td>
<td>23.4 (17.0)</td>
<td>75th: 35.2</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Max: 80.6</td>
</tr>
<tr>
<td>Chan et al. (2005a)</td>
<td>Taipei, Taiwan</td>
<td>1-h</td>
<td>21.9 (15.4)</td>
<td>Max: 114.9</td>
</tr>
<tr>
<td>Wu et al. (2010)</td>
<td>Taipei, Taiwan</td>
<td>Working period</td>
<td>24.9 (14.0)</td>
<td>Max: 59.2</td>
</tr>
<tr>
<td>Chuang et al. (2007a)</td>
<td>Taipei, Taiwan</td>
<td>24-h</td>
<td>28.4 (12.1)</td>
<td>Max: 49.3</td>
</tr>
<tr>
<td></td>
<td></td>
<td>48-h</td>
<td>33.3 (8.9)</td>
<td>Max: 47.8</td>
</tr>
<tr>
<td></td>
<td></td>
<td>72-h</td>
<td>33.8 (7.1)</td>
<td>Max: 48.3</td>
</tr>
<tr>
<td>Chuang et al. (2007b)</td>
<td>Taipei, Taiwan</td>
<td>1-h</td>
<td>35.1</td>
<td>Max: 192.0</td>
</tr>
</tbody>
</table>

*Note: Median presented (information on mean not given); studies presented in order of first appearance in the text of this section.
An additional follow-up evaluation using the Normative Aging Study examined the potential for effect modification by chronic lead (Pb) exposure on the relationship between air pollution and HRV (Park et al., 2008). Authors observed graded reductions in HF and LF of HRV in relation to O₃ (and sulfate) concentrations across increasing quartiles of tibia and patella lead (HF: percent change 32.3% [95% CI: -32.5, 159.3] for the first quartile of tibia Pb and -59.1 [95% CI: -77.3, -26.1] for the fourth quartile of tibia Pb per 30 ppb increase in 4-h avg O₃ concentrations; LF: percent change 8.0% [95% CI: -36.9, 84.9] for the first quartile of tibia Pb and -59.3 [95% CI: -74.6, -34.8] for the fourth quartile of tibia Pb per 30 ppb increase in 4-h avg O₃ concentrations). In addition, associations were similar when education and cumulative traffic-adjusted bone Pb levels were used in analyses. Authors indicate the possibility that O₃ (which has low indoor concentrations) was acting as a proxy for sulfate (correlation coefficient for O₃ and sulfate = 0.57). Investigators of a more recent follow-up to the Normative Aging Study hypothesized that the relationships between short-term air pollution exposures and ventricular repolarization, as measured by changes in the heart-rate corrected QT interval (QTc), would be modified by participant characteristics (e.g., obesity, diabetes, smoking history) and genetic susceptibility to oxidative stress (Baja et al., 2010). No evidence of an association between O₃ concentration (using a quadratic constrained distributed lag model and hourly exposure lag models over a 10-hour time window preceding the visit) and QTc was reported (change in mean QTc -0.74 [95% CI: -3.73, 2.25]); therefore, potential effect modification of personal and genetic characteristics with O₃ concentration was not assessed (Baja et al., 2010). Collectively, the results from studies that examined the Normative Aging Study cohort found an association between increases in short-term O₃ concentration and decreases in HRV (Park et al., 2008; Park et al., 2007; Park et al., 2005b) although not consistently in all of the studies (Baja et al., 2010). Further, observed relationships appear to be stronger among those with ischemic heart disease, hypertension, and elevated bone lead levels, as well as when air masses arrive from the west (the coal belt). However, it is not clear if O₃ concentration is acting as a proxy for other, secondary particle pollutants (such as sulfate) (Park et al., 2008). In addition, since the Normative Aging Study participants were older, predominately white men, results may not be generalizable to the a large proportion of the U.S. population.

Additional studies of populations not limited to the Normative Aging Study have also examined associations between O₃ exposure and HRV. A panel study among 18 individuals with COPD and 12 individuals with recent myocardial infarction (MI) was conducted in Atlanta, Georgia (Wheeler et al., 2006). HRV was assessed for each participant on 7 days in fall 1999 and/or spring 2000. Ozone concentrations were not associated with HRV (SDNN) among all subjects (percent change of 2.36% [95% CI: -10.8%, 17.5%] per 30 ppb 4-hour O₃ increase) or when stratified by disease type (COPD, recent MI, and baseline FEV₁) (Wheeler et al., 2006).

HRV and air pollution was assessed in a panel study among 46 predominately white male patients (study population: 80.4% male, 93.5% white) aged 43-75 years in Boston, Massachusetts, with coronary artery disease (Zanobetti et al., 2010). Up to four home visits were made to assess HRV over the year following the index event.
Pollution lags used in analyses ranged between 30 minutes to a few hours and up to 5 days prior to the HRV assessments, calculated from hourly O₃ measurements averaged over three monitoring sites in Boston. Decreases in r-MSSD were reported for all averaging times of O₃ concentration (percent change of -5.18% [95% CI: -7.89, -2.30] per 20 ppb of 5-day moving average of O₃ concentration), but no evidence of an association between O₃ concentration and HF was observed (quantitative results not provided). In two-pollutant models with O₃ and either PM$_{2.5}$ or BC, O₃ associations remained robust.

A few recent studies were conducted outside of the U.S. in Europe (Hampel et al., 2012; Ruidavets et al., 2005a) and Asia (Wu et al., 2010; Chuang et al., 2007b; Chuang et al., 2007a; Chan et al., 2005a; Ruidavets et al., 2005a) that also examined the relationship between air pollution concentrations and heart rate and HRV. No consistent relationships were identified between O₃ concentration and resting heart rate among middle-aged (35-64 years) participants residing in Toulouse, France (Ruidavets et al., 2005a). A negative trend was reported for the 3-day cumulative (lag days 1-3) concentration of 8-h max O₃ with heart rate (p for trend = 0.02); however, the individual odds ratios comparing quintiles of exposure showed no association (OR for O₃ concentration of 0.93 [95% CI: 0.86, 1.01] for the highest quintile of resting heart rate compared to the lowest). When stratified by current smoking status, non-smokers had a decreased trend with increased 3-day cumulative O₃ concentrations but none of the quintiles for heart rate were statistically significant. In a panel study conducted in Augsburg, Germany, Hampel et al. (2012) examined the effect of short-term O₃ exposures on measures of heart rate and repolarization in individuals with type 2 diabetes or impaired glucose tolerance and healthy individuals with a potential genetic predisposition. A ~1% increase in HR was observed for individuals with type 2 diabetes and impaired glucose tolerance at concurrent and lag 1-4 hours for an approximate 10 ppb increase in O₃ concentrations\(^1\); no effect was observed for healthy individuals. These associations remained robust in copollutants models with sulfate, PM, and ultrafine particles. Additionally, there was evidence of T-wave flattening across all lags in healthy individuals and those with type 2 diabetes and impaired glucose intolerance, with the effect strongest in these individuals at concurrent (-1.31% [95% CI: -2.19, -0.42]) and lag 1 hour (-1.32% [95% CI: -2.19, 0.45]). Similarly, there was evidence of an increase in T-wave complexity for all participants across all lags examined, with the strongest effects again for individuals with type 2 diabetes and impaired glucose tolerance, but at lags 1 and 2 hours. An increase in T wave complexity for healthy participants at lags of 3 and 4 hours. Hampel et al. (2012) also found evidence of effect modification for each of the heart rate and repolarization metrics when taking into consideration the location and season in which the ECG recordings were obtained, with greater effects occurring when measurements were taken outdoors during the summer.

\(^1\) These results were not standardized to a 1-h max O₃ concentration of 40 ppb because the study examined hourly changes in heart rate parameters. Using an increment of 40 ppb would not appropriately represent the potential hourly change in O₃ concentrations.
In a study conducted in Taipei, Taiwan no associations were reported between O$_3$ concentration and HRV among CHD patients and patients with one or more major CHD risk factors (Chan et al., 2005a). Another study in Taipei, Taiwan examined mail carriers and reported O$_3$ concentration measured using personal monitors. No association was observed between O$_3$ concentration and the measures of HRV (percent change for SDNN: 0.57 [95% CI: -21.27, 28.46], r-MSSD: -7.10 [95% CI: -24.24, 13.92], HF: -1.92 [95% CI: -23.68, 26.02], LF: -4.82 [95% CI: -25.34, 21.35] per 40 ppb O$_3$) (Wu et al., 2010). A panel study was conducted in Taiwan to assess the relationship between air pollutants and inflammation, oxidative stress, blood coagulation, and autonomic dysfunction (Chuang et al., 2007b; Chuang et al., 2007a). Participants were apparently healthy college students (aged 18-25 years) who were living in a university dormitory in metropolitan Taipei. Health endpoints were measured three times from April to June in 2004 or 2005. Ozone concentration was assessed in statistical models using the average of the 24, 48, and 72 hours before the hour of each blood sampling. Decreases in HRV (measured as SDNN, r-MSSD, LF, and HF) were associated with increases in O$_3$ concentrations in single-pollutant models (percent change for SDNN: -13.45 [95% CI: -16.26, -10.60], r-MSSD -13.76 [95% CI: -21.62, -5.44], LF -9.16 [95% CI: -13.29, -4.95], HF -10.76 [95% CI: -18.88, -2.32] per 20 ppb cumulative 3-day avg O$_3$ concentrations) and remained associated with 3-day O$_3$ concentrations in two-pollutant models with sulfate. Another study in Taiwan recruited individuals with CHD or at risk for cardiovascular disease from outpatient clinics during the study period (two weeks in February) (Chuang et al., 2007b). No association was observed between O$_3$ concentration and HRV measures (SDNN, r-MSSD, LF, HF) (numerical results not provided in publication).

Overall, studies of O$_3$ concentration and HRV report inconsistent results. Multiple studies conducted in Boston, MA, observed positive associations but the authors of many of these studies postulated that O$_3$ concentration was possibly acting as a proxy for other pollutants. The majority of other studies, both in the U.S. and internationally, report null findings. The inconsistencies observed are further complicated by the different HRV measures and averaging times used by the studies.

### 6.3.2.3 Stroke

The 2006 O$_3$ AQCD did not identify any studies that examined the association between short-term O$_3$ exposure and stroke. However, recent studies have attempted to examine this relationship. Lisabeth et al. (2008) used a time-series approach to assess the relationship between daily counts of ischemic stroke and transient ischemic attack (TIA) with O$_3$ concentrations in a southeast Texas community among residents 45 years and older (2001-2005; median age of cases, 72 years). The median O$_3$ concentration (hourly average per 24-hour time-period) was 25.6 ppb (IQR 18.1-33.8). The associations between same-day O$_3$ concentrations and stroke/TIA risk were positive (RR: 1.03 [95% CI: 0.96, 1.10] per 20 ppb increase in 24-h avg O$_3$ concentrations) and previous-day (RR: 1.05 [95% CI: 0.99, 1.12] per
20 ppb increase in 24-h avg O₃ concentrations). Associations were robust to adjustment for PM₂.₅.

A case-crossover design was used in a study conducted in Dijon, France between March 1994 and December 2004, among those 40 years of age and older who presented with first-ever stroke (Henrotin et al., 2007). The mean O₃ concentration, calculated over 8-hour daytime periods, was 14.95 ppb (IQR: 6-22 ppb).

No association was observed between O₃ concentration at any of the single-day lags examined (i.e., 0-3 days) and hemorrhagic stroke. However, an association between ischemic stroke occurrence and O₃ concentrations with a 1-day lag was observed (OR: 1.54 [95% CI: 1.14, 2.09] per 30 ppb increase in 8-h max O₃ concentrations).

The observed association between short-term O₃ exposure and ischemic stroke persisted in two-pollutant models with PM₁₀, SO₂, NO₂, or CO. This association was stronger among men (OR: 2.12 [95% CI: 1.36, 3.30] per 30 ppb increase in 8-h max O₃ concentrations) than among women (OR: 1.17 [95% CI: 0.77, 1.78] per 30 ppb increase in 8-h max O₃ concentrations) in single pollutant models. When stroke was examined by subtype among men, an association was observed for ischemic strokes of large arteries and for transient ischemic attacks, but not for cardioembolic or lacunar ischemic strokes. The subtype analysis was not performed for women. Additionally, for men a linear exposure-response was observed when O₃ concentration was assessed based on quintiles (p for trend = 0.01) (Figure 6-21).

A potential limitation of this study is that 67.4% of the participating men were smokers compared to 9.3% of the women.

Another case-crossover study performed in Dijon, France examined the association between O₃ concentration and incidence of fatal and non-fatal ischemic cerebrovascular events (ICVE) (Henrotin et al., 2010). Mean 8-hour O₃ concentration was 19.1 ppb (SD 12.2 ppb). A positive association was observed between recurrent ICVE and 8-hour O₃ concentration with a 3-day lag (OR: 1.92 [95% CI 1.17, 3.12]), but not for other lags (0, 1, 2, 4) or cumulative days (0-1, 0-2, 1-2, 1-3). Although some ORs for incident ICVEs were elevated, none were statistically significant. Results for associations using the maximum daily 1-hour O₃ concentrations were similar to the 8-hour results but slightly attenuated. ORs were similar in two pollutant models with SO₂, NO₂, CO, and PM₁₀ (data not given). In stratified analyses, the association between 1-day lagged O₃ concentration and incident and recurrent ICVE was greater among individuals with diabetes or individuals with multiple pre-existing vascular conditions.
Source: Henrotin et al. (2007).

**Figure 6-21**  Odds ratio (95% confidence interval) for ischemic stroke by quintiles of O$_3$ exposure.

### 6.3.2.4 Biomarkers

An increasing number of studies have examined the relationship between air pollution and biomarkers in an attempt to elucidate the biological mechanisms linking air pollution and cardiovascular disease. A wide range of markers assessed as well as different types of study designs and locations chosen make comparisons across studies difficult. **Table 6-32** provides an overview of the O$_3$ concentrations reported in each of the studies evaluated.
Table 6-32  Characterization of O$_3$ concentrations (in ppb) from studies of biomarkers.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Averaging Time</th>
<th>Mean Concentration (Standard Deviation)</th>
<th>Upper Range of Concentration</th>
</tr>
</thead>
<tbody>
<tr>
<td>Liao et al. (2005)</td>
<td>3 U.S. counties</td>
<td>8-h</td>
<td>40 (20)</td>
<td></td>
</tr>
<tr>
<td>Thompson et al. (2010)</td>
<td>Toronto, Ontario</td>
<td>1-h / 1 yr</td>
<td>21.94 (15.78)</td>
<td></td>
</tr>
<tr>
<td>Rudez et al. (2009)</td>
<td>Rotterdam, the Netherlands</td>
<td>24-h</td>
<td>22*</td>
<td>75th: 31.5 Max: 90</td>
</tr>
<tr>
<td>Chuang et al. (2007a)</td>
<td>Taipei, Taiwan</td>
<td>24-h</td>
<td>28.4 (12.1)</td>
<td>Max: 49.3</td>
</tr>
<tr>
<td></td>
<td></td>
<td>48-h</td>
<td>33.3 (8.9)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>72-h</td>
<td>33.8 (7.1)</td>
<td>Max: 48.3</td>
</tr>
<tr>
<td>Steinvil et al. (2008)</td>
<td>Tel-Aviv, Israel</td>
<td>0.5-h</td>
<td>29.2 (9.7)</td>
<td>75th: 36</td>
</tr>
<tr>
<td>Chen et al. (2007a)</td>
<td>Los Angeles and San Francisco, CA</td>
<td>8-h / 2 weeks</td>
<td>30.8*</td>
<td>Max: 47.9</td>
</tr>
<tr>
<td></td>
<td></td>
<td>8-h / 1 mo</td>
<td>28.3*</td>
<td></td>
</tr>
<tr>
<td>Wellenius et al. (2007)</td>
<td>Boston, MA</td>
<td>1-h / 24-h</td>
<td>25.1 (12.9)</td>
<td></td>
</tr>
<tr>
<td>Goldberg et al. (2008)</td>
<td>Montreal, Quebec</td>
<td>24-h</td>
<td>NS</td>
<td></td>
</tr>
<tr>
<td>Baccarelli et al. (2007)</td>
<td>Lombardia, Italy</td>
<td>1-h</td>
<td>18.3*</td>
<td>75th: 35.1 Max: 202.3</td>
</tr>
<tr>
<td>Chuang et al. (2010)</td>
<td>Taiwan</td>
<td></td>
<td>26.83 (9.7)</td>
<td>Max: 62.1</td>
</tr>
</tbody>
</table>

*Note: Median presented (information on mean not given); studies presented in order of first appearance in the text of this section.

Hemostasis and coagulation markers

Multiple studies used various markers to examine if associations were present between short-term O$_3$ exposure and hemostasis and coagulation. Some of the markers included in these studies were as follows: fibrinogen, von Willebrand factor (vWF), plasminogen activator fibrinogen inhibitor-1 (PAI-1), tissue-type plasminogen activator (tPA), platelet aggregation, and thrombin generation.

A population-based study in the United States was conducted to assess the relationship between short-term exposure to air pollution and markers of blood coagulation using the Atherosclerosis Risk in Communities (ARIC) study cohort (Liao et al., 2005). Significant curvilinear associations were observed for O$_3$ (1 day prior to blood draw) and fibrinogen and vWF (quantitative results not provided for regression models although adjusted means [SE] of vWF were given as 118% [0.79%] for O$_3$ concentrations <40 ppb, 117% [0.86%] for O$_3$ concentrations 40-70 ppb, and 124% [1.97%] for O$_3$ concentrations of 70 ppb). The association between short-term O$_3$ exposure and fibrinogen was more pronounced among those with a history of cardiovascular disease (CVD) and was statistically significant.
among only this subgroup of the population. The curvilinear relationship between concentration and outcome suggested stronger relationships at higher concentrations of O₃. The authors note that the most pronounced associations occurred when the pollutant concentrations were 2-3 standard deviations above the mean. The results from this relatively large-scale cross-sectional study suggest weak associations with between short-term O₃ exposure and increases in fibrinogen (among those with a history of CVD) and vWF. A retrospective repeated measures analysis was performed in Toronto, Canada among adults aged 18-40 years (n = 45) between the years of 1999 and 2006 (Thompson et al., 2010). Single pollutant models were used with moving averages up to 7 days. No evidence of an association was observed between short-term O₃ exposure and increases in fibrinogen.

A repeated measures study was conducted among 40 healthy individuals living or working in the city center of Rotterdam, the Netherlands to assess the relationship between air pollution and markers of hemostasis and coagulation (platelet aggregation, thrombin generation, and fibrinogen) (Rudez et al., 2009). Each participant provided between 11 and 13 blood samples throughout a 1-year period (498 samples on 197 days). Examined lags ranged from 6 hours to 3 days prior to blood sampling. No consistent evidence of an association was observed between O₃ concentration and any of the biomarkers (percent change of max platelet aggregation: -6.87 [95% CI: -21.46, 7.70] per 20 ppb increase in 24-h avg O₃ concentration at 4-day average; percent change of endogenous thrombin potential: 0.95 [95% CI: -3.05, 4.95] per 20 ppb increase in 24-h avg O₃ concentration at 4-day avg; percent change of fibrinogen: -0.57 [95% CI: -3.05, 2.00] per 20 ppb increase in 24-h avg O₃ concentration at lag 1-day). Some associations with O₃ were in the opposite direction to that hypothesized which may be explained by the negative correlation between O₃ and other pollutants (correlation coefficients ranged from -0.4 to -0.6).

The statistically significant inverse effects observed in single-pollutant models with O₃ were no longer apparent when PM₁₀ was included in the model (Rudez et al., 2009).

A panel study in Taiwan measured health endpoints using blood samples from healthy individuals (n = 76) at three times from April to June in 2004 or 2005 (Chuang et al., 2007a). Increases in fibrinogen and PAI-1 were associated with increases in O₃ concentrations in single-pollutant models (percent change in fibrinogen: 11.76 [95% CI: 4.03, 19.71] per 20 ppb 3-day cumulative avg O₃ concentration; percent change in PAI-1: 6.08 [95% CI: 38.91, 84.27] per 20 ppb 3-day cumulative avg O₃ concentration). These associations were also observed at 1 and 2 day averaging times. Associations between PAI-1 and 3-day O₃ concentrations remained robust in two-pollutant models with sulfate. No association was observed between O₃ concentration and tPA, a fibrinolytic factor (percent change 16.15 [95% CI: -4.62, 38.34] per 20 ppb 3-day avg O₃ concentration).

A study in Israel examined the association between pollutant concentrations and fibrinogen among 3,659 apparently healthy individuals (Steinvil et al., 2008). In single pollutant models, O₃ was associated with an increase in fibrinogen at a 4-day lag among men and a same-day O₃ concentration among women but results for
other lags (0 through 7 days) were mixed (i.e., some positive and some negative; none statistically significant).

**Inflammatory markers**

Potential associations between short-term exposures to air pollution and inflammatory markers (C-reactive protein [CRP], white blood cell [WBC] count, albumin, and Interleukin-6 [IL-6]) were also examined in several studies.

The ARIC study cohort, which included men and women aged 45-64 years old at the start of the study, was utilized to assess the association between O\textsubscript{3} concentrations and markers of inflammation, albumin and WBC count (Liao et al., 2005). No association was observed between O\textsubscript{3} concentrations and albumin or WBC count.

Thompson et al. (2010) assessed ambient air pollution exposures and IL-6. This retrospective repeated measures analysis was conducted among 45 adults (18-40 years of age) in Toronto, Canada between the years of 1999 and 2006. Single pollutant models were used to analyze the repeated-measures data using moving averages up to 7 days. A positive association was observed between IL-6 and short-term 1-hour O\textsubscript{3} exposure with the strongest effects observed for the average of lags 0-3 days (quantitative results not provided). No association was observed for shorter averaging times (average lags of <1 day). When examined by season using 2-day moving averages, the association between short-term O\textsubscript{3} exposure and IL-6 was positive during only the spring and summer.

In Rotterdam, the Netherlands, a repeated measures study of healthy individuals living or working in the city center reported no association between short-term O\textsubscript{3} exposure and CRP (Rudez et al., 2009). Each of the 40 participants provided between 11 and 13 blood samples throughout a 1-year period (498 samples on 197 days). No consistent evidence of an association was observed between O\textsubscript{3} concentration and CRP (percent change: -0.48 [95% CI: -14.05, 13.10] per 20 ppb increase in 24-h avg O\textsubscript{3} concentration at lag 1-day). Additionally, no association was observed with 2 or 3 day lags.

The relationship between pollutant concentrations and one-time measures of inflammatory biomarkers was assessed in sex-stratified analyses among 3,659 apparently healthy individuals in Tel Aviv, Israel (Steinvil et al., 2008). No evidence of an association was observed between O\textsubscript{3} concentration and CRP or WBC for men and women.

A panel study of healthy individuals (n = 76) was conducted in Taiwan to assess the relationship between air pollutants and inflammation (Chuang et al., 2007a). Health endpoints were measured three times from April to June in 2004 or 2005. Ozone effects were assessed in statistical models using the average of the 24 hours (1 day), 48 hours (2 days), and 72 hours (3 days) before the hour of each blood sampling. Increases in CRP were associated with increases in O\textsubscript{3} concentrations in single-pollutant models (percent change in CRP: 244.38 [95% CI: 4.54, 585.15] per 20 ppb
3-day avg $O_3$ concentration). The association was also observed using a 2-day cumulative averaging time, but no association was present with a 1-day averaging time.

**Oxidative stress markers**

A few studies have reported on the relationships between short-term $O_3$ exposure and increases in markers of oxidative stress. The association between $O_3$ concentration and markers of lipid peroxidation and antioxidant capacity was examined among 120 nonsmoking healthy college students, aged 18-22 years, from the University of California, Berkeley (February-June 2002) (Chen et al., 2007a). By design, students were chosen that had experienced different geographic concentrations of $O_3$ over their lifetimes and during recent summer vacation in either greater Los Angeles (LA) or the San Francisco Bay Area (SF). Long-term (based on lifetime residential history) and shorter-term (based on the moving averages of 8-h max concentrations 1-30 days prior to the day of blood collection) $O_3$ concentration were estimated (lifetime exposure results are presented in Chapter 7). A marker of lipid peroxidation, 8-isoprostane (8-iso-PGF), was assessed. This marker is formed continuously under normal physiological conditions but has been found at elevated concentrations in response to environmental exposures. A marker of overall antioxidant capacity, ferric reducing ability of plasma (FRAP), was also measured. Levels of 8-iso-PGF were associated with 2-week ($\beta = 0.035 \ [\text{pg/mL}/8\text{-hour ppb } O_3], p = 0.007$) and 1-month ($\beta = 0.031 \ [\text{pg/mL}/8\text{-hour ppb } O_3], p = 0.006$) estimated $O_3$ concentrations. No evidence of association was observed between short-term $O_3$ exposure and increases in FRAP. A chamber study performed among a subset of study participants supported the primary study results. The concentrations of 8-iso-PGF increased immediately after the 4-hour controlled $O_3$ exposure ended ($p = 0.10$). However, levels returned to near baseline by 18 hours without further exposure. The authors note that $O_3$ was highly correlated with PM$_{10-2.5}$ and NO$_2$ in this study population; however, $O_3$ associations remained robust in copollutant models.

Using blood samples collected between April and June of 2004 or 2005 in Taiwan, the association between short-term $O_3$ exposure and a marker of oxidative stress (i.e., 8-hydroxy-2’-deoxyguanosine (8-OHdG)) was studied among healthy individuals ($n = 76$) (Chuang et al., 2007a). Increases in 8-OHdG were associated with increases in $O_3$ concentrations in single-pollutant models (percent change in 8-OHdG: 2.46 [95% CI: 1.01, 3.92] per 20 ppb increase in 24-h avg $O_3$). The association did not persist with 2- or 3-day cumulative averaging times.

**Markers of overall cardiovascular health**

Multiple studies used markers that assess overall cardiovascular well-being. Wellenius et al. (2007) examined B-type natriuretic peptide (BNP), a marker of heart failure, in a repeated-measures study conducted in Boston, MA, among 28 patients with congestive heart failure and impaired systolic function. The authors found no
evidence of an association between BNP and short-term O₃ exposures at lags 0-3 days (quantitative results not provided). BNP was chosen because it is directly associated with cardiac hemodynamics and symptom severity among those with heart failure and is considered a marker of functional status. However, the authors conclude that the use of BNP may not be useful in studies of the health effects of ambient air pollutants due to the large amount of within-person variability in BNP levels observed in this population.

The relationship between air pollution and oxygen saturation and pulse rate, markers of physiological well-being, was examined in a 2-month panel study among 31 congestive heart failure patients (aged 50-85 years) in Montreal, Canada from July 2002 to October 2003 (Goldberg et al., 2008). All participants had limited physical functioning (New York Heart Association Classification ≥ II) and an ejection fraction (the fraction of blood pumped out of the heart per beat) less than or equal to 35% (normal is above 55%). Daily mean O₃ concentrations were calculated based on hourly measures at 10 monitoring stations. There was an inverse association between O₃ concentration (lag-0) and oxygen saturation when adjustment was made for temporal trends. In the models incorporating personal covariates and weather factors, the association remained but was not statistically significant. The associations of O₃ concentration with a lag of 1 day or a 3-day mean were not statistically significant. No evidence of association was observed between O₃ concentration and pulse rate.

Total homocysteine (tHcy) is an independent risk factor for vascular disease and measurement of this marker after oral methionine load is used to identify individuals with mild impairment of homocysteine metabolism. The effects of air pollution on fasting and postmethionine-load tHcy levels were assessed among 1,213 apparently healthy individuals from Lombardia, Italy from January 1995 to September 2005 (Baccarelli et al., 2007). A 20-ppb increase in the 24-h avg O₃ concentrations was associated with an increase in fasting tHcy (percent change 6.25 [95% CI: 0.84, 11.91]) but no association was observed with postmethionine-load tHcy (percent change 3.36 [95% CI: -1.30, 8.39]). In addition, no evidence of an association was observed between 7-day cumulative averaged O₃ concentrations and tHcy (percent change for fasting tHcy 4.16 [95% CI: -1.76, 10.42] and percent change for postmethionine-load tHcy -0.65 [95% CI: -5.66, 4.71] per 20 ppb increase in 24-h avg O₃ concentrations). No evidence of effect modification by smoking was observed.

**Blood lipids and glucose metabolism markers**

Chuang et al. (2010) conducted a population-based cross-sectional analysis of data collected on 7,778 participants during the Taiwanese Survey on Prevalence of Hyperglycemia, Hyperlipidemia, and Hypertension in 2001. Apolipoprotein B (ApoB), the primary apolipoprotein among low-density lipoproteins, was associated with 3-day avg O₃ concentration at the p <0.10 level. The 5-day mean O₃ concentration was associated with an increase in triglycerides at p <0.10. In addition, the 1-, 3-, and 5-day mean O₃ concentrations were associated with increased HbA1c
levels (a marker used to monitor the degree of control of glucose metabolism) at the p < 0.05 level. The 5-day mean O₃ concentration was associated with increased fasting glucose levels (p < 0.10). No association was observed between O₃ concentration and ApoA1.

### 6.3.2.5 Myocardial Infarction (MI)

The 2006 O₃ AQCD did not report consistent results indicating an association between short-term O₃ exposure and MI. One study reported a positive association between current day O₃ concentration and acute MI, especially among the oldest age group (55 to 64 year-olds) (Ruidavets et al., 2005b). No association was observed in a case-crossover study of O₃ concentration during the surrounding hours and MI (Peters et al., 2001). Since the 2006 O₃ AQCD, a few recent epidemiologic studies have examined the association between O₃ concentration and MI (Henrotin et al., 2010; Rich et al., 2010), arterial stiffness (Wu et al., 2010) and ST-segment depression (Delfino et al., 2011).

One of the studies conducted in the U.S. examined hospital admissions for first MI and reported no association with O₃ concentration (Rich et al., 2010). More details on this study are reported in the section on hospital admissions (Section 6.3.2.7). A study performed in Dijon, France examined the association between O₃ concentration and incident and recurrent MI (Henrotin et al., 2010). The mean 8-hour O₃ concentration was 19.1 ppb (SD 12.2 ppb). Odds ratios for the association between cumulative O₃ concentrations and recurrent MIs were elevated but none of the results were statistically significant (OR: 1.71 [95% CI: 0.91, 3.20] per 20 ppb increase in 24-h avg O₃ concentration for a cumulative lag of 1-3 days). No association was observed for incident MIs. In analyses stratified by vascular risk factors, positive associations were observed between 1-day lagged O₃ concentration and MIs (incident and recurrent combined) among those who reported having hypercholesterolaemia (OR: 1.52 [95% CI: 1.08, 2.15] per 20 ppb increase in 24-h avg O₃ concentration) and a slight inverse association was observed among those who reported not having hypercholesterolaemia (OR: 0.69 [95% CI: 0.50, 0.94] per 20 ppb increase in 24-h avg O₃ concentration). In other stratified analyses combining different vascular factors, only those containing individuals with hypercholesterolaemia demonstrated a positive association; none were inverse associations.

Wu et al. (2010) examined mail carriers aged 25-46 years and measured exposure to O₃ concentrations through personal monitors [mean O₃ 24.9 (SD 14.0) ppb]. Ozone concentration was positively associated with arterial stiffness (percent change 11.24% [95% CI: 3.67, 19.62] per 40 ppb O₃) and was robust to adjustment for ultrafine PM.

A study performed in the Los Angeles basin reported on the association between O₃ concentration and ST-segment depression, a measure representing cardiac ischemia (Delfino et al., 2011). Study participants were nonsmokers, at least 65 years old, had
a history of coronary artery disease, and were living in a retirement community. Study periods included five consecutive days in both July to mid-October and mid-October to February. Mean 24-hour O$_3$ concentrations were 27.1 ppb (SD 11.5 ppb). No association was observed between O$_3$ concentration and ST-segment depression of at least 1.0 mm during any of the exposure periods (i.e., 1-hour, 8-hours, 1-day, 2-day avg, 3-day avg, 4-day avg).

### 6.3.2.6 Blood Pressure

In the 2006 O$_3$ AQCD, no epidemiologic studies examined O$_3$-related effects on blood pressure (BP). Recent studies have been conducted to evaluate this relationship and overall the findings are inconsistent. The O$_3$ concentrations for these studies are listed in Table 6-33.

#### Table 6-33 Characterization of O$_3$ concentrations (in ppb) from studies of blood pressure.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Averaging Time</th>
<th>Mean Concentration (Standard Deviation)</th>
<th>Upper Range of Concentration</th>
</tr>
</thead>
<tbody>
<tr>
<td>Zanobetti et al. (2004)</td>
<td>Boston, Massachusetts</td>
<td>1-h</td>
<td>20</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>5-days</td>
<td>24</td>
<td></td>
</tr>
<tr>
<td>Delfino et al. (2010b)</td>
<td>Los Angeles, California</td>
<td>24-h</td>
<td>27.1 (11.5)</td>
<td>Max: 60.7</td>
</tr>
<tr>
<td>Choi et al. (2007)</td>
<td>Incheon, South Korea</td>
<td>8-h (warm season)</td>
<td>26.6 (11.8)</td>
<td>75th: 34.8, Max: 62.4</td>
</tr>
<tr>
<td></td>
<td></td>
<td>8-h (cold season)</td>
<td>17.5 (7.3)</td>
<td>75th: 22.9, Max: 33.9</td>
</tr>
<tr>
<td>Chuang et al. (2010)</td>
<td>Taiwan</td>
<td></td>
<td>26.83 (9.7)</td>
<td>Max: 62.1</td>
</tr>
</tbody>
</table>

*Note: Studies presented in order of first appearance in the text of this section.

Zanobetti et al. (2004) examined the relationship between air pollutants and BP from May 1999 to January 2001 for 631 repeat visits among 62 Boston, MA, residents with CVD. In single-pollutant models, higher resting diastolic blood pressure (DBP) was associated with the 5-day (0-4 days) averages of O$_3$ concentration (RR: 1.03 [95% CI: 1.00, 1.05] per 20 ppb increase in 24-hour O$_3$ concentrations). However, this effect was no longer apparent when PM$_{2.5}$ was included in the model (data were not presented) (Zanobetti et al., 2004). Delfino et al. (2010b) examined 64 subjects 65 years and older with coronary artery disease, no tobacco smoke exposure, and living in retirement communities in the Los Angeles air basin with hourly (up to 14-hours/day) ambulatory BP monitoring for 5 days during a warm period (July-mid-October) and 5 days during a cool period (mid-October-February). Investigators assessed lags of 1, 4, and 8 hours, 1 day, and up to 9 days before each BP measure; no evidence of an association was observed for O$_3$ (change in BP associated with a
20 ppb increase in 24-h avg O₃ concentration was 0.67 [95% CI: -1.16, 2.51] for systolic BP (SBP) and -0.25 [95% CI: -1.25, 0.75] for DBP) (Delfino et al., 2010b). Choi et al. (2007) conducted a cross-sectional study to investigate the relationship between air pollutants and BP among 10,459 participants of the Inha University Hospital health examination from 2001 to 2003. These individuals had no medical history of cardiovascular disease or hypertension. Ozone concentration was associated with an increase in SBP for 1-day lag in the warm season and similar effect estimates were observed during the cold season but were not statistically significant (quantitative results not provided). Associations between O₃ concentration and DBP were present in the cold season but not the warm season (quantitative results not provided). Chuang et al. (2010) conducted a similar type of study among 7,578 participants of the Taiwanese Survey on Prevalence of Hyperglycemia, Hyperlipidemia, and Hypertension in 2001. Investigators examined 1-, 3-, and 5-day avg O₃ concentrations. An increase in DBP was associated with the 3-day mean O₃ concentration (change in BP for a 20 ppb increase in 24-h avg O₃ concentration was 0.61 [95% CI: 0.07, 1.14]) (Chuang et al., 2010). Associations were not observed for other days or with SBP.

6.3.2.7 Hospital Admissions and Emergency Department Visits

Upon evaluating the collective evidence for O₃-related cardiovascular hospital admissions and emergency department (ED) visits, the 2006 O₃ AQCD concluded that “a few studies observed positive O₃ associations, largely in the warm season. Overall, however, the currently available evidence is inconclusive regarding any association between ambient O₃ exposure on cardiovascular hospitalizations” (U.S. EPA, 2006b). Table 6-34 provides information on the O₃ concentrations reported in each of the recent hospital admission and ED visit studies evaluated.
<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Averaging Time</th>
<th>Mean Concentration (Standard Deviation)</th>
<th>Upper Range of Concentration</th>
</tr>
</thead>
<tbody>
<tr>
<td>Peel et al. (2007)</td>
<td>Atlanta, GA</td>
<td>8-h max warm season</td>
<td>55.6 (23.8)</td>
<td></td>
</tr>
<tr>
<td>Tolbert et al. (2007)</td>
<td>Atlanta, GA</td>
<td>8-h max warm season</td>
<td>53.0</td>
<td>75th: 67.0, Max: 147.5</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>12 Canadian cities</td>
<td>1-h</td>
<td>6.7-8.3*</td>
<td>75th: 8.4-12.4</td>
</tr>
<tr>
<td></td>
<td>8 European cities</td>
<td>1-h</td>
<td>11.0-38.1*</td>
<td>75th: 15.3-49.4</td>
</tr>
<tr>
<td></td>
<td>14 United States cities</td>
<td>1-h</td>
<td>34.9-60.0*</td>
<td>75th: 46.8-68.8</td>
</tr>
<tr>
<td>Rich et al. (2010)</td>
<td>New Jersey</td>
<td>24-h</td>
<td>NR</td>
<td></td>
</tr>
<tr>
<td>Cakmak et al. (2006a)</td>
<td>10 Canadian cities</td>
<td>1-h max</td>
<td>17.4</td>
<td></td>
</tr>
<tr>
<td>Stieb et al. (2009)</td>
<td>7 Canadian cities</td>
<td>24-h</td>
<td>18.4</td>
<td></td>
</tr>
<tr>
<td>Szyszkowicz (2008)</td>
<td>Edmonton, Canada</td>
<td>24-h</td>
<td>18.6 (9.3)</td>
<td></td>
</tr>
<tr>
<td>Villeneuve et al. (2006a)</td>
<td>Edmonton, Canada</td>
<td>24-h</td>
<td>17 (9.1)</td>
<td>75th: 23.5</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>21.8 (8)</td>
<td>75th: 27.0</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>12.2 (7.4)</td>
<td>75th: 17.0</td>
</tr>
<tr>
<td>Symons et al. (2006)</td>
<td>Baltimore, MD</td>
<td>8-h warm season</td>
<td>31.0 (20.0)</td>
<td>Max: 120.0</td>
</tr>
<tr>
<td>Wellenius et al. (2005)</td>
<td>Allegheny County, PA</td>
<td>24-h</td>
<td>24.3 (12.2)</td>
<td>75th: 32.0</td>
</tr>
<tr>
<td>Zanobetti and Schwartz (2006)</td>
<td>Boston, MA</td>
<td>24-h</td>
<td>22.4*</td>
<td>75th: 31.0</td>
</tr>
<tr>
<td>Lee et al. (2007)</td>
<td>Kaohsiung, Taiwan</td>
<td>24-h</td>
<td>26.5</td>
<td>75th: 35.5, Max: 83.0</td>
</tr>
<tr>
<td>Chan et al. (2006)</td>
<td>Taipei, Taiwan</td>
<td>1-h max</td>
<td>50.9 (26.4)</td>
<td>Max: 150.3</td>
</tr>
<tr>
<td>Chiu and Yang (2009)</td>
<td>Taipei, Taiwan</td>
<td>24-h</td>
<td>23.0</td>
<td>75th: 28.7, Max: 62.8</td>
</tr>
<tr>
<td>Lee et al. (2008a)</td>
<td>Taipei, Taiwan</td>
<td>24-h</td>
<td>21.0</td>
<td>75th: 26.4, Max: 62.8</td>
</tr>
<tr>
<td>Wong et al. (2009)</td>
<td>Hong Kong</td>
<td>8-h</td>
<td>18.5 (11.5)</td>
<td>75th: 25.4, Max: 48.3</td>
</tr>
<tr>
<td>Bell et al. (2008)</td>
<td>Taipei, Taiwan</td>
<td>24-h</td>
<td>21.4</td>
<td>Max: 53.4</td>
</tr>
<tr>
<td>Buadong et al. (2009)</td>
<td>Bangkok, Thailand</td>
<td>1-h</td>
<td>14.4 (3.2)</td>
<td>Max: 41.9</td>
</tr>
<tr>
<td>Lee et al. (2003b)</td>
<td>Seoul, Korea</td>
<td>1-h max</td>
<td>36.0 (18.6)</td>
<td>75th: 44.9</td>
</tr>
<tr>
<td>Azevedo et al. (2011)</td>
<td>Portugal</td>
<td>1-h</td>
<td>NR</td>
<td></td>
</tr>
<tr>
<td>Linares and Diaz (2010)</td>
<td>Madrid, Spain</td>
<td>24-h</td>
<td>17.4 (8.9)</td>
<td></td>
</tr>
<tr>
<td>Middleton et al. (2008)</td>
<td>Nicosia, Cyprus</td>
<td>8-h max</td>
<td>28.7 - 54.9</td>
<td></td>
</tr>
</tbody>
</table>
Multiple recent studies of O$_3$ concentration and cardiovascular hospital admissions and ED visits have been conducted in the U.S. and Canada. Peel et al. (2007) used a case-crossover framework (using a time-stratified approach matching on day of the week in the calendar month of the event) to assess the relationship between air pollutants and cardiovascular disease ED visits among those with and without secondary comorbid conditions (hypertension, diabetes, chronic obstructive pulmonary disease [COPD], congestive heart failure [CHF], and dysrhythmia). Data on over 4 million ED visits from 31 hospitals were collected from January 1993 to August 2000. Ozone was monitored from March to October. This study was a re-analysis of a time series study conducted to assess the main effects of air pollutants on cardiovascular ED visits in Atlanta, GA (Tolbert et al., 2007; Metzger et al., 2004). In the initial study, no evidence of associations was observed between O$_3$ concentration and all CVD visits or visits for CVD subgroups, such as dysrhythmia, CHF, ischemic heart disease (IHD), and peripheral vascular and cerebrovascular disease. The relative risk for all CVD visits was 1.01 (95% CI: 0.98, 1.04) for a 30 ppb increase in the 3-day moving avg (lags 0-2 days) of 8-hour O$_3$ concentration (Metzger et al., 2004). Similar to the initial investigation using a time-series analysis, no evidence of an association was observed between short-term O$_3$ exposure and CVD visits at lag 0-2 among the entire population using the case-crossover design (Peel et al., 2007). However, the relationship between O$_3$ concentration and

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<table>
<thead>
<tr>
<th>Study$^a$</th>
<th>Location</th>
<th>Averaging Time</th>
<th>Mean Concentration (Standard Deviation)</th>
<th>Upper Range of Concentration</th>
</tr>
</thead>
<tbody>
<tr>
<td>Turner et al. (2007)</td>
<td>Sydney, Australia</td>
<td>24-h</td>
<td>28</td>
<td>75th: 33</td>
</tr>
<tr>
<td>Ballester et al. (2006)</td>
<td>14 Spanish cities</td>
<td>8-h warm season</td>
<td>24.2 - 44.3</td>
<td></td>
</tr>
<tr>
<td>De Pablo et al. (2006)</td>
<td>Castilla-Leon, Spain</td>
<td>24-h</td>
<td>23.2-33.6</td>
<td></td>
</tr>
<tr>
<td>Von Klot et al. (2005)</td>
<td>5 European cities</td>
<td>8 h max warm season</td>
<td>16.4 - 28.0</td>
<td></td>
</tr>
<tr>
<td>Oudin et al. (2010)</td>
<td>Scania, Sweden</td>
<td>24-h</td>
<td>30.5</td>
<td></td>
</tr>
<tr>
<td>Halonen et al. (2009)</td>
<td>Helsinki, Finland</td>
<td>8-h max warm season</td>
<td>35.7*</td>
<td>75th: 42.1 Max: 79.6</td>
</tr>
<tr>
<td>Larrieu et al. (2007)</td>
<td>8 French cities</td>
<td>8-h max warm season</td>
<td>34.2 - 53.1</td>
<td></td>
</tr>
<tr>
<td>Barnett et al. (2006)</td>
<td>4 Australian cities</td>
<td>8-h</td>
<td>19.0-28.5</td>
<td>Max: 58.4-86.8</td>
</tr>
<tr>
<td>Hinwood et al. (2006)</td>
<td>Perth, Australia</td>
<td>8-h max</td>
<td>25.9 (6.5)</td>
<td></td>
</tr>
<tr>
<td>Lanki et al. (2006)</td>
<td>5 European cities</td>
<td>8-h max warm season</td>
<td>31.7 - 57.2*</td>
<td></td>
</tr>
<tr>
<td>Hosseinpoor et al. (2005)</td>
<td>Tehran, Iran</td>
<td>8-h max</td>
<td>4.9 (4.8)</td>
<td>75th: 7.2 Max: 99.0</td>
</tr>
<tr>
<td>Simpson et al. (2005)</td>
<td>4 Australian cities</td>
<td>1-h max</td>
<td>24.4-33.8</td>
<td>Max: 96.0-111.5</td>
</tr>
<tr>
<td>Dennekamp et al. (2010)</td>
<td>Melbourne, Australia</td>
<td>24-h</td>
<td>13.34</td>
<td>75th: 16.93</td>
</tr>
<tr>
<td>Silverman et al. (2010)</td>
<td>New York City, NY</td>
<td>8-h max</td>
<td>28*</td>
<td>75th: 40</td>
</tr>
</tbody>
</table>

$^a$Notes: Median presented (information on mean not given); NR: Not reported; Studies presented in order of first appearance in the text of this section.
peripheral and cerebrovascular disease visits was stronger among patients with comorbid COPD (OR: 1.29 [95% CI: 1.05-1.59] per 30 ppb, lag 0-2 days) as compared to patients without COPD (OR: 1.01 [95% CI: 0.96-1.06] per 30 ppb, lag 0-2 days). The same research group expanded upon the number of Atlanta hospitals providing ED visit data (41 hospitals) as well as the length of the study period (1993-2004) (Tolbert et al., 2007). Again, models assessing the health effects of O₃ concentration utilized data collected from March through October. Similar to the results presented by Metzger et al. (2004) and Peel et al. (2007) among the entire study population, no evidence of associations was observed for O₃ concentration and CVD visits (Tolbert et al., 2007).

Existing multicity studies in North America and Europe were evaluated under a common framework in the Air Pollution and Health: A European and North American Approach (APHENA) study (Katsouyanni et al., 2009). One component of the study examined the relationship between short-term O₃ exposure and CVD hospital admissions among individuals 65 years of age and older. The study presented multiple models but this section focuses on the results for the models that used 8 df to account for temporal trends and natural splines (see Section 6.2.7.2 for additional explanation). Across the study locations, no associations were observed between O₃ concentration and CVD hospital admissions at lags 0-1, lag 1, or a distributed lag of 0-2. Additionally, there was no evidence of an association when restricting the analysis to the summer months.

A study of hospital admissions for MI was performed using a statewide registry from New Jersey between January 2004 and December 2006 (Rich et al., 2010). Using a case-crossover design, the association between the previous 24-hours O₃ concentration and transmural infarction (n = 1,003) was examined. No association was observed (OR: 0.94 [95% CI: 0.79, 1.13] per 20 ppb increase in 24-h avg O₃ concentration) and this did not change with the inclusion of PM₂.₅ in the model.

Cakmak et al. (2006a) investigated the relationship between gaseous air pollutants and cardiac hospitalizations in 10 large Canadian cities using a time-series approach. A total of 316,234 hospital discharge records for primary diagnosis of congestive heart failure, ischemic heart disease, or dysrhythmia were obtained from April 1993 through March 2000. Correlations between pollutants varied substantially across cities, which could partially explain discrepancies in effect estimates observed across the cities. In addition, pollutant lags differed across cities; the average lag for O₃ was 2.9 days. The pooled effect estimate for a 20 ppb increase in the daily 1-h max O₃ concentration and the percent change in hospitalizations among all 10 cities was 2.3 (95% CI: 0.11, 4.50) in an all-year analysis. The authors reported no evidence of effect modification by sex, neighborhood-level education, or neighborhood-level income. A similar multicity time-series study was conducted using nearly 400,000 ED visits to 14 hospitals in seven Canadian cities from 1992 to 2003 (Stieb et al., 2009). Primary analyses considered daily O₃ single day lags of 0-2 days; in addition, sub-daily lags of 3-h avg concentrations up to 12 hours before presentation to the ED were considered. Seasonal variation was assessed by stratifying analyses by warm and cold seasons. No evidence of associations between short-term O₃ exposure and
CVD ED visits was observed. One negative, statistically significant association was reported between a 1-day lag of O₃ concentration and visits for angina/myocardial infarction. Ozone concentration was negatively correlated with many of the other pollutants, particularly during the cold season.

The effect of air pollution on daily ED visits for ischemic stroke (n = 10,881 visits) in Edmonton, Canada was assessed from April 1992 through March 2002 (Szyszkowicz, 2008). A 26.4% (95% CI: 3.16-54.5) increase in stroke ED visits was associated with a 20 ppb increase in 24-hour average O₃ concentration at lag 1 among men aged 20-64 years in the warm season. No associations were present among women or among men age 65 and older. In addition, no associations were observed for the cold season or for other lags (lag 0 or lag 2). A similar investigation over the same time period in Edmonton, Canada, assessed the relationship between air pollutants and ED visits for stroke (ischemic stroke, hemorrhagic stroke, and transient ischemic attack) among those 65 years of age and older using a case-crossover framework (Villeneuve et al., 2006a). No evidence of association was reported for O₃ concentration and stroke hospitalization in single or copollutant models (Villeneuve et al., 2006a).

Additional studies in the U.S. reported no evidence of an association between O₃ concentrations and ED visits, hospitalizations, or symptoms leading to hospitalization (Symons et al., 2006; Zanobetti and Schwartz, 2006; Wellenius et al., 2005). Symons et al. (2006) used a case-crossover framework to assess the relationship between air pollutants and the onset of symptoms (dyspnea) severe enough to lead to hospitalization (through the ED) for congestive heart failure. The study was conducted from April to December of 2002 in Baltimore, Maryland. Exposures were assigned using 3 index times: 8-hour and 24-hour periods prior to symptom onset and date of hospital admission. No evidence of association was reported for O₃ concentrations. Although seasonal variation was not assessed, the time frame for the study did not involve an entire year (April to December). Wellenius et al. (2005) investigated the association between air pollutants and congestive heart failure hospitalization among Medicare beneficiaries in Pittsburgh, Pennsylvania from 1987 to 1999 utilizing a case-crossover framework. A total of 55,019 admissions from the emergency room with a primary discharge diagnosis of CHF were collected. No evidence of an association was reported for O₃ concentration and CHF hospitalization (Wellenius et al., 2005). Finally, Zanobetti and Schwartz (2006) assessed the relationship between air pollutants and hospital admissions through the ED for MI and pneumonia among patients aged 65 and older residing in the greater Boston, MA, area (1995-1999) using a case-crossover framework with control days in the same month matched on temperature. Pollution exposures were assigned for the same day and for the mean of the exposure the day of and the day before the admission. Ozone concentration was not associated with MI admissions in all-year and seasonal analyses.

Several recent studies have examined the relationship between air pollution and CVD hospital admissions and/or emergency department visits in Asia. Of note, some areas of Asia have a more tropical climate than the U.S. and do not experience similar
seasonal changes. In Taiwan, fairly consistent positive associations have been reported for O₃ concentration and congestive heart failure hospital admissions (for single- and copollutant models) in Taipei on warm days (Yang, 2008) and in Kaohsiung (Lee et al., 2007); cerebrovascular disease ED visits (for lag 0 single- and two-pollutant models but not other lags) in Taipei (Chan et al., 2006); and arrhythmia ED visits in Taipei among those without comorbid conditions (Chiu et al., 2009; Lee et al., 2008a) and in Taipei on warm days among those with and without comorbid conditions (Lee et al., 2008a). However, one study in Taiwan did not show an association. Bell et al. (2008) reported no evidence of an association between O₃ concentration and hospital admissions for ischemic heart disease or cerebrovascular disease. Studies based in Asia but outside Taiwan were also performed. A Hong Kong-based investigation (Wong et al., 2009) reported no consistent evidence of a modifying effect of influenza on the relationship between O₃ concentration and CVD admissions. Among elderly populations in Thailand, O₃ concentration was associated with CVD visits, but this association was not detected among younger age groups (15-64) (Buadong et al., 2009). Also, a study performed in Seoul, Korea reported a positive association between O₃ concentration and hospital admissions for ischemic heart disease; the association was slightly greater among those over 64 years of age (Lee et al., 2003b).

Positive associations between short-term O₃ exposure and CVD hospital admissions and/or ED visits have been reported in other areas of the world as well (Azevedo et al., 2011; Linares and Diaz, 2010; Middleton et al., 2008; Turner et al., 2007; Ballester et al., 2006; De Pablo et al., 2006; Von Klot et al., 2005), although not consistently; some studies reported no association (Oudin et al., 2010; Halonen et al., 2009; Larrieu et al., 2007; Barnett et al., 2006; Hinwood et al., 2006; Lanki et al., 2006; Hosseinpoor et al., 2005; Simpson et al., 2005).

A couple of studies (U.S. and Australia) have examined cardiac arrests where emergency services attempted treatment/resuscitation. No evidence of an association between O₃ concentration and out-of-hospital cardiac arrest was observed (Dennekamp et al., 2010; Silverman et al., 2010).

An increasing number of air pollution studies have investigated the relationship between O₃ concentrations and CVD hospital admissions and/or ED visits. As summarized in the 2006 O₃ AQCD, some, especially those reporting results stratified by season (or temperature) or comorbid conditions have reported positive associations. However, even studies performing these stratified analyses are not consistent and the overall evidence remains inconclusive regarding the association between short-term O₃ exposure and CVD hospital admissions and ED visits. The Hospital Admission (HA) and ED visit studies evaluated in this section are summarized in Figure 6-22 through Figure 6-26, which depict the associations for studies in which quantitative data were presented. Table 6-35 through Table 6-39 provide the numerical results displayed in the figures.
Note: Change in O₃ standardized to 20 ppb for 24-h avg period, 30 ppb for 8-h avg period, and 40 ppb for 1-h avg period (see Section 2.5). Ozone concentrations in ppb. Seasons depicted by colors – black: all year; red: warm season; light blue: cold season. Age groups of study populations were not specified or were adults with the exception of Katsouyanni et al. (2009), Fung et al. (2005), Wong et al. (1999b), and Prescott et al. (1998), which included only individuals aged 65+. Studies organized by outcome and season and then listed in descending order of publication date.

**Figure 6-22** Effect estimate (95% CI) per increment ppb increase in O₃ for over all cardiovascular ED visits or hospital admissions.
Table 6-35  Effect estimate (95% CI) per increment ppb increase in O3 for overall cardiovascular ED visits or hospital admissions in studies presented in Figure 6-22.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Outcome</th>
<th>Averaging Time</th>
<th>Effect Estimate (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atkinson et al. (1999)</td>
<td>London, England</td>
<td>Cardiovascular disease</td>
<td>8-h</td>
<td>1.03 (1.00, 1.05)</td>
</tr>
<tr>
<td>Ballester et al. (2006)</td>
<td>14 Spanish cities</td>
<td>Cardiovascular disease</td>
<td>8-h warm season</td>
<td>1.04 (1.02, 1.06)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Cardiac disease</td>
<td>8-h warm season</td>
<td>1.04 (1.01, 1.07)</td>
</tr>
<tr>
<td>Ballester et al. (2006)</td>
<td>Valencia, Spain</td>
<td>Cardiovascular disease</td>
<td>8-h</td>
<td>0.94 (0.84, 1.06)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Cardiac disease</td>
<td>8-h</td>
<td>0.88 (0.75, 1.03)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Cerebrovascular disease</td>
<td>8-h</td>
<td>0.86 (0.72, 1.04)</td>
</tr>
<tr>
<td>Bell et al. (2008)</td>
<td>Taipei, Taiwan</td>
<td>Cerebrovascular disease</td>
<td>24-h</td>
<td>0.94 (0.87, 1.02)</td>
</tr>
<tr>
<td>Buadong et al. (2009)</td>
<td>Bangkok, Thailand</td>
<td>Cardiovascular disease</td>
<td>1-h</td>
<td>1.01 (1.00, 1.02)</td>
</tr>
<tr>
<td>Cakmak et al. (2006)a</td>
<td>10 Canadian cities</td>
<td>Cardiac disease</td>
<td>1-h max</td>
<td>1.02 (1.00, 1.04)</td>
</tr>
<tr>
<td>Chan et al. (2006)</td>
<td>Taipei, Taiwan</td>
<td>Cerebrovascular disease</td>
<td>1-h max</td>
<td>1.02 (1.01, 1.03)</td>
</tr>
<tr>
<td>Chang et al. (2005)</td>
<td>Taipei, Taiwan</td>
<td>Cardiovascular disease</td>
<td>24-h warm season</td>
<td>1.42 (1.33, 1.50)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h cold season</td>
<td>1.15 (1.04, 1.27)</td>
</tr>
<tr>
<td>Fung et al. (2005)</td>
<td>Windsor, Canada</td>
<td>Cardiovascular disease</td>
<td>1-h</td>
<td>1.02 (0.92, 1.13)</td>
</tr>
<tr>
<td>Halonen et al. (2009)</td>
<td>Helsinki, Finland</td>
<td>Cardiovascular disease</td>
<td>8-h max warm season</td>
<td>1.05 (0.96, 1.14)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-h max</td>
<td>1.01 (0.99, 1.03)</td>
</tr>
<tr>
<td>Katsouyanni et al.</td>
<td>14 U.S. cities</td>
<td>Cardiovascular disease</td>
<td>1-h max warm season</td>
<td>1.00 (0.97, 1.03)</td>
</tr>
<tr>
<td>(2009)</td>
<td></td>
<td></td>
<td>1-h max</td>
<td>1.00 (0.95, 1.04)</td>
</tr>
<tr>
<td></td>
<td>12 Canadian cities</td>
<td>Cardiovascular disease</td>
<td>1-h max warm season</td>
<td>0.98 (0.94, 1.02)</td>
</tr>
<tr>
<td></td>
<td>8 European cities</td>
<td>Cardiovascular disease</td>
<td>1-h max warm season</td>
<td>0.99 (0.96, 1.02)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1-h max</td>
<td>0.98 (0.94, 1.03)</td>
</tr>
<tr>
<td>Larrieu et al. (2007)</td>
<td>8 French cities</td>
<td>Cardiac disease</td>
<td>8-h max warm season</td>
<td>1.01 (0.98, 1.04)</td>
</tr>
<tr>
<td>Linn et al. (2000)</td>
<td>Los Angeles, California</td>
<td>Cardiovascular disease</td>
<td>24-h</td>
<td>0.99 (0.98, 1.00)</td>
</tr>
<tr>
<td>Middleton et al. (2008)</td>
<td>Nicosia, Cyprus</td>
<td>Cardiovascular disease</td>
<td>8-h max</td>
<td>1.09 (1.00, 1.18)</td>
</tr>
<tr>
<td>Morgan et al. (1998)</td>
<td>Sydney, Australia</td>
<td>Cardiac disease</td>
<td>1-h max</td>
<td>1.02 (0.99, 1.05)</td>
</tr>
<tr>
<td>Peel et al. (2007)</td>
<td>Atlanta, GA</td>
<td>Cardiovascular disease</td>
<td>8-h warm season</td>
<td>1.00 (0.98, 1.02)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Cerebrovascular disease</td>
<td>8-h warm season</td>
<td>1.03 (0.97, 1.08)</td>
</tr>
<tr>
<td>Petroeschevsky et al.</td>
<td>Brisbane, Australia</td>
<td>Cardiovascular disease</td>
<td>8-h</td>
<td>0.96 (0.92, 1.01)</td>
</tr>
<tr>
<td>(2001)</td>
<td></td>
<td></td>
<td>8-h</td>
<td>0.97 (0.93, 1.01)</td>
</tr>
<tr>
<td>Polaniecki et al. (1997)</td>
<td>London, England</td>
<td>Cardiovascular disease</td>
<td>8-h</td>
<td>0.98 (0.95, 1.02)</td>
</tr>
<tr>
<td>Study*</td>
<td>Location</td>
<td>Outcome</td>
<td>Averaging Time</td>
<td>Effect Estimate (95% CI)</td>
</tr>
<tr>
<td>---------------------</td>
<td>-------------------</td>
<td>--------------------------</td>
<td>----------------------</td>
<td>--------------------------</td>
</tr>
<tr>
<td>Prescott et al. (1998)</td>
<td>Edinburgh, Scotland</td>
<td>Cardiovascular disease</td>
<td>24-h</td>
<td>0.89 (0.78, 1.00)</td>
</tr>
<tr>
<td>Von Klot et al. (2005)</td>
<td>5 European cities</td>
<td>Cardiac disease</td>
<td>8-h max warm season</td>
<td>1.11 (1.00, 1.22)</td>
</tr>
<tr>
<td>Wong et al. (1999b)</td>
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<td>Cardiovascular disease</td>
<td>24-h</td>
<td>1.08 (1.03, 1.13)</td>
</tr>
<tr>
<td>Wong et al. (1999b)</td>
<td>Hong Kong</td>
<td>Cerebrovascular disease</td>
<td>24-h</td>
<td>0.95 (0.90, 1.01)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Cardiovascular disease</td>
<td>24-h warm season</td>
<td>1.01 (0.96, 1.06)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Cerebrovascular disease</td>
<td>24-h cold season</td>
<td>1.06 (1.02, 1.11)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Cardiovascular disease</td>
<td>24-h</td>
<td>0.99 (0.95, 1.04)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Cerebrovascular disease</td>
<td>24-h warm season</td>
<td>0.98 (0.90, 1.08)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Cerebrovascular disease</td>
<td>24-h cold season</td>
<td>1.02 (0.96, 1.10)</td>
</tr>
<tr>
<td>Yang et al. (2004)</td>
<td>Kaohsiung, Taiwan</td>
<td>Cardiovascular disease</td>
<td>24-h warm season</td>
<td>1.33 (1.26, 1.40)</td>
</tr>
<tr>
<td>Yang et al. (2004)</td>
<td>Kaohsiung, Taiwan</td>
<td>Cardiovascular disease</td>
<td>24-h cold season</td>
<td>1.05 (0.96, 1.15)</td>
</tr>
</tbody>
</table>

*Studies included in Figure 6-22.

Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Age groups of study populations were not specified or were adults with the exception of Katsouyanni et al. (2009), Fung et al. (2005), Wong et al. (1999a), and Prescott et al. (1998), which included only individuals aged 65+. Studies listed in alphabetical order.

Warm season defined as: March-October (Peel et al., 2007), May-October (Ballester et al., 2005; Wong et al., 1999a), May-September (Halonen et al., 2009), April-September (Katsouyanni et al., 2009; Larrieu et al., 2007; Von Klot et al., 2005), ≥ 20°C (Chang et al., 2005) and ≥ 25°C (Yang et al., 2004). Cold season defined as: November-April (Wong et al., 1999a), <20°C (Chang et al., 2005) and <25°C (Yang et al., 2004). December-March (Wong et al., 1999b).
Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Seasons depicted by colors: black: all year; red: warm season; light blue: cold season. Outcomes were all congestive heart failure, with the exception of Symons et al. (2006), which examined onset of congestive heart failure symptoms leading to a heart attack. Age groups of study populations were not specified or were adults with the exception of Wellenius et al. (2005) and Wong et al. (1999a), which included only individuals aged 65+. Studies organized by outcome and season and then listed in descending order of publication date.

Figure 6-23 Effect estimate (95% CI) per increment ppb increase in O₃ for congestive heart failure ED visits or hospital admissions.
Table 6-36  Effect estimate (95% CI) per increment ppb increase in O₃ for congestive heart failure ED visits or hospital admissions for studies in Figure 6-23.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Outcome</th>
<th>Averaging Time</th>
<th>Effect Estimate (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lee et al. (2007)</td>
<td>Kaohsiung, Taiwan</td>
<td>Congestive heart failure</td>
<td>24-h warm season</td>
<td>1.25 (1.15, 1.36)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Congestive heart failure</td>
<td>24-h cold season</td>
<td>1.24 (1.09, 1.41)</td>
</tr>
<tr>
<td>Peel et al. (2007)</td>
<td>Atlanta, GA</td>
<td>Congestive heart failure</td>
<td>8-h warm season</td>
<td>0.94 (0.89, 1.00)</td>
</tr>
<tr>
<td>Polaniecki et al. (1997)</td>
<td>London, England</td>
<td>Congestive heart failure</td>
<td>8-h</td>
<td>0.99 (0.95, 1.03)</td>
</tr>
<tr>
<td>Stieb et al. (2009)</td>
<td>7 Canadian cities</td>
<td>Congestive heart failure</td>
<td>24-h</td>
<td>1.03 (0.98, 1.07)</td>
</tr>
<tr>
<td>Symons et al. (2006)</td>
<td>Baltimore, MD</td>
<td>Onset of congestive heart failure symptoms leading to heart attack</td>
<td>8-h warm season</td>
<td>0.83 (0.49, 1.41)</td>
</tr>
<tr>
<td>Wellenius et al. (2005)</td>
<td>Allegheny county, PA</td>
<td>Congestive heart failure</td>
<td>24-h</td>
<td>0.98 (0.96, 1.01)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Congestive heart failure</td>
<td>24-h</td>
<td>1.11 (1.04, 1.80)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h warm season</td>
<td>1.09 (0.96, 1.23)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h cold season</td>
<td>1.16 (1.06, 1.27)</td>
</tr>
<tr>
<td>Yang (2008)</td>
<td>Taipei, Taiwan</td>
<td>Congestive heart failure</td>
<td>24-h warm season</td>
<td>1.39 (1.27, 1.51)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Congestive heart failure</td>
<td>24-h cold season</td>
<td>0.61 (0.52, 0.73)</td>
</tr>
<tr>
<td>Wong et al. (1999b)</td>
<td>Hong Kong</td>
<td>Congestive heart failure</td>
<td>24-h</td>
<td>1.25 (1.11, 1.41)</td>
</tr>
</tbody>
</table>

*Studies include those from Figure 6-23.

Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Outcomes were all congestive heart failure, with the exception of Symons et al. (2006), which examined onset of congestive heart failure symptoms leading to a heart attack. Age groups of study populations were not specified or were adults with the exception of Wellenius et al. (2005) and Wong et al. (1999a), which included only individuals aged 65+. Studies listed in alphabetical order.

Warm season defined as: March-October (Peel et al., 2007), April-November (Symons et al., 2006), May-October (Wong et al., 1999a) ≥ 20°C (Yang, 2008), and >25°C (Lee et al., 2007). Cold season defined as: November-April (Wong et al., 1999a), <20°C (Yang, 2008), and <25°C (Lee et al., 2007).
Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Seasons depicted by colors: black: all year; red: warm season; light blue: cold season. Age groups of study populations were not specified or were adults with the exception of Wong et al. (1999a) and Atkinson et al. (1999), which included only individuals aged 65+. Studies organized by outcome and season and then listed in descending order of publication date.

Figure 6-24  Effect estimate (95% CI) per increment ppb increase in O₃ for ischemic heart disease, coronary heart disease, myocardial infarction, and angina pectoris ED visits or hospital admissions.
Table 6-37  Effect estimate (95% CI) per increment ppb increase in O₃ for ischemic heart disease, coronary heart disease, myocardial infarction, and angina pectoris ED visits or hospital admissions for studies presented in Figure 6-24.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Outcome</th>
<th>Averaging Time</th>
<th>Effect Estimate (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atkinson et al. (1999)</td>
<td>London, England</td>
<td>Ischemic heart disease</td>
<td>8-h</td>
<td>0.97 (0.94, 1.01)</td>
</tr>
<tr>
<td>Bell et al. (2008)</td>
<td>Taipei, Taiwan</td>
<td>Ischemic heart disease</td>
<td>24-h</td>
<td>1.01 (0.91, 1.12)</td>
</tr>
<tr>
<td>Buadong et al. (2009)</td>
<td>Bangkok, Thailand</td>
<td>Ischemic heart disease</td>
<td>1-h</td>
<td>1.00 (0.98, 1.02)</td>
</tr>
<tr>
<td>Buadong et al. (2009)</td>
<td>Bangkok, Thailand</td>
<td>Myocardial infarction</td>
<td>1-h</td>
<td>0.97 (0.94, 1.01)</td>
</tr>
<tr>
<td>Halonen et al. (2009)</td>
<td>Helsinki, Finland</td>
<td>Coronary heart disease</td>
<td>8-h max warm season</td>
<td>0.99 (0.79, 1.25)</td>
</tr>
<tr>
<td>Hosseinpoor et al. (2005)</td>
<td>Tehran, Iran</td>
<td>Angina</td>
<td>8-h max</td>
<td>0.80 (0.70, 0.92)</td>
</tr>
<tr>
<td>Lanki et al. (2006)</td>
<td>5 European cities</td>
<td>Myocardial infarction</td>
<td>8-h max warm season</td>
<td>0.96 (0.92, 1.01)</td>
</tr>
<tr>
<td>Larrieu et al. (2007)</td>
<td>8 French cities</td>
<td>Ischemic heart disease</td>
<td>8-h max warm season</td>
<td>1.02 (0.98, 1.07)</td>
</tr>
<tr>
<td>Lee et al. (2003b)</td>
<td>Seoul, Korea</td>
<td>Ischemic heart disease</td>
<td>1-h max</td>
<td>1.07 (1.02, 1.13)</td>
</tr>
<tr>
<td>Peel et al. (2007)</td>
<td>Atlanta, GA</td>
<td>Ischemic heart disease</td>
<td>8-h warm season</td>
<td>1.00 (0.96, 1.05)</td>
</tr>
<tr>
<td>Poloniecki et al. (1997)</td>
<td>London, England</td>
<td>Myocardial infarction</td>
<td>8-h</td>
<td>0.98 (0.94, 1.02)</td>
</tr>
<tr>
<td>Rich et al. (2010)</td>
<td>New Jersey</td>
<td>Myocardial infarction</td>
<td>24-h</td>
<td>0.94 (0.79, 1.13)</td>
</tr>
<tr>
<td>Stieb et al. (2009)</td>
<td>7 Canadian cities</td>
<td>Myocardial infarction</td>
<td>2-h</td>
<td>1.00 (0.96, 1.04)</td>
</tr>
<tr>
<td>Von Klot et al. (2005)</td>
<td>5 European cities</td>
<td>Myocardial infarction</td>
<td>8-h max warm season</td>
<td>1.00 (0.83, 1.21)</td>
</tr>
<tr>
<td>Von Klot et al. (2005)</td>
<td>5 European cities</td>
<td>Angina</td>
<td>8-h max warm season</td>
<td>1.19 (1.05, 1.35)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Ischemic heart disease</td>
<td>24-h warm season</td>
<td>1.02 (0.94, 1.11)</td>
</tr>
<tr>
<td>Wong et al. (1999b)</td>
<td>Hong Kong</td>
<td>Ischemic heart disease</td>
<td>24-h cold season</td>
<td>1.02 (0.95, 1.09)</td>
</tr>
<tr>
<td>Zanobetti and Schwartz (2006)</td>
<td>Boston, MA</td>
<td>Myocardial infarction</td>
<td>24-h</td>
<td>0.98 (0.92, 1.03)</td>
</tr>
</tbody>
</table>

*Studies included from Figure 6-24.

Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Age groups of study populations were not specified or were adults with the exception of Wong et al. (1999a) and Atkinson et al. (1999), which included only individuals aged 65+. Studies listed in alphabetical order.

Warm season defined as: March-October (Peel et al., 2007), June-August (Lee et al., 2003b), May-September (Halonen et al., 2009), May-October (Buadong et al., 2009), and April-September (Larrieu et al., 2007; Lanki et al., 2006; Von Klot et al., 2005). Cold season defined as: November-April (Buadong et al., 2009).
Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Seasons depicted by colors: black: all year; red: warm season; light blue: cold season. Age groups of study populations were not specified or were adults with the exception of Villeneuve et al. (2006a), which included only individuals aged 65+, and Chan et al. (2006), which included only individuals aged 50+. Studies organized by outcome and season and then listed in descending order of publication date.

**Figure 6-25** Effect estimate (95% CI) per increment ppb increase in O₃ for stroke ED visits or hospital admissions.
### Table 6-38  Effect estimate (95% CI) per increment ppb increase in O₃ for stroke ED visits or hospital admissions for studies presented in Figure 6-25.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Outcome</th>
<th>Averaging Time</th>
<th>Effect Estimate (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chan et al. (2006)</td>
<td>Taipei, Taiwan</td>
<td>All/non-specified stroke</td>
<td>1-h max</td>
<td>1.01 (0.99, 1.03)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Ischemic stroke</td>
<td>1-h max</td>
<td>1.03 (0.99, 1.07)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Hemorrhagic stroke</td>
<td>1-h max</td>
<td>0.99 (0.92, 1.06)</td>
</tr>
<tr>
<td>Halonen et al. (2009)</td>
<td>Helsinki, Finland</td>
<td>All/non-specified stroke</td>
<td>8-h max warm season</td>
<td>1.08 (0.83, 1.41)</td>
</tr>
<tr>
<td>Larrieu et al. (2007)</td>
<td>8 French cities</td>
<td>All/non-specified stroke</td>
<td>8-h max warm season</td>
<td>0.98 (0.93, 1.02)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>Ischemic stroke</td>
<td>24-h</td>
<td>1.00 (0.88, 1.13)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h warm season</td>
<td>1.09 (0.91, 1.32)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h cold season</td>
<td>0.98 (0.80, 1.18)</td>
</tr>
<tr>
<td>Villeneuve et al. (2006a)</td>
<td>Edmonton, Canada</td>
<td>Hemorrhagic stroke</td>
<td>24-h</td>
<td>1.12 (0.88, 1.43)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h warm season</td>
<td>0.97 (0.76, 1.22)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Transient ischemic stroke</td>
<td>24-h</td>
<td>0.98 (0.87, 1.10)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h warm season</td>
<td>0.85 (0.70, 1.01)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h cold season</td>
<td>1.11 (0.93, 1.32)</td>
</tr>
</tbody>
</table>

*Studies included from Figure 6-25.

Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Age groups of study populations were not specified or were adults with the exception of Villeneuve et al. (2006a), which included only individuals aged 65+, and Chan et al. (2006), which included only individuals aged 50+. Studies listed in alphabetical order.

Warm season defined as: May-September (Halonen et al., 2009), and April-September (Larrieu et al., 2007; Villeneuve et al., 2006a). Cold season defined as: October-March (Villeneuve et al., 2006a).
Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Seasons depicted by colors: black: all year; red: warm season; light blue: cold season. Age groups of study populations were not specified or were adults with the exception of Wong et al. (1999a), which included only individuals aged 65+. Studies organized by outcome and season and then listed in descending order of publication date.

Figure 6-26  Effect estimate (95% CI) per increment ppb increase in O₃ for arrhythmia and dysrhythmia ED visits or hospital admissions.
Table 6-39  Effect estimate (95% CI) per increment ppb increase in O₃ for arrhythmia and dysrhythmia ED visits or hospital admissions for studies presented in Figure 6-26.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Outcome</th>
<th>Averaging Time</th>
<th>Effect Estimate (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Buadong et al. (2009)</td>
<td>Bangkok, Thailand</td>
<td>Arrhythmia</td>
<td>1-h</td>
<td>0.99 (0.95, 1.04)</td>
</tr>
<tr>
<td>Halonen et al. (2009)</td>
<td>Helsinki, Finland</td>
<td>Arrhythmia</td>
<td>8-h max warm season</td>
<td>1.04 (0.80, 1.35)</td>
</tr>
<tr>
<td>Peel et al. (2007)</td>
<td>Atlanta, GA</td>
<td>Dysrhythmia</td>
<td>8-h warm season</td>
<td>1.01 (0.97, 1.06)</td>
</tr>
<tr>
<td>Poloniecki et al. (1997)</td>
<td>London, England</td>
<td>Arrhythmia</td>
<td>8-h</td>
<td>1.02 (0.96, 1.07)</td>
</tr>
<tr>
<td>Stieb et al. (2009)</td>
<td>7 Canadian cities</td>
<td>Dysrhythmia</td>
<td>24-h</td>
<td>1.02 (0.95, 1.09)</td>
</tr>
<tr>
<td>Wong et al. (1999a)</td>
<td>Hong Kong</td>
<td>Arrhythmia</td>
<td>24-h</td>
<td>1.06 (0.99, 1.12)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h warm season</td>
<td>1.10 (0.96, 1.26)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>24-h cold season</td>
<td>1.11 (1.01, 1.23)</td>
</tr>
</tbody>
</table>

*Studies included from Figure 6-26.

Note: Change in O₃ standardized to 20 ppb for 24-hour averaging period, 30 ppb for 8-hour averaging period, and 40 ppb for 1-hour averaging period (see Section 2.5). Ozone concentrations in ppb. Age groups of study populations were not specified or were adults with the exception of (Wong et al., 1999a), which included only individuals aged 65+. Studies listed in alphabetical order. Warm season defined as: March-October (Peel et al., 2007), May-October (Wong et al., 1999a) and May-September (Halonen et al., 2009). Cold season defined as: November-April (Wong et al., 1999a).

6.3.2.8  Cardiovascular Mortality

As discussed within this section (Section 6.3), epidemiologic studies provide inconsistent evidence of an association between short-term O₃ exposure and cardiovascular effects. However, toxicological studies have demonstrated O₃-induced cardiovascular effects, specifically enhanced atherosclerosis and altered vascular function, which could lead to death. The 2006 O₃ AQCD provided evidence, primarily from single-city studies, of consistent positive associations between short-term O₃ exposure and cardiovascular mortality. Recent multicity studies conducted in the U.S., Canada, and Europe further support the association between short-term O₃ exposure and cardiovascular mortality.

As discussed in Section 6.2.7.2, the APHENA study (Katsouyanni et al., 2009) also examined associations between short-term O₃ exposure and mortality and found consistent positive associations for cardiovascular mortality in all-year analyses. However, in analyses restricted to the summer season, results were more variable with no evidence of an association in the Canadian dataset in the population <75 years of age, and evidence of associations persisting or increasing in magnitude in the Canadian (population ≥ 75 years of age), U.S., and European datasets. Additional multicity studies from the U.S. (Zanobetti and Schwartz, 2008b), Europe (Samoli et al., 2009), Italy (Stafoggia et al., 2010), and Asia (Wong et al., 2010) that conducted summer season and/or all-year analyses provide additional support for an association between short-term O₃ exposure and cardiovascular mortality (Figure 6-27).
Of the studies evaluated, only the APHENA study (Katsouyanni et al., 2009) and the Italian multicity study (Stafoggia et al., 2010) conducted an analysis of the potential for copollutant confounding of the O$_3$-cardiovascular mortality relationship. In the European dataset, when focusing on the natural spline model with 8 df/year (Section 6.2.7.2) and lag 1 results in order to compare results across study locations (Section 6.6.2.1), cardiovascular mortality risk estimates were robust to the inclusion of PM$_{10}$ in copollutant models in all-year analyses with more variability in the Canadian and U.S. datasets (i.e., cardiovascular O$_3$ mortality risk estimates were reduced or increased in copollutant models). In summer season analyses, cardiovascular O$_3$ mortality risk estimates were robust in the European dataset and attenuated but remained positive in the U.S. dataset. Similarly, in the Italian multicity study (Stafoggia et al., 2010), which was limited to the summer season, cardiovascular mortality risk estimates were robust to the inclusion of PM$_{10}$ in copollutant models. Based on the APHENA and Italian multicity results, O$_3$ cardiovascular mortality risk estimates appear to be robust to inclusion of PM$_{10}$ in copollutant models. However, in the U.S. and Canadian datasets there was evidence that O$_3$ cardiovascular mortality risk estimates are moderately to substantially sensitive (e.g., increased or attenuated) to PM$_{10}$. The mostly every-6th-day sampling schedule for PM$_{10}$ in the Canadian and U.S. datasets greatly reduced their sample size and limits the interpretation of these results.

### 6.3.2.9 Summary of Epidemiologic Studies

Overall, the available body of evidence examining the relationship between short-term exposures to O$_3$ concentrations and cardiovascular morbidity is inconsistent. Across studies, different definitions, i.e., ICD-9 diagnostic codes were used for both all-cause and cause-specific cardiovascular morbidity (Table 6-35, Table 6-36, Table 6-37, Table 6-38, and Table 6-39), which may contribute to inconsistency in results. However, within diagnostic categories, no consistent pattern of association was found with O$_3$. Generally, the studies summarized in this section used nearest air monitors to assess O$_3$ concentrations, with a few exceptions that used modeling or personal exposure monitors (these exceptions were noted throughout the previous sections). The inconsistencies in the associations observed between short-term O$_3$ and CVD morbidities are unlikely to be explained by the different exposure assignment methods used (see Section 4.6). The wide variety of biomarkers considered and the lack of consistency among definitions used for specific cardiovascular disease endpoints (e.g., arrhythmias, HRV) make comparisons across studies difficult. Despite the inconsistent evidence for an association between O$_3$ concentration and CVD morbidity, mortality studies indicate a consistent positive association between short-term O$_3$ exposure and cardiovascular mortality in multicity studies and a multicontinent study.
6.3.3 Toxicology

In the previous O\textsubscript{3} AQCDs (U.S. EPA, 2006b, 1996a) experimental animal studies have reported relatively few cardiovascular system alterations after exposure to O\textsubscript{3} and other photochemical oxidants. The limited amount of research directed at examining O\textsubscript{3}-induced cardiovascular effects has primarily found alterations in heart rate (HR), heart rhythm, and BP after O\textsubscript{3} exposure. Although O\textsubscript{3} induced changes in HR and core temperature (T\textsubscript{CO}) in a number of rat studies, these responses have not been reported or extensively studied in humans exposed to O\textsubscript{3} and may be unique to rodents.

According to recent animal toxicology studies, short-term O\textsubscript{3} exposure induces vascular oxidative stress and proinflammatory mediators, alters HR and HRV, and disrupts the regulation of the pulmonary endothelin system (study details are provided in Table 6-40). A number of these effects were variable between strains examined, suggesting a genetic component to development of O\textsubscript{3} induced cardiovascular effects. Further, recent studies provide evidence that extended O\textsubscript{3} exposure enhances the risk of ischemia-reperfusion (I/R) injury and atherosclerotic lesion development. Still, few studies have investigated the role of O\textsubscript{3} reaction products in these processes, but more evidence is provided for elevated inflammatory and reduction-oxidation (redox) cascades known to initiate these cardiovascular pathologies.

**Heart Rate, Rhythm, and Heart Rate Variability**

Studies (Arito et al., 1992; Arito et al., 1990; Uchiyama and Yokoyama, 1989; Yokoyama et al., 1989; Uchiyama et al., 1986) report O\textsubscript{3} exposure (0.2-1.0 ppm, 3 hours to 3 days) in rats decreased T\textsubscript{CO}, HR, and mean arterial pressure (MAP). In addition, O\textsubscript{3} exposure (0.1 – 1.0 ppm, 3 hours to 3 days) in rats induced arrhythmias, including increased PR interval and QRS complex, premature atrial contraction, and incomplete A-V block (Arito et al., 1990; Yokoyama et al., 1989; Uchiyama et al., 1986). The effects were more pronounced in adult and awake rats than in younger or sleeping animals, whereas no sex-related differences were noted in these O\textsubscript{3} induced outcomes (Uchiyama et al., 1986). However, these cardiovascular responses to O\textsubscript{3}, including decreased T\textsubscript{CO} and HR, could be attenuated by increased ambient temperatures and environmental stress and exhibited adaptation (Watkinson et al., 2003; Watkinson et al., 1993). These studies suggest that these responses to O\textsubscript{3} were the result of the rodent hypothermic response, which serves as a physiological and behavioral defense mechanism to minimize the irritant effects of O\textsubscript{3} inhalation, (Iwasaki et al., 1998; Arito et al., 1997). As humans do not appear to exhibit decreased HR, MAP, and T\textsubscript{CO} with routine environmental (Section 6.3.2) or controlled laboratory (Section 6.3.1) exposures to O\textsubscript{3}, caution must be used in extrapolating the results of these animal studies to humans.

Other studies have shown that O\textsubscript{3} can increase BP in animal models. Rats exposed to 0.6 ppm O\textsubscript{3} for 33 days had increased systolic pressure and HR (Revis et al., 1981).
Increased BP triggers the release of atrial natriuretic factor (ANF), which has been found in increased levels in the heart, lungs, and circulation of O₃ exposed (0.5 ppm) rats (Vesely et al., 1994a, b, c). Exposures to high concentrations of O₃ (1.0 ppm) have also been found to lead to heart and lung edema (Friedman et al., 1983), which could be the result of increased ANF levels. Thus, O₃ may increase blood pressure and HR, leading to increased ANF and tissue edema.

Recent studies report strain differences in HR and HRV in response to a 2-hour O₃ pretreatment followed by exposure to carbon black (CB) in mice (C3H/HeJ [HeJ], C57BL/6J [B6], and C3H/HeOuJ [OuJ]) (Hamade and Tankersley, 2009; Hamade et al., 2008). These mice strains were chosen from prior studies on lung inflammatory and hyperpermeability responses to be at increased risk (B6 and OuJ) or resistant (HeJ) to O₃-induced health effects (Kleeberger et al., 2000). HR decreased during O₃ pre-exposure for all strains, but recovered during the CB exposure (Hamade et al., 2008). Percent change in HRV parameters, SDNN (indicating total HRV) and rMSSD (indicating beat-to-beat HRV), were increased in both C3H mice strains, but not B6 mice, during O₃ pre-exposure and recovered during CB exposure when compared to the filtered air group. The two C3H strains differ by a mutation in the Toll-like receptor 4 (TLR4) gene, but these effects did not seem to be related to this mutation since similar responses were observed. Hamade et al. (2008) speculate that the B6 and C3H strains differ in mechanisms of HR response after O₃ exposure between withdrawal of sympathetic tone and increase of parasympathetic tone; however, no direct evidence for this conclusion was reported. The strain differences observed in HR and HRV suggest that genetic variability affects cardiac responses after acute air pollutant exposures.

Hamade and Tankersley (2009) continued this investigation of gene-environment interactions on cardiopulmonary adaptation of O₃ and CB induced changes in HR and HRV using the previously described (Hamade et al., 2008) daily exposure scheme for 3 consecutive days. By comparing day-1 interim values it is possible to observe that O₃ exposure increased SDNN and rMSSD, but decreased HR in all strains. Measures of HR and HRV in B6 and HeJ mice recovered to levels consistent with filtered air treated mice by day 3; however, these responses in OuJ mice remained suppressed. B6 mice had no change in respiratory rate (RR) after O₃ treatment, whereas HeJ mice on days 1 and 2 had increased RR and OuJ mice on days 2 and 3 exhibited increased RR. $V_T$ did not change with treatment among the strains. Overall, B6 mice were mildly responsive with rapid adaptation, whereas C3 mice were highly responsive with adaptation only in HeJ mice with regards to changes in cardiac and respiratory responses. HR and HRV parameters were not equally correlated with $V_T$ and RR between the three mice strains, which suggest that strains vary in the integration of the cardiac and respiratory systems. These complex interactions could help explain variability in interindividual responses to air pollution.

Hamade et al. (2010) expanded their investigation to explore the variation of these strain dependent cardiopulmonary responses with age. As was observed previously, all experimental mouse strains (B6, HeJ, and OuJ) exhibited decreased HR and
increased HRV after O$_3$ exposure. Younger O$_3$-exposed mice had a significantly lower HR compared to older exposed mice, indicating an attenuation of the bradycardic effect of O$_3$ with age. Younger mice also had a greater increase in rMSSD in HeJ and OuJ strains and SDNN in HeJ mice. Conversely, B6 mice had a slightly greater increase in SDNN in aged mice compared to the young mice. No change was observed in the magnitude of the O$_3$ induced increase of SDNN in OuJ mice or rMSSD in B6 mice. The B6 and HeJ mice genetically vary in respect to the nuclear factor erythroid 2-related factor 2 (Nrf-2). The authors propose that the genetic differences between the mice strains could be altering the formation of ROS, which tends to increase with age, thus modulating the changes in cardiopulmonary physiology after O$_3$ exposure.

Strain and age differences in HR and heart function were further investigated in B6 and 129S1/SvImJ (129) mice in response to a sequential O$_3$ and filtered air or CB exposure (Tankersley et al., 2010). Young 129 mice showed a decrease in HR after O$_3$ or O$_3$ and CB exposure. This bradycardia was not observed in B6 or older animals in this study, suggesting a possible alteration or adaptation of the autonomic nervous system activity with age. However, these authors did previously report bradycardia in similarly aged young B6 mice (Hamade et al., 2010; Hamade and Tankersley, 2009; Hamade et al., 2008). Ozone exposure in 129 mice also resulted in an increase in left ventricular chamber dimensions at end diastole (LVEDD) in young and old mice and a decrease in left ventricular posterior wall thickness at end systole (PWTES) in older mice. The increase in LVEDD caused a decrease in fractional shortening, which can be used as a rough indicator of left ventricular function. Regression analysis revealed a significant interaction between age and strain on HR and PWTES, which implies that aging affects HR and heart function in response to O$_3$ differently between mouse strains.

**Vascular Disease and Injury**

A recent study in young mice (C57Bl/6) and rhesus monkeys examined the effects of short-term O$_3$ exposure (0.5 ppm, 1 or 5 days) on a number of cardiovascular endpoints (Chuang et al., 2009). Mice exposed to O$_3$ for 5 days had increased HR as well as mean and diastolic blood pressure. This is in contrast to the bradycardia that was reported in 18-20 week-old B6 mice treated with O$_3$, as described above (Hamade and Tankersley, 2009; Hamade et al., 2008). Increased blood pressure could be explained by the inhibition in endothelial-dependent (acyethylcholine) vasorelaxation from decreased bioavailability of aortic nitric oxide (NO). Ozone caused a decrease in aortic NOX (nitrite and nitrate levels) and a decrease in total, but not phosphorylated, endothelial nitric oxide synthase (eNOS). Ozone also increased vascular oxidative stress in the form of increased aortic and lung lipid peroxidation (F2-isoprostane), increased aortic protein nitration (3-nitrotyrosine), decreased aortic superoxide dismutase (SOD2) protein and activity, and decreased aortic aconitase activity, indicating specific inactivation by O$_2^-$ and ONOO$^-$ Mitochondrial DNA (mtDNA) damage was also used as a measure of oxidative and nitrative stress in mice and infant rhesus monkeys exposed to O$_3$. Chuang et al. (2009) observed that
mtDNA damage accumulated in the lung and aorta of mice after 1 and 5 days of O₃ exposure and in the proximal and distal aorta of O₃ treated nonhuman primates. Additionally, genetically hyperlipidemic mice exposed to O₃ (0.5 ppm) for 8 weeks had increased aortic atherosclerotic lesion area (Section 7.3.1), which may be associated with the short-term exposure changes discussed. Overall, this study suggests that O₃ initiates an oxidative environment by increasing O₂⁻ production, which leads to mtDNA damage and ·NO consumption, known to perturb endothelial function (Chuang et al., 2009). Endothelial dysfunction is characteristic of early and advanced atherosclerosis and coincides with impaired vasodilation and blood pressure regulation.

Vascular occlusion resulting from atherosclerosis can block blood flow causing ischemia. The restoration of blood flow in the vessel or reperfusion can cause injury to the tissue from subsequent inflammation and oxidative damage. Perepu et al. (2010) observed that O₃ exposure (0.8 ppm, 28 or 56 days) enhanced the sensitivity to myocardial I/R injury in Sprague-Dawley rats while increasing oxidative stress levels and pro-inflammatory mediators and decreasing production of anti-inflammatory proteins. Ozone was also found to decrease the left ventricular developed pressure, rate of change of pressure development, and rate of change of pressure decay while increasing left ventricular end diastolic pressure in isolated perfused hearts. In this ex vivo heart model, O₃ induced oxidative stress by decreasing SOD enzyme activity and increasing malondialdehyde levels. Ozone also elicited a proinflammatory state which was evident by an increase in TNF-α and a decrease in the anti-inflammatory cytokine IL-10. Perepu et al. (2010) concluded that O₃ exposure may result in a greater I/R injury.

**Effects on Cardiovascular-Related Proteins**

Increased BP, changes in HRV, and increased atherosclerosis may be related to increases in the vasoconstrictor peptide, endothelin-1 (amino acids 1-21, ET-1[1-21]). Regulation of the pulmonary endothelin system can be affected in rats by inhalation of PM (0, 5, 50 mg/m³, EHC-93) and O₃ (Thomson et al., 2006; Thomson et al., 2005). Exposure to either O₃ (0.8 ppm) or PM increased plasma ET-1[1-21], ET-3[1-21], and the ET-1 precursor peptide, bigET-1. Increases in circulating ET-1[1-21] could be a result of a transient increase in the gene expression of lung preproET-1 and endothelin converting enzyme-1 (ECE-1) immediately following inhalation of O₃ or PM. These latter gene expression changes (e.g., preproET-1 and ECE-1) were additive with co-exposure to O₃ and PM. Conversely, preproET-3 decreased immediately after O₃ exposure, suggesting the increase in ET-3[1-21] was not through de novo production. A recent study also found increased ET-1 gene expression in the aorta of O₃-exposed rats (Kodavanti et al., 2011). These rats also exhibited an increase in ETₐR after O₃ exposure; however, they did not demonstrate increased biomarkers for vascular inflammation, thrombosis, or oxidation.

Ozone can oxidize protein functional groups and disturb the affected protein. For example, the soluble plasma protein fibrinogen is oxidized by O₃ (0.01-0.03 ppm) in
vitro, creating fibrinogen and fibrin aggregates, characteristically similar to defective fibrinogen (Rosenfeld et al., 2009; Rozenfeld et al., 2008). In these studies, oxidized fibrinogen retained the ability to form fibrin gels that are involved in coagulation, however the aggregation time increased and the gels were rougher than normal with thicker fibers. Oxidized fibrinogen also developed the ability to self assemble creating fibrinogen aggregates that may play a role in thrombosis. Since O3 does not readily translocate past the ELF and pulmonary epithelium and fibrinogen is primarily a plasma protein, it is uncertain if O3 would have the opportunity to react with plasma fibrinogen. However, fibrinogen can be released from the basolateral face of pulmonary epithelial cells during inflammation, where the deposition of fibrinogen could lead to lung injury (Lawrence and Simpson-Haidaris, 2004).

**Studies on Ozone Reaction Products**

Although toxicological studies have demonstrated O3-induced effects on the cardiovascular system, it remains unclear if the mechanism is through a reflex response or the result of effects from O3 reaction products (U.S. EPA, 2006b, 1996a). Oxysterols derived from cholesterol ozonation, such as β-epoxide and 5β,6β-epoxycholesterol (and its metabolite cholestan-6-oxo-3,5-diol), have been implicated in inflammation associated with cardiovascular disease (Pulfer et al., 2005; Pulfer and Murphy, 2004). Two other cholesterol ozonolysis products, atheronal-A and -B (e.g., cholesterol secoaldehyde), have been found in human atherosclerotic plaques and shown in vitro to induce foam cell formation and induce cardiomyocyte apoptosis and necrosis (Sathishkumar et al., 2005; Wentworth et al., 2003); however, these products have not been found in the lung compartment or systemically after O3 exposure. The ability to form these cholesterol ozonation products in the circulation in the absence of O3 exposure complicates their implication in O3 induced cardiovascular disease.

Although it has been proposed that O3 reaction products released after the interaction of O3 with ELF constituents (see Section 5.2.3) on O3 interaction with ELF) are responsible for systemic effects, it is not known whether they gain access to the vascular space. Alternatively, extrapulmonary release of diffusible mediators, such as cytokines or endothelins, may initiate or propagate inflammatory responses in the vascular or systemic compartments (Cole and Freeman, 2009) (Section 5.3.8). Ozone reacts within the lung to amplify ROS production, induce pulmonary inflammation, and activate inflammatory cells, resulting in a cascading proinflammatory state and extrapulmonary release of diffusible mediators that could lead to cardiovascular injury.

A recent study that examined O3 reaction byproducts has shown that cholesterol secoaldehyde (e.g., atheronal A) induces apoptosis in vitro in mouse macrophages (Gao et al., 2009b) and cardiomyocytes (Sathishkumar et al., 2009). Additionally, atheronal-A and -B has been found to induce in vitro macrophage and endothelial cell proinflammatory events involved in the initiation of atherosclerosis (Takeuchi et al., 2006). These O3 reaction products when complexed with low density lipoprotein
upregulate scavenger receptor class A and induce dose-dependent macrophage chemotaxis. Atheronal-A increases expression of the adhesion molecule, E-selectin, in endothelial cells, while atheronal-B induces monocyte differentiation. These events contribute to both monocyte recruitment and foam cell formation in atherosclerotic vessels. It is unknown whether these OЗ reaction products gain access to the vascular space from the lungs. Alternative explanations include the extrapulmonary release of diffusible mediators that may initiate or propagate inflammatory responses in the vascular or systemic compartments.
## Table 6-40  Characterization of study details for Section 6.3.3.

<table>
<thead>
<tr>
<th>Study(^a)</th>
<th>Model</th>
<th>(O_3) (ppm)</th>
<th>Exposure Duration</th>
<th>Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chuang et al. (2009)</td>
<td>Mice; C57Bl/6; M; 6 weeks</td>
<td>0.5</td>
<td>1 or 5 days, 8-h/day</td>
<td>Increased HR and blood pressure. Initiated an oxidative environment by increasing vascular (O_3^+) production, which lead to mtDNA damage and (\cdot)NO consumption, known to perturb endothelial function.</td>
</tr>
<tr>
<td></td>
<td>Monkey; rhesus Macaca mulatta; M; Infant (180 days old)</td>
<td>0.5</td>
<td>5 days, 8-h/day</td>
<td>Increased aortic mtDNA damage.</td>
</tr>
<tr>
<td>Perepu et al. (2010)</td>
<td>Rat; Sprague-Dawley; 50-75 g</td>
<td>0.8</td>
<td>28 days, 8-h/day</td>
<td>Enhanced the sensitivity to myocardial I/R injury while increasing oxidative stress and pro-inflammatory mediators and decreasing production of anti-inflammatory proteins.</td>
</tr>
<tr>
<td>Hamade et al. (2008)</td>
<td>Mice; C57Bl/6J, C3H/HeJ, and C3H/HeOuJ; M; 18-20 weeks</td>
<td>0.6</td>
<td>2-h followed by 3 h of CB</td>
<td>Decreased HR. Strain differences observed in HRV suggest that genetic variability affects cardiac responses.</td>
</tr>
<tr>
<td>Hamade and Tankersley (2009)</td>
<td>Mice; C57Bl/6J, C3H/HeJ, and C3H/HeOuJ; M; 18-20 weeks</td>
<td>0.6 (subsequent CB exposure, 536 (\mu)g/m(^3))</td>
<td>3 days, 2-h/day followed by 3-h of CB</td>
<td>Strains varied in integration of the cardiac and respiratory systems, implications in interindividual variability. B6 mice were mildly responsive with rapid adaptation, whereas C3 mice were highly responsive with adaptation only in HeJ mice with regards to changes in cardiac and respiratory responses.</td>
</tr>
<tr>
<td>Hamade et al. (2010)</td>
<td>Mice; C57Bl/6J, C3H/HeJ, and C3H/HeOuJ; M; 5 or 12 mo old</td>
<td>0.6 (subsequent CB exposure, 536 (\mu)g/m(^3))</td>
<td>2-h followed by 3-h of CB</td>
<td>Aged mice exhibited attenuated changes in cardiopulmonary physiology after (O_3) exposure. Genetic differences between mice strains could be altering formation of ROS, which tends to increase with age, thus modulating (O_3) induced effects.</td>
</tr>
<tr>
<td>Tankersley et al. (2010)</td>
<td>Mice; C57Bl/6J, 129S1/SvlmJ; Mi/F; 5 or 18 mo old</td>
<td>0.6 (subsequent CB exposure, 556 (\mu)g/m(^3))</td>
<td>2-h followed by 3-h of CB</td>
<td>Significant interaction between age and strain on HR and PWTES, which implies that aging affects the HR and function in response to (O_3) differently between mouse strains.</td>
</tr>
<tr>
<td>Thomson et al. (2005)</td>
<td>Rat; Fischer-344; M; 200-250 g</td>
<td>0.4 or 0.8</td>
<td>4-h</td>
<td>Activation of the vasoconstricting ET system. Increased plasma ET-1 through higher production and slower clearance.</td>
</tr>
<tr>
<td>Thomson et al. (2006)</td>
<td>Rat; Fischer-344; M; 200-250 g</td>
<td>0.8</td>
<td>4-h</td>
<td>Increased plasma ET-3 not due to de novo synthesis, unlike ET-1.</td>
</tr>
<tr>
<td>Kodavanti et al. (2011)</td>
<td>Rat; Wistar; M; 10-12 weeks</td>
<td>0.5 or 1.0</td>
<td>2 days, 5-h/day</td>
<td>No changes to aortic genes of thrombosis, inflammation, or proteolysis, except ET-1 and ETBR (1.0 ppm).</td>
</tr>
</tbody>
</table>

\(^a\)Results from previous studies are presented in Annex Table AX5-14 of the 2006 \(O_3\) AQCD and Table 6-23 of the 1996 \(O_3\) AQCD.

\(^*\)Study details for Section 6.3.3.
Summary of Toxicological Studies

Overall, animal studies suggest that O\textsubscript{3} exposure may result in O\textsubscript{3} induced cardiovascular effects. Studies provide evidence for both increased and decreased HR, however it is uncertain if O\textsubscript{3}-induced bradycardia would also occur in humans or if it is due solely to a rodent hypothermic response. Animal studies also provide evidence for increased HRV, arrhythmias, vascular disease, and injury following short-term O\textsubscript{3} exposure. In addition, a series of studies highlight the role of gene-environment interactions and age in the induction of effects and attenuation of responses to O\textsubscript{3} exposure.

Biologically plausible mechanisms are present for the cardiovascular effects observed in animal exposure studies. Further discussion of the modes of action that may lead to cardiovascular effects can be found in Section 5.3.8. Recent studies suggest that O\textsubscript{3} exposure may disrupt both the NO\textsuperscript{•} and endothelin systems, which can result in an increase in HR, HRV, and ANF. The observed bradycardia following O\textsubscript{3} exposure may be the result of reflex reactions, including the trigeminocardiac reflex, evoked following the stimulation of sensory receptors lining the nose and RT. These mechanisms of parasympathetically-derived cardiac effects are described in more detail in Section 5.3.2. Additionally, O\textsubscript{3} may increase oxidative stress and vascular inflammation promoting the progression of atherosclerosis and leading to increased susceptibility to I/R injury. As O\textsubscript{3} reacts quickly with the ELF and does not translocate to the heart and large vessels, studies suggest that the cardiovascular effects exhibited could be caused by reaction byproducts of O\textsubscript{3} exposure. However, direct evidence of translocation of O\textsubscript{3} reaction products to the cardiovascular system has not been demonstrated in vivo. Alternatively, extrapulmonary release of diffusible mediators, such as cytokines or endothelins, may initiate or propagate inflammatory responses in the vascular or systemic compartments leading to the reported cardiovascular pathologies.

6.3.4 Summary and Causal Determination

In previous O\textsubscript{3} reviews (U.S. EPA, 2006b, 1996a), very few studies were described which examined the effect of short-term O\textsubscript{3} exposure on the cardiovascular system. More recently, the body of scientific evidence available that has examined the effect of O\textsubscript{3} on the cardiovascular system has expanded.

Toxicological studies, although limited in number, provide evidence of O\textsubscript{3}-induced cardiovascular effects. These include enhanced I/R injury, disrupted NO-induced vascular reactivity, decreased cardiac function, increased vascular disease, and increased HRV following short-term O\textsubscript{3} exposure. A number of these effects have also been observed following long-term O\textsubscript{3} exposure (see Section 7.3.1.2). Results of studies investigating the role of O\textsubscript{3} in heart rate regulation are mixed with both bradycardic and tachycardic responses observed in animal models.

The cardiovascular effects of O\textsubscript{3} found in animals may, in part, correspond to alteration of the autonomic nervous system or to the development and maintenance
of systemic oxidative stress and a proinflammatory environment that may result from pulmonary inflammation.

Controlled human exposure studies also suggest cardiovascular effects in response to short-term O₃ exposure and provide some coherence with evidence from animal toxicology studies. Increases and decreases in high frequency HRV have been reported following relatively low (120 ppb during rest) and high (300 ppb with exercise) O₃ exposures, respectively. These changes in cardiac function observed in animal and human studies provide preliminary evidence for O₃-induced modulation of the autonomic nervous system through the activation of neural reflexes in the lung (see Section 5.3.2). Controlled human exposure studies also support the animal toxicology studies by demonstrating O₃-induced effects on blood biomarkers of systemic inflammation and oxidative stress as well as changes in biomarkers suggestive of a prothrombogenic response to O₃.

The experimental evidence provides initial biological plausibility for the consistently positive associations observed in epidemiologic studies of short-term O₃ exposure and cardiovascular mortality. These include studies reviewed in the 2006 O₃ AQCD, recent multicity studies, and the multicontinent APHENA study. The few studies that examined copollutant confounding found that associations with cardiovascular mortality remain robust in copollutant models with PM. However, epidemiologic studies generally do not observe associations between short-term exposure to O₃ and cardiovascular morbidity; studies of cardiovascular-related hospital admissions and ED visits and other various cardiovascular effects did not find consistent evidence of a relationship with O₃ exposure. The lack of coherence between the results from studies that examined associations between short-term O₃ exposure and cardiovascular morbidity and cardiovascular mortality complicate the interpretation of the overall evidence for O₃-induced cardiovascular effects.

In conclusion, animal toxicological studies demonstrate O₃-induced cardiovascular effects, and support the strong body of evidence indicating O₃-induced cardiovascular mortality. Animal toxicological and controlled human exposure studies provide evidence for biologically plausible mechanisms underlying these O₃-induced cardiovascular effects. However, a lack of coherence with epidemiologic studies of cardiovascular morbidity remains an important uncertainty. Taken together, the overall body of evidence across disciplines is sufficient to conclude that there is likely to be a causal relationship between relevant short-term exposures to O₃ and cardiovascular effects.

6.4 Central Nervous System Effects

The 2006 O₃ AQCD (U.S. EPA, 2006b) included toxicological evidence indicating that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short and long term memory, and sleep patterns. Additionally, histological signs of neurodegeneration have been observed. Reports of headache, dizziness, and irritation of the nose with O₃ exposure are common complaints in humans, and some
behavioral changes in animals may be related to these symptoms rather than indicative of neurotoxicity. Peterson and Andrews (1963) and Tepper et al. (1983) showed that mice would alter their behavior to avoid O₃ exposure. Murphy et al. (1964) and Tepper et al. (1982) showed that running-wheel behavior was suppressed, and Tepper et al. (1985) subsequently demonstrated the effects of a 6-hour exposure to O₃ on the suppression of running-wheel behavior in rats and mice, with the lowest effective concentration being about 0.12 ppm O₃ in the rat and about 0.2 ppm in the mouse. The suppression of active behavior by 6 hours of exposure to 0.12 ppm O₃ has recently been confirmed by Martrette et al. (2011) in juvenile female rats, and the suppression of three different active behavior parameters was found to become more pronounced after 15 days of exposure. A table of studies examining the effects of O₃ on behavior can be found on p 6-128 of the 1996 O₃ AQCD. Generally speaking, transient changes in behavior in rodent models appear to be dependent on a complex interaction of factors such as (1) the type of behavior being measured, with some behaviors increased and others suppressed; (2) the factors motivating that behavior (differences in reinforcement); and (3) the sensitivity of the particular behavior (e.g., active behaviors are more affected than more sedentary behaviors). Many behavioral changes are likely to result from avoidance of irritation, but more recent studies indicate that O₃ also directly affects the CNS.

Research in the area of O₃-induced neurotoxicity has notably increased over the past few years, with the majority of the evidence coming from toxicological studies that examined the association between O₃ exposure, neuropathology, and neurobehavioral effects. As discussed below, these studies demonstrated that exposure to O₃ can produce a variety of CNS effects including behavioral deficits, morphological changes, and oxidative stress in the brains of rodents. In these rodent studies, interestingly, CNS effects were reported at O₃ concentrations that were generally lower than those concentrations commonly observed to produce pulmonary or cardiac effects in rats. A recent epidemiologic study provides new evidence, which is coherent with the toxicological evidence indicating that ambient O₃ exposure may result in cognitive function decrements. This study is discussed in detail in Chapter 7 (Section 7.5.1) because it focuses on long-term exposures to O₃.

A number of new studies demonstrate various perturbations in neurologic function or histology, including changes similar to those observed with Parkinson’s and Alzheimer’s disease pathologies occurring in similar regions of the brain (Table 6-41). Many of these include exposure durations ranging from short-term to long-term, and as such are discussed here and in Chapter 7 with emphasis on the effects resulting from exposure durations relevant to the respective chapter. Several studies assess short- and long-term memory acquisition via passive avoidance behavioral testing and find decrements in test performance after O₃ exposure, consistent with the aforementioned observation made in humans by Chen and Schwartz (2009). Impairment of long-term memory has been previously described in rats exposed to 0.2 ppm O₃ for 4 hours (Rivas-Arancibia et al., 1998) and in other studies of 4-hour exposures at concentrations of 0.7 to 1 ppm (Dorado-Martinez et al., 2001; Rivas-Arancibia et al., 2000; Avila-Costa et al., 1999). More recently,
statistically significant decreases in both short and long-term memory were observed in rats after 15 days of exposure to 0.25 ppm O₃ (Rivas-Arancibia et al., 2010).

The central nervous system is very sensitive to oxidative stress, due in part to its high content of polyunsaturated fatty acids, high rate of oxygen consumption, and low antioxidant enzyme capacity. Oxidative stress has been identified as one of the pathophysiological mechanisms underlying neurodegenerative disorders such as Parkinson’s and Alzheimer’s disease, among others (Simonian and Coyle, 1996). It is also believed to play a role in altering hippocampal function, which causes cognitive deficits with aging (Vanguilder and Freeman, 2011). A particularly common finding in studies of O₃-exposed rats is lipid peroxidation in the brain, especially in the hippocampus, which is important for higher cognitive function including contextual memory acquisition. Performance in passive avoidance learning tests is impaired when the hippocampus is injured, and the observed behavioral effects are well correlated with histological and biochemical changes in the hippocampus, including reduction in spine density in the pyramidal neurons (Avila-Costa et al., 1999), lipoperoxidation (Rivas-Arancibia et al., 2010; Dorado-Martinez et al., 2001), progressive neurodegeneration, and activated and phagocytic microglia (Rivas-Arancibia et al., 2010). The hippocampus is also one of the main regions affected by age-related neurodegenerative diseases, including Alzheimer’s disease, and it may be more sensitive to oxidative damage in aged rats. In a study of young (47 days) and aged (900 days) rats exposed to 1 ppm O₃ for 4 hours, O₃-induced lipid peroxidation occurred to a greater extent in the striatum of young rats, whereas it was highest in the hippocampus in aged rats (Rivas-Arancibia et al., 2000). Martínez-Canabal and Angora-Perez (2008) showed exposure of rats to 0.25 ppm, 4h/day, for 7, 15, or 30 days increased lipoperoxides in the hippocampus. This effect was observed at day 7 and continued to increase with time, indicating cumulative oxidative damage. O₃-induced changes in lipid peroxidation, neuronal death, and COX-2 positive cells in the hippocampus could be significantly inhibited by daily treatment with growth hormone (GH), which declines with age in most species. The protective effect of GH on -induced oxidative stress was greatest at 15 days of exposure and was non-significant at day 30. Consistent with these findings, lipid peroxidation in the hippocampus of rats was observed to increase significantly after a 30-day exposure to 0.25 ppm, but not after a single 4-hour exposure to the same concentration (Mokoena et al., 2010). However, 4 hours of exposure was sufficient to cause significant increases in lipid peroxidation when the concentration was increased to 0.7 ppm, and another study observed lipid peroxidation after a 4-hour exposure to 0.4 ppm (Dorado-Martinez et al., 2001).

Other commonly affected areas of the brain include the striatum, substantia nigra, cerebellum, olfactory bulb, and frontal/prefrontal cortex. The striatum and substantia nigra are particularly sensitive to oxidative stress because the metabolism of dopamine, central to their function, is an oxidative process perturbed by redox imbalance. Oxidative stress has been implicated in the premature death of substantia nigra dopamine neurons in Parkinson’s disease. Angoa-Pérez et al. (2006) have shown progressive lipoperoxidation in the substantia nigra and a decrease in nigral dopamine neurons in ovariectomized female rats exposed to 0.25 ppm O₃, 4h/day,
for 7, 15, or 30 days. Estradiol, an antioxidant, attenuated O₃-induced oxidative stress and nigral neuronal death, and the authors note that in humans, estrogen therapy can ameliorate symptoms of Parkinson’s disease, which is more prevalent in men. Progressive oxidative stress has also been observed in the striatum and substantia nigra of rats after 15 and 30 days of exposure to 0.25 ppm O₃ for 4 hours/day, along with a loss of dopaminergic neurons from the substantia nigra (Pereyra-Muñoz et al., 2006). Decreases in motor activity were also observed at 15 and 30 days of exposure, consistent with other reports (Martrette et al., 2011; Dorado-Martínez et al., 2001). Using a similar O₃ exposure protocol, Santiago-López et al. (2010) also observed a progressive loss of dopaminergic neurons within the substantia nigra, accompanied by alterations in the morphology of remaining cells and an increase in p53 levels and nuclear translocation.

The olfactory bulb also undergoes oxidative damage in O₃ exposed animals, in some cases altering olfactory-dependent behavior. Lipid peroxidation was observed in the olfactory bulbs of ovariectomized female rats exposed to 0.25 ppm O₃ (4 hours/day) for 30 or 60 days (Guevara-Guzmán et al., 2009). Ozone also induced decrements in a selective olfactory recognition memory test, and the authors note that early deficits in odor perception and memory are components of human neurodegenerative diseases. The decrements in olfactory memory were not due to damaged olfactory perception based on other tests. However, deficits in olfactory perception emerged with longer exposures (discussed in Chapter 7). As with the study by Angoa-Pérez et al. (2006) described above, a protective effect for estradiol was demonstrated for both lipid peroxidation and olfactory memory defects. The role of oxidative stress in memory deficits and associated morphological changes has also been demonstrated via attenuation by other antioxidants as well, such as α-tocopherol (Guerrero et al., 1999) and taurine (Rivas-Arancibia et al., 2000). It is unclear how persistent these effects might be. One study of acute exposure, using 1 ppm O₃ for 4 hours, observed morphological changes in the olfactory bulb of rats at 2 hours, and 1 and 10 days, but not 15 days, after exposure (Colín-Barenque et al., 2005).

Other acute studies also report changes in the CNS. Lipid peroxidation was observed in multiple regions of the brain after a 1- to 9-hour exposure to 1 ppm O₃ (Escalante-Membrillo et al., 2005). Ozone has also been shown to alter gene expression of endothelin-1 (pituitary) and inducible nitric oxide synthase (cerebral hemisphere) after a single 4-hour exposure to 0.8 ppm O₃, indicating potential cerebrovascular effects. This concentration-dependent effect was not observed at 0.4 ppm O₃ (Thomson et al., 2007). Vascular endothelial growth factor was upregulated in astroglial cells in the central respiratory areas of the brain of rats exposed to 0.5 ppm O₃ for 3 hours (Araneda et al., 2008). The persistence of CNS changes after a single exposure was also examined and the increase in vascular endothelial growth factor was present after a short (3 hours) recovery period. Thus, there is evidence that O₃-induced CNS effects are both concentration- and time-dependent.

Because O₃ can produce a disruption of the sleep-wake cycle (U.S. EPA, 2006b), Alfaro-Rodriguez and González-Piña (2005) examined whether acetylcholine in a region of the brain involved in sleep regulation was altered by O₃. After a 24-hour
exposure to 0.5 ppm O₃, the acetylcholine concentration in the medial preoptic area was decreased by 58% and strongly correlated with a disruption in paradoxical sleep. Such behavioral-biochemical effects of O₃ are confirmed by a number of studies which have demonstrated morphological and biochemical changes in rats.

CNS effects have also been demonstrated in newborn and adult rats whose only exposure to O₃ occurred in utero. Several neurotransmitters were assessed in male offspring of dams exposed to 1 ppm O₃ during the entire pregnancy (Gonzalez-Pina et al., 2008). The data showed that catecholamine neurotransmitters were affected to a greater degree than indole-amine neurotransmitters in the cerebellum. CNS changes, including behavioral, cellular, and biochemical effects, have also been observed after in utero exposure to 0.5 ppm O₃ for 12 hours/day from gestational days 5-20 (Boussouar et al., 2009). Tyrosine hydroxylase labeling in the nucleus tractus solatarius was increased after in utero exposure to O₃ whereas Fos protein labeling did not change. When these offspring were challenged by immobilization stress, neuroplasticity pathways, which were activated in air-exposed offspring, were inhibited in O₃-exposed offspring. Although an O₃ exposure C-R was not studied in these two in utero studies, it has been examined in one study. Santucci et al. (2006) investigated behavioral effects and gene expression after in utero exposure of mice to as little as 0.3 ppm O₃. Increased defensive/submissive behavior and reduced social investigation were observed in both the 0.3 and 0.6 ppm O₃ groups. Changes in gene expression of brain-derived neurotrophic factor (BDNF, increased in striatum) and nerve growth factor (NGF, decreased in hippocampus) accompanied these behavioral changes. Thus, these three studies demonstrate that CNS effects can occur as a result of in utero exposure to O₃, and although the mode of action of these effects is not known, it has been suggested that circulating lipid peroxidation products may play a role (Boussouar et al., 2009). Importantly, these CNS effects occurred in rodent models after in utero only exposure to relevant concentrations of O₃.
Table 6-41 Central nervous system and behavioral effects of short-term O₃ exposure in rats.

<table>
<thead>
<tr>
<th>Study</th>
<th>Model</th>
<th>O₃ (ppm)</th>
<th>Exposure Duration</th>
<th>Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Martrette et al. (2011)</td>
<td>Rat; Wistar; F; Weight: 152g; 7 weeks old</td>
<td>0.12</td>
<td>1-15 days, 6 h/day</td>
<td>Significant decrease in rearing, locomotor activity, and jumping activity at day 1, with a further decrease in these activities by day 15.</td>
</tr>
<tr>
<td>Angoa-Pérez et al. (2006)</td>
<td>Rat; Wistar; F; Weight: 300g; ovariectomized</td>
<td>0.25</td>
<td>7 to 60 days, 4h/day, 5 days/week</td>
<td>Progressive lipid peroxidation and loss of tyrosine hydroxylase-immunopositive neurons in the substantia nigra starting at 7 days.</td>
</tr>
<tr>
<td>Guevara-Guzmán et al. (2009)</td>
<td>Rat; Wistar; F; 264g; ovariectomized</td>
<td>0.25</td>
<td>30 and 60 days, 4h/day</td>
<td>Estradiol treatment protected against lipid peroxidation and decreases in estrogen receptors and dopamine β-hydroxylase in olfactory bulbs along with deficits in olfactory recognition memory.</td>
</tr>
<tr>
<td>Martínez-Canabal and Angora-Perez (2008)</td>
<td>Rat; Wistar; M; Weight: 300g</td>
<td>0.25</td>
<td>7 to 30 days, 4h/day</td>
<td>Growth hormone inhibited O₃-induced increases in lipoperoxidation and COX-2 positive cells in the hippocampus.</td>
</tr>
<tr>
<td>Pereyra-Muñoz et al. (2006)</td>
<td>Rat; Wistar; M; 250-300g</td>
<td>0.25</td>
<td>15 and 30 days, 4h/day</td>
<td>Decreased motor activity, increased lipid peroxidation, altered morphology, and loss of dopamine neurons in substantia nigra and striatum, increased expression of DARPP-32, iNOS, and SOD.</td>
</tr>
<tr>
<td>Rivas-Arancibia et al. (2010)</td>
<td>Rat; Wistar; M; 250-300g</td>
<td>0.25</td>
<td>15 to 90 days, 4h/day</td>
<td>Ozone produced significant increases in lipid peroxidation in the hippocampus, and altered the number of p53 positive immunoreactive cells, activated and phagocytic microglia cells, GFAP immunoreactive cells, and doublecortine cells, and short- and long-term memory-retention latency.</td>
</tr>
<tr>
<td>Santiago-López et al. (2010)</td>
<td>Rat; Wistar; M; 250-300g</td>
<td>0.25</td>
<td>15, 30, and 60 days, 4h/day</td>
<td>Progressive loss of dopamine reactivity in the substantia nigra, along with morphological changes. Increased p53 levels and nuclear translocation.</td>
</tr>
<tr>
<td>Thomson et al. (2007)</td>
<td>Rat; Fischer-344; M; 200-250g</td>
<td>0.4; 0.8</td>
<td>4-h; assays at 0 and 24 h postexposure</td>
<td>At 0.8 ppm, O₃ produced rapid perturbations in the ET-NO pathway gene expression in the brain. Ozone induced a small but significant time- and concentration-dependent increase in prepro-endothelin-1 mRNA levels in the cerebral hemisphere and pituitary, whereas TNFa and iNOS mRNA levels were decreased at 0 h and unchanged or increased, respectively, at 24 h.</td>
</tr>
<tr>
<td>Alfaro-Rodríguez and González-Piña (2005)</td>
<td>Rat; Wistar; M; 292g</td>
<td>0.5</td>
<td>24-h</td>
<td>During the light phase, O₃ caused a significant decrease in paradoxical sleep accompanied by a significant decrease in Ach levels in the hypothalamic median preoptic area. The same effects occurred during the dark phase exposure to O₃ in addition to a significant increase in slow-wave sleep and decrease in wakefulness.</td>
</tr>
<tr>
<td>Araneda et al. (2008)</td>
<td>Rats; Sprague-Dawley; M; 280-320g</td>
<td>0.5</td>
<td>3h (measurements taken at 0 h and 3 h after exposure)</td>
<td>Ozone upregulated VEGF in astroglial cells located in the respiratory center of the brain. VEGF co-located with IL-6 and TNF in cells near blood vessel walls, and blood vessel area was markedly increased.</td>
</tr>
<tr>
<td>Study</td>
<td>Model</td>
<td>O₃ (ppm)</td>
<td>Exposure Duration</td>
<td>Effects</td>
</tr>
<tr>
<td>-----------------------</td>
<td>--------------------------------------------</td>
<td>----------</td>
<td>-------------------</td>
<td>--------------------------------------------------------------------------</td>
</tr>
<tr>
<td>Boussouar et al. (2009)</td>
<td>Rat; Sprague-Dawley; M; adult offspring of prenatally exposed dams; 403-414g</td>
<td>0.5</td>
<td>From embryonic day E5 to E20 for 1-h/day; immobilization stress</td>
<td>Prenatal O₃ exposure had a long term impact on the nucleus tractus solitarius of adult rats, as revealed during immobilization stress.</td>
</tr>
<tr>
<td>Soulage et al. (2004)</td>
<td>Rat; Sprague-Dawley; M; Approx. 7 weeks old</td>
<td>0.7</td>
<td>5-h</td>
<td>Ozone produced differential effects on peripheral and central components of the sympato-adrenal system. While catecholamine biosynthesis was increased in portions of the brain, the catecholamine turnover rate was significantly increased in the heart and cerebral cortex and inhibited in the lung and striatum.</td>
</tr>
<tr>
<td>Calderón Guzmán et al. (2006); Calderón Guzmán et al. (2005)</td>
<td>Rat; Wistar; M; 21 days old; well-nourished and malnourished groups</td>
<td>0.75</td>
<td>15 successive days for 4-h/day</td>
<td>A significant decrease in body weight was observed in both well nourished (WN) and malnourished (MN) rats after O₃ exposure. Localized ATPase, TBARS, and GSH levels changed in response to O₃ in certain brain areas and the O₃-induced changes were dependent on nutritional condition.</td>
</tr>
<tr>
<td>Colin-Barenque et al. (2005)</td>
<td>Rats; Wistar; M; 250-300g</td>
<td>1.0</td>
<td>4-h; assays at 2-h, 24-h, 10 days, and 15 days after exposure</td>
<td>A significant loss of dendritic spines in granule cells of the olfactory bulb occurred at 2 hrs to 10 days after exposure. Cytological and ultrastructural changes returned toward normal morphology by 15 days.</td>
</tr>
<tr>
<td>Escalante-Membrillo et al. (2005)</td>
<td>Rats; Wistar; M; 280-320g</td>
<td>1.0</td>
<td>1-, 3-, 6-, or 9-h</td>
<td>Significant increases in TBARS occurred in hypothalamus, cortex, striatum, midbrain, thalamus, and pons. Partial but significant recovery was observed by 3 h after the 9 h exposure.</td>
</tr>
<tr>
<td>Gonzalez-Pina et al. (2008)</td>
<td>Rat; Wistar; M;</td>
<td>1</td>
<td>12-h/day, 21 days of gestation; assays at 0, 5, &amp; 10 days postnatal</td>
<td>Prenatal O₃ exposure produced significant decreases in cerebellar monoamine but not indolamine content at 0 and 5 days after birth with a partial recovery by 10 days. 5-hydroxy-indole-acetic acid levels were significantly increased at 10 days.</td>
</tr>
</tbody>
</table>

6.4.1 Neuroendocrine Effects

According to the 2006 O₃ AQCD, early studies suggested an interaction of O₃ with the pituitary-thyroid-adrenal axis, because thyroidectomy, hypophysectomy, and adrenalectomy protected against the lethal effects of O₃. Concentrations of 0.7-1.0 ppm O₃ for a 1-day exposure in male rats caused changes in the parathyroid, thymic atrophy, decreased serum levels of thyroid hormones and protein binding, and increased prolactin. Increased toxicity to O₃ was reported in hyperthyroid rats and T3 supplementation was shown to increase metabolic rate and pulmonary injury in the lungs of O₃-treated animals. The mechanisms by which O₃ affects neuroendocrine function are not well understood, but previous work suggests that high ambient levels of O₃ can produce marked neural disturbances in structures involved in the integration of chemosensory inputs, arousal, and motor control, effects that may be
responsible for some of the behavioral effects seen with O$_3$ exposure. A more recent study exposing immature female rats to 0.12 ppm O$_3$ demonstrated significantly increased serum levels of the thyroid hormone free T$_3$ after 15 days of exposure, whereas free T$_4$ was unchanged (Martrette et al., 2011). These results are in contrast to those previously presented whereby 1 ppm O$_3$ for 1 day significantly decreased T$_3$ and T$_4$ (Clemons and Garcia, 1980), although comparisons are made difficult by highly disparate exposure regimens along with sex differences. Martrette et al. (2011) also demonstrated significantly increased corticosterone levels after 15 days of exposure, suggesting a stress related response.

### 6.4.2 Summary and Causal Determination

In rodents, O$_3$ exposure has been shown to cause physicochemical changes in the brain indicative of oxidative stress and inflammation. Newer toxicological studies add to earlier evidence that acute exposures to O$_3$ can produce a range of effects on the central nervous system and behavior. Previously observed effects, including neurodegeneration, alterations in neurotransmitters, short and long term memory, and sleep patterns, have been further supported by recent studies. In instances where pathology and behavior are both examined, animals exhibit decrements in behaviors tied to the brain regions or chemicals found to be affected or damaged. For example, damage in the hippocampus, which is important for memory acquisition, was correlated with impaired performance in tests designed to assess memory. Thus the brain is functionally affected by O$_3$ exposure. The single epidemiologic study conducted showed an association between O$_3$ exposure and memory deficits in humans as well, albeit on a long-term exposure basis. Notably, exposure to O$_3$ levels as low as 0.25 ppm for 7 days has resulted in progressive neurodegeneration and deficits in both short and long-term memory in rodents. Examination of changes in the brain at lower exposure concentrations or at 0.25 ppm for shorter durations has not been reported, but 0.12 ppm O$_3$ has been shown to alter behavior. It is possible that some behavioral changes may reflect avoidance of irritation as opposed to functional changes in brain morphology or chemistry, but in many cases functional changes are related to oxidative stress and damage. In some instances, changes were dependent on the nutritional status of the rats (high versus low protein diet). For example, O$_3$ produced an increase in glutathione in the brains of rats fed the high protein diet but decreases in glutathione in rats fed low protein chow (Calderón Guzmán et al., 2006). The hippocampus, one of the main regions affected by age-related neurodegenerative diseases, appears to be more sensitive to oxidative damage in aged rats (Rivas-Arancibia et al., 2000), and growth hormone, which declines with age in most species, may be protective (Martínez-Canabal and Angora-Perez, 2008). Developing animals may also be sensitive, as changes in the CNS, including biochemical, cellular, and behavioral effects, have been observed in juvenile and adult animals whose sole exposure occurred in utero, at levels as a low as 0.3 ppm. A number of studies demonstrate O$_3$-induced changes that are also observed in human neurodegenerative disorders such as Alzheimer’s and Parkinson’s disease, including signs of oxidative stress, loss of neurons/neuronal death, reductions in
dopamine levels, increased COX-2 expression, and increases in activated microglia in important regions of the brain (hippocampus, substantia nigra).

Thus, evidence for neurological effects from epidemiologic and controlled human exposure studies is lacking. However, the toxicological evidence for the impact of O₃ on the brain and behavior is strong, and suggestive of a causal relationship between O₃ exposure and effects on the central nervous system.

### 6.5 Effects on Other Organ Systems

#### 6.5.1 Effects on the Liver and Xenobiotic Metabolism

Early investigations of the effects of O₃ on the liver centered on xenobiotic metabolism, and the prolongation of drug-induced sleeping time, which was observed at 0.1 ppm O₃ (Graham et al., 1981). In some species, only adults and especially females were affected. In rats, high (1.0-2.0 ppm for 3 hours) acute O₃ exposures caused increased production of NO by hepatocytes and enhanced protein synthesis (Laskin et al., 1996; Laskin et al., 1994). Except for the earlier work on xenobiotic metabolism, the responses occurred only after very high acute O₃ exposures. One study, conducted at 1 ppm O₃ exposure, has been identified (Last et al., 2005) in which alterations in gene expression underlying O₃-induced cachexia and downregulation of xenobiotic metabolism were examined. A number of the downregulated genes are known to be interferon (IFN) dependent, suggesting a role for circulating IFN. A more recent study by Aibo et al. (2010) demonstrates exacerbation of acetaminophen-induced liver injury in mice after a single 6-hour exposure to 0.25 or 0.5 ppm O₃. Data indicate that O₃ may worsen drug-induced liver injury by inhibiting hepatic repair. The O₃-associated effects shown in the liver are thought to be mediated by inflammatory cytokines or other cytotoxic mediators released by activated macrophages or other cells in the lungs (Laskin and Laskin, 2001; Laskin et al., 1998; Vincent et al., 1996a). Recently, increased peroxidated lipids were detected in the plasma of O₃ exposed animals (Santiago-López et al., 2010).

In summary, mediators generated by O₃ exposure may cause effects on the liver in laboratory rodents. Ozone exposures as low as 0.1 ppm have been shown to affect drug-induced sleeping time, and exposure to 0.25 ppm can exacerbate liver injury induced by a common analgesic. However, very few studies at relevant concentrations have been conducted, and no data from controlled human exposure or epidemiologic studies are currently available. Therefore the collective evidence is inadequate to determine if a causal relationship exists between short-term O₃ exposure and effects on the liver and metabolism.
6.5.2 Effects on Cutaneous and Ocular Tissues

In addition to the lungs, the skin is highly exposed to O₃ and contains O₃ reactive targets (polyunsaturated fatty acids) that can produce lipid peroxides. The 2006 O₃ AQCD (U.S. EPA, 2006b) reported that although there is evidence of oxidative stress at near ambient O₃ concentrations, skin and eyes are only affected at high concentrations (greater than 1-5 ppm). Ozone exposure (0.8 ppm for 7 days) induces oxidative stress in the skin of hairless mice, along with proinflammatory cytokines (Valacchi et al., 2009). A recent study demonstrated that 0.25 ppm O₃ differentially alters expression of metalloproteinases in the skin of young and aged mice, indicating that age may potentially increase risk of oxidative stress (Fortino et al., 2007). In young mice, healing of skin wounds is not significantly affected by O₃ exposure (Lim et al., 2006). However, exposure to 0.5 ppm O₃ for 6 hours/day significantly delays wound closure in aged mice. As with effects on the liver described above, the effects of O₃ on the skin and eyes have not been widely studied, and information from controlled human exposure or epidemiologic studies is not currently available. Therefore the collective evidence is inadequate to determine if a causal relationship exists between short-term O₃ exposure and effects on cutaneous and ocular tissues.

6.6 Mortality

6.6.1 Summary of Findings from 2006 O₃ AQCD

The 2006 O₃ AQCD (U.S. EPA, 2006b) reviewed a large number of time-series studies consisting of single- and multicity studies, and meta-analyses. In the large U.S. multicity studies that examined all-year data, summary effect estimates corresponding to single-day lags ranged from a 0.5-1% increase in all-cause (nonaccidental) mortality per a standardized unit increase in O₃ of 20 ppb for 24-h avg, 30 ppb for 8-h max, and 40 ppb for 1-h max as discussed in Section 2.5. The association between short-term O₃ exposure and mortality was substantiated by a collection of meta-analyses and international multicity studies. The studies evaluated found some evidence for heterogeneity in O₃ mortality risk estimates across cities and studies. Studies that conducted seasonal analyses, although more limited in number, reported larger O₃ mortality risk estimates during the warm or summer season. Overall, the 2006 O₃ AQCD identified robust associations between various measures of daily ambient O₃ concentrations and all-cause mortality, with additional evidence for associations with cardiovascular mortality, which could not be readily explained by confounding due to time, weather, or copollutants. However, it was noted that multiple uncertainties remain regarding the O₃-mortality relationship including: the extent of residual confounding by copollutants; factors that modify the O₃-mortality association; the appropriate lag structure for identifying O₃-mortality effects (e.g., single-day lags versus distributed lag model); the shape of
the O₃-mortality C-R function and whether a threshold exists; and the identification of populations at-risk to O₃-related health effects. Collectively, the 2006 O₃ AQCD concluded that “the overall body of evidence is highly suggestive that O₃ directly or indirectly contributes to non-accidental and cardiopulmonary-related mortality.”

6.6.2 Associations of Mortality and Short-Term O₃ Exposure

Recent studies that examined the association between short-term O₃ exposure and mortality further confirmed the associations reported in the 2006 O₃ AQCD. New multicontinent and multicity studies reported consistent positive associations between short-term O₃ exposure and all-cause mortality in all-year analyses, with additional evidence for larger mortality risk estimates during the warm or summer months (Figure 6-27 [and Table 6-42]). These associations were reported across a range of ambient O₃ concentrations that were in some cases quite low (Table 6-43).

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Lag</th>
<th>All-Year</th>
<th>Summer</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gryparis et al. (2004)</td>
<td>APHEA2 (23 cities)</td>
<td>0-1</td>
<td>0-1</td>
<td>0-1</td>
</tr>
<tr>
<td>Bell et al. (2007)</td>
<td>98 U.S. communities</td>
<td>0-6</td>
<td>0-6</td>
<td>0-6</td>
</tr>
<tr>
<td>Schwartz (2005)</td>
<td>98 U.S. communities</td>
<td>0-6</td>
<td>0-6</td>
<td>0-6</td>
</tr>
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<td>Bell et al. (2004a)</td>
<td>APHEA-Canada</td>
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<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
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<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Katsoyanni et al. (2009)</td>
<td>APHENA-Europe</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Bell et al. (2005a)</td>
<td>U.S. and Non-U.S.</td>
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<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Ito et al. (2005)</td>
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<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
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<td>Wong et al. (2010)</td>
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<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009b)</td>
<td>APHA (4 cities)</td>
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</tr>
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<td>Samoli et al. (2009)</td>
<td>APHEA-Canada</td>
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</tr>
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<td>Bell et al. (2004a)</td>
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<td>0-6</td>
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</tr>
<tr>
<td>Schwartz (2005)</td>
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<td>Zanobetti and Schwartz (2008)</td>
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<td>95 U.S. communities</td>
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<td>0-1</td>
</tr>
<tr>
<td>Gryparis et al. (2004)</td>
<td>APHEA2 (21 cities)</td>
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</tr>
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<td>Levy et al. (2005a)</td>
<td>APHENA-Canada</td>
<td>DL(0-2)</td>
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<td>DL(0-2)</td>
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<td>DL(0-2)</td>
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<td>APHENA-Canada</td>
<td>DL(0-2)</td>
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</tr>
<tr>
<td>Stafoggia et al. (2010)</td>
<td>U.S. and Non-U.S.</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
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<td>APHENA-Canada</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Zanobetti and Schwartz (2008)</td>
<td>95 U.S. communities</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Franklin and Schwartz (2008)</td>
<td>48 U.S. cities</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
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</tr>
<tr>
<td>Gryparis et al. (2004)</td>
<td>APHEA2 (23 cities)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Malamitsi-Ramon and Schwartz (2008)</td>
<td>APHENA-Europe</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
</tr>
<tr>
<td>Katsoyanni et al. (2009)</td>
<td>U.S. and Non-U.S.</td>
<td>DL(0-2)</td>
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</tr>
<tr>
<td>Bell et al. (2005a)</td>
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<td>DL(0-2)</td>
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<td>Ito et al. (2005)</td>
<td>APHENA-U.S.</td>
<td>DL(0-2)</td>
<td>DL(0-2)</td>
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<td>Katsouyanni et al. (2009)</td>
<td>U.S. and Non-U.S.</td>
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<td>Stafoggia et al. (2010)</td>
<td>10 Italian cities</td>
<td>DL(0-2)</td>
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</tr>
</tbody>
</table>

Note: Effect estimates are for a 40 ppb increase in 1-h max, 30 ppb increase in 8-h max, and 20 ppb increase in 24-h avg O₃ concentrations. An “a” represent multicity studies and meta-analyses from the 2006 O₃ AQCD. Bell et al. (2005), Ito et al. (2005), and Levy et al. (2005) used a range of lag days in the meta-analysis: Lag 0, 1, 2, or average 0-1 or 1-2; single-day lags from 0 to 3; and lag 0 and 1-2; respectively. A “b” represents risk estimates from APHENA-Canada standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O₃ concentrations (see explanation in Section 6.2.7.2).

Figure 6-27 Summary of mortality risk estimates for short-term O₃ exposure and all-cause (nonaccidental) mortality from all-year and summer season analyses.
Table 6-42  Corresponding effect estimates for Figure 6-27.

<table>
<thead>
<tr>
<th>Study*</th>
<th>Location</th>
<th>Lag</th>
<th>Avg Time</th>
<th>% Increase (95% CI)</th>
</tr>
</thead>
<tbody>
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<td><strong>All-year</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Gryparis et al. (2004)</td>
<td>APHEA2 (23 cities)</td>
<td>0-1</td>
<td>1-h max</td>
<td>0.24 (-0.86, 1.98)</td>
</tr>
<tr>
<td>Bell et al. (2007)</td>
<td>98 U.S. communities</td>
<td>0-1</td>
<td>24-h avg</td>
<td>0.64 (0.34, 0.92)</td>
</tr>
<tr>
<td>Schwartz (2005a)</td>
<td>14 U.S. cities</td>
<td>0</td>
<td>1-h max</td>
<td>0.76 (0.13, 1.40)</td>
</tr>
<tr>
<td>Bell and Dominici (2008)</td>
<td>98 U.S. communities</td>
<td>0-6</td>
<td>24-h avg</td>
<td>1.04 (0.56, 1.55)</td>
</tr>
<tr>
<td>Bell et al. (2004)*</td>
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<td>0-6</td>
<td>24-h avg</td>
<td>1.04 (0.54, 1.55)</td>
</tr>
<tr>
<td>Levy et al. (2005)*a</td>
<td>U.S. and Non-U.S.</td>
<td>---</td>
<td>24-h avg</td>
<td>1.64 (1.25, 2.03)</td>
</tr>
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<td>Katsouyanni et al. (2009)</td>
<td>APHENA-europe</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>1.66 (0.47, 2.94)</td>
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<td>Bell et al. (2005)*</td>
<td>U.S. and Non-U.S.</td>
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<td>24-h avg</td>
<td>1.75 (1.10, 2.37)</td>
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<td>Ito et al. (2008)*</td>
<td>U.S. and Non-U.S.</td>
<td>---</td>
<td>24-h avg</td>
<td>2.20 (0.80, 3.60)</td>
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<tr>
<td>Wong et al. (2010)</td>
<td>PAPA (4 cities)</td>
<td>0-1</td>
<td>8-h avg</td>
<td>2.26 (1.36, 3.16)</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-U.S.</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>3.02 (1.10, 4.89)</td>
</tr>
<tr>
<td>Cakmak et al. (2011)</td>
<td>7 Chilean cities</td>
<td>DL(0-6)</td>
<td>8-h max</td>
<td>3.35 (1.07, 5.75)</td>
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<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-Canada</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>5.87 (1.82, 9.81)</td>
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<tr>
<td>Katsouyanni et al. (2009)b</td>
<td>APHENA-Canada</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>0.73 (0.23, 1.20)</td>
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<tr>
<td><strong>Summer</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Samoli et al. (2009)</td>
<td>21 European cities</td>
<td>0-1</td>
<td>8-h max</td>
<td>0.66 (0.24, 1.05)</td>
</tr>
<tr>
<td>Bell et al. (2004)*</td>
<td>95 U.S. communities</td>
<td>0-6</td>
<td>24-h avg</td>
<td>0.78 (0.26, 1.30)</td>
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<td>8-h max</td>
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<td>Zanobetti and Schwartz (2008b)</td>
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<td>8-h max</td>
<td>1.60 (0.84, 2.33)</td>
</tr>
<tr>
<td>Franklin and Schwartz (2008)</td>
<td>18 U.S. communities</td>
<td>0</td>
<td>24-h avg</td>
<td>1.79 (0.90, 2.68)</td>
</tr>
<tr>
<td>Gryparis et al. (2004)</td>
<td>APHEA2 (21 cities)</td>
<td>0-1</td>
<td>8-h max</td>
<td>1.80 (0.99, 3.06)</td>
</tr>
<tr>
<td>Medina-Ramón and Schwartz (2008)</td>
<td>48 U.S. cities</td>
<td>0-2</td>
<td>8-h max</td>
<td>1.96 (1.14, 2.82)</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-europe</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>2.38 (0.87, 3.91)</td>
</tr>
<tr>
<td>Bell et al. (2005)*</td>
<td>U.S. and Non-U.S.</td>
<td>---</td>
<td>24-h avg</td>
<td>3.02 (1.45, 4.63)</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-Canada</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>3.34 (1.26, 5.38)</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-Canada</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>0.42 (0.16, 0.67)</td>
</tr>
<tr>
<td>Levy et al. (2005)*</td>
<td>U.S. and Non-U.S.</td>
<td>---</td>
<td>24-h avg</td>
<td>3.38 (2.27, 4.42)</td>
</tr>
<tr>
<td>Ito et al. (2005)*</td>
<td>U.S. and Non-U.S.</td>
<td>---</td>
<td>24-h avg</td>
<td>3.50 (2.10, 4.90)</td>
</tr>
<tr>
<td>Katsouyanni et al. (2009)</td>
<td>APHENA-U.S.</td>
<td>DL(0-2)</td>
<td>1-h max</td>
<td>3.83 (1.90, 5.79)</td>
</tr>
<tr>
<td>Stafoggia et al. (2010)</td>
<td>10 Italian cities</td>
<td>DL(0-5)</td>
<td>8-h max</td>
<td>9.15 (5.41, 13.0)</td>
</tr>
</tbody>
</table>

*Studies included from Figure 6-27.

*aMulticity studies and meta-analyses from the 2006 O3 AQCD. Bell et al. (2005)*b, Ito et al. (2005)ab, and Levy et al. (2005)ab used a range of lag days in the meta-analysis: Lag 0, 1, 2, or average 0-1 or 1-2; Single-day lags from 0-3; and Lag 0 and 1-2, respectively.

bRisk estimates from APHENA-Canada standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O3 concentrations (see explanation in Section 6.2.7.2).
### Table 6-43  Range of mean and upper percentile O₃ concentrations in previous and recent multicity studies.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Years</th>
<th>Averaging Time</th>
<th>Mean Concentration (ppb)a</th>
<th>Upper Percentile Concentrations (ppb)a</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bell and Dominici (2008)</td>
<td>98 U.S. communities</td>
<td>1987-2000</td>
<td>24-h avg</td>
<td>All year: 26.8 May-September: 30.0 Maximum: All year: 37.3 May-September: 47.2</td>
<td></td>
</tr>
<tr>
<td>and Schwartz (2008)</td>
<td>cities (APHEA2)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(2008)</td>
<td>(May-Sept)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Samoli et al. (2009)</td>
<td>21 European cities</td>
<td>1990-1997</td>
<td>8-h max</td>
<td>20.0-62.8</td>
<td>75th: 27.2-74.8</td>
</tr>
<tr>
<td>and Schwartz (2010)</td>
<td>(APHEA2)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Stafoggia et al. (2010)</td>
<td>10 Italian cities</td>
<td>2001-2005</td>
<td>8-h max</td>
<td>41.2-58.9</td>
<td>75th: 47.0-71.6</td>
</tr>
<tr>
<td>and Schwartz (2011)</td>
<td>7 Chilean cities</td>
<td>1997-2007</td>
<td>8-h max</td>
<td>59.0-87.6</td>
<td>NR</td>
</tr>
<tr>
<td>Wong et al. (2010)</td>
<td>PAPA (4 cities)</td>
<td>1999-2003</td>
<td>8-h avg</td>
<td>18.7-43.7</td>
<td>75th: 38.4 - 60.4 Max: 92.1 - 131.8</td>
</tr>
<tr>
<td>and Schwartz (2008b)</td>
<td>48 U.S. cities</td>
<td>1989-2000</td>
<td>8-h max</td>
<td>15.1-62.8</td>
<td>Max: 34.3-146.2 75th: 19.8-75.9</td>
</tr>
</tbody>
</table>

*aPPb, parts per billion; **PPb, parts per billion.
In addition to examining the relationship between short-term O₃ exposure and all-cause mortality, recent studies attempted to address the uncertainties that remained upon the completion of the 2006 O₃ AQCD. As a result, given the robust associations between short-term O₃ exposure and mortality presented across studies in the 2006 O₃ AQCD and supported in the new multicity studies (Figure 6-27), the following sections primarily focus on the examination of previously identified uncertainties in the O₃-mortality relationship, specifically: confounding, effect modification (i.e., sources of heterogeneity in risk estimates across cities), the O₃-mortality C-R relationship including lag structure (e.g., multiday effects and mortality displacement), and O₃ associations with cause-specific mortality. Focusing specifically on these uncertainties allows for a more detailed characterization of the relationship between short-term O₃ exposure and mortality.

### 6.6.2.1 Confounding

Recent epidemiologic studies examined potential confounders of the O₃-mortality relationship. These studies specifically focused on whether PM and its constituents or seasonal trends confounded the association between short-term O₃ exposure and mortality.

#### Confounding by PM and PM Constituents

An important question in the evaluation of the association between short-term O₃ exposure and mortality is whether the relationship is confounded by particulate matter, particularly the PM chemical components that are found in the “summer haze” mixture which also contains O₃. However, because of the temporal correlation...
among these PM components and O₃, and their possible interactions, the interpretation of results from copollutant models that attempt to disentangle the health effects associated with each pollutant is challenging. Further complicating the interpretation of copollutant results, at times, is the every-3rd or -6th day PM sampling schedule employed in most locations, which limits the number of days where both PM and O₃ data is available.

The potential confounding effects of PM₁₀ and PM₂.₅ on the O₃-mortality relationship were examined by Bell et al. (2007) using data on 98 U.S. urban communities for the years 1987-2000 from the National Morbidity, Mortality, and Air Pollution Study (NMMAPS). In this analysis the authors included PM as a covariate in time-series models, and also examined O₃-mortality associations on days when O₃ concentrations were below a specified value. This analysis was limited by the small fraction of days when both PM and O₃ data were available, due to the every-3rd - or 6th -day sampling schedule for the PM indices, and the limited amount of city-specific data for PM₂.₅ because it was only collected in most cities since 1999. As a result, of the 91 communities with PM₂.₅ data, only 9.2% of days in the study period had data for both O₃ and PM₂.₅, resulting in the use of only 62 communities in the PM₂.₅ analysis. An examination of the correlation between PM (PM₁₀ and PM₂.₅) and O₃ across various strata of daily PM₁₀ and PM₂.₅ concentrations found that neither PM size fraction was highly correlated with daily O₃ concentrations across any of the strata examined. These results were also observed when using 8-h max and 1-h max O₃ exposure metrics. National and community-specific effect estimates of the association between short-term O₃ exposure and mortality were robust to inclusion of PM₁₀ or PM₂.₅ in time-series models through the range of O₃ concentrations (i.e., <10 ppb, 10-20, 20-40, 40-60, 60-80, and >80 ppb). Even with the small number of days in which both PM₂.₅ and O₃ data was available, the percent increases in nonaccidental deaths per 10 ppb increase 24-h avg O₃ concentrations at lag 0-1 day were 0.22% (95% CI: -0.22, 0.65) without PM₂.₅ and 0.21% (95% CI: -0.22, 0.64) with PM₂.₅ in 62 communities.

Although strong correlations between PM and O₃ were not reported by Bell et al. (2007) the patterns observed suggest regional differences in their correlation (Table 6-44). Both PM₁₀ and PM₂.₅ show positive correlations with O₃ in the Industrial Midwest, Northeast, Urban Midwest, and Southeast, especially in the summer months, presumably, because of the summer peaking sulfate. However, the mostly negative or weak correlations between PM and O₃ in the summer in the Southwest, Northwest, and southern California could be due to winter-peaking nitrate. Thus, the potential confounding effect of PM on the O₃-mortality relationship could be influenced by the relative contribution of sulfate and nitrate, which varies regionally and seasonally.
Table 6-44 Correlations between PM and O₃ by season and region.

<table>
<thead>
<tr>
<th>Region</th>
<th>No. of Communities</th>
<th>Winter</th>
<th>Spring</th>
<th>Summer</th>
<th>Fall</th>
<th>Yearly</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>PM₁₀</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Industrial Midwest</td>
<td>19</td>
<td>0.37</td>
<td>0.44</td>
<td>0.44</td>
<td>0.39</td>
<td>0.41</td>
</tr>
<tr>
<td>Northeast</td>
<td>15</td>
<td>0.34</td>
<td>0.44</td>
<td>0.36</td>
<td>0.44</td>
<td>0.40</td>
</tr>
<tr>
<td>Urban Midwest</td>
<td>6</td>
<td>0.24</td>
<td>0.25</td>
<td>0.22</td>
<td>0.26</td>
<td>0.24</td>
</tr>
<tr>
<td>Southwest</td>
<td>9</td>
<td>0.00</td>
<td>0.02</td>
<td>-0.02</td>
<td>0.10</td>
<td>0.03</td>
</tr>
<tr>
<td>Northwest</td>
<td>11</td>
<td>-0.17</td>
<td>-0.20</td>
<td>-0.13</td>
<td>-0.11</td>
<td>-0.16</td>
</tr>
<tr>
<td>Southern California</td>
<td>7</td>
<td>0.19</td>
<td>0.08</td>
<td>0.12</td>
<td>0.19</td>
<td>0.14</td>
</tr>
<tr>
<td>Southeast</td>
<td>25</td>
<td>0.33</td>
<td>0.35</td>
<td>0.31</td>
<td>0.31</td>
<td>0.32</td>
</tr>
<tr>
<td>U.S.</td>
<td>93</td>
<td>0.23</td>
<td>0.26</td>
<td>0.24</td>
<td>0.26</td>
<td>0.25</td>
</tr>
<tr>
<td><strong>PM₂.₅</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Industrial Midwest</td>
<td>19</td>
<td>0.18</td>
<td>0.39</td>
<td>0.43</td>
<td>0.44</td>
<td>0.36</td>
</tr>
<tr>
<td>Northeast</td>
<td>13</td>
<td>0.05</td>
<td>0.26</td>
<td>0.16</td>
<td>0.43</td>
<td>0.25</td>
</tr>
<tr>
<td>Urban Midwest</td>
<td>4</td>
<td>0.22</td>
<td>0.31</td>
<td>0.15</td>
<td>0.32</td>
<td>0.20</td>
</tr>
<tr>
<td>Southwest</td>
<td>9</td>
<td>-0.15</td>
<td>-0.08</td>
<td>-0.17</td>
<td>-0.15</td>
<td>-0.14</td>
</tr>
<tr>
<td>Northwest</td>
<td>11</td>
<td>-0.32</td>
<td>-0.34</td>
<td>-0.39</td>
<td>-0.24</td>
<td>-0.31</td>
</tr>
<tr>
<td>Southern California</td>
<td>7</td>
<td>-0.25</td>
<td>-0.22</td>
<td>-0.25</td>
<td>-0.15</td>
<td>-0.23</td>
</tr>
<tr>
<td>Southeast</td>
<td>26</td>
<td>0.38</td>
<td>0.47</td>
<td>0.30</td>
<td>0.37</td>
<td>0.39</td>
</tr>
<tr>
<td>U.S.</td>
<td>90</td>
<td>0.09</td>
<td>0.21</td>
<td>0.12</td>
<td>0.22</td>
<td>0.16</td>
</tr>
</tbody>
</table>

Source: Bell et al. (2007).
Figure 6-28 Scatter plots of O$_3$ mortality risk estimates with versus without adjustment for PM$_{10}$ in NMMAPS cities.

In an attempt to reassess a number of issues associated with the O$_3$-mortality relationship, including confounding, Smith et al. (2009b) re-analyzed the publicly available NMMAPS database for the years 1987-2000. Similar to Bell et al. (2007), the PM$_{10}$ data used in the Smith et al. (2009b) analysis consisted primarily of every-6th day data. In analyses conducted to examine the potential confounding effects of PM$_{10}$, the authors reported that, in most cases, O$_3$ mortality risk estimates were reduced by between 22% and 33% in copollutant models. This is further highlighted in Figure 6-28, which shows scatter plots of O$_3$-mortality risk estimates with adjustment for PM$_{10}$ versus without adjustment for PM$_{10}$. Smith et al. (2009b) point out that a larger fraction (89 out of 93) of the posterior estimates lie below the diagonal line (i.e., estimates are smaller with PM$_{10}$ adjustment) compared to the raw estimates (56 out of 93). This observation could be attributed to both sets of posterior estimates being calculated by “shrinking towards the mean” along with the small number of days where both PM$_{10}$ and O$_3$ data was available. However, the most prominent feature of these plots is that the variation of O$_3$-mortality risk estimates across cities is much larger than the impact of PM$_{10}$ adjustment on the O$_3$-mortality relationship.
Franklin and Schwartz (2008) examined the sensitivity of O3 mortality risk estimates to the inclusion of PM2.5 or PM chemical components associated with secondary aerosols (e.g., sulfate [SO4\(^{2-}\)], organic carbon [OC], and nitrate [NO3\(^{-}\)]) in copollutant models. This analysis consisted of between 3 and 6 years of data from May through September 2000-2005 from 18 U.S. communities. The association between O3 and non-accidental mortality was examined in single-pollutant models and after adjustment for PM2.5, sulfate, organic carbon, or nitrate concentrations. The single-city effect estimates were combined into an overall estimate using a random-effects model. In the single-pollutant model, the authors found a 0.89% (95% CI: 0.45, 1.33%) increase in nonaccidental mortality with a 10 ppb increase in same-day 24-hour summertime O3 concentrations across the 18 U.S. communities. Adjustment for PM2.5 mass, which was available for 84% of the days, decreased the O3-mortality risk estimate only slightly (from 0.88% to 0.79%), but the inclusion of sulfate in the model reduced the risk estimate by 31% (from 0.85% to 0.58%). However, sulfate data were only available for 18% of the days. Therefore, a limitation of this study is the limited amount of data for PM2.5 chemical components due to the every-3rd-day or every-6th-day sampling schedule. For example, when using a subset of days when organic carbon measurements were available (i.e., 17% of the available days), O3 mortality risk estimates were reduced to 0.51% (95% CI: -0.36 to 1.36) in a single-pollutant model.

Consistent with the studies previously discussed, the results from Franklin and Schwartz (2008) also demonstrate that the interpretation of the potential confounding effects of copollutants on O3 mortality risk estimates is not straightforward as a result of the PM sampling schedule employed in most cities. However, Franklin and Schwartz (2008) find that O3-mortality risk estimates, although attenuated in some cases (i.e., sulfate), remain positive. As presented in Figure 6-29, the regional and city-to-city variations in O3 mortality risk estimates appear greater than the impact of adjusting for copollutants. In addition, in some cases, a negative O3 mortality risk estimate becomes even more negative with the inclusion of sulfate (e.g., Seattle) in a copollutant model, or a null O3 mortality risk estimate becomes negative when sulfate is included (e.g., Dallas and Detroit). Thus, the reduction in the overall O3 mortality risk estimate (i.e., across cities) needs to be assessed in the context of the heterogeneity in the single-city estimates.
Figure 6-29  Community-specific $O_3$-mortality risk estimates for nonaccidental mortality per 10 ppb increase in same-day 24-h average summertime $O_3$ concentrations in single-pollutant models and copollutant models with sulfate.

In the APHENA study, the investigators from the U.S. (NMMAPS), Canadian, and European (APHEA2) multicity studies collaborated and conducted a joint analysis of PM$_{10}$ and $O_3$ using each of these datasets (Katsouyanni et al., 2009). For mortality, each dataset consisted of a different number of cities and years of air quality data: U.S. encompassed 90 cities with daily $O_3$ data from 1987-1996 of which 36 cities had summer only $O_3$ measurements; Europe included 23 cities with 3-7 years of daily $O_3$ data during 1990-1997; and Canada consisted of 12 cities with daily $O_3$ data from 1987 to 1996. As discussed in Section 6.2.7.2, the APHENA study conducted extensive sensitivity analyses, of which the 8 df/year results for both the penalized spline (PS) and natural spline (NS) models are presented in the text for comparison purposes, but only the NS results are presented in figures because alternative spline models have previously been shown to result in similar effect estimates (HEI, 2003).
Additionally, for the Canadian results, figures contain risk estimates standardized to a 40 ppb increment for 1-h max O₃ concentrations, consistent with the rest of the ISA, but also standardized to the approximate IQR across the Canadian cities as discussed previously (Section 6.2.7.2).

In the three datasets, the authors found generally positive associations between short-term O₃ exposure and all-cause, cardiovascular, and respiratory mortality. The estimated excess risks for O₃ were larger for the Canadian cities than for the U.S. and European cities. When examining the potential confounding effects of PM₁₀ on O₃ mortality risk estimates, the sensitivity of the estimates varied across the data sets and age groups. In the Canadian dataset, O₃ risk estimates were modestly reduced, but remained positive, when adjusting for PM₁₀ for all-cause mortality for all ages in the PS (4.5% [95% CI: 2.2, 6.7%]) and NS (4.2% [95% CI: 1.9, 6.5%]) models to 3.8% (95% CI: -1.4, 9.8%) and 3.2% (95% CI: -2.2, 9.0%), respectively, at lag 1 for a 40 ppb increase in 1-h max O₃ concentrations (Figure 6-30 [and Table 6-45]). However, adjusting for PM₁₀ reduced O₃ mortality risk estimates in the ≥75-year age group, but increased the risk estimates in the <75-year age group. For cardiovascular and respiratory mortality more variable results were observed with O₃ risk estimates being reduced and increased, respectively, in copollutant models with PM₁₀ (Figure 6-30 [and Table 6-45]). Unlike the European and U.S. datasets, the Canadian dataset only conducted copollutant analyses at lag 1; as a result, to provide a comparison across study locations only the lag 1 results are presented for the European and U.S. datasets in this section.

In the European data, O₃ risk estimates were robust when adjusting for PM₁₀ in the year-round data for all-cause, cardiovascular and respiratory mortality. When restricting the analysis to the summer months moderate reductions were observed in O₃ risk estimates for all-cause mortality with more pronounced reductions in respiratory mortality. In the U.S. data, adjusting for PM₁₀ moderately reduced O₃ risk estimates for all-cause mortality in a year-round analysis at lag 1 (e.g., both the PS and NS models were reduced from 0.18% to 0.13%) (Figure 6-30 [and Table 6-45]). Similar to the European data, when restricting the analysis to the summer months, in the United States. Ozone mortality risk estimates were moderately reduced, but remained positive, when adjusting for PM₁₀ for all-cause mortality. However, when examining cause-specific mortality risk estimates, consistent with the results from the Canadian dataset, which employed a similar PM sampling strategy (i.e., every-6th-day sampling), O₃ risk estimates for cardiovascular and respiratory mortality were more variable (i.e., reduced or increased in all-year and summer analyses). Overall, the estimated O₃ risks appeared to be moderately or substantially sensitive to inclusion of PM₁₀ in copollutant models. Despite the multicity approach, the mostly every-6th-day sampling schedule for PM₁₀ in the Canadian and U.S. datasets greatly reduced the sample size and limits the interpretation of these results.
Note: Effect estimates are for a 40 ppb increase in 1-h max $O_3$ concentrations at lag 1. All estimates are for the 8 df/year model with natural splines. Circles represent all-year analysis results while diamonds represent summer season analysis results. Open circles and diamonds represent copollutant models with $PM_{10}$. Black = all-cause mortality; red = cardiovascular mortality; and blue = respiratory mortality.

Risk estimates from APHENA-Canada standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in $O_3$ concentrations (see explanation in Section 6.2.7.2).

Figure 6-30  Percent increase in all-cause (nonaccidental) and cause-specific mortality from natural spline models with 8 df/yr from the APHENA study for single- and copollutant models.
Table 6-45  Corresponding effect estimates for Figure 6-30.

<table>
<thead>
<tr>
<th>Location*</th>
<th>Mortality</th>
<th>Ages</th>
<th>Season</th>
<th>Copollutant</th>
<th>% Increase (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>APHENA-U.S.</strong></td>
<td>All-Cause</td>
<td>All-year</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>1.42 (0.08, 2.78)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>1.02 (-1.40, 3.50)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Summer</td>
<td></td>
<td>PM$_{10}$</td>
<td>4.31 (2.22, 6.45)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>1.90 (-0.78, 4.64)</td>
</tr>
<tr>
<td></td>
<td>Cardiovascular</td>
<td>All-year</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>1.10 (-1.33, 3.67)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>0.47 (-4.61, 5.79)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>≥ 75</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>-0.16 (-3.02, 2.86)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>&lt;75</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>1.34 (3.63, 6.61)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>≥ 75</td>
<td>Summer</td>
<td>PM$_{10}$</td>
<td>3.58 (0.87, 6.37)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>&lt;75</td>
<td>Summer</td>
<td>PM$_{10}$</td>
<td>-1.17 (-6.18, 4.07)</td>
</tr>
<tr>
<td></td>
<td>Respiratory</td>
<td>All-year</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>2.46 (-1.87, 6.68)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>6.04 (1.18, 11.1)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>7.03 (-3.48, 18.5)</td>
</tr>
<tr>
<td><strong>APHENA-Canada</strong></td>
<td>All-Cause</td>
<td>All-year</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>0.52 (0.24, 0.80)*</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>3.18 (-2.18, 8.96)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>0.40 (-0.28, 1.10)*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>≥ 75</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>5.62 (0.95, 10.7)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>0.70 (0.12, 1.30)*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>≥ 75</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>1.90 (-9.03, 14.1)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>0.24 (-1.20, 1.70)*</td>
</tr>
<tr>
<td></td>
<td>Cardiovascular</td>
<td>&lt;75</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>1.10 (-4.08, 6.61)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>0.14 (-0.53, 0.82)*</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>-2.64 (-14.7, 11.5)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>-0.34 (-2.00, 1.40)*</td>
</tr>
<tr>
<td></td>
<td>Respiratory</td>
<td>All-year</td>
<td>All-year</td>
<td>PM$_{10}$</td>
<td>0.87 (-6.40, 8.96)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>22.3 (-12.6, 71.3)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM$_{10}$</td>
<td>2.60 (-1.70, 7.10)*</td>
</tr>
<tr>
<td>Location*</td>
<td>Mortality</td>
<td>Ages</td>
<td>Season</td>
<td>Copollutant</td>
<td>% Increase (95% CI)</td>
</tr>
<tr>
<td>-----------</td>
<td>-------------------</td>
<td>-------</td>
<td>--------</td>
<td>-------------</td>
<td>---------------------</td>
</tr>
<tr>
<td>APHENA-Europe</td>
<td>All-Cause</td>
<td>All</td>
<td>All-year</td>
<td>PM₁₀</td>
<td>1.02 (0.39, 1.66)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>1.26 (0.47, 1.98)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>2.06 (1.10, 2.94)</td>
</tr>
<tr>
<td></td>
<td>Cardiovascular</td>
<td>≥ 75</td>
<td>All-year</td>
<td>PM₁₀</td>
<td>1.10 (-0.47, 2.70)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>&lt;75</td>
<td>All-year</td>
<td>PM₁₀</td>
<td>1.34 (-0.24, 2.94)</td>
</tr>
<tr>
<td></td>
<td>Respiratory</td>
<td>All</td>
<td>All-year</td>
<td>PM₁₀</td>
<td>1.42 (-1.02, 3.83)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>PM₁₀</td>
<td>4.31 (1.66, 7.11)</td>
</tr>
</tbody>
</table>

*Effect estimates from Figure 6-30.

Risk estimates from APHENA-Canada standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O₃ concentrations (see explanation in Section 6.2.7.2).

Stafoggia et al. (2010) examined the potential confounding effects of PM₁₀ on the O₃-mortality relationship in individuals 35 years of age and older in 10 Italian cities from 2001 to 2005. In a time-stratified case-crossover analysis, using data for the summer months (i.e., April-September), the authors examined O₃-mortality associations across each city, and then obtained a pooled estimate through a random-effects meta-analysis. Stafoggia et al. (2010) found a strong association with nonaccidental mortality (9.2% [95% CI: 5.4, 13.0%] for a 30 ppb increase in 8-h max O₃ concentrations) in an unconstrained distributed lag model (lag 0-5) that persisted in copollutant models with PM₁₀ (9.2% [95% CI: 5.4, 13.7%]). Additionally, when examining cause-specific mortality, the authors found positive associations between short-term O₃ exposure and cardiovascular (14.3% [95% CI: 6.7, 22.4%]), cerebrovascular (8.5% [95% CI: 0.1, 16.3%]), and respiratory (17.6% [95% CI: 1.8, 35.6%]) mortality in single-pollutant models. In copollutant models, O₃-mortality effect estimates for cardiovascular and cerebrovascular mortality were robust to the inclusion of PM₁₀ (9.2% [95% CI: 5.4, 13.7%]) and 7.3% [95% CI: -1.2, 16.3%], respectively), and attenuated, but remained positive, for respiratory mortality (9.2% [95% CI: -6.9, 28.8%]). Of note, the correlations between O₃ and PM₁₀ across cities were found to be generally low, ranging from (-0.03 to 0.49). The authors do not specify the sampling strategy used for PM₁₀ in this analysis.
Confounding by Seasonal Trend

The APHENA study (Katsouyanni et al., 2009), mentioned above, also conducted extensive sensitivity analyses to identify the appropriate: (1) smoothing method and basis functions to estimate smooth functions of time in city-specific models; and (2) degrees of freedom to be used in the smooth functions of time, to adjust for seasonal trends. Because O3 peaks in the summer and mortality peaks in the winter, not adjusting or not sufficiently adjusting for the seasonal trend would result in an apparent negative association between the O3 and mortality time-series. Katsouyanni et al. (2009) examined the effect of the extent of smoothing for seasonal trends by using models with 3 df/year, 8 df/year (the choice for their main model), 12 df/year, and df/year selected using the sum of absolute values of partial autocorrelation function of the model residuals (PACF) (i.e., choosing the degrees of freedom that minimizes positive and negative autocorrelations in the residuals). Table 6-46 presents the results of the degrees of freedom analysis using alternative methods to calculate a combined estimate: the Berkey et al. (1998) meta-regression and the two-level normal independent sampling estimation (TLNISE) hierarchical method. The results show that the methods used to combine single-city estimates did not influence the overall results, and that neither 3 df/year nor choosing the df/year by minimizing the sum of absolute values of PACF of regression residuals was sufficient to adjust for the seasonal negative relationship between O3 and mortality. However, it should be noted, the majority of studies in the literature that examined the mortality effects of short-term O3 exposure, particularly the multicity studies, used 7 or 8 df/year to adjust for seasonal trends, and in both methods a positive association was observed between O3 exposure and mortality.

Table 6-46  Sensitivity of O3 risk estimates per 10 µg/m3 increase in 24-h average O3 concentrations at lag 0-1 to alternative methods for adjustment of seasonal trend, for all-cause mortality using Berkey MLE and TLNISE Hierarchical Models.

<table>
<thead>
<tr>
<th>Seasonality Control</th>
<th>Berkey</th>
<th>TLNISE</th>
</tr>
</thead>
<tbody>
<tr>
<td>3 df/year</td>
<td>-0.54 (-0.88, 0.20)</td>
<td>-0.55 (-0.88, -0.22)</td>
</tr>
<tr>
<td>8 df/year</td>
<td>0.30 (0.11, 0.50)</td>
<td>0.31 (0.09, 0.52)</td>
</tr>
<tr>
<td>12 df/year</td>
<td>0.34 (0.15, 0.53)</td>
<td>0.33 (0.12, 0.54)</td>
</tr>
<tr>
<td>PACF</td>
<td>-0.62 (-1.01, -0.22)</td>
<td>-0.62 (-0.98, -0.27)</td>
</tr>
</tbody>
</table>

Source: Reprinted with permission of Health Effects Institute (Katsouyanni et al., 2009).

6.6.2.2 Effect Modification

Epidemiologic studies have examined potential effect modifiers of the O3-mortality relationship through the use of either: (1) time-invariant factors or (2) time-variant factors. There have been several multicity studies that examined potential effect
modifiers, or time-invariant factors, which may modify O₃ mortality risk estimates. These effect modifiers can be categorized into either individual-level or community-level characteristics, which are traditionally examined in second stage regression models. The results from these analyses also inform upon whether certain populations are at greater risk of an O₃-related health effect (Chapter 8). In addition to potentially modifying the association between short-term O₃ exposure and mortality, both individual-level and community-level characteristics may contribute to the geographic pattern of spatial heterogeneity in O₃ mortality risk estimates. As a result, the geographic pattern of O₃ mortality risk estimates is also evaluated in this section. Although less common, this section also evaluates those studies that examine effect modification by using time-variant factors, such as temperature and copollutants that are included in first stage time-series regression models.

**Time-Invariant Factors**

**Individual-Level Characteristics**

Medina-Ramón and Schwartz (2008) conducted a case-only study in 48 U.S. cities to identify populations potentially at increased risk to O₃-related mortality for the period 1989-2000 (May through September of each year [i.e., warm season]). A case-only design predicts the occurrence of time-invariant characteristics among cases as a function of the exposure level (Armstrong, 2003). For each potential effect modifier (time-invariant individual-level characteristics), city-specific logistic regression models were fitted, and the estimates were pooled across all cities. Furthermore, the authors examined potential differences in individual effect modifiers according to several city characteristics (e.g., mean O₃ level, mean temperature, households with central air conditioning, and population density) in a meta-regression. Across cities, the authors found a 1.96% (95% CI: 1.14-2.82%) increase in mortality at lag 0-2 for a 30 ppb increase in 8-h max O₃ concentrations. Additionally, Medina-Ramón and Schwartz (2008) examined a number of individual-level characteristics (e.g., age, race) and chronic conditions (e.g., secondary causes of death) as effect modifiers of the association between short-term O₃ exposure and mortality. The authors found that older adults (i.e., ≥ 65), women >60 years of age, black race, and secondary atrial fibrillation showed the greatest additional percent change in O₃-related mortality (Table 6-47). When examining city-level characteristics, the authors found that older adults, black race, and secondary atrial fibrillation had a larger effect on O₃ mortality risk estimates in cities with lower mean O₃ concentrations. Of note, a similar case-only study (Schwartz, 2005b) examined potential effect modifiers of the association between temperature and mortality, which would be expected to find results consistent with the Medina-Ramón and Schwartz (2008) study due to the high correlation between temperature and O₃. However, when stratifying days by temperature Schwartz (2005b) found strong evidence that diabetes modified the temperature-mortality association on hot days, which was not as evident when examining the O₃-mortality association in Medina-Ramón and Schwartz (2008). This difference could be due to the study design and populations included in both studies, a multicity study including all ages (Medina-Ramón and Schwartz, 2008) compared
to a single-city study of individuals ≥ 65 years of age (Schwartz, 2005b). However, when examining results stratified by race, nonwhites were found to have higher mortality risks on both hot and cold days, which provide some support for the additional risk found for black race in Medina-Ramón and Schwartz (2008).

Individual-level factors that may result in increased risk of O$_3$-related mortality were also examined by Stafoggia et al. (2010). As discussed above, using a time-stratified case-crossover analysis, the authors found an association between short-term O$_3$ exposure and nonaccidental mortality in an unconstrained distributed lag model in 10 Italian cities (9.2% [95% CI: 5.4, 13.0%; lag 0-5 for a 30 ppb increase in 8-h max O$_3$ concentrations). Stafoggia et al. (2010) conducted additional analyses to examine whether age, sex, income level, location of death, and underlying chronic conditions increased the risk of O$_3$-related mortality, but data were only available for nine of the cities for these analyses. Of the individual-level factors examined, the authors found the strongest evidence for increased risk of O$_3$-related mortality in individuals ≥ 85 years of age (22.4% [95% CI: 15.0, 30.2%]), women (13.7% [95% CI: 8.5, 19.7%]), and out-of-hospital deaths (13.0% [95% CI: 6.0, 20.4%]). When focusing specifically on out-of hospital deaths and the subset of individuals with chronic conditions, Stafoggia et al. (2010) found the strongest association for individuals with diabetes, which is consistent with the potentially increased risk of diabetics on hot days observed in Schwartz (2005b).
### Table 6-47 Additional percent change in O₃-related mortality for individual-level characteristics.

<table>
<thead>
<tr>
<th>Socio-demographic characteristics</th>
<th>Percentage</th>
<th>(95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age 65 yr or older</td>
<td>1.10</td>
<td>0.44, 1.77</td>
</tr>
<tr>
<td>Women</td>
<td>0.58</td>
<td>0.18, 0.98</td>
</tr>
<tr>
<td>Women &lt;60 yr old</td>
<td>-0.09</td>
<td>-0.76, 0.58</td>
</tr>
<tr>
<td>Women ≥ 60 yr old</td>
<td>0.60</td>
<td>0.25, 0.96</td>
</tr>
<tr>
<td>Black race</td>
<td>0.53</td>
<td>0.19, 0.87</td>
</tr>
<tr>
<td>Low education</td>
<td>-0.29</td>
<td>-0.81, 0.23</td>
</tr>
</tbody>
</table>

**Chronic conditions (listed as secondary cause)**

<table>
<thead>
<tr>
<th>Respiratory system diseases</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Asthma</td>
<td>1.35</td>
<td>-0.31, 3.03</td>
</tr>
<tr>
<td>COPD</td>
<td>0.01</td>
<td>-0.49, 0.52</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Circulatory system diseases</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Atherosclerosis</td>
<td>-0.72</td>
<td>-1.89, 0.45</td>
</tr>
<tr>
<td>Atherosclerotic CVD</td>
<td>0.74</td>
<td>-0.86, 2.37</td>
</tr>
<tr>
<td>Atherosclerotic heart disease</td>
<td>-0.38</td>
<td>-1.70, 0.96</td>
</tr>
<tr>
<td>Congestive heart disease</td>
<td>-0.04</td>
<td>-0.39, 0.30</td>
</tr>
<tr>
<td>Atrial fibrillation</td>
<td>1.66</td>
<td>0.03, 3.32</td>
</tr>
<tr>
<td>Stroke</td>
<td>0.17</td>
<td>-0.28, 0.62</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Other diseases</th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Diabetes</td>
<td>0.19</td>
<td>-0.46, 0.84</td>
</tr>
<tr>
<td>Inflammatory diseases</td>
<td>0.18</td>
<td>-1.09, 1.46</td>
</tr>
</tbody>
</table>

*These estimates represent the additional percent change in mortality for persons who had the characteristic being examined compared to persons who did not have the characteristic, when the mean O₃ level of the previous 3 days increased 10 ppb. These values were not standardized because they do not represent the actual effect estimate for the characteristic being evaluated, but instead, the difference between effect estimates for persons with versus without the condition.

bCompared with males in the same age group.

Source: Reprinted with permission of Lippincott Williams & Wilkins (Medina-Ramón and Schwartz, 2008).

Additionally, Cakmak et al. (2011) examined the effect of individual-level characteristics that may modify the O₃-mortality relationship in 7 Chilean cities. In a time-series analysis using a constrained distributed lag of 0-6 days, Cakmak et al. (2011) found evidence for larger O₃ mortality effects in individuals ≥75 years of age compared to younger ages, which is similar to Medina-Ramón and Schwartz (2008) and Stafoggia et al. (2010). Unlike the studies discussed above O₃-mortality risk estimates were found to be slightly larger in males (3.71% [95% CI: 0.79, 6.66] for a 40 ppb increase in max 8-h avg O₃ concentrations), but were not significantly different than those observed for females (3.00% [95% CI: 0.43, 5.68]). The major focus of Cakmak et al. (2011) is the examination of the influence of SES indicators (i.e., educational attainment, income level, and employment status) on the O₃-
mortality relationship. The authors found the largest risk estimates in the lowest SES categories for each of the indicators examined this includes: primary school not completed when examining educational attainment; the lowest quartile of income level; and unemployed individuals when comparing employment status.

Overall, uncertainties exist in the interpretation of the potential effect modifiers identified in Medina-Ramón and Schwartz (2008), Stafoggia et al. (2010), and Cakmak et al. (2011) of the O₃-mortality relationship due to the heterogeneity in O₃-mortality risk estimates across cities as highlighted in Smith et al. (2009b) (Figure 6-28) and Franklin and Schwartz (2008) (Figure 6-29). In addition, it is likely that individual-level factors identified in Medina-Ramón and Schwartz (2008), Stafoggia et al. (2010), and Cakmak et al. (2011) only modify the O₃-mortality relationship and do not entirely explain the observed regional heterogeneity in O₃-mortality risk estimates.

Community-level Characteristics

Several studies also examined city-level (i.e., ecological) variables in an attempt to explain the observed city-to-city variation in estimated O₃-mortality risk estimates. Bell and Dominici (2008) investigated whether community-level characteristics, such as race, income, education, urbanization, transportation use, PM and O₃ concentrations, number of O₃ monitors, weather, and air conditioning use could explain the heterogeneity in O₃-mortality risk estimates across cities. The authors analyzed 98 U.S. urban communities from NMMAPS for the period 1987-2000. In the all-year regression model that included no community-level variables, a 20 ppb increase in 24-h avg O₃ concentrations during the previous week was associated with a 1.04% (95% CI: 0.56, 1.55) increase in mortality. Bell and Dominici (2008) found that higher O₃-mortality effect estimates were associated with an increase in: percent unemployment, fraction of the population Black/African-American, percent of the population that take public transportation to work; and with a reduction in: temperatures and percent of households with central air conditioning (Figure 6-31). The modification of O₃-mortality risk estimates reported for city-specific temperature and prevalence of central air conditioning in this analysis confirm the result from the meta-analyses reviewed in the 2006 O₃ AQCD.

The APHENA project (Katsouyanni et al., 2009) examined potential effect modification of O₃ risk estimates in the Canadian, European, and U.S. data sets using a consistent set of city-specific variables. Table 6-48 presents the results from all age analyses for all-cause mortality using all-year O₃ data for the average of lag 0-1 day. While there are several significant effect modifiers in the U.S. data, the results are mostly inconsistent with the results from the Canadian and European data sets. The positive effect modification by percentage unemployed and the negative effect modification by mean temperature (i.e., a surrogate for air conditioning rate) are consistent with the results reported by Bell and Dominici (2008) discussed above. However, the lack of consistency across the data sets, even between the Canadian and U.S. data, makes it difficult to interpret the results. Some of these associations
may be due to coincidental correlations with other unmeasured factors that vary regionally (e.g., mean SO₂ tend to be higher in the eastern U.S.).
Table 6-48  Percent change in all-cause mortality, for all ages, associated with a 40ppb increase in 1-h max O$_3$ concentrations at Lag 0–1 at the 25th and 75th percentile of the center-specific distribution of selected effect modifiers.

<table>
<thead>
<tr>
<th>Effect Modifier</th>
<th>25th Percentile Estimate (95% CI)</th>
<th>75th Percentile Estimate (95% CI)</th>
<th>t-value</th>
<th>25th Percentile Estimate (95% CI)</th>
<th>75th Percentile Estimate (95% CI)</th>
<th>t-value</th>
<th>25th Percentile Estimate (95% CI)</th>
<th>75th Percentile Estimate (95% CI)</th>
<th>t-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>NO$_2$ CV</td>
<td>3.10 (1.90, 4.40)</td>
<td>3.99 (2.38, 5.62)</td>
<td>1.33</td>
<td>1.66 (0.71, 2.62)</td>
<td>-0.49 (0.47, 1.98)</td>
<td>0.08</td>
<td>-2.87 (0.78, 0.95)</td>
<td>1.66 (0.39, 2.86)</td>
<td>0.16</td>
</tr>
<tr>
<td>Mean SO$_2$</td>
<td>2.22 (0.71, 3.83)</td>
<td>4.72 (2.94, 6.61)</td>
<td>2.16</td>
<td>1.58 (0.47, 2.62)</td>
<td>0.16 (0.47, 1.42)</td>
<td>1.98</td>
<td>2.79 (1.10, 2.94)</td>
<td>1.58 (0.39, 2.86)</td>
<td>0.16</td>
</tr>
<tr>
<td>O$_3$ CV</td>
<td>2.86 (0.79, 5.05)</td>
<td>3.50 (2.14, 4.89)</td>
<td>0.60</td>
<td>2.62 (1.50, 3.75)</td>
<td>-2.65 (0.70, 1.10)</td>
<td>1.50</td>
<td>2.68 (0.71, 2.22)</td>
<td>1.50 (0.71, 2.22)</td>
<td>2.64</td>
</tr>
<tr>
<td>Mean NO$<em>2$/PM$</em>{10}$</td>
<td>3.91 (2.54, 5.29)</td>
<td>2.54 (0.95, 4.15)</td>
<td>-1.58</td>
<td>1.74 (0.87, 2.70)</td>
<td>-0.43 (-1.02, 0.95)</td>
<td>1.26</td>
<td>2.64 (0.47, 2.06)</td>
<td>0.00 (-0.78, 0.79)</td>
<td>-4.40</td>
</tr>
<tr>
<td>Mean Temperature</td>
<td>2.86 (0.95, 4.72)</td>
<td>3.50 (2.22, 4.89)</td>
<td>0.83</td>
<td>1.58 (0.39, 2.86)</td>
<td>-0.04 (1.34, 2.94)</td>
<td>0.00</td>
<td>2.64 (0.47, 2.06)</td>
<td>0.00 (-0.78, 0.79)</td>
<td>-4.40</td>
</tr>
<tr>
<td>% ≥ 75 yr</td>
<td>2.22 (0.79, 3.58)</td>
<td>4.23 (3.02, 5.54)</td>
<td>2.66</td>
<td>1.50 (0.55, 2.46)</td>
<td>0.52 (0.24, 1.90)</td>
<td>1.02</td>
<td>2.64 (0.47, 2.06)</td>
<td>1.02 (0.31, 1.74)</td>
<td>-0.02</td>
</tr>
<tr>
<td>Age-standardized Mortality</td>
<td>2.62 (0.79, 4.48)</td>
<td>4.07 (2.22, 5.87)</td>
<td>1.14</td>
<td>1.10 (-0.16, 2.38)</td>
<td>1.07 (-0.94, 0.87)</td>
<td>1.58</td>
<td>2.64 (0.47, 2.06)</td>
<td>1.58 (0.87, 2.38)</td>
<td>3.81</td>
</tr>
<tr>
<td>% Unemployed</td>
<td>2.78 (1.42, 4.07)</td>
<td>3.75 (2.54, 4.89)</td>
<td>1.88</td>
<td>1.42 (-0.47, 3.34)</td>
<td>-0.07 (-0.78, 1.18)</td>
<td>1.50</td>
<td>2.64 (0.47, 2.06)</td>
<td>1.50 (0.71, 2.30)</td>
<td>2.45</td>
</tr>
</tbody>
</table>

Source: Adapted with permission of Health Effects Institute; Katsouyanni et al. (2009).

Regional Pattern of Ozone-Mortality Risk Estimates

In addition to examining whether individual- and community-level factors modify the O$_3$-mortality association, studies have also examined whether these associations varied regionally within the United States. Bell and Dominici (2008), in the study discussed above, also noted that O$_3$-mortality risk estimates were higher in the Northeast (1.44% [95% CI: 0.78, 2.10%]) and Industrial Midwest (0.73% [95% CI: 0.11, 1.35%]), while null associations were observed in the Southwest and Urban Midwest (Table 6-49). The regional heterogeneity in O$_3$-mortality risk estimates was further reflected by Bell and Dominici (2008) in a map of community-specific Bayesian O$_3$-mortality risk estimates (Figure 6-32). It is worth noting that in the analysis of PM$_{10}$ using the same data set, Peng et al. (2005) also found that both the Northeast and Industrial Midwest showed particularly elevated effects, especially during the summer months. As mentioned above, although no evidence for confounding of O$_3$ mortality risk estimates by PM$_{10}$ was observed, Bell et al. (2007) did find regional differences in the correlation between O$_3$ and PM$_{10}$. Thus, the heterogeneity in O$_3$ mortality risk estimates may need to be examined as a function of the correlation between PM and O$_3$. 
Smith et al. (2009b), as discussed earlier, also examined the regional difference in O₃ mortality risk estimates across the same seven regions and similarly found evidence for regional heterogeneity. In addition, Smith et al. (2009b) constructed spatial maps of the risk estimates by an extension of a hierarchical model that allows for spatial auto-correlation among the city-specific random effects. Figure 6-33 presents the spatial map of O₃ mortality coefficients from the Smith et al. (2009b) analysis that used 8-h max O₃ concentrations during the summer. The results from the Bell and Dominici (2008) analysis (Figure 6-32) shows much stronger apparent heterogeneity in O₃-mortality risk estimates across cities than the smoothed map from Smith et al. (2009b) (Figure 6-33), but both maps generally show larger risk estimates in the eastern region of the U.S.
Table 6-49  Percentage increase in daily mortality for a 10 ppb increase in 24-h average O$_3$ concentrations during the previous week by geographic region in the U.S., 1987-2000.

<table>
<thead>
<tr>
<th>Regional results</th>
<th>No. of Communities</th>
<th>Regional Estimate</th>
<th>95% PI*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Industrial Midwest</td>
<td>20</td>
<td>0.73</td>
<td>0.11, 1.35</td>
</tr>
<tr>
<td>Northeast</td>
<td>16</td>
<td>1.44</td>
<td>0.78, 2.10</td>
</tr>
<tr>
<td>Northwest</td>
<td>12</td>
<td>0.08</td>
<td>-0.92, 1.09</td>
</tr>
<tr>
<td>Southern California</td>
<td>7</td>
<td>0.21</td>
<td>-0.46, 0.88</td>
</tr>
<tr>
<td>Southeast</td>
<td>26</td>
<td>0.38</td>
<td>-0.07, 0.85</td>
</tr>
<tr>
<td>Southwest</td>
<td>9</td>
<td>-0.06</td>
<td>-0.92, 0.81</td>
</tr>
<tr>
<td>Urban Midwest</td>
<td>7</td>
<td>-0.05</td>
<td>-1.28, 1.19</td>
</tr>
<tr>
<td>National results</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>All continental communities</td>
<td>97</td>
<td>0.51</td>
<td>0.27, 0.76</td>
</tr>
<tr>
<td>All communities</td>
<td>98</td>
<td>0.52</td>
<td>0.28, 0.77</td>
</tr>
</tbody>
</table>

*PI, posterior interval.
Source: Reprinted with permission of Johns Hopkins Bloomberg School of Public Health (Bell and Dominici, 2008).

Figure 6-32  Community-specific Bayesian O$_3$-mortality risk estimates in 98 U.S. communities.
Figure 6-33 Map of spatially dependent O₃-mortality coefficients for 8-h max O₃ concentrations using summer data.

Time-Variant Factors

To date, only a few time-series studies have investigated the potential interaction between O₃ exposure and copollutants or weather variables in first stage regression models. This can be attributed to the moderate to high correlation between O₃ and these covariates, which makes such investigations methodologically challenging.

Ren et al. (2008) examined the possible synergistic effect between O₃ and temperature on mortality in the 60 largest eastern U.S. communities from the NMMAPS data during the warm months (i.e., April to October) from 1987-2000. This analysis was restricted to the eastern areas of the U.S. (i.e., Northeast, Industrial Midwest and Southeast) because a previous study which focused specifically on the eastern U.S. found that temperature-mortality patterns differ between the northeast and southeast regions possibly due to climatic differences (Curriero et al., 2002). To examine possible geographic differences in the interaction between temperature and O₃, Ren et al. (2008) further divided the NMMAPS regions into the Northeast, which included the Northeast and Industrial Midwest regions (34 cities), and the Southeast, which included the Southeast region (26 cities). The potential synergistic effects between O₃ and temperature were examined using two different models. Model 1 included an interaction term in a Generalized Additive Model (GAM) for O₃ and maximum temperature (3-day avg values were used for both terms) to examine the bivariate response surface and the pattern of interaction between the two variables in each community. Model 2 consisted of a Generalized Linear Model (GLM) that used interaction terms to stratify by “low,” “moderate,” and “high”
temperature days using the first and third quartiles of temperature as cut-offs to examine the percent increase in mortality in each community. Furthermore, a two-stage Bayesian hierarchical model was used to estimate the overall percent increase in all-cause mortality associated with short-term O₃ exposure across temperature levels and each region using model 2. The same covariates were used in both model 1 and 2. The bivariate response surfaces from model 1 suggest possible interactive effects between O₃ and temperature although the interpretation of these results is not straightforward due to the high correlation between these terms. The apparent interaction between temperature and O₃ as evaluated in model 2 varied across geographic regions. In the northeast region, a 20 ppb increase in 24-h avg O₃ concentrations at lag 0-2 was associated with an increase of 4.49% (95% posterior interval [PI]: 2.39, 6.36%), 6.21% (95% PI: 4.47, 7.66%) and 12.8% (95% PI: 9.77, 15.7%) in mortality at low, moderate and high temperature levels, respectively. The corresponding percent increases in mortality in the southeast region were 2.27% (95% PI: -2.23, 6.46%) for low temperature, 3.02% (95% PI: 0.44, 5.70%) for moderate temperature, and 2.60% (95% PI: -0.66, 6.01%) for high temperature.

When examining the relationship between temperature and O₃-related mortality, the results reported by Ren et al. (2008) (i.e., higher O₃-mortality risks on days with higher temperatures) may appear to contradict the results of Bell and Dominici (2008) described earlier (i.e., communities with higher temperature have lower O₃-mortality risk estimates). However, the observed difference in results can be attributed to the interpretation of effect modification in a second-stage regression which uses long-term average temperatures, as was performed by Bell and Dominici (2008), compared to a first-stage regression that examines the interaction between daily temperature and O₃-related mortality. In this case, the second-stage regression results from Bell and Dominici (2008) indicate that a city with lower temperatures, on average, tend to show a stronger O₃ mortality effect, whereas, in the first-stage regression performed by Ren et al. (2008), the days with higher temperature tend to show a larger O₃-mortality effect. This observed difference may in part reflect the higher air conditioning use in communities with higher long-term average temperatures. Therefore, the findings from Ren et al. (2008) indicating generally lower O₃ risk estimates in the southeast region where the average temperature is higher than in the northeast region is consistent with the regional results reported by Bell and Dominici (2008). As demonstrated by the results from both Ren et al. (2008) and Bell and Dominici (2008) caution is required when interpreting results from studies that examined interactive effects using two different approaches because potential effect modification as suggested in a second-stage regression generally does not provide evidence for a short-term interaction examined in a first-stage regression. Overall, further examination of the potential interactive (synergistic) effects of O₃ and covariates in time-series regression models is required to more clearly understand the factors that may influence O₃ mortality risk estimates.
Evaluation of the O$_3$-Mortality C-R Relationship and Related Issues

Evaluation of the O$_3$-mortality C-R relationship is not straightforward because the evidence from multicity studies (using log-linear models) suggests that O$_3$-mortality associations are highly heterogeneous across regions. In addition, there are numerous issues that may influence the shape of the O$_3$-mortality C-R relationship and the observed association between short-term O$_3$ exposure and mortality that warrant examination including: multi-day effects (distributed lags), mortality displacement (i.e., hastening of death by a short period), potential adaptation, and the exposure metric used to compute risks (e.g., 1-hour daily max versus 24-h avg). The following section presents the recent studies identified that conducted an initial examination of these issues.

Multiday Effects, Mortality Displacement, and Adaptation

The pattern of positive lagged associations followed by negative associations in a distributed lag model may be considered an indication of “mortality displacement” (i.e., deaths are occurring in frail individuals and exposure is only moving the day of death to a day slightly earlier). Zanobetti and Schwartz (2008b) examined this issue in 48 U.S. cities during the warm season (i.e., June-August) for the years 1989-2000. In an initial analysis, the authors applied a GLM to examine same-day O$_3$-mortality effects, and in the model included an unconstrained distributed lag for apparent temperature to take into account the effect of temperature on the day death occurred and the previous 7 days. To examine mortality displacement Zanobetti and Schwartz (2008b) refit models using two approaches: an unconstrained and a smooth distributed lag each with 21-day lags for O$_3$. In this study, all-cause mortality as well as cause-specific mortality (i.e., cardiovascular, respiratory, and stroke) were examined for evidence of mortality displacement. The authors found a 0.96% (95% CI: 0.60, 1.30%) increase in all-cause mortality across all 48 cities for a 30 ppb increase in 8-h max O$_3$ concentrations at lag 0 whereas the combined estimate of the unconstrained distributed lag model (lag 0-20) was 1.54% (95% CI: 0.15, 2.91%). Similarly, when examining the cause-specific mortality results (Table 6-50), larger risk estimates were observed for the distributed lag model compared to the lag 0 day estimates. However, for stroke a slightly larger effect was observed at lags 4-20 compared to lags 0-3 suggesting a larger window for O$_3$-induced stroke mortality. This is further supported by the sum of lags 0 through 20 days showing the greatest effect. Overall, these results suggest that estimating the mortality risk using a single day of O$_3$ exposure may underestimate the public health impact, but the extent of multi-day effects appear to be limited to a few days. This is further supported by the shape of the combined smooth distributed lag (Figure 6-34). It should be noted that the proportion of total variation in the effect estimates due to the between-cities heterogeneity, as measured by I$^2$ statistic, was relatively low (4% for the lag 0 estimates and 21% for the distributed lag), but 21 out of the 48 cities exhibited null or negative estimates. As a result, the estimated shape of the distributed lag cannot be
interpreted as a general form of lag structure of associations applicable to all the cities included in this analysis.

Samoli et al. (2009) also investigated the temporal pattern of mortality effects in response to short-term exposure to O\textsubscript{3} in 21 European cities that were included in the APHEA2 project. Using a method similar to Zanobetti and Schwartz (2008b), the authors applied unconstrained distributed lag models with lags up to 21 days in each city during the summer months (i.e., June through August) to examine the effect of O\textsubscript{3} on all-cause, cardiovascular, and respiratory mortality. They also applied a generalized additive distributed lag model to obtain smoothed distributed lag coefficients. However, unlike Zanobetti and Schwartz (2008b), Samoli et al. (2009) controlled for temperature using a linear term for humidity and an unconstrained distributed lag model of temperature at lags 0-3 days. The choice of 0- through 3-day lags of temperature was based on a previous European multicity study (Baccini et al., 2008), which suggested that summer temperature effects last only a few days. Upon combining the individual city estimates across cities in a second stage regression, Samoli et al. (2009) found that the estimated effects on respiratory mortality were extended for a period of two weeks. However, for all-cause and cardiovascular mortality, the 21-day distributed lag models yielded null or (non-significant) negative estimates (Table 6-51). Figure 6-35 shows the distributed lag coefficients for all-cause mortality, which exhibit a declining trend and negative coefficients beyond 5-day lags. The authors’ interpretation of these results was that “using single-day exposures may have overestimated the effects on all-cause and cardiovascular mortality, but underestimated the effects on respiratory mortality.” Thus, the results in part suggest evidence of mortality displacement for all-cause and cardiovascular mortality.
Table 6-50  Estimated effect of a 10 ppb increase in 8-h max O₃ concentrations on mortality during the summer months for single-day and distributed lag models.

<table>
<thead>
<tr>
<th></th>
<th>% (Percentage)</th>
<th>95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Total mortality</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.32</td>
<td>0.20, 0.43</td>
</tr>
<tr>
<td>Sum lags 0-20</td>
<td>0.51</td>
<td>0.05, 0.96</td>
</tr>
<tr>
<td>Sum lags 0-3</td>
<td>0.53</td>
<td>0.28, 0.77</td>
</tr>
<tr>
<td>Sum lags 4-20</td>
<td>-0.02</td>
<td>-0.35, 0.31</td>
</tr>
<tr>
<td><strong>Cardiovascular mortality</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.47</td>
<td>0.30, 0.64</td>
</tr>
<tr>
<td>Sum lags 0-20</td>
<td>0.49</td>
<td>-0.01, 1.00</td>
</tr>
<tr>
<td>Sum lags 0-3</td>
<td>0.80</td>
<td>0.48, 1.13</td>
</tr>
<tr>
<td>Sum lags 4-20</td>
<td>-0.23</td>
<td>-0.67, 0.22</td>
</tr>
<tr>
<td><strong>Respiratory mortality</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.54</td>
<td>0.26, 0.81</td>
</tr>
<tr>
<td>Sum lags 0-20</td>
<td>0.61</td>
<td>-0.41, 1.65</td>
</tr>
<tr>
<td>Sum lags 0-3</td>
<td>0.83</td>
<td>0.38, 1.28</td>
</tr>
<tr>
<td>Sum lags 4-20</td>
<td>-0.24</td>
<td>-1.08, 0.60</td>
</tr>
<tr>
<td><strong>Stroke</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.37</td>
<td>0.01, 0.74</td>
</tr>
<tr>
<td>Sum lags 0-20</td>
<td>2.20</td>
<td>0.76, 3.67</td>
</tr>
<tr>
<td>Sum lags 0-3</td>
<td>0.92</td>
<td>0.26, 1.59</td>
</tr>
<tr>
<td>Sum lags 4-20</td>
<td>1.26</td>
<td>0.05, 2.49</td>
</tr>
</tbody>
</table>

Figure 6-34  Estimated combined smooth distributed lag for 48 U.S. cities during the summer months.
Table 6-51 Estimated percent increase in cause-specific mortality (and 95% CIs) for a 10-µg/m³ increase in 8-h daily max O₃ during June-August.

<table>
<thead>
<tr>
<th></th>
<th>Fixed effects % (95% CI)</th>
<th>Random effects % (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Total mortalityᵃ</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.28 (0.11, 0.45)</td>
<td>0.28 (0.07, 0.48)</td>
</tr>
<tr>
<td>Average lags 0-1</td>
<td>0.24 (0.15, 0.34)</td>
<td>0.22 (0.08, 0.35)</td>
</tr>
<tr>
<td>Sum lags 0-20, unconstrained</td>
<td>0.01 (-0.40, 0.41)</td>
<td>-0.54 (-1.28, 0.20)</td>
</tr>
<tr>
<td>Sum lags 0-20, penalized</td>
<td>0.01 (-0.41, 0.42)</td>
<td>-0.56 (-1.30, 0.19)</td>
</tr>
<tr>
<td><strong>Cardiovascular mortalityᵃ</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.43 (0.18, 0.69)</td>
<td>0.37 (0.05, 0.69)</td>
</tr>
<tr>
<td>Average lags 0-1</td>
<td>0.33 (0.19, 0.48)</td>
<td>0.25 (0.03, 0.47)</td>
</tr>
<tr>
<td>Sum lags 0-20, unconstrained</td>
<td>-0.33 (-0.93, 0.29)</td>
<td>-0.62 (-1.47, 0.24)</td>
</tr>
<tr>
<td>Sum lags 0-20, penalized</td>
<td>-0.32 (-0.92, 0.28)</td>
<td>-0.57 (-1.39, 0.26)</td>
</tr>
<tr>
<td><strong>Respiratory mortalityᵃ</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lag 0</td>
<td>0.36 (-0.21, 0.94)</td>
<td>0.36 (-0.21, 0.94)</td>
</tr>
<tr>
<td>Average lags 0-1</td>
<td>0.40 (0.11, 0.70)</td>
<td>0.40 (0.11, 0.70)</td>
</tr>
<tr>
<td>Sum lags 0-20, unconstrained</td>
<td>3.35 (1.90, 4.83)</td>
<td>3.35 (1.90, 4.83)</td>
</tr>
<tr>
<td>Sum lags 0-20, penalized</td>
<td>3.66 (2.25, 5.08)</td>
<td>3.66 (2.25, 5.08)</td>
</tr>
</tbody>
</table>

ᵃAnalysis for the same day (lag 0), the average of the same and previous day (lag 0-1), the unconstrained distributed lag model for the sum of 0-20 days and the penalized distributed lag model (lag 0-20)

Source: Used with permission of BMJ Group (Samoli et al., 2009).

Although the APHENA project (Katsouyanni et al., 2009) did not specifically investigate mortality displacement and therefore did not consider longer lags (e.g., lag >3 days), the study did present O₃ risk estimates for lag 0-1, lag 1, and a distributed lag model of 0-2 days in the Canadian, European, and U.S. datasets. Katsouyanni et al. (2009) found that the results vary somewhat across the regions, but, in general, there was no indication that the distributed lag model with up to a 2-day lag yielded meaningfully larger O₃ mortality risk estimates than the lag 0-1 and lag 1 results. For example, for all-cause mortality, using the model with natural splines and 8 df/year to adjust for seasonal trends, the reported percent excess risk for mortality for a 40 ppb increase in 1-h max O₃ concentrations for lag 0-1, lag 1, and the distributed lag model (lag 0-2) was 2.70% (95% CI: 1.02, 4.40%), 1.42% (95% CI: 0.08, 2.78%), and 3.02% (95% CI: 1.10, 4.89%), respectively. Thus, the observed associations appear to occur over a short time period, (i.e., a few days). Similarly, the Public Health and Air Pollution in Asia (PAPA) study (Wong et al., 2010) also examined multiple lag days (i.e., lag 0, lag 0-1, and lag 0-4), and although it did not specifically examine mortality displacement it does provide additional evidence regarding the timing of mortality effects proceeding O₃ exposure. In a combined analysis using data from all four cities examined (Bangkok, Hong Kong, Shanghai, and Wuhan), excess risk estimates at lag 0-4 were larger than those at lag...
0 or lag 0-1 in both fixed and random effect models (results not presented quantitatively). The larger risk estimates at lag 0-4 can primarily be attributed to the strong associations observed in Bangkok and Shanghai. However, it is worth noting that Bangkok differs from the three Chinese cities included in this analysis in that it has a tropical climate and does not exhibit seasonal patterns of mortality. As a result, Wong et al. (2010) examined the O₃-mortality associations at lag 0-1 in only the three Chinese cities and found that risk estimates were slightly reduced from 2.26% (95% CI: 1.36, 3.16) in the 4 city analysis to 1.84% (0.77, 2.86) in the 3 city analysis for a 30 ppb increase in 8-h max O₃ concentrations. Overall, the PAPA study further supports the observation of the APHENA study that associations between O₃ and mortality occur over a relatively short-time period, but also indicates that it may be difficult to interpret O₃-mortality associations across cities with different climates and mortality patterns.

When comparing the studies that explicitly examined the potential for mortality displacement in the O₃-mortality relationship, the results from Samoli et al. (2009), which provide evidence that suggests mortality displacement, are not consistent with those reported by Zanobetti and Schwartz (2008b). However, the shapes of the estimated smooth distributed lag associations are similar (Figure 6-34 and Figure 6-35). A closer examination of these figures shows that in the European data beyond a lag of 5 days the estimates remain negative whereas in the U.S. data the results remain near zero for the corresponding lags. These observed difference could be due to the differences in the model specification between the two studies, specifically the use of: an unconstrained distributed lag model for apparent temperature up to 7 previous days (Zanobetti and Schwartz, 2008b) versus a linear term for humidity and an unconstrained distributed lag model of temperature up to 3 previous days (Samoli et al., 2009); and natural cubic splines with 2 df per season (Zanobetti and Schwartz, 2008b) versus dummy variables per month per year to adjust for season (Samoli et al., 2009). It is important to note that these differences in model specification may have also influenced the city-to-city variation in risk estimates observed in these two studies (i.e., homogenous estimates across cities in Zanobetti and Schwartz (2008b) and heterogeneous estimates across cities in Samoli et al. (2009). Overall, the evidence of mortality displacement remains unclear, but Samoli et al. (2009), Zanobetti and Schwartz (2008b), and Katsouyanni et al. (2009) all suggest that the positive associations between O₃ and mortality are observed mainly in the first few days after exposure.
Controlled human exposure studies have demonstrated an adaptive response to O₃ exposure for respiratory effects, such as lung function decrements, but this issue has not been examined in the epidemiologic investigation of mortality effects of O₃. Zanobetti and Schwartz (2008a) examined if there was evidence of an adaptive response in the O₃-mortality relationship in 48 U.S. cities from 1989 to 2000 (i.e., the same data analyzed in Zanobetti and Schwartz (2008b)). The authors examined all-cause mortality using a case-crossover design to estimate the same-day (i.e., lag 0) effect of O₃, matched on referent days from every-3rd-day in the same month and year as the case. Zanobetti and Schwartz (2008a) examined O₃-mortality associations by: season, month in the summer season (i.e., May through September), and age categories in the summer season (Table 6-52). The estimated O₃ mortality risk estimate at lag 0 was found to be highest in the summer (1.51% [95% CI: 1.14, 1.87%]; lag 0 for a 30 ppb increase in 8-h max O₃ concentrations), and, within the warm months, the association was highest in July (1.96% [95% CI: 1.42, 2.48%);
Upon further examination of the summer months, the authors also observed diminished effects in August (0.84% [95% CI: 0.33, 1.39%]; lag 0). Based on these results, the authors concluded that the mortality effects of O₃ appear diminished later in the O₃ season.

Table 6-52  Percent excess all-cause mortality per 10 ppb increase in daily 8-h max O₃ on the same day, by season, month, and age groups.

<table>
<thead>
<tr>
<th></th>
<th>%</th>
<th>95% CI</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>By Season</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Winter</td>
<td>-0.13</td>
<td>-0.56, 0.29</td>
</tr>
<tr>
<td>Spring</td>
<td>0.35</td>
<td>0.16, 0.54</td>
</tr>
<tr>
<td>Summer</td>
<td>0.50</td>
<td>0.38, 0.62</td>
</tr>
<tr>
<td>Fall</td>
<td>0.05</td>
<td>-0.14, 0.24</td>
</tr>
<tr>
<td><strong>By Month</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>May</td>
<td>0.48</td>
<td>0.28, 0.68</td>
</tr>
<tr>
<td>June</td>
<td>0.46</td>
<td>0.24, 0.68</td>
</tr>
<tr>
<td>July</td>
<td>0.65</td>
<td>0.47, 0.82</td>
</tr>
<tr>
<td>August</td>
<td>0.28</td>
<td>0.11, 0.46</td>
</tr>
<tr>
<td>September</td>
<td>-0.09</td>
<td>-0.35, 0.16</td>
</tr>
<tr>
<td><strong>By Age Group</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0-20</td>
<td>0.08</td>
<td>-0.42, 0.57</td>
</tr>
<tr>
<td>21-30</td>
<td>0.10</td>
<td>-0.67, 0.87</td>
</tr>
<tr>
<td>31-40</td>
<td>0.07</td>
<td>-0.38, 0.52</td>
</tr>
<tr>
<td>41-50</td>
<td>0.08</td>
<td>-0.27, 0.43</td>
</tr>
<tr>
<td>51-60</td>
<td>0.54</td>
<td>0.19, 0.89</td>
</tr>
<tr>
<td>61-70</td>
<td>0.38</td>
<td>0.16, 0.61</td>
</tr>
<tr>
<td>71-80</td>
<td>0.50</td>
<td>0.32, 0.67</td>
</tr>
<tr>
<td>80</td>
<td>0.29</td>
<td>0.13, 0.44</td>
</tr>
</tbody>
</table>


To further evaluate the potential adaptive response observed in Zanobetti and Schwartz (2008a) the distribution of the O₃ concentrations across the 48 U.S. cities during July and August was examined. Both July and August were found to have comparable means of 48.6 and 47.9 ppb with a reported maximum value of 97.9 and 96.0 ppb, respectively. Thus, the observed reduction in O₃-related mortality effect estimates in August (0.84%) compared to July (1.96%) appears to support the existence of an adaptive response. However, unlike an individual’s adaptive response

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1 These values have been standardized to the increment used throughout the ISA for max 8-h avg increase in O₃ concentrations of 30 ppb. These values differ from those presented in Table 6-52; from Zanobetti and Schwartz (2008a) because the authors presented values for a 10 ppb increase in max 8-h avg O₃ concentrations.
to decrements in lung function from short-term O_3 exposure, an examination of mortality prevents a direct observation of adaptation. Rather, for mortality the adaptation hypothesis is tested with a tacit assumption that, whatever the mechanism for O_3-induced mortality, the risk of death from short-term O_3 exposure is reduced over the course of the summer months through repeated exposures. This idea would translate to a smaller population that would die from O_3 exposure toward the end of summer. This may complicate the interpretation of the distributed lag coefficients with long lag periods because the decreased coefficients may reflect diminished effects of the late summer, rather than diminished effects that are constant across the summer. These intertwined issues need to be investigated together in future research.

**Exposure Metric**

When examining the association between short-term O_3 exposure and mortality it is also important to consider the exposure metric used (i.e., 24-h avg, 8-h max, and 1-h max). To date, only a few studies have conducted analyses to examine the impact of different exposure metrics on O_3 mortality risk estimates. In Smith et al. (2009b), the authors examined the effect of different exposure metrics (i.e., 24-h avg, 8-h max, and 1-h max) on O_3-mortality regression coefficients. When examining whether there are differences in city-specific risk estimates when using different exposure metrics, Smith et al. (2009b) found a rather high correlation (r = 0.7 – 0.8) between risk estimates calculated using 24-h avg versus 8-h max and 1-h max versus 8-h max averaging times. These results are consistent with the correlations reported by Darrow et al. (2011a) (Section 6.2.7.3) between the 8-h max and 24-avg exposure metrics.

In addition to these recent studies published since the 2006 O_3 AQCD, Gryparis et al. (2004) also supports the high correlation between 1-h max and 8-h max O_3 concentrations reported in Smith et al. (2009b) and Darrow et al. (2011a) and the subsequent high degree of similarity between mortality risk estimates calculated using either metric. Although only a limited number of studies have examined the effect of different exposure metrics on O_3-mortality risk estimates, these studies suggest relatively comparable results across the exposure metrics used.

**Ozone-Mortality C-R Relationship and Threshold Analyses**

Several of the recent studies evaluated have applied a variety of statistical approaches to examine the shape of the O_3-mortality C-R relationship and whether a threshold exists. The approach used by Bell et al. (2006) consisted of applying four statistical models to the NMMAPS data, which included 98 U.S. communities for the period 1987-2000. These models included: a linear analysis (i.e., any change in O_3 concentration can be associated with mortality) (Model 1); a subset analysis (i.e., examining O_3-mortality relationship below a specific 24-avg concentration, ranging from 5 to 60 ppb) (Model 2); a threshold analysis (i.e., assuming that an association between O_3 and mortality is observed above a specific concentration and
not below it, using the threshold values set at an increment of 5 ppb between 0 to 60 ppb and evaluating a presence of a local minima in AICs computed at each increment) (Model 3); and nonlinear models using natural cubic splines with boundary knots placed at 0 and 80 ppb, and interior knots placed at 20 and 40 ppb (Model 4). A two-stage Bayesian hierarchical model was used to examine these models and O₃-mortality risk estimates at the city-level in the first stage analysis and aggregate estimates across cities in the 2nd stage analysis using the average of 0- and 1-day lagged 24-h avg O₃ concentrations. The results from all of these models suggest that if a threshold exists it does so well below the current O₃ NAAQS. When restricting the analysis to all days when the 1997 O₃ NAAQS 8-hour standard (i.e., 84 ppb daily 8-h max) is met in each community, Bell et al. (2006) found there was still a 0.60% (95% PI: 0.30, 0.90%) increase in mortality per 20 ppb increase in 24-h avg O₃ concentrations at lag 0-1. Figure 6-36 shows the combined C-R curve obtained using the nonlinear model (Model 4). Although these results suggest the lack of threshold in the O₃-mortality relationship, it is difficult to interpret such a curve because: (1) there is uncertainty around the shape of the C-R curve at 24-h avg O₃ concentrations generally below 20 ppb, and (2) the C-R curve does not take into consideration the heterogeneity in O₃-mortality risk estimates across cities.
Using the same NMMAPS dataset as Bell et al. (2006), Smith et al. (2009b) further examined the O₃-mortality C-R relationship. Similar to Bell et al. (2006), Smith et al. (2009b) conduct a subset analysis, but instead of restricting the analysis to days with O₃ concentrations below a cutoff the authors only include days above a defined cutoff in the analysis. The results of this “reversed subset” approach are in line with those reported by Bell et al. (2006); consistent positive associations at all cutoff points up to a defined concentration where the total number of days with 24-h avg O₃ concentrations above a value are so limited that the variability around the central estimate is increased. In the Smith et al. (2009b) analysis this observation was initially observed at 45 ppb, with the largest variability at 60 ppb; however, unlike Bell et al. (2006) where 73% of days are excluded when subsetting the data to less than 20 ppb, the authors do not detail the number of days of data included in the subset analyses at higher concentrations. In addition to the subset analysis, Smith et al. (2009b) examined the shape of the C-R curve using a piecewise linear approach with cutpoints at 8-h avg concentrations of 40 ppb, 60 ppb, and 80 ppb. Smith et al. (2009b) found that the shape of the C-R curve is similar to that reported by Bell et al. (2006) (Figure 6-36), but argue that slopes of the β for each piece of the curve are highly variable with the largest variation in the 60-80 ppb range. However, the larger
variability around the β between 60-80 would be expected due to the small number of days with O₃ concentrations within that range in an all-year analysis. This result is consistent with that observed by Bell et al. (2006), which is presented in Figure 6-36.

The APHENA project (Katsouyanni et al., 2009) also analyzed the Canadian and European datasets (the U.S. data were analyzed for PM₁₀ only) for evidence of a threshold, using the threshold analysis method (Model 3) applied in Bell et al. (2006) study described above. There was no evidence of a threshold in the Canadian data (i.e., the pattern of AIC values for each increment of a potential threshold value varied across cities, most of which showed no local minima). Likewise, the threshold analysis conducted using the European data also showed no evidence of a threshold.

The PAPA study, did not examine whether a threshold exists in the O₃-mortality C-R relationship, but instead the shape of the C-R curve individually for each city (Bangkok, Hong Kong, Shanghai, and Wuhan) (Wong et al., 2010). Using a natural spline smoother with 3df for the O₃ term, Wong et al. (2010) examined whether non-linearity was present by testing the change in deviance between the smoothed, non-linear model and an unsmoothed linear model with 1 df. For each of the cities, both across the full range of the O₃ distribution and specifically within the range of the 25th to 75th percentile of each city’s 24-h avg O₃ concentrations (i.e., a range of 9.7 ppb to 60.4 ppb across the cities) there was no evidence of a non-linear relationship in the O₃-mortality C-R curve. It should be noted that the range of the 25th to 75th percentiles of O₃ concentrations in all of the cities, except Wuhan, was at the lower end of the distribution observed in the U.S. using all-year data, where the range from the 25th to 75th percentiles is 30 ppb to 50 ppb (Table 3-6).

Additional threshold analyses were conducted using NMMAPS data, by Xia and Tong (2006) and Stylianou and Nicolich (2009). Both studies used a new statistical approach developed by Xia and Tong (2006) to examine thresholds in the O₃ mortality C-R relationship. The approach consisted of an extended GAM model, which accounted for the cumulative and nonlinear effects of air pollution using a weighted cumulative sum for each pollutant, with the weights (non-increasing further into the past) derived by a restricted minimization method. The authors did not use the term distributed lag model, but their model has the form of distributed lag model, except that it allows for nonlinear functional forms. Using NMMAPS data for 1987-1994 for 3 U.S. cities (Chicago, Pittsburgh, and El Paso), Xia and Tong (2006) found that the extent of cumulative effects of O₃ on mortality were relatively short. While the authors also note that there was evidence of a threshold effect around 24-h avg concentrations of 25 ppb, the threshold values estimated in the analysis were sometimes in the range where data density was low. Thus, this threshold analysis needs to be replicated in a larger number of cities to confirm this observation. It should be noted that the model used in this analysis did not include a smooth function of days to adjust for unmeasured temporal confounders, and instead adjusted for season using a temperature term. As a result, these results need to be viewed with caution because some potential temporal confounders (e.g., influenza) do not always follow seasonal patterns of temperature.
Stylianou and Nicolich (2009) examined the existence of thresholds following an approach similar to Xia and Tong (2006) for all-cause, cardiovascular, and respiratory mortality using data from NMMAPS for nine major U.S. cities (i.e., Baltimore, MD, Chicago, IL, Dallas/Fort Worth, TX, Los Angeles, CA, Miami, FL, New York, NY, Philadelphia, PA, Pittsburgh, PA, and Seattle, WA) for the years 1987-2000. The authors found that PM$_{10}$ and O$_3$ were the two important predictors of mortality. Stylianou and Nicolich (2009) found that the estimated O$_3$-mortality risks varied across the nine cities with the models exhibiting apparent thresholds, in the 10-45 ppb range for O$_3$ (3-day accumulation). However, given the city-to-city variation in risk estimates, combining the city-specific estimates into an overall estimate complicates the interpretation of a threshold. Unlike the Xia and Tong (2006) analysis, Stylianou and Nicolich (2009) included a smooth function of time to adjust for seasonal/temporal confounding, which could explain the difference in results between the two studies.

In conclusion, the evaluation of the O$_3$-mortality C-R relationship did not find any evidence that supports a threshold in the relationship between short-term exposure to O$_3$ and mortality within the range of O$_3$ concentrations observed in the United States. Additionally, recent evidence suggests that the shape of the O$_3$-mortality C-R curve remains linear across the full range of O$_3$ concentrations. However, the studies evaluated demonstrated that the heterogeneity in the O$_3$-mortality relationship across cities (or regions) complicates the interpretation of a combined C-R curve and threshold analysis. Given the effect modifiers identified in the mortality analyses that are also expected to vary regionally (e.g., temperature, air conditioning prevalence), a national or combined analysis may not be appropriate to identify whether a threshold exists in the O$_3$-mortality C-R relationship. Overall, the studies evaluated support a linear O$_3$-mortality C-R relationship and continue to support the conclusions from the 2006 O$_3$ AQCD, which stated that “if a population threshold level exists in O$_3$ health effects, it is likely near the lower limit of ambient O$_3$ concentrations in the United States” (U.S. EPA, 2006b).

6.6.2.4 Associations of Cause-Specific Mortality and Short-term O$_3$ Exposure

In the 2006 O$_3$ AQCD, an evaluation of studies that examined cause-specific mortality found consistent positive associations between short-term O$_3$ exposure and cardiovascular mortality, with less consistent evidence for associations with respiratory mortality. The majority of the evidence for associations between O$_3$ exposure and cause-specific mortality were from single-city studies, which had small daily mortality counts and subsequently limited statistical power to detect associations.

New multicity studies evaluated in this review build upon and confirm the associations between short-term O$_3$ exposure and cause-specific mortality identified in the 2006 O$_3$ AQCD (U.S. EPA, 2006b) (Figure 6-37 [and Table 6-53]).
In APHENA, a multicontinent study that consisted of the NMMAPS, APHEA2 and Canadian multiplicity datasets, consistent positive associations were reported for both cardiovascular and respiratory mortality in all-year analyses when focusing on the natural spline model with 8 df/year (Figure 6-37 [and Table 6-53]). The associations between O₃ exposure and cardiovascular and respiratory mortality in all-year analyses were further supported by the multiplicity PAPA study (Wong et al., 2010). The magnitude of cardiovascular mortality associations were primarily larger in analyses restricted to the summer season compared to those observed in all-year analyses (Figure 6-37 [and Table 6-53]). Additional multiplicity studies from the U.S. (Zanobetti and Schwartz, 2008b) and Europe (Stafoggia et al., 2010; Samoli et al., 2009) that conducted summer season analyses provide evidence supporting associations between O₃ exposure and cardiovascular and respiratory mortality that are similar or larger in magnitude compared to those observed in all-year analyses.

Of the studies evaluated, only the APHENA study (Katsouyanni et al., 2009) and an Italian multiplicity study (Stafoggia et al., 2010) conducted an analysis of the potential for copollutant confounding of the O₃ cause-specific mortality relationship. When focusing on the natural spline model with 8 df/year and lag 1 results (as discussed in Section 6.6.2.1), the APHENA study found that O₃ cause-specific mortality risk estimates were fairly robust to the inclusion of PM₁₀ in copollutant models in the European dataset with more variability in the U.S. and Canadian datasets (i.e., copollutant risk estimates increased and decreased for respiratory and cardiovascular mortality). In summer season analyses cardiovascular O₃ mortality risk estimates were robust in the European dataset and attenuated but remained positive in the U.S. datasets; whereas, respiratory O₃ mortality risk estimates were attenuated in the European dataset and robust in the U.S. dataset. The authors did not examine copollutant models during the summer season in the Canadian dataset (Figure 6-30 [and Table 6-45]). Interpretation of these results requires caution; however, due to the different PM sampling schedules employed in each of these study locations (i.e., primarily every-6th day in the U.S. and Canadian datasets and every-day in the European dataset). The results of the summer season analyses from the APHENA study (Katsouyanni et al., 2009) are consistent with those from a study of 10 Italian cities during the summer months (Stafoggia et al., 2010). Stafoggia et al. (2010) found that cardiovascular (14.3% [95% CI: 6.7, 22.4%]) and cerebrovascular (8.5% [95% CI: 0.06, 16.3%]) mortality O₃ effect estimates were robust to the inclusion of PM₁₀ in copollutant models (14.3% [95% CI: 6.7, 23.1%] and 7.3% [95% CI: -1.2, 16.3], respectively), while respiratory mortality O₃ effects estimates (17.6% [95% CI: 1.8, 35.5%]) were attenuated, but remained positive (9.2% [95% CI: -6.9, 28.8%]).
Effect estimates are for a 20 ppb increase in 24-h avg; 30 in 8-h max; and 40 ppb increase in 1-h max O₃ concentrations.

Red = cardiovascular; blue = respiratory; closed circles = all-year analysis; and open circles = summer-only analysis. An “a” represents studies from the 2006 O₃ AQCD. A “b” represents risk estimates from APHENA-Canada standardized to an approximate IQR of 5.1 ppb for a 1-h max increase in O₃ concentrations (Section 6.2.7.2).

Figure 6-37  Percent increase in cause-specific mortality.
Table 6-53  Corresponding effect estimates for Figure 6-37.

<table>
<thead>
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<th>Study*</th>
<th>Location</th>
<th>Ages</th>
<th>Lag</th>
<th>Avg Time</th>
<th>%Increase (95% CI)</th>
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<td></td>
<td>All-year - Cardiovascular</td>
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<td>2.20 (0.06, 4.37)</td>
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<td></td>
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<td>8-h max</td>
<td>2.70 (1.29, 4.32)</td>
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<td>8-h max</td>
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<td>8-h max</td>
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<td>DL(0-5)</td>
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<td>14.33 (6.65, 22.4)</td>
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<tr>
<td>Bell et al. (2005)*</td>
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<td>24-h avg</td>
<td>0.94 (-1.02, 2.96)</td>
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<td>8-h max</td>
<td>2.02 (-0.41, 4.49)</td>
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<td>2.54 (-3.32, 8.79)</td>
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Collectively, the results from the new multicity studies provide evidence of associations between short-term O₃ exposure and cardiovascular and respiratory mortality with additional evidence indicating these associations persist, and in some cases the magnitude of associations are increased, in the summer season. Although copollutant analyses of cause-specific mortality are limited, the APHENA study found that O₃ cause-specific mortality risk estimates were fairly robust to the inclusion of PM₁₀ in copollutant models when focusing on the dataset with daily PM₁₀ data (i.e., the European dataset), which is supported by the results from Stafoggia et al. (2010). Additionally, APHENA found that O₃ cause-specific mortality risk estimates were moderately to substantially sensitive (e.g., increased or attenuated) to inclusion of PM₁₀ in the U.S. and Canadian datasets. However, the mostly every-6th-day sampling schedule for PM₁₀ in the U.S. and Canadian datasets greatly reduced their sample size and limits the interpretation of these results.

6.6.3 Summary and Causal Determination

The evaluation of new multicity studies that examined the association between short-term O₃ exposure and mortality found evidence which supports the conclusions of the 2006 O₃ AQCD. These new studies reported consistent positive associations between short-term O₃ exposure and all-cause (nonaccidental) mortality, with associations persisting or increasing in magnitude during the warm season, and provide additional support for associations between O₃ exposure and cardiovascular and respiratory mortality.
Recent studies further examined potential confounders (e.g., copollutants and seasonality) of the O₃-mortality relationship. Because the PM-O₃ correlation varies across regions, due to the difference in PM chemical constituents, interpretation of the combined effect of PM on the relationship between O₃ and mortality is not straightforward. Unlike previous studies that were limited to primarily examining the confounding effects of PM₁₀, the new studies expanded their analyses to include multiple PM indices (e.g., PM₁₀, PM₂.₅, and PM components). An examination of copollutant models found evidence that associations between O₃ and all-cause mortality were robust to the inclusion of PM₁₀ or PM₂.₅ (Stafoggia et al., 2010; Katsouyanni et al., 2009; Bell et al., 2007), while other studies found evidence for a modest reduction (~20-30%) when examining PM₁₀ (Smith et al., 2009b). Additional evidence suggests potential sensitivity (e.g., increases and attenuation) of O₃ mortality risk estimates to copollutants by age group or cause-specific mortality (e.g., respiratory and cardiovascular) (Stafoggia et al., 2010; Katsouyanni et al., 2009). An examination of PM components, specifically sulfate, found evidence for reductions in O₃-mortality risk estimates in copollutant models (Franklin and Schwartz, 2008). Overall, across studies, the potential impact of PM indices on O₃-mortality risk estimates tended to be much smaller than the variation in O₃-mortality risk estimates across cities suggesting that O₃ effects are independent of the relationship between PM and mortality. However, interpretation of the potential confounding effects of PM on O₃-mortality risk estimates requires caution. This is because the PM-O₃ correlation varies across regions, due to the difference in PM components, complicating the interpretation of the combined effect of PM on the relationship between O₃ and mortality. Additionally, the limited PM or PM component datasets used as a result of the every-3rd- and 6th-day PM sampling schedule instituted in most cities limits the overall sample size employed to examine whether PM or one of its components confounds the O₃-mortality relationship.

An examination of potential seasonal confounding of the O₃-mortality relationship found that the extent of smoothing or the methods used for adjustment can influence O₃ risk estimates when not applying enough degrees of freedom to control for temporal/season trends (Katsouyanni et al., 2009). This is because of the opposing seasonal trends between O₃ and mortality.

The multicity studies evaluated within this section also examined the regional heterogeneity observed in O₃-mortality risk estimates. These studies provide evidence which suggests generally higher O₃-mortality risk estimates in northeastern U.S. cities with some regions showing no associations between O₃ exposure and mortality (e.g., Southwest, Urban Midwest) (Smith et al., 2009b; Bell and Dominici, 2008). Multicity studies that examined individual- and community-level characteristics identified characteristics that may explain the observed regional heterogeneity in O₃-mortality risk estimates as well as characteristics of populations potentially at greatest risk for O₃-related health effects. An examination of community-level characteristics found an increase in the O₃-mortality risk estimates in cities with higher unemployment, percentage of the population Black/African-American, percentage of the working population that uses public transportation, lower temperatures, and lower prevalence of central air conditioning (Medina-Ramón...
and Schwartz, 2008). Additionally, a potential interactive, or synergistic, effect on the O₃-mortality relationship was observed when examining differences in the O₃-mortality association across temperature levels (Ren et al., 2008).

An examination of individual-level characteristics found evidence that older age, female sex, Black race, having atrial fibrillation, SES indicators (i.e., educational attainment, income level, and employment status), and out-of hospital deaths, specifically in those individuals with diabetes, modify O₃-mortality associations (Cakmak et al., 2011; Stafoggia et al., 2010; Medina-Ramón and Schwartz, 2008), and lead to increased risk of O₃-related mortality. Overall, additional research is warranted to further confirm whether these characteristics, individually or in combination, can explain the observed regional heterogeneity.

Additional studies were evaluated that examined factors that may influence the shape of the O₃-mortality C-R curve, such as multi-day effects, mortality displacement, adaptation, the use of different exposure metrics (i.e., 24-h avg, 8-h max or 1-h max), and whether a threshold exists in the O₃-mortality relationship. An examination of multiday effects in a U.S. and European multicity study found conflicting evidence for mortality displacement, but both studies suggest that the positive associations between O₃ and mortality are observed mainly in the first few days after exposure (Samoli et al., 2009; Zanobetti and Schwartz, 2008b). A U.S. multicity study found evidence of an adaptive response to O₃ exposure, with the highest risk estimates earlier in the O₃ season (i.e., July) and diminished effects later (i.e., August) (Zanobetti and Schwartz, 2008a). However, the evidence of adaptive effects has an implication for the interpretation of multi-day effects, and requires further analysis. The limited number of studies conducted that examined the effect of using different exposure metrics (i.e., 1-h max, 8-h max, and 24-h avg) when examining the O₃-mortality relationship found relatively comparable O₃-mortality risk estimates across the exposure metrics used (Smith et al., 2009b; Gryparis et al., 2004). Analyses that specifically focused on the O₃-mortality C-R relationship supported a linear O₃-mortality relationship and found no evidence of a threshold within the range of O₃ concentrations in the U.S., but did observe evidence for potential differences in the C-R relationship across cities (Katsouyanni et al., 2009; Stylianou and Nicolich, 2009; Bell et al., 2006). Collectively, these studies support the conclusions of the 2006 O₃ AQCD that “if a population threshold level exists in O₃ health effects, it is likely near the lower limit of ambient O₃ concentrations in the U.S.”

Studies that examined the association between short-term O₃ exposure and cause-specific mortality confirm the associations with both cardiovascular and respiratory mortality reported in the 2006 O₃ AQCD (Stafoggia et al., 2010; Wong et al., 2010; Katsouyanni et al., 2009; Samoli et al., 2009; Zanobetti and Schwartz, 2008b). These associations were primarily larger in magnitude during the summer season compared to all-year analyses. Of the studies that examined the potential confounding effects of PM [i.e., Stafoggia et al. (2010); Katsouyanni et al. (2009)], O₃ mortality associations remained relatively robust in copollutant models, but interpretation of these studies was complicated by the different PM sampling schedules (e.g., every 6th-day) employed in each study. Overall, the strong evidence for respiratory effects due to short-term O₃ exposure (Section 6.2) are consistent across disciplines and
provides coherence for the respiratory mortality associations observed across studies. The strong evidence for O$_3$-induced cardiovascular mortality is supported by controlled human exposure and animal toxicological studies that provide initial evidence for a biologically plausible mechanism for O$_3$-induced cardiovascular mortality. However, a lack of coherence with epidemiologic studies of cardiovascular morbidity that do not demonstrate consistent evidence of O$_3$-induced cardiovascular effects complicate the evidence for a biological pathway of events leading to mortality (Section 6.3).

In conclusion, the recent epidemiologic studies build upon and confirm the associations between short-term O$_3$ exposure and all-cause and cause-specific mortality reported in the 2006 O$_3$ AQCD. However, there is a lack of coherence across disciplines and consistency across health outcomes for O$_3$-induced cardiovascular morbidity (Section 6.3) which do not support the relatively strong epidemiologic evidence for O$_3$-related cardiovascular mortality. Overall, recent studies have provided additional information regarding key uncertainties (previously identified - including the potential confounding effects of copollutants and seasonal trend), individual- and community-level factors that may lead to increased risk of O$_3$-induced mortality and the heterogeneity in O$_3$-mortality risk estimates, and continued evidence of a linear no-threshold C-R relationship. Although some uncertainties still remain, the collective body of evidence is sufficient to conclude there is likely to be a causal relationship between short-term O$_3$ exposure and total mortality.

### 6.7 Overall Summary

The evidence reviewed in this chapter describes the recent findings regarding the health effects of short-term exposure to ambient O$_3$ concentrations. Table 6-54 provides an overview of the causal determinations for each of the health categories evaluated.

<table>
<thead>
<tr>
<th>Health Category</th>
<th>Causal Determination</th>
</tr>
</thead>
<tbody>
<tr>
<td>Respiratory Effects</td>
<td>Causal relationship</td>
</tr>
<tr>
<td>Cardiovascular Effects</td>
<td>Likely to be a causal relationship</td>
</tr>
<tr>
<td>Central Nervous System Effects</td>
<td>Suggestive of a causal relationship</td>
</tr>
<tr>
<td>Effects on Liver and Xenobiotic Metabolism</td>
<td>Inadequate to infer a causal relationship</td>
</tr>
<tr>
<td>Effects on Cutaneous and Ocular Tissues</td>
<td>Inadequate to infer a causal relationship</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>Likely to be a causal relationship</td>
</tr>
</tbody>
</table>
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7 INTEGRATED HEALTH EFFECTS OF LONG-TERM OZONE EXPOSURE

7.1 Introduction

This chapter reviews, summarizes, and integrates the evidence on relationships between health effects and long-term exposures to O₃. Both epidemiologic and toxicological studies provide a basis for examining long-term O₃ exposure health effects for respiratory effects, cardiovascular effects, reproductive and developmental effects, central nervous system effects, cancer outcomes, and mortality. Long-term exposure has been defined as a duration of approximately 30 days (1 month) or longer. However, in order to characterize the weight of evidence for the effects of O₃ on reproductive and developmental effects in a consistent, cohesive and integrated manner, results from both short-term and long-term exposure periods are included in that section, and are identified accordingly in the text and tables.

Conclusions from the 2006 O₃ AQCD (U.S. EPA, 2006b) are summarized briefly at the beginning of each section, and the evaluation of evidence from recent studies builds upon what was available during the previous review. For each health outcome (e.g., respiratory disease, lung function), results are summarized for studies from the specific scientific discipline, i.e., epidemiologic and toxicological studies. The major sections (i.e., respiratory, cardiovascular, mortality, reproductive/developmental, cancer) conclude with summaries of the evidence for the various health outcomes within that category and integration of the findings that lead to conclusions regarding causality based upon the framework described in the Preamble to this ISA. Determination of causality is made for the overall health effect category, such as respiratory effects, with coherence and plausibility being based on evidence from across disciplines and also across the suite of related health outcomes, including cause-specific mortality.

As mentioned in Chapter 2 (Section 2.3), epidemiologic studies generally present O₃-related effect estimates for mortality and morbidity health outcomes based on an incremental change in exposure. Studies traditionally present the relative risk per an incremental change equal to the interquartile range in O₃ concentrations or some other arbitrary value (e.g., 10 ppb). Additionally, various exposure metrics are used in O₃ epidemiologic studies, with the three most common being the maximum 1-h average within a 24-hour period (1-h max), the maximum 8-h average within a 24-hour period (8-h max), and 24-h average (24-h avg). For the purpose of presenting results from studies that use different exposure metrics, EPA consistently applies the same O₃ increments to facilitate comparisons between the results of various studies that may present results for different incremental changes. Differences due to the use of varying exposure metrics (e.g., 1-h max, 24-h avg)
become less apparent when averaged across longer exposure periods, because levels are typically lower and less variable. As such, throughout this chapter an increment of 10 ppb was consistently applied across studies, regardless of exposure metric, to facilitate comparisons between the results from these studies.

### 7.2 Respiratory Effects

Studies reviewed in the 2006 O₃ AQCD examined evidence for relationships between long-term O₃ exposure (several months to yearly) and effects on respiratory health outcomes including declines in lung function, increases in inflammation, and development of asthma in children and adults. Animal toxicology data provided a clearer picture indicating that long-term O₃ exposure may have lasting effects. Chronic exposure studies in animals have reported biochemical and morphological changes suggestive of irreversible long-term O₃ impacts on the lung. In contrast to supportive evidence from chronic animal studies, the epidemiologic studies on longer-term (annual) lung function declines, inflammation, and new asthma development remained inconclusive.

Several studies reviewed in the 2006 O₃ AQCD (Horak et al., 2002; Frischer et al., 1999) collectively indicated that O₃ exposure averaged over several summer months was associated with smaller increases in lung function growth in children. For longer averaging periods (annual), the definitive analysis in the Children’s Health Study (CHS) reported by Gauderman et al. (2004) provided little evidence that such long-term exposure to ambient O₃ was associated with significant deficits in the growth rate of lung function in children in contrast to the effects observed with other pollutants such as acid vapor, NO₂, and PM₂.₅. Limited epidemiologic research examined the relationship between long-term O₃ exposures and inflammation. Consistent with evidence of inflammation and allergic responses reported in experimental studies, an association between 30-day average O₃ and increased eosinophil levels was observed in an Austrian study (Frischer et al., 2001). The cross-sectional studies available for the 2006 O₃ AQCD detected no associations between long-term O₃ exposures and asthma prevalence, asthma-related symptoms or allergy to common aeroallergens in children after controlling for covariates. However, longitudinal studies provided evidence that long-term O₃ exposure influences the risk of asthma development in children (McConnell et al., 2002) and adults (McDonnell et al., 1999a; Greer et al., 1993).

New evidence presented below reports interactions between genetic variants and long-term O₃ exposure in effects on new onset asthma in U.S. cohorts in multi-community studies where protection by specific oxidant gene variants was restricted to children living in low O₃ communities. Related studies report coherent relationships between respiratory symptoms among asthmatics and long-term O₃ exposure. This evidence for respiratory effects associated with long-term O₃ exposure is supported by a large evidence base indicating associations of short-term exposure to O₃ with increases in respiratory symptoms and asthma medication use in
children with asthma (Section 6.2.4.1) and asthma hospitalizations in children (Section 6.2.7.2). A new line of evidence reports a positive concentration-response relationship between first asthma hospitalization and long-term O₃ exposure. Results from recent studies examining pulmonary function, inflammation, and allergic responses are also presented.

### 7.2.1 Asthma

#### 7.2.1.1 New Onset Asthma

Asthma is a heterogeneous disease with a high degree of temporal variability. Its progression and symptoms can vary within an individual’s experience over time. The course of asthma may vary markedly between young children, older children and adolescents, and adults. This variation is probably more dependent on age than on symptoms (NHLBI, 2007). Longitudinal cohort studies have examined associations between long-term O₃ exposures and the onset of asthma in adults and children (McConnell et al., 2002; McDonnell et al., 1999a; Greer et al., 1993), with results indicating a direct effect of long-term O₃ exposure on asthma risk in adults and effect modification by O₃ in children.

Associations between long-term O₃ exposure and new cases of asthma were reported in a cohort of nonsmoking adults in California (McDonnell et al., 1999a; Greer et al., 1993). The Adventist Health and Smog (AHSMOG) study cohort of 3,914 (age 27 to 87 years, 36% male) was drawn from nonsmoking, non-Hispanic white California Seventh Day Adventists, who were surveyed in 1977, 1987, and 1992. To be eligible, subjects had to have lived 10 or more years within 5 miles of their current residence in 1977. Residences from 1977 onward were followed and linked in time and space to interpolate concentrations of O₃, PM₂.₅, SO₄²⁻, SO₂, and NO₂. New asthma cases were defined as self-reported doctor-diagnosed asthma at either the 1987 or 1992 follow-up questionnaire among those who had not reported having asthma upon enrollment in 1977. During the 10-year follow-up (1977 to 1987), the incidence of new asthma was 2.1% for males and 2.2% for females (Greer et al., 1993). Ozone concentration data were not provided. A relative risk of 3.12 (95% CI: 1.16, 5.85) per 10-ppb increase in annual mean O₃ (exposure metric not stated) was observed in males, compared to a relative risk of 0.94 (95% CI: 0.65, 1.34) in females.

In the 15-year follow-up study (1977-1992), 3.2% of the eligible males and a slightly greater 4.3% of the eligible females developed adult asthma (McDonnell et al., 1999a). The mean 20-year average (1973-1992) for 8-h avg O₃ (9 a.m. to 5 p.m.) was 46.5 ppb (SD 15.3). For males, the relative risk of developing asthma was 1.31 (95% CI: 1.01, 1.71) per 10-ppb increase in 8-h avg O₃. Once again, there was no evidence of a positive association between O₃ and new-onset asthma in females (relative risk of 0.94 [95% CI: 0.87, 1.02]). The lack of an association does not necessarily indicate no effect of O₃ on the development of asthma among females.
For example, differences between females and males in time-activity patterns may influence relative exposures to ambient $O_3$. During summer 1992, the mean (SD) hours per week spent outdoors for male and female asthma cases were 13.8 (10.6) and 11.4 (10.9), respectively, indicating potential greater misclassification of exposure in females. None of the other pollutants ($PM_{10}$, $SO_4^{2-}$, $SO_2$, and $NO_2$) were associated with development of asthma in either males or females. Adjusting for copollutants did not diminish the association between $O_3$ and asthma incidence for males. In no case was the $O_3$ coefficient reduced by more than 10% in the two-pollutant models compared to the model containing $O_3$ alone. The consistency of the results in the two studies with different follow-up times, as well as the independent and robust association between annual mean $O_3$ concentrations and asthma incidence, provide supportive evidence that long-term $O_3$ exposure may be associated with the development of asthma in adult males. However, because the AHSMOG cohort was drawn from a narrow subject definition, the representativeness of this cohort to the general U.S. population may be limited.

In children, the relationship between long-term $O_3$ exposure and new onset asthma has been extensively investigated in the CHS. In this cohort, evidence provides stronger support for long-term $O_3$ exposure modifying the risk of new onset asthma associated with other potential risk factors than having a main effect on new onset asthma. Initiated in the early 1990s, the CHS was originally designed to examine whether long-term exposure to ambient pollutants was related to chronic respiratory outcomes in children in 12 communities in southern California (Peters et al., 1999b; Peters et al., 1999a). New-onset asthma was classified as having no prior history of asthma at study entry with subsequent report of physician-diagnosed asthma at follow-up with the date of onset assigned to be the midpoint of the interval between the interview date when asthma diagnosis was first reported and the previous interview date. In a cohort recruited during 2002-2003 and followed for three years beginning in kindergarten or first grade, McConnell et al. (2010) reported a hazard ratio for new onset asthma of 0.76 (95% CI: 0.38, 1.54) comparing the communities with the highest (59.8 ppb) and lowest (29.5 ppb) annual average of 8-h avg (10 a.m.-6 p.m.) $O_3$. With adjustment for school and residential modeled non-freeway traffic-related exposure, the estimated HR for $O_3$ was 1.01 (95% CI: 0.49, 2.11).

Similarly in a cohort recruited in 1993, asthma risk was not higher for residents of the six high-$O_3$ communities versus residents of the six low-$O_3$ communities (McConnell et al., 2002). In this study, 3,535 initially nonasthmatic children (ages 9 to 16 years at enrollment) were followed for up to 5 years, during which 265 cases of new-onset asthma were identified. Communities were stratified by 4-year average 1-h max $O_3$ levels, with six high-$O_3$ communities (mean 75.4 ppb [SD 6.8]) and six low-$O_3$ communities (mean 50.1 ppb [SD 11.0]). Within the high-$O_3$ communities, asthma risk was 3.3 (95% CI: 1.9, 5.8) times greater for children who played three or more sports as compared with children who played no sports. None of the children who lived in high-$O_3$ communities and played three or more sports had a family history of asthma. In models with individual sports entered as dummy variables, only tennis was significantly associated with asthma and only in the high $O_3$ communities. This association was absent in the low-$O_3$ communities (relative risk of 0.8 [95% CI:...
The overall observed pattern of effects of sports participation on asthma risk was robust to adjustment for SES, history of allergy, family history of asthma, insurance, maternal smoking, and BMI.

Analyses aimed at distinguishing the effects of O₃ from effects of other pollutants indicated that in communities with high O₃ and low levels of other pollutants there was a 4.2-fold (95% CI: 1.6, 10.7) increased risk of asthma in children playing three or more sports, compared to children who played no sports. The relative risk in children playing three or more sports was slightly lower (3.3 [95% CI: 1.6, 6.9]) in communities with a combination of high levels of O₃ and other pollutants. Ozone concentrations were not strongly correlated with PM₁₀, PM₂.₅, NO₂, or inorganic acid vapors, and no associations with asthma were found for these other pollutants. These results provide additional support that the effects of physical activity on asthma are modified by long-term O₃ exposure. Overall, the results from McConnell et al. (2002) suggest that playing sports may indicate greater outdoor activity when O₃ levels are higher and an increased ventilation rate, which may lead to increased O₃ exposure. It should be noted, however, that these findings were based on a small number of new asthma cases (n = 29 among children who played three or more sports) and were not based on a priori hypotheses.

Recent studies from the CHS provide evidence for gene-environment interactions in effects on new-onset asthma by indicating that the lower risks associated with specific genetic variants are found in children who live in lower O₃ communities (Islam et al., 2009; Islam et al., 2008; Oryszczyn et al., 2007; Lee et al., 2004b; Gilliland et al., 2002). Risk for new-onset asthma is related in part to genetic susceptibility, behavioral factors and environmental exposure (Gilliland et al., 1999). Gene-environment interactions in asthma have been well discussed in the literature (von Mutius, 2009; Holgate et al., 2007; Martinez, 2007a, b; Rahman et al., 2006; Hoffjan et al., 2005; Kleeberger and Peden, 2005; Ober, 2005). Complex chronic diseases, such as asthma, are partially the result of a sequence of biochemical reactions involving exposures to various environmental agents metabolized by a number of different genes (Conti et al., 2003). Oxidative stress has been proposed to underlie these mechanistic hypotheses (Gilliland et al., 1999). Genetic variants may impact disease risk directly or modify disease risk by affecting internal dose of pollutants and other environmental agents and/or their reaction products or by altering cellular and molecular modes of action. Understanding the relation between genetic polymorphisms and environmental exposure can help identify high-risk subgroups in the population and provide better insight into pathway mechanisms for these complex diseases.

CHS analyses have found that asthma risk is related to interactions between O₃ and variants in genes for enzymes such as heme-oxygenase (HO-1), arginases (ARG1 and 2), and glutathione S transferase P1 (GSTP1) (Himes et al., 2009; Islam et al., 2008; Li et al., 2008; Hanene et al., 2007; Ercan et al., 2006; Li et al., 2006a; Tamer et al., 2004; Gilliland et al., 2002). Biological plausibility for these findings is provided by evidence that these enzymes have antioxidant and/or anti-inflammatory activity and participate in well recognized modes of action in asthma pathogenesis.
Further, several lines of evidence demonstrate that secondary oxidation products of \( \text{O}_3 \) initiate the key modes of action that mediate downstream health effects (Section 5.3.2). For example, \( \text{HO}-1 \) has been found to respond rapidly to oxidants, have anti-inflammatory and anti-oxidant effects (Exner et al., 2004), relax airway smooth muscle, and be induced in airways during asthma (Carter et al., 2004). The GSTP1 Val/Val genotype has been associated with increased risk of having atopic asthma (Tamer et al., 2004). Gene-environment interactions are discussed in greater detail in Section 5.4.2.1.

Islam et al. (2008) found that functional polymorphisms of the heme oxygenase-1 gene (HMOX-1, [(GT)n repeat]) influenced the risk of new-onset asthma, depending on ethnicity and long-term community \( \text{O}_3 \) concentrations. Ozone-gene interactions were not found for variants in other antioxidant genes: catalase (CAT [-262C >T -844C >T0]) or and manganese superoxide dismutase (MNSOD, [Ala-9Val]). Analyses were restricted to children of Hispanic (\( n = 576 \)) or non-Hispanic white ethnicity (\( n = 1,125 \)) and were conducted with long-term pollutant levels averaged from 1994 to 2003. The effect of ambient air pollution on the relationship between genetic polymorphism and new-onset asthma was assessed using Cox proportional hazard regression models where the community specific average air pollution levels were fitted as a continuous variable together with the appropriate interaction terms for genes and air pollutants and a random effect of community (Berhane et al., 2004).

Over the follow-up period, 160 new cases of asthma were diagnosed (Islam et al., 2008). For HMOX-1, the interaction (\( p = 0.003 \)) indicated a greater protective effect of the S-allele (short, <23 (GT)n repeats) compared to the L-allele (long, >23 repeats) among non-Hispanic white children who lived in the low \( \text{O}_3 \) community (nonparallelism presented in Figure 7-1). Among children residing in low-\( \text{O}_3 \) communities, the hazard ratio (HR) of new onset asthma associated with the S-allele was 0.44 (95% CI: 0.23, 0.83) compared to non-Hispanic white children who lived in low \( \text{O}_3 \) communities and had no S-alleles. Biological plausibility for these results is provided by evidence that the S-allele variant of HMOX-1 is more readily induced than those with more numerous repeats. The S-allele was found to have a less protective effect in non-Hispanic white children who resided in high \( \text{O}_3 \) communities (HR = 0.88; [95% CI: 0.33, 2.34] compared to non-Hispanic white children in low \( \text{O}_3 \) communities with no S-allele). Because HMOX-1 variants were not associated with asthma risk in Hispanic children, effect modification by \( \text{O}_3 \) was not investigated. No significant interactions were observed between \( \text{PM}_{10} \) or other pollutants and the HMOX-1 gene; quantitative results were not presented. Average \( \text{O}_3 \) levels showed low correlation with the other monitored pollutants. The authors did not consider the lack of adjustment for multiple testing to be a concern in this analysis because the selection of the genes was based on a priori hypotheses defined by a well-studied biological pathway, in which oxidative stress serves as the link among \( \text{O}_3 \) exposure, enzyme activity, and asthma.

Collectively, results from Islam et al. (2008) indicate that a variant in HMOX-1 that produces a more readily inducible enzyme is associated with lower risk of new-onset asthma in children who live in low \( \text{O}_3 \) communities. Results were not presented for
the main effects relating new-onset asthma to O$_3$ exposure. However, they do indicate that in environments of low ambient O$_3$, enzymes with greater antioxidative activity may have the capacity to counter any temporary imbalance in an oxidant-antioxidant relationship. However, in the presence of high background O$_3$, the protective effect may be attenuated because with higher exposure to oxidants, the antioxidant genes may be at their maximal level of inducibility, and variation in promoters no longer affects levels of expression. Supporting evidence is provided by Schroer et al. (2009), who found that infants with multiple environmental exposures were at increased risk of wheeze regardless of variant in GSTP1, which encodes a gene with antioxidant activity.

Note: An interaction p-value of 0.003 was obtained from the hierarchical two stage Cox proportional hazard model fitting the community specific O$_3$ and controlling for random effect of the communities. The interaction indicates there is a greater protective effect of having a heme-oxygenase S-allele compared to having the L-allele among children living in communities with lower long-term ambient O$_3$ concentrations. The HRs are off-set as opposed to overlapping in the figure to allow clearer presentation of the results.

Source: Developed by EPA with data from Islam et al. (2008) (data used with permission of American Thoracic Society).

**Figure 7-1** Interaction of heme-oxygenase genetic variants and O$_3$ level on the Hazard Ratio (HR) of new-onset asthma in the 12 Children’s Health Study communities.
Expanding on the results of McConnell et al. (2002), Islam et al. (2009) provided evidence that variants in GSTM1 and GSTP1 may influence associations between outdoor exercise and new onset asthma. A primary conclusion that the authors (Islam et al., 2009) reported was that the GSTP1 Ile/Ile and GSTM1 null genotypes increased risk of new onset asthma during adolescence. The highest risk was found for participation in three or more team sports (compared to no sports) among children with GSTP1 Ile/Ile genotype living in high-O3 communities (HR: 6.15, [95% CI: 2.2, 7.4]). No three-way interaction was found for GSTM1. These results demonstrate the potential importance of a combination of genetic variability, O3 exposure, and outdoor activity on asthma risk. It is important to note that while some studies have found a modifying role of air pollution on the association between GSTP1 Ile/Ile and asthma in children (Lee et al., 2004b), others have found that the GSTP1 Val/Val variant to be associated with greater asthma prevalence and increased risk of O3-associated respiratory morbidity (see discussion in Section 6.2.4.1).

The CHS also provided evidence of interactions between O3 exposure and variants in genes for arginase (Salam et al., 2009). Arginase catalyzes the conversion of L-arginine. Because L-arginine is a precursor of NO, higher arginase activity can limit production of NO and subsequent nitrosative stress. Epidemiologic evidence of associations of arginase variants with asthma are limited (Li et al., 2006a); however, asthmatic subjects have been found to have higher arginase activity than non-asthmatic subjects (Morris et al., 2004). The modifying effect of O3 and atopy on the association between ARG1 and ARG2 haplotypes and asthma were evaluated using likelihood ratio tests with appropriate interaction terms. Having more copies of the ARG1h4 haplotype (compared to having zero copies) was associated with lower odds of asthma, particularly among children with atopic asthma living in high O3 communities (OR: 0.12; [95% CI: 0.04, 0.43]). Having more copies of the ARG2h3 haplotype (compared to having zero copies) was associated with increased risk of childhood-onset asthma among children in both low and high O3 communities. The implications of findings are somewhat limited because the functional relevance of the ARG1 and ARG2 variants is not clear.

### 7.2.1.2 Prevalence of Asthma and Asthma Symptoms

Some cross-sectional studies reviewed in the 2006 O3 AQCD observed positive relationships between chronic exposure to O3 and prevalence of asthma and asthmatic symptoms in school children (Ramadour et al., 2000; Wang et al., 1999) while others (Kuo et al., 2002; Charpin et al., 1999) did not. Recent studies provide additional evidence.

In a cross-sectional nationwide study of 32,672 Taiwanese school children, Hwang et al. (2005) assessed the effects of air pollutants on the risk of asthma. The study population was recruited from elementary and middle schools within 1 km of air monitoring stations. The risk of asthma was related to O3 in the one-pollutant model. The addition of other pollutants (NOX, CO, SO2, and PM10), in two-pollutant and
three-pollutant models, increased the O₃ risk estimates. The prevalence of childhood asthma was assessed in Portugal by contrasting the risk of asthma between a high O₃ rural area and an area with low O₃ levels (Sousa et al., 2011; Sousa et al., 2009; Sousa et al., 2008). The locations were selected to provide a difference in O₃ levels without the confounding effects of other pollutants. Both evaluation for asthma symptoms and FEV₁ suggested that O₃ increased asthma prevalence. Clark et al. (2010) investigated the effect of exposure to ambient air pollution in utero and during the first year of life on risk of subsequent incidence asthma diagnosis up to 3-4 years of age in a population-based nested case-control study for all children born in southwestern British Columbia in 1999 and 2000 (n = 37,401; including 3,482 [9.3%] with asthma). Air pollution exposure for each subject was estimated based on their residential address history using regulatory monitoring data, land use regression modeling, and proximity to stationary pollutant sources. Daily values from the three closest monitors within 50 km were used to calculate exposures. Traffic-related pollutants were associated with the highest risk. Ozone was inversely correlated with the primary traffic-related pollutants (r = -0.7 to -0.9). The low reliability of asthma diagnosis in infants makes this study difficult to interpret (Martinez et al., 1995). In a cross-sectional analysis, Akinbami et al. (2010) examined the association between chronic exposure to outdoor pollutants (12-month avg levels by county) and asthma outcomes in a national sample of children ages 3-17 years living in U.S. metropolitan areas (National Health Interview Survey, N = 34,073). A 5-ppb increase in estimated 8-h max O₃ concentration (annual average) yielded a positive association for both currently having asthma and for having at least 1 asthma attack in the previous year, while the adjusted odds ratios for other pollutants were not statistically significant. Models in which pollutant value ranges were divided into quartiles produced comparable results. Multipollutant models (SO₂ and PM) produced similar results. The median value for 12-month avg O₃ levels was 39.5 ppb and the IQR was 35.9-43.7 ppb. The adjusted odds for current asthma for the highest quartile (49.9-59.5 ppb) of estimated O₃ exposure was 1.56 (95% CI: 1.15, 2.10) with a positive concentration-response relationship apparent from the lowest quartile to the highest. Thus, this cross-sectional analysis and Hwang et al. (2005) provided further evidence relating O₃ exposure and the risk of asthma.

Relationships between long-term exposure and respiratory symptoms in asthmatic children also were examined in the CHS. McConnell et al. (1999) examined the association between O₃ levels and the prevalence of chronic lower respiratory tract symptoms in 3,676 cohort children with asthma. In this cross-sectional study, bronchitis, phlegm, and cough were not associated with annual mean 1-h max O₃ concentrations in children with asthma or wheeze. All other pollutants examined (PM₁₀, PM₂.₅, NO₂, and gaseous acid) were associated with an increase in phlegm but not cough. The mean annual average 1-h max O₃ concentration was 65.6 ppb (range 35.5 to 97.5) across the 12 communities. In another CHS analysis, McConnell et al. (2003) evaluated relationships between air pollutants and bronchitic symptoms among 475 children with asthma. The mean 4-year average 8-h avg O₃ (10 a.m.-6 p.m.) concentration was 47.2 ppb (range 28.3 to 65.8) across the 12 communities. For a 10-ppb increase in 8-h avg O₃ averaged over 4 years, the between-community odds ratio was 0.90 (95% CI: 0.82, 1.00) whereas the within-community
(i.e., difference between one- and four-year average) odds ratio was larger, i.e., 1.79 (95% CI: 1.00, 3.21). The authors commented that if the larger within-community effect estimates were correct, then other cross-sectional (between-community) studies might have underestimated the true effect of air pollution on bronchitic symptoms in children. These differences might be attributable to confounding by poorly measured or unmeasured risk factors that vary between communities. Within community effects may more accurately represent risk associated with pollutant exposure because the analyses characterize health effects associated with changing pollutant concentrations within a community, thereby minimizing potential confounding by factors that are constant over time within a community. PM$_{2.5}$, NO$_2$, and organic carbon also were associated with bronchitic symptoms. In two-pollutant models, the within-community effect estimates for O$_3$ were markedly reduced and no longer statistically significant in some cases.

CHS also examined interactions between TNF-α 308 genotype and long-term O$_3$ exposure in the occurrence of bronchitic symptoms among children with asthma (Lee et al., 2009b). Increased airway levels of the cytokine TNF-α has been related to inflammation, and the GG genotype has been linked to lower expression of TNF-α. Asthmatic children with the GG genotype had a lower prevalence of bronchitic symptoms compared with children carrying at least one A-allele (e.g., GA or AA genotype). Low-versus high-O$_3$ strata were defined as less than or greater than 50-ppb O$_3$ avg. Asthmatic children with TNF-308 GG genotype had a significantly reduced risk of bronchitic symptoms with low-O$_3$ exposure (OR: 0.53 [95% CI: 0.31, 0.91]). The risk was not reduced in children living in high-O$_3$ communities (OR: 1.42 [95% CI: 0.75, 2.70]). The difference in genotypic effects between low- and high-O$_3$ environments was statistically significant among asthmatics (P for interaction = 0.01), but not significant among non-asthmatic children. Figure 7-2 presents adjusted O$_3$ community-specific regression coefficients plotted against ambient O$_3$ concentration, using weights proportional to the inverse variance. Investigators further reported no substantial differences in the effect of the GG genotype on bronchitic symptoms by long-term exposure to PM$_{10}$, PM$_{2.5}$, NO$_2$, acid vapor, or second-hand smoke.
Using indicator variables for each category of genotype and O₃ exposure, investigators calculated effect estimates for TNF-α GG genotype on the occurrence of bronchitic symptoms among children with asthma.

Note: Reprinted with permission of John Wiley & Sons, (Lee et al., 2009b).

**Figure 7-2** Ozone modifies the effect of TNF GG genotype on bronchitic symptoms among children with asthma in the CHS.

Another CHS analyses reported interrelationships between variants in CAT and myeloperoxidase (MPO) genes, ambient pollutants, and respiratory-related school absences for 1,136 Hispanic and non-Hispanic white cohort children (Wenten et al., 2009). A related study (Gilliland et al., 2001), found increased O₃ exposure to be related to greater school absenteeism due to respiratory illness but did not consider genetic variants. Wenten et al. (2009) hypothesized that variation in the level or function of antioxidant enzymes would modulate respiratory illness risk, especially under high levels of oxidative stress expected from high ambient O₃ exposure. The joint effect of variants in these two genes (genetic epistasis) on respiratory illness was examined because the enzyme products operate on the same substrate within the same biological pathway. Risk of respiratory-related school absences was elevated for children with CAT GG plus MPO GA or AA genotypes (RR: 1.35 [95% CI: 1.03, 1.77] compared to GG for both genes) and reduced for children with CAT GA or AA plus MPO GA or AA (RR: 0.81 [95% CI: 0.55, 1.19] compared to GG for both genes). Both CAT GG and MPO GA (or AA) genotypes produced a less activity enzyme. In analyses that stratified communities into high and low O₃ exposure groups by median levels (46.9 ppb), the protective effect of CAT GA or AA plus MPO GA or AA genotype was largely limited to children living in communities with high ambient O₃ levels (RR: 0.42 [95% CI: 0.20, 0.89]).

The association of respiratory-illness absences with functional variants in CAT and
MPO that differed by air pollution levels illustrates the need to consider genetic epistasis in assessing gene-environment interactions.

Collective evidence from CHS provides an important demonstration of gene-environment interactions. In the complex gene-environment setting a modifying effect might not be reflected in an exposure main effect. The simultaneous occurrence of main effect and interaction effect can occur. The study of gene-environment interactions helps to dissect disease mechanisms in humans by using information on susceptibility genes to focus on the biological pathways that are most relevant to that disease (Hunter, 2005).

The French Epidemiology study on Genetics and Environment of Asthma (EGEA) investigated the relationship between ambient air pollution and asthma severity in a cohort in five French cities (Paris, Lyon, Marseille, Montpellier, and Grenoble) (Rage et al., 2009b). In this cross-sectional study, asthma severity over the past 12 months was assessed among 328 adult asthmatics using two methods: (1) a four-class severity score that integrated clinical events and type of treatment; and (2) a five-level asthma score based only on symptoms. Two measures of exposure were also assessed: (1) [first method] closest monitor data from 1991 to 1995 where a total of 93% of the subjects lived within 10 km of a monitor, but where 70% of the O₃ concentrations were back-extrapolated values; and (2) [second method] a validated spatial model that used geostatistical interpolations and then assigned air pollutants to the geocoded residential addresses of all participants and individually assigned exposure to ambient air pollution estimates. Higher asthma severity scores were significantly related to both the 8-h avg O₃ during April-September and the number of days with 8-h O₃ averages above 55 ppb. Both exposure assessment methods and severity score methods resulted in very similar findings. Effect estimates of O₃ were similar in three-pollutant models. No PM data were available. Since these estimates were not sensitive to the inclusion of ambient NO₂ in the three-pollutant models, the authors viewed the findings not to be explained by particles which usually have substantial correlations between PM and NO₂. Effect estimates for O₃ in three-pollutant models including O₃, SO₂, and NO₂ yielded OR for O₃-days of 2.74 (95% CI: 1.68, 4.48) per IQR days of 10-28 (+18) ppb. The effect estimates for SO₂ and NO₂ in the three-pollutant model were 1.33 (95% CI: 0.85, 2.11) and 0.94 (95% CI: 0.68, 1.29), respectively. Taking into account duration of residence did not change the result. This study suggests that a higher asthma severity score is related to long-term O₃ exposure.

An EGEA follow-up study (Jacquemin et al., 2012), examines the relationship between asthma and O₃, NO₂, and PM₁₀. New aspects considered include: (1) examination of three domains of asthma control (symptoms, exacerbations, and lung function); (2) levels of asthma control (controlled, partially controlled, and uncontrolled asthma); and (3) PM₁₀ and multipollutant analysis. In this cross-sectional analysis, EGEA2 studied 481 adult subjects with current asthma from 2003 to 2007. The IQRs were 11 (41-52) µg/m³ for annual O₃ and 13 (25-38) µg/m³ for summer (April-September) O₃. The association between asthma control and air pollutants was expressed by ORs (reported for one IQR of the pollutant), derived
from multinomial logistic regression. For each factor, the simultaneous assessment of
the risk for uncontrolled asthma and for partly controlled asthma was compared with
controlled asthma using a composite of the three domains. In crude and adjusted
models, O₃-sum and PM₁₀ were positively associated with partly controlled and
uncontrolled asthma, with a clear gradient from controlled, partly controlled
(OR = 1.53, 95% CI: 1.01, 2.33) and uncontrolled (OR = 2.14, 95% CI: 1.34, 3.43)
(from the multinomial logistic regression).

Separately, they used a composite asthma control classification that used the ordinal
logistic regression for risk comparing controlled to partly controlled asthma and
comparing partly controlled to uncontrolled asthma. For these two pollutants, the
ORs assessed using the ordinal logistic regression were significant (ORs were 1.69
(95% CI: 1.22, 2.34) and 1.35 (95% CI: 1.13, 1.64) for O₃-sum and PM₁₀,
respectively). For two pollutant models using the ordinal logistic regression, the
adjusted ORs for O₃-sum and PM₁₀ included simultaneously in a unique model were
1.50 (95% CI: 1.07, 2.11) for O₃-sum and 1.28 (95% CI: 1.06, 1.55) for PM₁₀,
respectively. This result suggests that the effects of both pollutants are independent.

The analysis of the associations between air pollution for all asthma subjects and
each one of the three asthma control domains showed the following: (1) for lung
function defined dichotomously as percent predicted FEV₁ value <=80
(OR = 1.35, 95% CI: 0.80, 2.28 for adjusted O₃-sum); (2) for symptoms defined as
asthma attacks or dyspnea or woken by asthma attack or shortness of breath in the
past three months (OR = 1.59, 95% CI: 1.10, 2.30 for adjusted O₃-sum); and (3) for
exacerbations defined at least one hospitalizations or ER visits in the last year or oral
corticosteroids in the past three months (OR = 1.58, 95% CI: 0.97, 2.59
for adjusted O₃-sum). Since the estimates for both pollutants were more stable and significant
when using the integrated measure of asthma control, this indicates that the results
are not driven by one domain. These results support an effect of long-term exposure
to O₃ on asthma control in adulthood in subjects with pre-existing asthma.

Goss et al. (2004) investigated the effect of O₃ on pulmonary exacerbations and lung
function in individuals over the age of 6 years with cystic fibrosis (n = 11,484).
The study included patients enrolled in the Cystic Fibrosis Foundation National
Patient Registry, which contains demographic and clinical data collected annually at
accredited centers for cystic fibrosis. For 1999 through 2000, the annual mean O₃
concentration, calculated from 1-h averages from 616 monitors in the U.S. EPA
Aerometric Information Retrieval System (AIRS), was 51.0 ppb (SD 7.3). Exposure
was assessed by linking air pollution values from AIRS with the patient’s home ZIP
code. No clear association was found between annual mean O₃ and lung function
parameters. However, a 10 ppb increase in annual mean O₃ was associated with a
10% (95% CI: 3, 17) increase in the odds of two or more pulmonary exacerbations.
Significant excess odds of pulmonary exacerbations also were observed with
increased annual mean PM₁₀ and PM₂.₅ concentrations. The O₃ effect was robust to
adjustment for PM₁₀ and PM₂.₅, 8% (95% CI: 1, 15) increase in odds of two or more
pulmonary exacerbations per 10 ppb increase in annual mean O₃.
7.2.2 Asthma Hospital Admissions and ED Visits

The studies on O\textsubscript{3}-related hospital discharges and emergency department (ED) visits for asthma and respiratory disease that were available in the 2006 O\textsubscript{3} AQCD mainly looked at the daily time metric. Collectively the short-term O\textsubscript{3} studies presented earlier in Section 6.2.7.5 indicate that there is evidence for increases in both hospital admissions and ED visits related to all respiratory outcomes including asthma with stronger associations in the warm months. New studies evaluated long-term O\textsubscript{3} exposure metrics, providing a new line of evidence that suggests a positive exposure-response relationship between first asthma hospital admission and long-term O\textsubscript{3} exposure.

An ecologic study (Moore et al., 2008) evaluated time trends in associations between declining warm-season O\textsubscript{3} concentrations and hospitalization for asthma in children in California’s South Coast Air Basin who ranged in age from birth to 19 years. Quarterly average concentrations from 195 spatial grids, 10×10 km, were used. Ozone was the only pollutant associated with increased hospital admissions over the study period. A linear relation was observed for asthma hospital discharges (Moore et al., 2008). A matched case-control study (Karr et al., 2007) was conducted on infant bronchiolitis (ICD 9, code 466.1) hospitalization and two measures of long-term pollutant exposure (the month prior to hospitalization and the lifetime average) for O\textsubscript{3} in the South Coast Air Basin of southern California among 18,595 infants born between 1995 and 2000. Ozone was associated with reduced risk in the single-pollutant model, but this relation did not persist in multipollutant models (CO, NO\textsubscript{2}, and PM\textsubscript{2.5}).

In a cross-sectional study, Meng et al. (2010) examined associations between air pollution and asthma morbidity in the San Joaquin Valley in California by using the 2001 California Health Interview Survey data from subjects ages 1 to 65+ who reported physician-diagnosed asthma (n = 1,502). Subjects were assigned annual average concentrations for O\textsubscript{3} based on residential ZIP code and the closet air monitoring station within 8 km but did not have data on duration of residence. Multipollutant models for O\textsubscript{3} and PM did not differ substantially from single-pollutant estimates, indicating that pollutant multi-collinearity is not a problem in these analyses. The authors reported increased asthma-related ED visits or hospitalizations for O\textsubscript{3} (OR = 1.49; [95\% CI: 1.05, 2.11] per 10 ppb) for all ages. Positive associations were obtained for symptoms, but 95\% confidence intervals included null values. Associations for symptoms for adults (ages 18+) were observed (OR = 1.40; [95\% CI: 1.02, 1.91] per 10 ppb).

Associations between air pollution and poorly controlled asthma among adults in Los Angeles and San Diego Counties were investigated using the California Health Interview Survey data collected between November 2000 and September 2001 (Meng et al., 2007). Each respondent was assigned an annual average concentration measured at the nearest station within 5 miles of the residential cross-street intersection. Poorly controlled asthma was defined as having daily or weekly asthma symptoms or at least one ED visit or hospitalization because of asthma during the
past 12 months. This cross-sectional study reports an OR of 3.34 (95% CI: 1.01, 11.09) for poorly controlled asthma when comparing those 65 years of age and older above the 90th percentile (28.7 ppb) level to those below that level. Copollutant (PM) analysis produced similar results.

Evidence associating long-term O₃ exposure to first asthma hospital admission in a concentration-response relationship is provided in a retrospective cohort study (Lin et al., 2008b). This study investigated the association between chronic exposure to O₃ and childhood asthma admissions (defined as a principal diagnosis of ICD9, code 493) by following a birth cohort of 1,204,396 eligible births born in New York State during 1995-1999 to first asthma admission or until 31 December 2000. There were 10,429 (0.87%) children admitted to the hospital for asthma between 1 and 6 years of age. The asthma hospitalization rate in New York State in 1993 was 2.87 per 1,000 (Lin et al., 1999). Three annual indicators (all 8-h max from 10:00 a.m. to 6:00 p.m.) were used to define chronic O₃ exposure: (1) mean concentration during the follow-up period (41.06 ppb); (2) mean concentration during the O₃ season (50.62 ppb); and (3) proportion of follow-up days with O₃ levels >70 ppb. In this study the authors aimed to predict the risk of having asthma admissions in a birth cohort, but the time to the first admission in children that is usually analyzed in survival models was not their primary interest. The effects of copollutants were assessed and controlled for using the Air Quality Index (AQI). Interaction terms were used to assess potential effect modifications. A positive association between chronic exposure to O₃ and childhood asthma hospital admissions was observed indicating that children exposed to high O₃ levels over time are more likely to develop asthma severe enough to be admitted to the hospital. The various factors were examined and differences were found for younger children (1-2 years), poor neighborhoods, Medicaid/self-paid births, geographic region and others. As shown in Figure 7-3, positive concentration-response relationships were observed. Asthma admissions were significantly associated with increased O₃ levels for all chronic exposure indicators (ORs, 1.16-1.68). When estimating the O₃ effect using the exceedance proportion, an increase was observed (OR = 1.68; [95% CI: 1.64, 1.73]) in hospital admissions with an IQR (2.51%) increase in O₃. A proportional hazards model for the New York City data was run as a sensitivity analysis and it yielded similar results between asthma admissions and chronic exposure to O₃ (Cox model: HR = 1.14, [95% CI: 1.124, 1.155] is similar to logistic model results: OR = 1.16 [95% CI: 1.15, 1.17]) (Lin, 2010). Thus, this study provides evidence associating long-term O₃ exposure to first asthma hospital admission in a concentration-response relationship.

In considering relationships between long-term pollutant exposure and chronic disease health endpoints, Künzli (2012) offers two hypotheses relevant to research on air pollution and chronic disease where chronic pathologies are found with acute expressions of the chronic disease: “H1: Exposure provides a basis for the development of the underlying chronic pathology, which increases the pool of people with chronic conditions prone to exacerbations; H2: Exposure triggers an acute event (or a state of frailty that results in an event with a delay of a few days or weeks) among those with the disease.” Künzli (2012) states if associations of pollution with events are much larger in the long-term studies, it provides some indirect evidence in
support of H1. If air pollution increases the pool of subjects with the chronic pathology (H1), more acute events are expected to be seen for higher exposures since events due to various causes are part of the chronic disease pathway.

Künzli (2012) makes such a comparison noting larger associations with long-term NO₂ exposures for adult asthma hospital admissions (Andersen et al., 2012) as compared to short-term NO₂ exposures for asthma hospital admissions (Peel et al., 2005). In a further example, Pope (2007) makes similar conclusions comparing long-term PM mortality study results to short-term PM mortality studies. The results of Lin et al. (2008b) for first asthma hospital admission, presented below, show effect estimates that are larger than those reported in a study of asthma hospital admissions in New York State by Silverman and Ito (2010), discussed in Chapter 6 (both studies are for young children). This provides some support for the hypothesis that O₃ exposure may not only have triggered the events but also increased the pool of asthmatics. However, caution is warranted in attributing associations in the Lin et al. (2008b) study to long-term exposures since there is potential for short-term exposures to contribute to the observed associations.

Note: Adjusted for child’s sex, age, birth weight, and gestational age; maternal race, ethnicity, age, education, insurance, and smoking status during pregnancy; and regional poverty level and temperature. ORs by low, medium, and high exposure are shown for New York City (NYC: low [37.3 ppb], medium [37.3-38.11 ppb], high [38.11+ ppb] and other New York State regions (Other NYS regions: low [42.58 ppb], medium [42.58-45.06 ppb], high [45.06+ ppb]) for first asthma hospital admission. Source: Lin (2010); Lin et al. (2008b)

**Figure 7-3** Ozone-asthma concentration-response relationship using the mean concentration during the entire follow-up period for first asthma hospital admission.
7.2.3 Pulmonary Structure and Function

7.2.3.1 Pulmonary Structure and Function: Evidence from Epidemiology Studies

The definitive 8-year follow-up analysis of the first cohort of the CHS, which is discussed in Section 7.2 (Gauderman et al., 2004), provided little evidence that long-term exposure to ambient O\textsubscript{3} was associated with significant deficits in the growth rate of lung function in children. A later CHS study (Islam et al., 2007) examined relationships between air pollution, lung function, and new onset asthma and reported no substantial differences in the effect of O\textsubscript{3} on lung function. Ozone concentrations from the least to most polluted communities (mean annual average of 8-h avg O\textsubscript{3}) ranged from 30 to 65 ppb, as compared to the ranges observed for the other pollutants, which had 4-fold- to 8-fold differences in concentrations. In a more recent CHS study, Breton et al. (2011) hypothesized that genetic variation in genes on the glutathione metabolic pathway may influence the association between ambient air pollutant exposures and lung function growth in children. They investigated whether genetic variation in glutathione genes GSS, GSR, GCLC, and GCLM was associated with lung function growth in healthy children using data collected on 2,106 children over an 8-year time-period as part of the Children’s Health Study. Breton et al. (2011) found that variation in the GSS locus was associated with differences in risk of children for lung function growth deficits associated with NO\textsubscript{2}, PM\textsubscript{10}, PM\textsubscript{2.5}, elemental carbon, organic carbon, and O\textsubscript{3}. The negative effects of air pollutants were largely observed within participants who had a particular GSS haplotype. The effects ranged from -124.2 to -149.1 mL for FEV\textsubscript{1}, -92.9 to -126.7 mL for FVC and -277.9 mL/sec for MMEF for all pollutants except O\textsubscript{3}, for which some positive associations were reported: 25.9 mL for FEV\textsubscript{1}; 0.1 mL for FVC, and 166.5 mL/sec for MMEF. Ozone was associated with larger decreases in lung function in children without this haplotype, when compared to the other pollutants with values of -76.6 mL for FEV\textsubscript{1}, -17.2 mL for FVC, and -200.3 mL/sec for MMEF, but only the association with MMEF was statistically significant.

As discussed in the 2006 O\textsubscript{3} AQCD, a study of freshman students at the University of California, Berkeley reported an interaction between lifetime exposure to O\textsubscript{3} and baseline FEF\textsubscript{25-75}/FVC ratio, a measure of intrinsic airway size for decreased measures of small airways (<2 mm) function (FEF\textsubscript{75} and FEF\textsubscript{25-75}) (Tager et al., 2005). Subjects with a small ratio (indicating an increased airway size relative to their lung volume) had decreases in FEF\textsubscript{75} and FEF\textsubscript{25-75} for increases in lifetime exposure to O\textsubscript{3}. Kinney and Lippmann (2000) examined 72 nonsmoking adults (mean age 20 years) from the second-year class of students at the U.S. Military Academy in West Point, NY, and reported results that appear to be consistent with a decline in lung function that may in part be due to O\textsubscript{3} exposures over a period of several summer months. Ihorst et al. (2004) examined 2,153 children with a median age of 7.6 years and reported pulmonary function results which indicated that significantly lower FVC and FEV\textsubscript{1} increases were associated with higher O\textsubscript{3}. 

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exposures over the medium-term of several summer months, but not over several months in the winter. Semi-annual mean O₃ concentrations ranged from 22 to 54 ppb during the summer months and 4 to 36 ppb during the winter months. Further, over the longer-term 3.5-year period Ihorst et al. (2004) found that higher mean summer months O₃ levels were not associated with growth rates in lung function and for FVC and FEV₁, in contrast to the significant medium-term effects. Frischer et al. (1999) found that higher O₃ over one summer season, one winter season, and greater increases from one summer to the next over a three-year period were associated with smaller increases in lung function growth, indicating both medium and longer-term effects.

Mortimer et al. (2008a, b) examined the association of prenatal and lifetime exposures to air pollutants with pulmonary function and allergen sensitization in a subset of asthmatic children (ages 6-11) included in the Fresno Asthmatic Children’s Environment Study (FACES). Monthly means of pollutant levels for the years 1989-2000 were created and averaged separately across several important developmental time-periods, including: the entire pregnancy, each trimester, the first 3 years of life, the first 6 years of life, and the entire lifetime. In the first analysis (Mortimer et al., 2008a), negative effects on pulmonary function were found for exposure to PM₁₀, NO₂, and CO during key neonatal and early life developmental periods. The authors did not find a negative effect of exposure to O₃ within this cohort. In the second analysis (Mortimer et al., 2008b), sensitization to at least one allergen was associated, in general, with higher levels of CO and PM₁₀ during the entire pregnancy and second trimester, and higher PM₁₀ during the first 2 years of life. Lower exposure to O₃ during the entire pregnancy or second trimester was associated with an increased risk of allergen sensitization. Although the pollutant metrics across time periods were correlated, the strongest associations with the outcomes were observed for prenatal exposures. Though it may be difficult to disentangle the effect of prenatal and postnatal exposures, the models from this group of studies suggest that each time period of exposure may contribute independently to different dimensions of school-aged children’s pulmonary function. For 4 of the 8 pulmonary-function measures (FVC, FEV₁, PEF, FEF₂₅₋₇₅), prenatal exposures were more influential on pulmonary function than early-lifetime metrics, while, in contrast, the ratio of measures (FEV₁/FVC and FEF₂₅₋₇₅/FVC) were most influenced by postnatal exposures. When lifetime metrics were considered alone, or in combination with the prenatal metrics, the lifetime measures were not associated with any of the outcomes. This suggests that the timing of the O₃ exposure may be more important than the overall dose, and prenatal exposures are not just markers for lifetime or current exposures.

Latzin et al. (2009) examined whether prenatal exposure to air pollution was associated with lung function changes in the newborn. Tidal breathing, lung volume, ventilation inhomogeneity and eNO were measured in 241 unsedated, sleeping neonates (age = 5 weeks). Consistent with the previous studies, no association was found for prenatal exposure to O₃ and lung function.
In a cross-sectional study of adults, Qian et al. (2005) examined the association of long-term exposure to O_3 and PM_{10} with pulmonary function from data of 10,240 middle-aged subjects who participated in the Atherosclerosis Risk in Communities (ARIC) study in four U.S. communities. A surrogate for long-term O_3 exposure from daily data was determined at the individual level. Ozone was significantly and negatively associated with measures of pulmonary function.

To determine the extent to which long-term exposure to outdoor air pollution accelerates adult decline in lung function, Forbes et al. (2009b) studied the association between chronic exposure to outdoor air pollution and lung function in approximately 42,000 adults aged 16 and older who were representatively sampled cross-sectionally from participants in the Health Survey for England (1995, 1996, 1997, and 2001). FEV_1 was not associated with O_3 concentrations. In contrast to the results for PM_{10}, NO_2, and SO_2; combining the results of all the survey years showed that a 5-ppb difference in O_3 was counter-intuitively associated with a higher FEV_1 by 22 mL.

In a prospective cohort study consisting of school-age, non-asthmatic children in Mexico City (n = 3,170) who were 8 years of age at the beginning of the study, Rojas-Martinez et al. (2007) evaluated the association between long-term exposure to O_3, PM_{10} and NO_2 and lung function growth every 6 months from April 1996 through May 1999. Exposure data were provided by 10 air quality monitor stations located within 2 km of each child’s school. Over the study period, 8-h O_3 concentrations ranged from 60 ppb (SD, ± 25) in the northeast area of Mexico City to 90 ppb (SD, ± 34) in the southwest, with an overall mean of 69.8 ppb.

In multipollutant models, an IQR increase in mean O_3 concentration of 11.3 ppb was associated with an annual deficit in FEV_1 of 12 mL in girls and 4 mL in boys. Single-pollutant models showed an association between ambient pollutants (O_3, PM_{10}, and NO_2) and deficits in lung function growth. While the estimates from copollutant models were not substantially different than single pollutant models, independent effects for pollutants could not be estimated accurately because the traffic-related pollutants were correlated. To reduce exposure misclassification, microenvironmental and personal exposure assessments were conducted in a randomly selected subsample of 60 children using passive O_3 samplers. Personal O_3 concentrations were correlated (p <0.05) with the measurements obtained from the fixed-site air monitoring stations.

In the 2006 O_3 AQCD, few studies had investigated the effect of chronic O_3 exposure on pulmonary function. The strongest evidence was for medium-term effects of extended O_3 exposures over several summer months on lung function (FEV_1) in children, i.e., reduced lung function growth being associated with higher ambient O_3 levels. Longer-term studies (annual), investigating the association of chronic O_3 exposure on lung function (FEV_1) such as the definitive 8-year follow-up analysis of the first cohort (Gauderman et al., 2004) provide little evidence that long-term exposure to ambient O_3 at current levels is associated with significant deficits in the growth rate of lung function in children. Analyses indicated that there was no evidence that either 8-h avg O_3 (10 a.m. to 6 p.m.) or 24-h avg O_3 was associated
with any measure of lung function growth over a 4-year (age 10 to 14 years; (Gauderman et al., 2000)) or 8-year (age 10 to 18 years; (Gauderman et al., 2004)) period. However, most of the other pollutants examined (including PM\textsubscript{2.5}, NO\textsubscript{2}, acid vapor, and elemental carbon) were found to be significantly associated with reduced growth in lung function. In addition, there was only about a 2- to 2.5-fold difference in O\textsubscript{3} concentrations from the least to most polluted communities (mean annual average of 8-h avg O\textsubscript{3} ranged from 30 to 65 ppb), versus the ranges observed for the other pollutants (which had 4- to 8-fold differences in concentrations).

Short-term O\textsubscript{3} exposure studies presented in Section 6.2.1.2 provide a cumulative body of epidemiologic evidence that strongly supports associations between ambient O\textsubscript{3} exposure and decrements in lung function among children. For new studies of long-term O\textsubscript{3} exposure relationship to pulmonary function, one study, where O\textsubscript{3} and other pollutant levels were higher (90 ppb at high end of the range) than those in the CHS, observes a relationship between O\textsubscript{3} concentration and pulmonary function declines in school-aged children. Two studies of adult cohorts provide mixed results where long- term exposures were at the high end of the range with levels of 49.5 ppb in one study and 27 ppb IQR in the other. Toxicological studies examining monkeys have provided data for airway resistance in an asthma model but this is difficult to compare to FEV\textsubscript{1} results. Thus there is little new evidence to build upon the very limited studies of pulmonary function (FEV\textsubscript{1}) from the 2006 O\textsubscript{3} AQCD.

### 7.2.3.2 Pulmonary Structure and Function: Evidence from Toxicological Studies and Nonhuman Primate Asthma Models

Long-term studies in animals allow for greater insight into the potential effects of prolonged exposure to O\textsubscript{3}, that may not be easily measured in humans, such as structural changes in the respiratory tract. As reviewed in the 1996 and 2006 O\textsubscript{3} AQCDs and Chapter 5 of this ISA, there are both qualitative and quantitative uncertainties in the extrapolation of data generated by rodent toxicology studies to the understanding of health effects in humans. Despite these uncertainties, epidemiologic studies observing functional changes in humans can attain biological plausibility, in conjunction with long-term toxicological studies, particularly O\textsubscript{3}-inhalation studies performed in non-human primates whose respiratory system most closely resembles that of the human. An important series of studies have used nonhuman primates to examine the effect of O\textsubscript{3} alone or in combination with an inhaled allergen, house dust mite antigen, on morphology and lung function. These animals exhibit the hallmarks of allergic asthma defined for humans, including: a positive skin test for HDMA with elevated levels of IgE in serum and IgE-positive cells within the tracheobronchial airway walls; impaired airflow which is reversible by treatment with aerosolized albuterol; increased abundance of immune cells, especially eosinophils, in airway exudates and bronchial lavage; and development of nonspecific airway responsiveness (NHLBI, 2007). Hyde et al. (2006) compared asthma models of rodents (mice) and the nonhuman primate model to responses in
humans and concluded that the unique responses to inhaled allergen shown in the rhesus monkeys make it the most appropriate animal model of human asthma. These studies and others have demonstrated changes in pulmonary function and airway morphology in adult and infant nonhuman primates repeatedly exposed to environmentally relevant concentrations of O₃ (Joad et al., 2008; Carey et al., 2007; Plopper et al., 2007; Fanucchi et al., 2006; Joad et al., 2006; Evans et al., 2004; Larson et al., 2004; Tran et al., 2004; Evans et al., 2003; Schelegle et al., 2003; Fanucchi et al., 2000; Hyde et al., 1989; Harkema et al., 1987a; Harkema et al., 1987b; Fujinaka et al., 1985). Many of the observations found in adult monkeys have also been noted in infant rhesus monkeys, although a direct comparison of the degree of effects between adult and infant monkeys has not been reported. The findings of these nonhuman primate studies have also been observed in rodent studies discussed at the end of this section and included in Table 7-1.

The initial observations in adult nonhuman primates have been expanded in a series of experiments using infant rhesus monkeys repeatedly exposed to 0.5 ppm O₃ starting at 1 month of age¹ (Plopper et al., 2007). The purpose of these studies, designed by Plopper and colleagues, was to determine if a cyclic regimen of O₃ inhalation would amplify the allergic responses and structural remodeling associated with allergic sensitization and inhalation in the infant rhesus monkey. In terms of pulmonary function changes, after several episodic exposures of infant monkeys to O₃, they observed a significant increase in the baseline airway resistance, which was accompanied by a small increase in airway responsiveness to inhaled histamine (Schelegle et al., 2003), although neither measurement was statistically different from filtered air control values. Exposure of animals to inhaled house dust mite antigen alone also produced small but not statistically significant changes in baseline airway resistance and airway responsiveness, whereas the combined exposure to both (O₃ + antigen) produced statistically significant and greater than additive changes in both functional measurements. This nonhuman primate evidence of an O₃-induced change in airway resistance and responsiveness supports the biologic plausibility of long-term exposure to O₃ contributing to the effects of asthma in children.

To understand which conducting airways and inflammatory mechanisms are involved in O₃-induced airway hyperresponsiveness in the infant rhesus monkey, a follow-up study examined airway responsiveness ex vivo in lung slices (Joad et al., 2006). Using video microscopy to morphometrically evaluate the response of bronchi and respiratory bronchioles to methacholine, (a bronchoconstricting agent commonly used to evaluate airway responsiveness in asthmatics), the investigators observed differential effects for the two airway sizes. While episodic exposure to O₃ alone (0.5 ppm) had little effect on ex vivo airway responsiveness in bronchi and respiratory bronchioles, exposure to dust mite antigen alone produced airway hyperresponsiveness in the large bronchi, whereas O₃ + antigen produced significant increases in airway hyperresponsiveness only in the respiratory bronchioles. These

¹ Schelegle et al. (2003) used a two-by-two block design. Twenty-four infant rhesus monkeys (30 days old) were exposed to 11 episodes (total of 6-months exposure period) of filtered air (FA), house dust mite allergen (HDMA), O₃ (5 days each followed by 9 days of FA). Ozone was delivered for 8h/day at 0.5 ppm. Twelve of the monkeys (HDMa, and HDMA + O₃ groups) were sensitized to house dust mite allergen (HDMa, confirmed by skin testing). To evaluate the potential for recovery, the 5 months of exposure were followed by another 6 months in FA until the monkeys were reevaluated at 12 months of age.
results suggest that effect of O₃ on airway responsiveness occurs predominantly in the smaller bronchioles, where dosimetric models indicate the dose would be higher.

The functional changes in the conducting airways of infant rhesus monkeys exposed to either O₃ alone or O₃ + antigen were accompanied by a number of cellular and morphological changes, including a significant 4-fold increase in eosinophils, (a cell type important in allergic asthma), in the bronchoalveolar lavage of infant monkeys exposed to O₃ alone. Thus, these studies demonstrate both functional and cellular changes in the lung of infant monkeys after cyclic exposure to 0.5 ppm O₃. This concentration, provides relevant information to understanding the potentially damaging effects of ambient O₃ exposure on the respiratory tract of humans. No concentration-response data, however, are available from these nonhuman primate studies.

In addition to these functional and cellular changes, significant structural changes in the respiratory tract have been observed in infant rhesus monkeys exposed to O₃. During normal respiratory tract development, conducting airways increase in diameter and length in the infant rhesus monkey. Exposure to O₃ alone (5 days of 0.5 ppm O₃ at 8 h/day, followed by 9 days of filtered air exposures for 11 cycles), however, markedly affected the growth pattern of distal conducting airways (Fanucchi et al., 2006). Whereas the first alveolar outpocketing occurred at airway generation 13 or 14 in filtered air-control infant monkeys, the most proximal alveolarized airways occurred at an average of 10 airway generations in O₃-exposed monkeys. Similarly, the diameter and length of the terminal and respiratory bronchioles were significantly decreased in O₃-exposed monkeys. Importantly, the O₃-induced structural pathway changes persisted after recovery in filtered air for 6 months after cessation of the O₃ exposures. These structural effects were accompanied by significant increases in mucus goblet cell mass, alterations in smooth muscle orientation in the respiratory bronchioles, epithelial nerve fiber distribution, and basement membrane zone morphometry. These latter effects are noteworthy because of their potential contribution to airway obstruction and airway hyperresponsiveness which are central features of asthma.

Because many cellular and biochemical factors are known to contribute to allergic asthma, the effect of exposure to O₃ alone or O₃ + antigen on immune system parameters was also examined in infant rhesus monkeys. Mast cells, which contribute to asthma via the release of potent proteases, were elevated in animals exposed to antigen alone but O₃ alone had little effect on mast cell numbers and the response of animals exposed to O₃ + antigen was not different from that of animals exposed to antigen alone; thus suggesting that mast cells played little role in the interaction between O₃ and antigen in this model of allergic asthma (Van Winkle et al., 2010). Increases in CD4+ and CD8+ lymphocytes were observed at 6 months of age in the blood and bronchoalveolar lavage fluid of infant rhesus monkeys exposed to O₃ + antigen but not in monkeys exposed to either agent alone (Miller et al., 2009). Activated lymphocytes (i.e., CD25+ cells) were morphometrically evaluated in the airway mucosa and significantly increased in infant monkeys exposed to antigen alone or O₃ + antigen. Although O₃ alone had no effect on CD25+ cells, it
did alter the anatomic distribution of CD25+ cells within the airways. Ozone had only a small effect on these sets of immune cells and did not produce a strong interaction with an inhaled allergen in this nonhuman primate model.

In addition to alterations in the immune system, nervous system interactions with epithelial cells are thought to play a contributing role to airway hyperresponsiveness. A critical aspect of postnatal lung development is the laying of nerve axons with specific connections serving to maintain lung homeostasis. Aberrant innervation patterns may underlie allergic airways disease pathology and long-term decrements in airway function. As noted in the 2006 O₃ AQCD, exposure of infant rhesus monkeys altered the normal development of neural innervation in the epithelium of the conducting airways (Larson et al., 2004). Significant mean reductions in nerve fiber density were observed in the midlevel airways of animals exposed to O₃ alone (49% reduction), and O₃ + antigen (55% reduction). Moreover, the morphology of nerve bundles was altered. The persistence of these effects was examined after a 6-month recovery period, and although nerve distribution remained atypical, there was a dramatic increase in airway nerve density (hyperinnervation) (Kajekar et al., 2007). Thus, in addition to structural, immune, and inflammatory effects, exposure to O₃ produces alterations in airway innervation which may contribute to O₃-induced exacerbation of asthma. Evaluation of the pathobiology of airway remodeling in growing lungs of neonates using an animal model where exposure to allergen generates reactive airway disease with all the hallmarks of asthma in humans illustrates that exposure to O₃ and allergen early in life produces a large number of disruptions of fundamental growth and differentiation processes.

A number of studies in both nonhuman primates and rodents demonstrate that O₃ exposure can increase collagen synthesis and deposition, inducing fibrotic-like changes in the lung (Last et al., 1994; Chang et al., 1992; Moffatt et al., 1987; Reiser et al., 1987; Last et al., 1984). Increased collagen content is often associated with elevated abnormal cross links that appear to be irreversible (Reiser et al., 1987). Generally changes in collagen content have been observed in rats exposed to 0.5 ppm O₃ or higher, although extracellular matrix thickening has been observed in the lungs of rats exposed to an urban pattern of O₃ with daily peaks of 0.25 ppm for 38 weeks (Chang et al., 1992; Chang et al., 1991). A more recent study using an urban pattern of exposure to 0.5 ppm O₃ demonstrated that O₃-induced collagen deposition in mice is dependent on the activity of TGF-β (Katre et al., 2011). Sex differences have been observed with respect to increased centriacinar collagen deposition and crosslinking, which was observed in female but not male rats exposed to 0.5 and 1.0 ppm O₃ for 20 months (Last et al., 1994). Few other long-term exposure morphological studies have presented sex differences and most only evaluated males.

As described in the 1996 and 2006 O₃ AQCDs, perhaps the largest chronic O₃ study was an NIEHS-NTP/HEI funded rodent study conducted by multiple investigators studying a number of different respiratory tract endpoints (Catalano et al., 1995b). Rats were exposed to 0.12, 0.5, or 1.0 ppm O₃ for 6 h/day and 5 d/week for 20 months. The most prominent changes were observed in the nasal cavity where a large fraction of O₃ is absorbed. Alterations in nasal function (increased mucous flow) and
structure (goblet cell metaplasia) were observed at 0.5 and 1.0 ppm but not 0.12 ppm O₃. In the lung, the centriacinar region (CAR) was the anatomical site most affected by O₃. The epithelial cell lining was changed to resemble that seen in respiratory bronchioles and the interstitial volume was increased. Biochemical analyses demonstrated increased collagen and glycoaminoglycans, an observation that supported the structural changes. As in the nose, these changes were observed only at the two highest exposure concentrations. Importantly, despite these morphologic and biochemical changes after 20 months of exposure, detailed pulmonary function testing revealed little to no measurable change in function. Thus, minor respiratory tract changes were observed after chronic exposure to O₃ up to 1.0 ppm in the F344 rat model.

It is unclear what the long-term impact of O₃-induced structural changes may be. Simulated seasonal (episodic) exposure studies suggest that such exposures might have cumulative impacts, and a number of studies indicate that structural changes in the respiratory system are persistent or irreversible. For example, O₃-induced hyperplasia was still evident in the nasal epithelia of rats 13 weeks after recovery from 0.5 ppm O₃ exposure (Harkema et al., 1999). In a study of episodic exposure to 0.25 ppm O₃, Chang et al. (1992) observed no reversal of basement membrane thickening in rat lungs up to 17 weeks post-exposure. Thickening of the sub-basement membrane is one of the persistent structural features observed in human asthmatics (NHLBI, 2007). Episodic exposure (0.25 ppm O₃, every other month) of young monkeys induced equivalent morphological changes compared to continuously exposed animals, even though they were exposed for half the time and evaluation occurred a month after exposure ceased as opposed to immediately (Tyler et al., 1988). Notably, episodic O₃ exposure increased total lung collagen content, chest wall compliance, and inspiratory capacity, suggesting a delay in lung maturation in episodically-exposed animals. These changes were in contrast to the continuously exposed group, which did not differ from the air exposed group in these particular parameters but did exhibit greater bronchiolitis than the episodically exposed animals. In a study by Harkema and colleagues (Harkema et al., 1993, 1987b), monkeys (both males and females) were acutely exposed for 8 h/day to 0.15 ppm O₃ (6 days) or chronically to 0.15 ppm or 0.3 ppm O₃ (90 days). For most endpoints in the nasal cavity, the observed morphologic changes and inflammation were greater in the monkeys exposed for 6 days compared to 90 days, whereas in the respiratory bronchioles of the same animals, there were no significant time or concentration dependent differences (increased epithelial thickness and proportion of cuboidal cells) between the 6 and 90 day exposure groups.

Stokinger (1962) reported that chronic bronchitis, bronchiolitis, and emphysematous and fibrotic changes develop in the lung tissues of mice, rats, hamsters, and guinea pigs exposed 6 h/day, 5 days/week for 14.5 months to a concentration slightly above 1 ppm O₃. Rats continuously exposed for 3 to 5 months to 0.8 ppm O₃ develop a disease that resembles emphysema, and they finally die of respiratory failure (Stephens et al., 1976). Ozone results in a greater response of fibroblasts in the lesion, thickening of the alveolar septae, and an increase in number of alveolar macrophages in the proximal alveoli.
Table 7-1  Respiratory effects in nonhuman primates and rodents resulting from long-term O$_3$ exposure.

<table>
<thead>
<tr>
<th>Study</th>
<th>Model</th>
<th>O$_3$ (ppm)</th>
<th>Exposure Duration</th>
<th>Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pinkerton et al. (1998); Harkema et al. (1997a); Harkema et al. (1997b); Catalano et al. (1995a); Chang et al. (1995); Pinkerton et al. (1995); Stockstill et al. (1995); Harkema et al. (1994); Last et al. (1994); Plopper et al. (1994)</td>
<td>Rat, male and female, Fischer F344, 6-8 weeks old</td>
<td>0.12 0.5 1.0</td>
<td>6 h/day, 5 days/week for 20 months</td>
<td>Effects similar to (or a model of) early fibrotic human disease were greater in the periacinar region than in terminal bronchioles. Thickened alveolar septa observed at 0.12 ppm O$_3$. Other effects (e.g., mucous cell metaplasia in the nose, mild fibrotic response in the parenchyma, and increased collagen in CAR of females) observed at 0.5 to 1.0 ppm. Some morphometric changes (epithelial thickening and bronchiolarization) occurred after 2 or 3 months of exposure to 1.0 ppm.</td>
</tr>
<tr>
<td>Herbert et al. (1996)</td>
<td>Mice, male and female, B6C3F1, 6-7 weeks old,</td>
<td>0.12 0.50 1.0</td>
<td>6 h/day, 5 days/week for 24 and 30 months</td>
<td>Similar to the response of rats in the same study (see rat above). Effects were seen in the nose and centriacinar region of the lung at 0.5 and 1.0 ppm.</td>
</tr>
<tr>
<td>Chang et al. (1991)</td>
<td>Rat, male, F344, 6 weeks old</td>
<td>Continuous: 0.12 or 0.25  Episodic/urban: baseline 0.06; peak 0.25</td>
<td>Continuous: 12 h/day for 6 weeks Simulated urban pattern; slow rise to peak 9 h/day, 5 days/week, 13 weeks</td>
<td>Increased Type 1 and 2 epithelial volume assessed by TEM. Linear relationship observed between increases in Type 1 epithelial cell volume and concentration x time product. Degree of injury not related to pattern of exposure (continuous or episodic).</td>
</tr>
<tr>
<td>Chang et al. (1992)</td>
<td>Rat, male, F344, 6 weeks old</td>
<td>baseline 0.06; peak 0.25</td>
<td>Slow rise to peak 9 h/day, 5 days/week, 13 and 78 weeks Recovery in filtered air for 6 or 17 weeks</td>
<td>Progressive epithelial hyperplasia, fibroblast proliferation, and interstitial matrix accumulation observed using TEM. Interstitial matrix thickening due to deposition of basement membrane and collagen fibers. Partial recovery of interstitial matrix during follow-up periods in air; but no resolution of basement membrane thickening.</td>
</tr>
<tr>
<td>Barry et al. (1985, 1983)</td>
<td>Rat, male, 1 day old or 6 weeks old</td>
<td>0.12 (adults only) 0.25</td>
<td>12 h/day for 6 weeks</td>
<td>Lung and alveolar development not significantly affected. Increased Type 1 and 2 epithelial cells and AM in CAR alveoli, thickened Type 1 cells with smaller volume and less surface coverage as assessed by TEM (adults and juveniles). In adults, smaller but statistically significant similar changes at 0.12 ppm, suggesting linear concentration-response relationship. No statistically significant age-related effects observed.</td>
</tr>
<tr>
<td>Tyler et al. (1988)</td>
<td>Monkey; male, Macaca fascicularis, 7 mo old</td>
<td>0.25</td>
<td>8 h/day, 7 days/week, Daily for 18 mo or episodically every other month for 18 mo Episodic group evaluated 1 mo postexposure</td>
<td>Increased collagen content, chest wall compliance, and inspiratory capacity in episodic group only. Respiratory bronchiolitis in both groups. Episodically exposed group incurred greater alterations in physiology and biochemistry and equivalent changes in morphometry even though exposed for half the time as the daily exposure group.</td>
</tr>
<tr>
<td>Harkema et al.</td>
<td>Rat, male, Fischer</td>
<td>0.25</td>
<td>8 h/day, 7 days/week</td>
<td>Mucous cell hyperplasia in nasal</td>
</tr>
</tbody>
</table>
Collectively, evidence from animal studies strongly suggests that chronic O$_3$ exposure is capable of damaging the distal airways and proximal alveoli, resulting in lung tissue remodeling and leading to apparent irreversible changes. Potentially, persistent inflammation and interstitial remodeling play an important role in the
progression and development of chronic lung disease. Further discussion of the modes of action that lead to O$_3$-induced morphological changes can be found in Section 5.3.7. The findings reported in chronic exposure animal studies offer insight into potential biological mechanisms for the suggested association between seasonal O$_3$ exposure and reduced lung function development in children as observed in epidemiologic studies (see Section 7.2.3). Discussion of mechanisms involved in lifestage susceptibility and developmental effects can be found in Section 5.4.2.4.

### 7.2.4 Pulmonary Inflammation, Injury, and Oxidative Stress

The 2006 O$_3$ AQCD stated that the extensive human clinical and animal toxicological evidence, together with the limited epidemiologic evidence available, suggests a causal role for O$_3$ in inflammatory responses in the airways. Short-term exposure epidemiologic studies discussed earlier in Section 6.2.3.2 show consistent associations of O$_3$ exposure and increased airway inflammation and oxidative stress. Further discussion of the mechanisms underlying inflammation and oxidative stress responses can be found in Section 5.3.3. Though the majority of recent studies focus on short-term exposures, several epidemiologic and toxicology studies of long-term exposure add to observations of O$_3$-induced inflammation and injury.

Inflammatory markers and peak expiratory pulmonary function were examined in 37 allergic children with physician-diagnosed mild persistent asthma in a highly polluted urban area in Italy and then again 7 days after relocation to a rural location with significantly lower pollutant levels (Renzetti et al., 2009). The authors observed a 4-fold decrease in nasal eosinophils and a statistically significant decrease in fractional exhaled nitric oxide along with an improvement in lower airway function. Several pollutants were examined, including PM$_{10}$, NO$_2$, and O$_3$, though pollutant-specific results were not presented. These results are consistent with studies showing that traffic-related exposures are associated with increased airway inflammation and reduced lung function in children with asthma and contribute to the notion that this negative influence may be rapidly reversible. Exhaled NO (eNO) has been shown to be a useful biomarker for airway inflammation in large population-based studies (Linn et al., 2009). Thus, while the time scale of 7 days between examinations for eNO needs to be evaluated for appropriateness, the results suggest that inflammatory responses are reduced when O$_3$ levels are decreased.

Chest radiographs (CXR) of 249 children in Mexico City who were chronically exposed to O$_3$ and PM$_{2.5}$ were analyzed by Calderón-Garciduenas et al. (2006). They reported an association between chronic exposures to O$_3$ and other pollutants and a significant increase in abnormal CXR’s and lung CTs suggestive of a bronchiolar, peribronchiolar, and/or alveolar duct inflammatory process, in clinically healthy children with no risk factors for lung disease. These CXR and CT results should be viewed with caution because it is difficult to attribute effects to O$_3$ exposure.

In a cross-sectional study, Wood et al. (2009) examined the association of outdoor air pollution with respiratory phenotype (PiZZ type) in alpha 1-antitrypsin deficiency
(α-ATD) from the U.K. α-ATD registry. This deficiency leads to exacerbated responses to inflammatory stimuli. In total, 304 PiZZ subjects underwent full lung function testing and quantitative high-resolution computed tomography to identify the presence and severity of COPD – emphysema. Mean annual air pollution data for 2006 was matched to the location of patients’ houses and used in regression models to identify phenotypic associations with pollution controlling for covariates. Relative trends in O₃ levels were assessed to validate use of a single year’s data to indicate long-term exposure and validation; data showed good correlations between modeled and measured data (Stedman and Kent, 2008). Regression models showed that estimated higher exposure to O₃ exposure was associated with worse gas transfer and more severe emphysema, albeit accounting for only a small proportion of the lung function variability. This suggests that a gene-specific group demonstrates a long-term O₃ exposure effect.

The similarities of nonhuman primates to humans make them attractive models in which to study the effects of O₃ on the respiratory tract. The nasal mucous membranes, which protect the more distal regions of the respiratory tract, are susceptible to injury from O₃. Carey et al. (2007) conducted a study of O₃ exposure in infant rhesus macaques, whose nasal airways closely resemble that of humans. Monkeys were exposed either acutely for 5 days (8 h/day) to 0.5 ppm O₃, or episodically for several biweekly cycles alternating 5 days of 0.5 ppm O₃ with 9 days of filtered air (0 ppm O₃), designed to mimic human exposure (70 days total). All monkeys acutely exposed to O₃ had moderate to marked necrotizing rhinitis, with focal regions of epithelial exfoliation, numerous infiltrating neutrophils, and some eosinophils. The distribution, character, and severity of lesions in episodically exposed monkeys were similar to that of acutely exposed animals. Neither group exhibited the mucous cell metaplasia proximal to the lesions, observed in adult monkeys exposed continuously to 0.3 ppm O₃ in another study (Harkema et al., 1987a). Adult monkeys also exhibited attenuation of inflammatory responses with continued daily exposure (Harkema et al., 1987a), but inflammation did not resolve over time in young episodically exposed monkeys (Carey et al., 2011). Inflammation in conducting airways has also been observed in rats chronically exposed to O₃. Using an agar-based technique to fill the alveoli so that only the rat bronchi are lavaged, a 90-day exposure of rats to 0.8 ppm O₃ (8 h/day) elicited significantly elevated pro-inflammatory eicosanoids PGE₂ and 12-HETE in the conducting airway compared to filtered air-exposed rats (Schmelzer et al., 2006).

Persistent inflammation and injury leading to interstitial remodeling may play an important role in the progression and development of chronic lung disease. Chronic airway inflammation is an important component of both asthma and COPD. The epidemiological evidence supporting an association between long-term exposure to O₃ and inflammation or injury is limited. However, animal studies clearly demonstrate O₃-induced inflammation and injury, which may or may not attenuate with chronic exposure depending on the model. Further discussion of how O₃ initiates inflammation can be found in Section 5.3.3.
7.2.5 Allergic Responses

The association of air pollutants with childhood respiratory allergies was examined in the U.S. using the 1999-2005 National Health Interview Survey of approximately 70,000 children, and ambient air pollution data from the U.S. EPA, with monitors within 20 miles of each child’s residential block (Parker et al., 2009). The authors examined the associations between the reporting of respiratory allergy or hay fever and medium-term exposure to O₃ over several summer months, controlling for demographic and geographic factors. Increased respiratory allergy/hay fever was associated with increased O₃ levels (adjusted OR per 10 ppb = 1.20; [95% CI: 1.15, 1.26]). These associations persisted after stratification by urban-rural status, inclusion of multiple pollutants (O₃, SO₂, NO₂, PM), and definition of exposure by differing exposure radii; smaller samples within 5 miles of monitors were remarkably similar to the primary results. No associations between the other pollutants and the reporting of respiratory allergy/hay fever were apparent. Ramadour et al. (2000) reported no relationship between O₃ levels and rhinitis symptoms and hay fever. Hwang et al. (2006) report the prevalence of allergic rhinitis (adjusted OR per 10 ppb = 1.05; [95% CI: 0.98, 1.12]) in a large cross-sectional study in Taiwan. In a large cross-sectional study in France, Penard-Morand et al. (2005) reported a positive relationship between lifetime allergic rhinitis and O₃ exposure in a two-pollutant model with NO₂. These studies related positive outcomes of allergic response and O₃ exposure but with variable strength for the effect estimates. A toxicological study reported that five weeks of continuous exposure to 0.4 ppm O₃ (but not 0.1 or 0.2 ppm O₃) augmented sneezing and nasal secretions in a guinea pig model of nasal allergy (Iijima and Kobayashi, 2004). Nasal eosinophils, which participate in allergic disease and inflammation, and allergic antibody levels in serum were also elevated by exposure to concentrations as low as 0.2 ppm (Iijima and Kobayashi, 2004).

Nasal eosinophils were observed to decrease by 4-fold in 37 atopic, mildly asthmatic children 7 days after relocation from a highly polluted urban area in Italy to a rural location with significantly lower pollutant levels (Renzetti et al., 2009). Inflammatory and allergic effects of O₃ exposure (30 day mean) such as increased eosinophil levels were observed in children in an Austrian study (Frischer et al., 2001). Episodic exposure of infant rhesus monkeys to 0.5 ppm O₃ for 5 months appears to significantly increase the number and proportion of eosinophils in the blood and airways (lavage) [protocol described above in Section 7.2.3.2 for Fanucchi et al. (2006)] (Maniar-Hew et al., 2011). These changes were not evident at 1 year of age (6 months after O₃ exposure ceased). Increased eosinophils levels have also been observed after acute or prolonged exposures to O₃ in adult bonnet and rhesus monkeys (Hyde et al., 1992; Eustis et al., 1981).

Total IgE levels were related to air pollution levels in 369 adult asthmatics in five French centers using generalized estimated equations (GEE) as part of the EGEA study described earlier (Rage et al., 2009a). Geostatistical models were performed on 4×4 km grids to assess individual outdoor air pollution exposure that was assigned to subject’s home address. Ozone concentrations were positively related to total IgE levels and an increase of 5 ppb of O₃ resulted in an increase of 20.4% (95% CI: 3.0,
40.7) in total IgE levels. Nearly 75% of the subjects were atopic. In two-pollutant models including O₃ and NO₂, the O₃ effect estimate was decreased by 25% while the NO₂ effect estimate was decreased by 57%. Associations were not sensitive to adjustment for covariates or the season of IgE measurements. These cross-sectional results suggest that exposure to O₃ may increase total IgE in adult asthmatics.

Although very few toxicological studies of long-term exposure examining allergy are available, short-term exposure studies in rodents and nonhuman primates demonstrate allergic skewing of immune responses and enhanced IgE production. Due to the persistent nature of these responses, the short-term toxicological evidence lends biological plausibility to the limited epidemiologic findings of an association between long-term O₃ exposure and allergic outcomes.

7.2.6 Host Defense

Short-term exposures to O₃ have been shown to cause decreases in host defenses against infectious lung disease in animal models. Acute O₃-induced suppression of alveolar phagocytosis and immune functions observed in animals appears to be transient and attenuated with continuous or repeated exposures, although chronic exposure (weeks, months) has been shown to slow alveolar clearance. In an important study investigating the effects of longer term O₃ exposure on alveolobronchiolar clearance, rats were exposed to an urban pattern of O₃ (continuous 0.06 ppm, 7 days/week with a slow rise to a peak of 0.25 ppm and subsequent decrease to 0.06 ppm over a 9 h period for 5 days/week) for 6 weeks and were exposed 3 days later to chrysotile asbestos, which can cause pulmonary fibrosis and neoplasia. After 30 days, the lungs of the O₃-exposed animals had twice the number and mass of asbestos fibers as the air-exposed rats. However, chronic exposures of 0.1 ppm do not cause greater effects on infectivity than short exposures, due to defense parameters becoming reestablished with prolonged exposures. No detrimental effects were seen with a 120-day exposure to 0.5 ppm O₃ on acute lung injury from influenza virus administered immediately before O₃ exposure started. However, O₃ was shown to increase the severity of postinfluenzal alveolitis and lung parenchymal changes. A recent study by Maniar-Hew et al. (2011) demonstrated that the immune system of infant rhesus monkeys episodically exposed to 0.5 ppm O₃ for 5 months appeared to be altered in ways that could diminish host defenses. Reduced numbers of circulating leukocytes were observed, particularly polymorphonuclear leukocytes (PMNs) and lymphocytes, which were decreased in the blood and airways. Challenge with LPS, a bacterial ligand that activates monocytes and other innate immune cells, elicited lower responses in O₃-exposed animals even though the relevant reactive cell population was increased. This was observed in both an in vivo inhalation challenge.

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1 Exposure protocol is described above in Section 7.2.3.2 for Fanucchi et al. (2006).
and an ex vivo challenge of peripheral blood mononuclear cells. Thus a decreased ability to respond to pathogenic signals was observed six months after O₃ exposure ceased, in both the lungs and periphery.

### 7.2.7 Respiratory Mortality

A limited number of epidemiologic studies have assessed the relationship between long-term exposure to O₃ and mortality. The 2006 O₃ AQCD concluded that an insufficient amount of evidence existed “to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans” (U.S. EPA, 2006b). Though total and cardio-pulmonary mortality were considered in these studies, respiratory mortality was not specifically considered. In the most recent follow-up analysis of the ACS cohort (Jerrett et al., 2009), cardiopulmonary deaths were subdivided into respiratory and cardiovascular, separately, as opposed to combined in the Pope et al. (2002) work. A 10-ppb increment in exposure to O₃ elevated the risk of death from respiratory causes and this effect was robust to the inclusion of PM₂.₅. The association between increased O₃ concentrations and increased risk of death from respiratory causes was insensitive to the use of a random-effects survival model allowing for spatial clustering within the metropolitan area and state of residence, and to adjustment for several ecologic variables considered individually. Additionally, a recent study (Zanobetti and Schwartz, 2011) observed an association between long-term exposure to O₃ and elevated risk of mortality among Medicare enrollees that had previously experienced an emergency hospital admission due to COPD.

### 7.2.8 Summary and Causal Determination

The epidemiologic studies reviewed in the 2006 O₃ AQCD detected no associations between long-term (annual) O₃ exposures and asthma-related symptoms, asthma prevalence, or allergy to common aeroallergens among children after controlling for covariates. Little evidence was available to relate long-term exposure to ambient O₃ concentrations with deficits in the growth rate of lung function in children. Additionally, limited evidence was available evaluating the relationship between long-term O₃ concentrations and pulmonary inflammation and other endpoints. From toxicological studies, it appeared that O₃-induced inflammation tapered off during long-term exposures, but that hyperplastic and fibrotic changes remained elevated and in some cases even worsened after a postexposure period in clean air. Episodic exposures were also known to cause more severe pulmonary morphologic changes than continuous exposure (U.S. EPA, 2006b).

The recent epidemiologic evidence base consists of studies using a variety of designs and analysis methods evaluating the relationship between long-term exposure to ambient O₃ concentrations and measures of respiratory health effects and mortality conducted by different research groups in different locations. See Table 7-2 for O₃
concentrations associated with selected studies. Table 7-2 is organized by longitudinal and cross-sectional studies both presented alphabetically. The positive results from various designs and locations support a relationship between long-term exposure to ambient O₃ concentrations and respiratory health effects and mortality.

Earlier studies reported associations of new-onset asthma and O₃ in an adult cohort in California (McDonnell et al., 1999a; Greer et al., 1993) but only in males. In the CHS cohort of children in 12 Southern California communities, long-term exposure to O₃ concentrations was not associated with increased risk of developing asthma (McConnell et al., 2010); however, greater outdoor exercise was associated with development of asthma in children living in communities with higher ambient O₃ concentrations (McConnell et al., 2002). Recent CHS studies examined interactions among genetic variants, long-term O₃ exposure, and new onset asthma in children. These prospective cohort studies are methodologically rigorous epidemiology studies, and evidence indicates gene-O₃ interactions. These studies have provided data supporting decreased risk of certain different genetic variants on new onset asthma (e.g., HMOX-1, ARG) that is limited to children either in low (Islam et al., 2008) or high (Salam et al., 2009) O₃ communities. Gene-environment interaction also was demonstrated with findings that greater outdoor exercise increased risk of asthma in GSTP1 Ile/Ile children living in high O₃ communities (Islam et al., 2009). Biological plausibility for these these gene-O₃ environment interactions is provided by evidence that these enzymes have antioxidant and/or anti-inflammatory activity and participate in well recognized modes of action in asthma pathogenesis. As O₃ is a source of oxidants in the airways, oxidative stress serves as the link among O₃ exposure, enzyme activity, and asthma.

Studies using a cross-sectional design provide support for a relationship between long-term O₃ exposure and health effects in asthmatics. A long-term O₃ exposure study relates bronchitic symptoms to TNF-308 genotype asthmatic children with ambient O₃ exposure in the CHS (Lee et al., 2009b). A study in five French cities reports effects on asthma severity related to long-term O₃ exposure (Rage et al., 2009b). A follow-up study of this cohort (Jacquemin et al., 2012) supports an effect of cumulative long-term O₃ exposure on asthma control in adulthood in subjects with pre-existing asthma. Akinbami et al. (2010) and Hwang et al. (2005) provide further evidence relating O₃ exposures and the risk of asthma. For the respiratory health of a cohort based on the general U.S. population, risk of respiratory-related school absences was elevated for children with the CAT and MPO variant genes related to communities with high ambient O₃ levels (Wenten et al., 2009).
Table 7-2  Summary of selected key new studies examining annual $O_3$ exposure and respiratory health effects.

<table>
<thead>
<tr>
<th>Study; Health Effect; Location</th>
<th>Annual Mean $O_3$ Concentration (ppb)</th>
<th>$O_3$ Range (ppb) Percentiles</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Longitudinal</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Islam et al. (2008); New-onset asthma; CHS</td>
<td>55.2 high vs. 38.4 low communities 10:00 a.m. to 6:00 p.m. average</td>
<td>See left</td>
</tr>
<tr>
<td>Islam et al. (2009); New-onset asthma; CHS</td>
<td>55.2 high vs. 38.4 low communities 10:00 a.m. to 6:00 p.m.</td>
<td>See left</td>
</tr>
<tr>
<td>Lin et al. (2008b); First asthma hospital admission; New York State - 10 regions</td>
<td>Range of mean $O_3$ concentrations over the 10 New York Regions 37.51 to 47.78 8-h max 10:00 a.m. to 6:00 p.m.</td>
<td>See left</td>
</tr>
<tr>
<td>Salam et al. (2009); Childhood onset asthma; CHS</td>
<td>$O_3$ greater than or less than 50 ppb</td>
<td>See left</td>
</tr>
<tr>
<td><strong>Cross-sectional</strong></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Akinbami et al. (2010); Current asthma U.S.</td>
<td>12 month median 39.8 8hr max</td>
<td>IQR 35.9 to 43.7</td>
</tr>
<tr>
<td>Hwang et al. (2005); Prevalence of asthma; Taiwan</td>
<td>Mean 23.14</td>
<td>Range 18.65 to 31.17</td>
</tr>
<tr>
<td>Jacquemin et al. (2012); Asthma control in adults; Five French cities</td>
<td>Median 46.9 ppb; 8-h average</td>
<td>25th-75th 41-52</td>
</tr>
<tr>
<td>Lee et al. (2009b); Bronchitic symptoms in asthmatic children; CHS</td>
<td>Above and below 50 ppb</td>
<td>See left</td>
</tr>
<tr>
<td>Meng et al. (2010); Asthma ED visits or hospitalizations; San Joaquin Valley, CA</td>
<td>Median 30.3 ppb Yearly based on hourly</td>
<td>25-75% range 27.1 to 34.0</td>
</tr>
<tr>
<td>Moore et al. (2008); Asthma hospital admissions; South Coast Basin</td>
<td>Median 87.8 ppb Quarterly 1hr daily max</td>
<td>Range 28.6 to 199.9</td>
</tr>
<tr>
<td>Rage et al. (2009a); Asthma severity; Five French cities</td>
<td>Mean 30 ppb 8-h average</td>
<td>25th-75th 21-36</td>
</tr>
<tr>
<td>Wenten et al. (2009); Respiratory school absence, U.S.</td>
<td>Median 46.9 ppb; 10a.m. – 6 p.m. average</td>
<td>Min-Max 27.6-65.3</td>
</tr>
</tbody>
</table>

Long-term $O_3$ exposure was related to first childhood asthma hospital admissions in a positive concentration-response relationship in a New York State birth cohort (Lin et al., 2008b). A separate hospitalization cross-sectional study in San Joaquin Valley, California reports similar findings (Meng et al., 2010). Another study relates asthma hospital admissions to quarterly average $O_3$ in the South Coast Air Basin of California (Moore et al., 2008).
Information from toxicological studies indicates that long term exposure to O₃ during gestation or development can result in irreversible morphological changes in the lung, which in turn can influence the function of the respiratory tract. Studies by Plopper and colleagues using an allergic asthma model have demonstrated changes in pulmonary function and airway morphology in adult and infant nonhuman primates repeatedly exposed to environmentally relevant concentrations of O₃ (Fanucchi et al., 2006; Joad et al., 2006; Schelegle et al., 2003; Harkema et al., 1987b). This nonhuman primate evidence of an O₃-induced change in airway responsiveness supports the biologic plausibility of long term exposure to O₃ contributing to effects of asthma in children. Results from epidemiologic studies examining long-term O₃ exposure and pulmonary function effects are inconclusive with some new studies relating effects at higher exposure levels. The definitive 8-year follow-up analysis of the first cohort of the CHS, which is discussed in Section 7.2 (Gauderman et al., 2004), provided little evidence that long-term exposure to ambient O₃ was associated with significant deficits in the growth rate of lung function in children. Other cross-sectional studies provide mixed results.

Several studies (see Table 7-3) provide results adjusted for potential confounders, presenting results for both O₃ and PM (single and multipollutant models) as well as other pollutants where PM effects were not provided. As shown in the table, O₃ associations are generally robust to adjustment for potential confounding by PM.
Table 7-3  Studies providing evidence concerning potential confounding by PM for available endpoints.

<table>
<thead>
<tr>
<th>Study and Endpoint</th>
<th>Exposure</th>
<th>Single Pollutant O₃</th>
<th>Single Pollutant PM</th>
<th>O₃ with PM</th>
<th>PM with O₃</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Asthma Related Health Effect Endpoint</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Akinbami et al. (2010)</td>
<td>IQR 35.9-43.7 ppb</td>
<td>1.56 (1.15, 2.10)</td>
<td>PM₂.₅ 1.43 (0.98, 2.10)</td>
<td>Adjusted for  SO₂, PM₂.₅, PM₁₀ 1.86 (1.02-3.40)</td>
<td>PM₂.₅ 1.24 (0.70-2.21)</td>
</tr>
<tr>
<td>Asthma prevalence in children</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Hwang et al. (2005)</td>
<td>10 ppb O₃</td>
<td>1.138 (1.001, 1.293)</td>
<td>0.934 (0.909, 0.960)</td>
<td>PM₁₀ 1.253 (1.089, 1.442)</td>
<td>0.925 (0.899, 0.952)</td>
</tr>
<tr>
<td>Asthma risk in children</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Jacquemin et al. (2012)</td>
<td>IQR 25-38 ppb O₃ summer</td>
<td>1.69 (1.22, 2.34)</td>
<td>1.33 (1.06, 1.67)</td>
<td>PM₁₀ 1.50 (1.07, 2.11)</td>
<td>1.28 (1.06, 1.55)</td>
</tr>
<tr>
<td>Asthma control in adults</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lee et al. (2009b)</td>
<td>High O₃ &gt;50 ppb</td>
<td>1.42 (0.75, 2.70)</td>
<td>NA</td>
<td>No substantial differences PM₁₀, PM₂.₅</td>
<td>NA</td>
</tr>
<tr>
<td>Bronchitic symptoms asthmatics</td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Lin et al. (2008b)</td>
<td>IQR 2.5%</td>
<td>1.16 (1.15, 1.17)</td>
<td>NA</td>
<td>Air Quality Index 1.24 (1.23, 1.25)</td>
<td>NA</td>
</tr>
<tr>
<td>Asthma admissions in children</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Meng et al. (2007)</td>
<td>1 ppm</td>
<td>1.70 (0.91, 3.18)</td>
<td>PM₁₀ 2.06 (1.17, 3.61) women</td>
<td>Did not differ</td>
<td>NA</td>
</tr>
<tr>
<td>Asthma control</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Meng et al. (2010)</td>
<td>10 ppb</td>
<td>1.49 (1.05, 2.11)</td>
<td>PM₁₀ 1.29 (0.99, 1.69)</td>
<td>Did not differ</td>
<td>NA</td>
</tr>
<tr>
<td>Asthma ED visits, Hospitalization</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rage et al. (2009b)</td>
<td>IQR 28.5-33.9 ppb</td>
<td>2.53 (1.69, 3.79)</td>
<td>NA</td>
<td>No PM data Three pollutant (O₃, NO₂, SO₂) 2.74 (1.68, 4.48)</td>
<td>NA</td>
</tr>
<tr>
<td>Asthma severity in adults</td>
<td></td>
<td></td>
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<tr>
<td><strong>Other Respiratory Health Effect Endpoints</strong></td>
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</tr>
<tr>
<td>Karr et al. (2007)</td>
<td>10 ppb</td>
<td>0.92 (0.88, 0.96)</td>
<td>1.09 (1.04, 1.14)</td>
<td>PM₂.₅ 1.02 (0.94, 1.10)</td>
<td>1.09 (1.03, 1.15)</td>
</tr>
<tr>
<td>Bronchiolitis Hospitalization</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Parker et al. (2009)</td>
<td>10 ppb</td>
<td>1.24 (1.15, 1.34)</td>
<td>1.23 (1.04, 1.46)</td>
<td>Multipollutant 1.18 (1.09, 1.27)</td>
<td>1.29 (1.07, 1.56)</td>
</tr>
<tr>
<td>Respiratory allergy</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rojas-Martinez et al. (2007)</td>
<td>11.3 ppb IQR</td>
<td>-24 (-30, -19)</td>
<td>PM₁₀ IQR 36.4 µg/m³ -29(-36,-21)</td>
<td>-17 (-23, -12)</td>
<td>-24 (-31,-16)</td>
</tr>
<tr>
<td>FEV₁ (mL) Deficit Girls</td>
<td></td>
<td></td>
<td></td>
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</tr>
</tbody>
</table>

The highest quartile is shown for all results
NA = not available
There is limited evidence for an association between long-term exposure to ambient O₃ concentrations and respiratory mortality (Jerrett et al., 2009) and this effect was robust to the inclusion of PM₂.₅. The association between increased O₃ concentrations and increased risk of death from respiratory causes was insensitive to a number of different model specifications. Additionally, there is evidence that long-term exposure to O₃ is associated with mortality among individuals that had previously experienced an emergency hospital admission due to COPD (Zanobetti and Schwartz, 2011).

Taken together, the recent epidemiologic studies of respiratory health effects (including respiratory symptoms, new-onset asthma and respiratory mortality) combined with toxicological studies in rodents and nonhuman primates, provide biologically plausible evidence that there is likely to be a causal relationship between long-term exposure to O₃ and respiratory effects. The epidemiologic evidence includes studies that evaluate the relationship between long-term O₃ exposure and respiratory effects such as studies that demonstrate interactions between exercise or different genetic variants and long-term measures of O₃ exposure on new-onset asthma in children; and increased respiratory symptom effects in asthmatics. Additional studies of respiratory health effects and a study of respiratory mortality provide a collective body of evidence supporting these relationships. Studies considering other pollutants provide data suggesting that the effects related to O₃ are independent from potential effects of the other pollutants. Some studies provide evidence for a positive concentration-response relationship. Short-term studies provide supportive evidence with increases in respiratory symptoms and asthma medication use, hospital admissions and ED visits for all respiratory outcomes and asthma, and decrements in lung function in children. The recent epidemiologic and toxicological data base provides a compelling case to support the hypothesis that a relationship exists between long-term exposure to ambient O₃ and measures of respiratory health effects.

7.3 Cardiovascular Effects

7.3.1 Cardiovascular Disease

7.3.1.1 Cardiovascular Epidemiology

Long-term exposure to O₃ and its effects on cardiovascular morbidity were not considered in the 2006 O₃ AQCD (U.S. EPA, 2006b). However, recent studies have assessed the chronic effects of O₃ concentration on cardiovascular morbidity (Chuang et al., 2011; Forbes et al., 2009a; Chen et al., 2007a). The association between O₃ concentration and markers of lipid peroxidation and antioxidant capacity was examined among 120 nonsmoking healthy college students, aged 18-22 years, from the University of California, Berkeley (February—June 2002) (Chen et al.,...
By design, students were chosen from geographic areas so they had experienced different concentrations of \( O_3 \) over their lifetimes and during recent summer vacation in either greater Los Angeles (LA) or the San Francisco Bay Area (SF). A marker of lipid peroxidation, 8-isoprostane (8-iso-PGF) in plasma, was assessed. This marker is formed continuously under normal physiological conditions but has been found at elevated concentrations in response to environmental exposures. A marker of overall antioxidant capacity, ferric reducing ability of plasma (FRAP), was also measured. The lifetime average \( O_3 \) concentration estimates (from estimated monthly averages) did not show much overlap between the two geographic areas [median (range): LA, 42.9 ppb (28.5-65.3); SF, 26.9 ppb (17.6-33.5)].

Estimated lifetime average \( O_3 \) concentration was related to 8-iso-PGF \( [\beta = 0.025 \text{ (pg/mL)/8-h ppb } O_3, \ p = 0.0007] \). For the 17-ppb lifetime \( O_3 \) concentration difference between LA and SF participants, there was a 17.41-pg/mL (95% CI: 15.43, 19.39) increase in 8-iso-PGF. No evidence of association was observed between lifetime \( O_3 \) concentration and FRAP \( [\beta = -2.21 \text{ (pg/mL)/8-h ppb } O_3, \ p = 0.45] \). The authors note that \( O_3 \) was highly correlated with PM\(_{10-2.5}\) and NO\(_2\) in this study population; however, their inclusion in the \( O_3 \) models did not substantially modify the magnitude of the associations with \( O_3 \). Because the average lifetime concentration results were supported by shorter-term exposure period results from analyses considering \( O_3 \) concentrations up to 30 days prior to sampling, the authors conclude that persistent exposure to \( O_3 \) can lead to sustained oxidative stress and increased lipid peroxidation. However, because there was not much overlap in average lifetime \( O_3 \) concentration estimates between LA and SF, it is possible that the risk estimates involving the lifetime \( O_3 \) exposures could be confounded by unmeasured factors related to other differences between the two cities.

Forbes et al. (2009a) used the annual average exposures to assess the relationship between chronic ambient air pollution and levels of fibrinogen and C-reactive protein (CRP) in a cross-sectional study conducted in England. Data were collected from the Health Survey of England for 1994, 1998, and 2003. The sampling strategy was designed to obtain a representative sample of the English population; however, due to small group sizes, only data from white ethnic groups were analyzed. For analyses, the annual concentrations of \( O_3 \) were averaged for the year of data collection and the previous year with the exception of 1994 (because pollutant data were not available for 1993). Median \( O_3 \) concentrations were 26.7 ppb, 25.4 ppb, and 28 ppb for 1994, 1998, and 2003, respectively. Year specific adjusted effect estimates were created and combined in a meta-analysis. No evidence of association was observed for \( O_3 \) and levels of fibrinogen or CRP (e.g., the combined estimates for the percent change in fibrinogen and CRP for a 10 ppb increase in \( O_3 \) were -0.28 [95% CI: -2.43, 1.92] and -3.05 [95% CI: -16.10, 12.02], respectively).

A study was performed in Taiwan to examine the association between long-term \( O_3 \) concentrations and blood pressure and blood markers using the Social Environment and Biomarkers of Aging Study (SEBAS) (Chuang et al., 2011). Individuals included in the study were 54 years of age and older. The mean annual \( O_3 \) concentration during the study period was 22.95 ppb (SD 6.76 ppb). Positive associations were observed between \( O_3 \) concentrations and both systolic and diastolic blood pressure...
[changes in systolic and diastolic blood pressure were 21.51 mmHg (95% CI: 16.90, 26.13) and 20.56 mmHg (95% CI: 18.14, 22.97) per 8.95 ppb increase in O₃, respectively]. Increased O₃ concentrations were also associated with increased levels of total cholesterol, fasting glucose, hemoglobin A₁c, and neutrophils. No associations were observed between O₃ concentrations and triglyceride and IL-6 levels. The observed associations were reduced when other pollutants were added to the models. Further research will be important for understanding the effects, if any, of chronic O₃ exposure on cardiovascular morbidity risk.

### 7.3.1.2 Cardiovascular Toxicology

Three new studies have investigated the cardiovascular effects of long-term exposure to O₃ in animal models (see Table 7-4 for study details). In addition to the short-term exposure effects described in Section 6.3.3, a recent study found that O₃ exposure in genetically hyperlipidemic mice enhanced aortic atherosclerotic lesion area compared to air exposed controls (Chuang et al., 2009). Chuang et al. (2009) not only provided evidence for increased atherogenesis in susceptible mice, but also reported an elevated vascular inflammatory and redox state in wild-type mice and infant primates (Section 6.3.3). This study is compelling in that it identifies biochemical and cellular events responsible for transducing the airway epithelial reactions of O₃ into proinflammatory responses that are apparent in the extrapulmonary vasculature (Cole and Freeman, 2009).

Another recent study provides further evidence for increased vascular inflammation and oxidation and long term effects in the extrapulmonary space. Rats episodically exposed to O₃ for 16 weeks presented marked increases in gene expression of biomarkers of oxidative stress, thrombosis, vasoconstriction, and proteolysis (Kodavanti et al., 2011). Ozone exposure upregulated aortic mRNA expression of heme oxygenase-1 (HO-1), tissue plasminogen activator (tPA), plasminogen activator inhibitor-1 (PAI-1), von Willebrand factor (vWF), thrombomodulin, endothelial nitric oxide synthase (eNOS), endothelin-1 (ET-1), matrix metalloprotease-2 (MMP-2), matrix metalloprotease-3 (MMP-3), and tissue inhibitor of matrix metalloprotease-2 (TIMP-2). In addition, O₃ exposure depleted some cardiac mitochondrial phospholipid fatty acids (C16:0 and C18:1), which may be the result of oxidative modifications. The authors speculate that oxidatively modified lipids and proteins produced in the lung and heart promote vascular pathology through activation of lectin-like oxidized-low density lipoprotein receptor-1 (LOX-1). Activated LOX-1 induces expression of a number of the biomarkers induced by O₃ exposure and is considered pro-atherogenic. Both LOX-1 mRNA and protein were increased in mouse aorta after O₃ exposure. This study provides a possible pathway and further support to the observed O₃ induced atherosclerosis.

Vascular occlusion resulting from atherosclerosis can block blood flow through vessels causing ischemia. The restoration of blood flow or reperfusion can cause injury to the tissue from subsequent inflammation and oxidative damage. Ozone
exposure enhanced the sensitivity to myocardial ischemia-reperfusion (I/R) injury in rats while increasing oxidative stress levels and pro-inflammatory mediators and decreasing production of anti-inflammatory proteins (Perepu et al., 2010). Both long- and short-term O₃ exposure decreased the left ventricular developed pressure, rate of change of pressure development, and rate of change of pressure decay and increased left ventricular end diastolic pressure in isolated perfused hearts (Section 6.3.3 for short-term exposure discussion). In this ex vivo heart model, O₃ induced oxidative stress by decreasing SOD enzyme activity and increasing malondialdehyde levels. Ozone also elicited a proinflammatory state evident by an increase in TNF-α and a decrease in the anti-inflammatory cytokine IL-10. The authors conclude that O₃ exposure will result in a greater I/R injury.

Overall, the few animal studies that have been conducted suggest that long-term O₃ exposure may result in cardiovascular effects. These studies demonstrate O₃-induced atherosclerosis and injury. In addition, evidence is presented for a potential mechanism for the development of vascular pathology that involves increased oxidative stress and proinflammatory mediators, activation of LOX-1 by O₃ oxidized lipids and proteins, and upregulation of genes responsible for proteolysis, thrombosis, and vasoconstriction. Further discussion of the mechanisms that may lead to cardiovascular effects from O₃ exposure can be found in Section 5.3.8.

### Table 7-4 Characterization of study details for Section 7.3.1.2.

<table>
<thead>
<tr>
<th>Study</th>
<th>Model</th>
<th>O₃ (ppm)</th>
<th>Exposure Duration</th>
<th>Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chuang et al. (2009)</td>
<td>Mice; ApoE−/−; M; 6 weeks</td>
<td>0.5</td>
<td>8 wks, 5 days/week, 8 h/day</td>
<td>Enhanced aortic atherosclerotic lesion area compared to air controls.</td>
</tr>
<tr>
<td>Kodavanti et al. (2011)</td>
<td>Rat; Wistar; M; 10-12 weeks</td>
<td>0.4</td>
<td>16 wks, 1 day/week, 5 h/day</td>
<td>Increased vascular inflammation and oxidative stress, possibly through activation of LOX-1 signaling.</td>
</tr>
<tr>
<td>Perepu et al. (2010)</td>
<td>Rat; Sprague-Dawley; Weight: 50-75 g</td>
<td>0.8</td>
<td>56 days, 8 h/day</td>
<td>Enhanced the sensitivity to myocardial I/R injury while increasing oxidative stress and pro-inflammatory mediators and decreasing production of anti-inflammatory proteins.</td>
</tr>
</tbody>
</table>

No previous studies investigated cardiovascular effects from long-term exposure to O₃. For details, see Section 7.3.1.2.

### 7.3.2 Cardiovascular Mortality

A limited number of epidemiologic studies have assessed the relationship between long-term exposure to O₃ and mortality. The 2006 O₃ AQCD concluded that an insufficient amount of evidence existed “to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans” (U.S. EPA, 2006b). Though total and cardio-pulmonary mortality were considered in these studies, cardiovascular mortality was not specifically considered. In the most recent follow-up analysis of the ACS cohort (Jerrett et al., 2009), cardiopulmonary deaths were
subdivided into respiratory and cardiovascular, separately, as opposed to combined in the Pope et al. (2002) work. A 10-ppb increment in exposure to O\textsubscript{3} elevated the risk of death from the cardiopulmonary, cardiovascular, and ischemic heart disease. Inclusion of PM\textsubscript{2.5} as a copollutant attenuated the association with exposure to O\textsubscript{3} for all of the cardiovascular endpoints to become null. Additionally, a recent study (Zanobetti and Schwartz, 2011) observed an association between long-term exposure to O\textsubscript{3} and elevated risk of mortality among Medicare enrollees that had previously experienced an emergency hospital admission due to congestive heart failure (CHF) or myocardial infarction (MI).

**7.3.3 Summary and Causal Determination**

Previous AQCDs did not address the cardiovascular effects of long-term O\textsubscript{3} exposure due to limited data availability. The evidence remains limited; however the emerging data are supportive of a role for O\textsubscript{3} in chronic cardiovascular diseases. Few epidemiologic studies have investigated cardiovascular morbidity after long-term O\textsubscript{3} exposure, and the majority only assessed cardiovascular disease related biomarkers. The studies used annual or multi-year averages of air monitoring data for exposure assessment. As described in Section 4.6, this exposure assignment method is typical of long-term epidemiologic studies, and analyses suggest that annual average concentrations are representative of exposure metrics accounting for residential mobility. A study on O\textsubscript{3} and cardiovascular mortality reported no association after adjustment for PM\textsubscript{2.5} levels. Further epidemiologic studies on cardiovascular morbidity and mortality after long-term exposure have not been published.

Toxicological evidence on long-term O\textsubscript{3} exposure is also limited but three strong toxicological studies have been published since the previous AQCD. These studies provide evidence for O\textsubscript{3} enhanced atherosclerosis and I/R injury, corresponding with development of a systemic oxidative, proinflammatory environment. Further discussion of the mechanisms that may lead to cardiovascular effects can be found in Section 5.3.8. Although questions exist for how O\textsubscript{3} inhalation causes systemic effects, a recent study proposes a mechanism for development of vascular pathology that involves activation of LOX-1 by O\textsubscript{3} oxidized lipids and proteins. This activation may also be responsible for O\textsubscript{3} induced changes in genes involved in proteolysis, thrombosis, and vasoconstriction. Taking into consideration the findings of toxicological studies, and the emerging evidence from epidemiologic studies, the generally limited body of evidence is suggestive of a causal relationship between long-term exposures to O\textsubscript{3} and cardiovascular effects.

**7.4 Reproductive and Developmental Effects**

Although the body of literature characterizing the health effects associated with exposure to O\textsubscript{3} is large and continues to grow, the research focusing on adverse birth outcomes is relatively small. Among these studies, various measures of birth weight
and fetal growth, such as low birth weight (LBW), small for gestational age (SGA), and intrauterine growth restriction (IUGR), and preterm birth (<37-week gestation; [PTB]) have received more attention in air pollution research, while congenital malformations are less studied. There are also recent studies on reproductive and developmental effects and infant mortality.

A major issue in studying environmental exposures and reproductive and developmental effects (including infant mortality) is selecting the relevant exposure period, since the biological mechanisms leading to these outcomes and the critical periods of exposure are poorly understood. To account for this, many epidemiologic studies evaluate multiple exposure periods, including long-term (months to years) exposure periods, such as entire pregnancy, individual trimesters or months of pregnancy, and short-term (days to weeks) exposure periods such as the days and weeks immediately preceding birth. Due to the length of gestation in rodents (18-24 days, on average), animal toxicological studies investigating the effects of O₃ generally utilize short-term exposure periods. Thus, an epidemiologic study that uses the entire pregnancy as the exposure period is considered to have a long-term exposure period (about 40 weeks, on average), while a toxicological study conducted with rats that also uses the entire pregnancy as the exposure period is considered to have a short-term exposure period (about 18-24 days, on average). In order to characterize the weight of evidence for the effects of O₃ on reproductive and developmental effects in a consistent, cohesive and integrated manner, results from both short-term and long-term exposure periods are included in this section and are identified accordingly in the text and tables throughout this section.

Due to the poorly understood biological mechanisms and uncertainty regarding relevant exposure studies, all of the studies of reproductive and developmental outcomes, including infant mortality, are evaluated in this section. Infant development processes, much like fetal development processes, may be particularly sensitive to O₃-induced health effects. Exposures proximate to the effect may be most relevant if exposure causes an acute effect. However, exposure occurring in early life might affect critical growth and development, with results observable later in the first year of life, or cumulative exposure during the first year of life may be the most important determinant. In dealing with the uncertainties surrounding these issues, studies have considered several exposure metrics based on different periods of exposure, including both short- and long-term exposure periods. In the toxicological literature, a challenge in interpreting data from studies that use very young murine pups, is that pups can have differential exposure to O₃ doses, versus their respective dams, because of the physiology and behavior associated with the early postnatal period. Namely, young pups tend to nuzzle close to their mothers and are often housed in cages with litter used in nest formation. Both the dam's fur and the bedding can absorb and react with O₃, decreasing the dose that a young animal might receive. The reproductive and developmental studies are characterized in this chapter, as they contribute to the weight of evidence for an effect of O₃ on reproductive and developmental effects.
Infants and fetal development processes may be particularly at-risk for O₃-induced health effects, and although the physical mechanisms are not fully understood, several hypotheses have been proposed; these include: oxidative stress, systemic inflammation, vascular dysfunction and impaired immune function (Section 5.3). Study of these outcomes can be difficult given the need for detailed exposure data and potential residential movement of mothers during pregnancy. Air pollution epidemiologic studies reviewed in the 2006 O₃ AQCD (U.S. EPA, 2006b) examined impacts on birth-related endpoints, including intrauterine, perinatal, postneonatal, and infant deaths; premature births; intrauterine growth retardation; very low birth weight (weight <1,500 grams) and low birth weight (weight <2,500 grams); and birth defects. However, in the limited number of studies that investigated O₃, no associations were found between O₃ and birth outcomes, with the possible exception of birth defects.

Several recent articles have reviewed methodological issues relating to the study of outdoor air pollution and adverse birth outcomes (Chen et al., 2010a; Woodruff et al., 2009; Ritz and Wilhelm, 2008; Slama et al., 2008). Some of the key challenges to interpretation of these study results include the difficulty in assessing exposure as most studies use existing monitoring networks to estimate individual exposure to ambient air pollution; the inability to control for potential confounders such as other risk factors that affect birth outcomes (e.g., smoking); evaluating the exposure window (e.g., trimester) of importance; and limited evidence on the physiological mechanism of these effects (Ritz and Wilhelm, 2008; Slama et al., 2008).

Overall, the evidence for an association between exposure to ambient O₃ and reproductive and developmental outcomes is growing, yet remains relatively small. Recently, an international collaboration was formed to better understand the relationships between air pollution and adverse birth outcomes and to examine some of these methodological issues through standardized parallel analyses in datasets from different countries (Woodruff et al., 2010). Initial results from this collaboration have examined PM and birth weight (Parker et al., 2011); work on O₃ has not yet been performed. Although early animal studies (Kavlock et al., 1980) found that exposure to O₃ in the late gestation of pregnancy in rats led to some abnormal neurological and behavioral performances for neonates, to date human studies have reported inconsistent results for the association of ambient O₃ concentrations and birth outcomes.

7.4.1 Effects on Sperm

A limited amount of research has been conducted to examine the association between air pollution and male reproductive outcomes, specifically semen quality. To date, the epidemiologic studies have considered various exposure durations before semen collection that encompass either the entire period of spermatogenesis (i.e., 90 days) or key periods of sperm development that correspond to epididymal storage, development of sperm motility, and spermatogenesis. In an analysis conducted as
part of the Teplice Program, 18-year-old men residing in the heavily polluted district of Teplice in the Czech Republic were found to be at greater risk of having abnormalities in sperm morphology and chromatin integrity than men of similar age residing in Prachaticce, a less polluted district (Selevan et al., 2000; Sram et al., 1999). A follow-up longitudinal study conducted on a subset of the same men from Teplice revealed associations between total episodic air pollution and abnormalities in sperm chromatin (Rubes et al., 2005). A limitation of these studies is that they did not identify specific pollutants or their concentrations.

More recent epidemiologic studies conducted in the U.S. have also reported associations between ambient air pollution and sperm quality for individual air pollutants, including O3 and PM2.5. In a repeated measures study in Los Angeles, CA, Sokol et al. (2006) reported a reduction in average sperm concentration during three exposure windows (short-term exposures of 0-9, 10-14, and 70-90 days before semen collection, as well as long-term exposures of 0-90 days before semen collection) associated with high ambient levels of O3 in healthy sperm donors. This effect persisted under a joint additive model for O3, CO, NO2 and PM10. The authors did not detect a reduction in sperm count. Hansen et al. (2010) investigated the effect of exposure to O3 and PM2.5 (using the same exposure windows used by Sokol et al. (2006) on sperm quality in three southeastern counties (Wake County, NC; Shelby County, TN; Galveston County, TX). Outcomes included sperm concentration and count, morphology, DNA integrity and chromatin maturity. Overall, the authors found both protective and adverse effects, although some results suggested adverse effects on sperm concentration, count and morphology.

The biological mechanisms linking ambient air pollution to decreased sperm quality have yet to be determined, though O3-induced oxidative stress, inflammatory reactions, and the induction of the formation of circulating toxic species have been suggested as possible mechanisms (see Section 5.3.8). Decremental effects on testicular morphology have been demonstrated in a toxicological study with histological evidence of O3-induced depletion of germ cells in testicular tissue and decreased seminiferous tubule epithelial layer. Jedlinska-Krakowska et al. (2006) demonstrated histopathological evidence of impaired spermatogenesis (round spermatids/ spermatocytes, giant spermatid cells, and focal epithelial desquamation with denudation to the basement membrane). The exposure protocol used five-month-old adult rats exposed to O3 as adults (long-term exposure, 0.5 ppm, 5 h/day for 50 days). This degeneration could be rescued by vitamin E administration, indicating an antioxidant effect. Vitamin C administration had no effect at low doses of ascorbic acid and exacerbated the O3-dependent damage at high doses, as would be expected as vitamin C can be a radical generator instead of an antioxidant at higher doses. In summary, this study provided toxicological evidence of impaired spermatogenesis with O3 exposure that was rescued with certain antioxidant supplementation.
Overall, there is limited epidemiologic evidence for an association with O\textsubscript{3} concentration and decreased sperm concentration. A recent toxicological study provides limited evidence for a possible biological mechanism (histopathology showing impaired spermatogenesis) for such an association.

### 7.4.2 Effects on Reproduction

Evidence suggests that exposure to air pollutants during pregnancy may be associated with adverse birth outcomes, which has been attributed to the increased sensitivity of the fetus due to physiologic immaturity. Gametes (i.e., ova and sperm) may be even more at-risk, especially outside of the human body, as occurs with assisted reproduction. Smokers require twice the number of in vitro fertilization (IVF) attempts to conceive as non-smokers (Feichtinger et al., 1997), suggesting that a preconception exposure can be harmful to pregnancy. A recent study used an established national-scale, log-normal kriging method to spatially estimate daily mean concentrations of criteria pollutants at addresses of women undergoing their first IVF cycle and at their IVF labs from 2000 to 2007 in the northeastern U.S. (Legro et al., 2010). Increasing O\textsubscript{3} concentration at the patient’s address during ovulation induction (short-term exposure, ~12 days) was significantly associated with an increased chance of live birth (OR = 1.13, [95% CI: 1.05, 1.22] per 10 ppb increase), but with decreased odds of live birth when exposed from embryo transfer to live birth (long-term exposure, ~200 days) (OR = 0.79, [95% CI: 0.69, 0.90] per 10 ppb increase). After controlling for NO\textsubscript{2} in a copollutant model, however, O\textsubscript{3} was no longer significantly associated with IVF failure. The results of this study suggest that short-term exposure to O\textsubscript{3} during ovulation was beneficial (perhaps due to early conditioning to O\textsubscript{3}), whereas long-term exposure to O\textsubscript{3} (e.g., during gestation) was detrimental, and reduced the likelihood of a live birth.

In most toxicological studies, reproductive success appears to be unaffected by O\textsubscript{3} exposure. Nonetheless, one study has reported that 25% of the BALB/c mouse dams in the highest O\textsubscript{3} exposure group (1.2 ppm, short-term exposure GD9-18) did not complete a successful pregnancy, a significant reduction (Sharkhuu et al., 2011). Ozone administration (continuous 0.4, 0.8 or 1.2 ppm O\textsubscript{3}) to CD-1 mouse dams during the majority of pregnancy (short-term exposure, PD7-17, which excludes the pre-implantation period), led to no adverse effects on reproductive success (proportion of successful pregnancies, litter size, sex ratio, frequency of still birth, or neonatal mortality) (Bignami et al., 1994). There was a nearly statistically significant increase in pregnancy duration (0.8 and 1.2 ppm O\textsubscript{3}). Initially, dam body weight (0.8 and 1.2 ppm O\textsubscript{3}), water consumption (0.4, 0.8 and 1.2 ppm O\textsubscript{3}) and food consumption (0.4, 0.8 and 1.2 ppm O\textsubscript{3}) during pregnancy were decreased with O\textsubscript{3} exposure but these deficits dissipated a week or two after the initial exposure (Bignami et al., 1994). The anorexigenic effect of O\textsubscript{3} exposure on the pregnant dam appears to dissipate with time; the dams seem to adapt to the O\textsubscript{3} exposure. In males, data exist showing morphological evidence of altered spermatogenesis in O\textsubscript{3} exposed animals (Jedlinska-Krakowska et al., 2006). Some evidence suggests that O\textsubscript{3} may
affect reproductive success when combined with other chemicals. Kavlock et al. (1979) showed that O₃ acted synergistically with sodium salicylate to increase the rate of pup resorptions after midgestational exposure (1.0 ppm O₃, short-term exposure, GD9-GD12). At low concentrations of O₃ exposure, toxicological studies show reproductive effects to include a transient anorexigenic effect of O₃ on gestational weight gain, and a synergistic effect of O₃ on salicylate-induced pup resorptions; other fecundity, pregnancy- and gestation-related outcomes appear unaffected by O₃ exposure.

Collectively, there is very little epidemiologic evidence for the effect of short- or long-term exposure to O₃ on reproductive success, and the reproductive success in rats appears to be unaffected in toxicological studies of short-term exposure to O₃.

### 7.4.3 Birth Weight

With birth weight routinely collected in vital statistics and being a powerful predictor of infant mortality, it is the most studied outcome within air pollution-birth outcome research. Air pollution researchers have analyzed birth weight as a continuous variable and/or as a dichotomized variable in the form of LBW (<2,500 g [5 lbs, 8 oz]).

Birth weight is primarily determined by gestational age and intrauterine growth, but also depends on maternal, placental and fetal factors as well as on environmental influences. In both developed and developing countries, LBW is the most important predictor for neonatal mortality and is a significant determinant of postneonatal mortality and morbidity. Studies report that infants who are smallest at birth have a higher incidence of diseases and disabilities, which continue into adulthood (Hack and Fanaroff, 1999).

The strongest evidence for an effect of O₃ on birth weight comes from the Children’s Health Study conducted in southern California. In this study, Salam et al. (2005) report that maternal exposure to 24-h avg O₃ concentrations averaged over the entire pregnancy was associated with reduced birth weight (39.3 g decrease [95% CI: -55.8, -22.8] in birth weight per 10 ppb and 8-h avg (19.2-g decrease [95% CI: -27.7, -10.7] in birth weight per 10 ppb). This effect was stronger for concentrations averaged over the second and third trimesters. PM₁₀, NO₂ and CO concentrations averaged over the entire pregnancy were not statistically significantly associated with birth weight, although CO concentrations in the first trimester and PM₁₀ concentrations in the third trimester were associated with a decrease in birth weight. Additionally, the authors observed a concentration-response relationship of birth weight with 24-h avg O₃ concentrations averaged over the entire pregnancy that was clearest above the 30-ppb level (see Figure 7-4). Relative to the lowest decile of 24-h avg O₃, estimates for the next 5 lowest deciles were approximately -40 g to -50 g, with no clear trend and with 95% confidence bounds that included zero. The highest four deciles of O₃ exposure showed an approximately linear decrease in birth weight, and all four 95% CIs
excluded zero, and ranged from mean decreases of 74 grams to decreases of 148 grams.

Figure 7-4 Birthweight deficit by decile of 24-h avg $O_3$ concentration averaged over the entire pregnancy compared with the decile group with the lowest $O_3$ exposure.

Several additional studies conducted in the U.S. and Canada also investigated the association between ambient $O_3$ concentrations and birth weight and report some weak evidence for an association. Morello-Frosch et al. (2010) estimated ambient $O_3$ concentrations throughout pregnancy and for each trimester in the neighborhoods of women who delivered term singleton births between 1996 and 2006 in California. A 10-ppb increase in the $O_3$ concentration averaged across the entire pregnancy was associated with a 5.7-g decrease (95% CI: -6.6, -4.9) in birth weight when exposures were calculated using monitors within 10 km of the maternal address at date of birth. When the distance from the monitor was restricted to 3 km, the decrease in birth weight associated with a 10-ppb increase in $O_3$ concentration was 8.9 g (95% CI: -10.6, -7.1). These results persisted in copollutant models and in models that stratified by trimester of exposure, SES, and race. Darrow et al. (2011b) did not observe an association with birth weight and $O_3$ concentrations during two exposure periods of interest (i.e., the first month and last trimester), but did find an association
with reduced birth weight when examining the cumulative air pollution concentration during the entire pregnancy period. Additionally, they observed effect modification by race and ethnicity, such that associations between birth weight and third-trimester O\textsubscript{3} concentrations were significantly stronger in Hispanics and non-Hispanic African Americans than in non-Hispanic whites. Chen et al. (2002) used 8-h avg O\textsubscript{3} concentrations to create exposure variables based on average maternal exposure for each trimester. Ozone was not found to be related to birth weight in single-pollutant models, though the O\textsubscript{3} effect during the third trimester was borderline statistically significant in a copollutant model with PM\textsubscript{10}.

Several studies found no association between ambient O\textsubscript{3} concentrations and birth weight. Wilhelm and Ritz (2005) extended previous analyses of term LBW (Ritz et al., 2000; Ritz and Yu, 1999) to include the period 1994-2000. The authors examined varying residential distances from monitoring stations to see if the distance affected risk estimation, exploring the possibility that effect attenuation may result from local pollutant heterogeneity inadequately captured by ambient monitors. As in their previous studies, the authors observed associations between elevated concentrations of CO and PM\textsubscript{10} both early and late in pregnancy and risk of term LBW. After adjusting for CO and/or PM\textsubscript{10} the authors did not observe associations between O\textsubscript{3} and term LBW in any of their models. Brauer et al. (2008) evaluated the impacts of air pollution (CO, NO\textsubscript{2}, NO, O\textsubscript{3}, SO\textsubscript{2}, PM\textsubscript{2.5}, PM\textsubscript{10}) on birth weight for the period 1999-2002 using spatiotemporal residential exposure metrics by month of pregnancy in Vancouver, BC. Quantitative results were not presented for the association between O\textsubscript{3} and LBW, though the authors observed associations that were largely protective. Dugandzic et al. (2006) examined the association between LBW and ambient levels of air pollutants by trimester of exposure among a cohort of term singleton births from 1988-2000. Though there was some indication of an association with SO\textsubscript{2} and PM\textsubscript{10}, there were no effects for O\textsubscript{3}.

Similarly, studies conducted in Australia, Latin America, and Asia report limited evidence for an association between ambient O\textsubscript{3} and measures of birth weight. In Sydney, Australia, Mannes et al. (2005) found that O\textsubscript{3} concentrations in the second trimester of pregnancy had small adverse effects on birth weight (7.5-g decrease; [95% CI: -13.8, 1.2] per 10 ppb), although this effect disappeared when the analysis was limited to births with a maternal address within 5 km of a monitoring station (87.7-g increase; [95% CI: 10.5, 164.9] per 10 ppb). Hansen et al. (2007) reported that trimester and monthly specific exposures to all pollutants were not statistically significantly associated with a reduction in birth weight in Brisbane, Australia. In Sao Paulo, Brazil, Gouveia et al. (2004) found that O\textsubscript{3} exhibited a small inverse relation with birth weight over the third trimester (6.0-g decrease; [95% CI: -30.8, 18.8] per 10 ppb). Lin et al. (2004b) reported a positive, though not statistically significant, exposure-response relationship for O\textsubscript{3} during the entire pregnancy in a Taiwanese study. In a study performed in Korea, Ha et al. (2001) reported no O\textsubscript{3} effect during the first trimester of pregnancy, but they found that during the third trimester of pregnancy O\textsubscript{3} was associated with LBW (RR = 1.05 [95% CI: 1.02, 1.08] per 10 ppb).
Table 7-5  Brief summary of epidemiologic studies of birth weight.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Sample Size</th>
<th>Mean O₃ (ppb)</th>
<th>Exposure assessment</th>
<th>Effect Estimateᵃ</th>
</tr>
</thead>
<tbody>
<tr>
<td>Salam et al. (2005)</td>
<td>California, U.S.</td>
<td>(n = 3,901)</td>
<td>24-h avg:</td>
<td>ZIP code level</td>
<td>Entire pregnancy:</td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>27.3</td>
<td></td>
<td>-39.3 g (-55.8, -22.8)</td>
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<td></td>
<td></td>
<td></td>
<td>8 h:</td>
<td></td>
<td>T1: -6.1 g (-16.8, 4.8)</td>
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<td></td>
<td></td>
<td></td>
<td>50.6</td>
<td></td>
<td>T2: -20.0 g (-31.7, -8.4)</td>
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<td></td>
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<td>T3: -20.7 g (-32.1, -9.3)</td>
</tr>
<tr>
<td>Morello-Frosch et al.</td>
<td>California, U.S.</td>
<td>(n = 3,545,177)</td>
<td>24-h avg:</td>
<td>Nearest Monitor (within 10, 5, 3 km)</td>
<td>Entire pregnancy:</td>
</tr>
<tr>
<td>(2010)</td>
<td></td>
<td></td>
<td>23.5</td>
<td></td>
<td>-5.7 g (-6.6, -4.9)</td>
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<td>T1: -2.1 g (-2.9, -1.4)</td>
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<td></td>
<td>T2: -2.3 g (-3.1, -1.5)</td>
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<td></td>
<td>T3: -1.3 g (-2.1, -0.6)</td>
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<tr>
<td>Darrow et al. (2011b)</td>
<td>Atlanta, GA</td>
<td>(N = 406,627)</td>
<td>8-h max:</td>
<td>Population-weighted spatial average</td>
<td>Entire pregnancy:</td>
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<td></td>
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<td>44.8</td>
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<td>-12.3 g (-17.8, -6.8)</td>
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<td>First 28 days</td>
<td>-0.5 g (-3.0, 2.1)</td>
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<td></td>
<td>T3: -0.9 g (-4.5, 2.8)</td>
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<tr>
<td>Chen et al. (2002)</td>
<td>Northern Nevada, U.S.</td>
<td>(n = 36,305)</td>
<td>8-h:</td>
<td>County level</td>
<td>Entire pregnancy:</td>
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<td></td>
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<td>27.2</td>
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<td>20.9 g (6.3, 35.5)</td>
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<td>T2: -19.4 g (-77.0, 38.2)</td>
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<td>T3: 7.7 g (-50.9, 66.3)</td>
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<tr>
<td>Wilhelm and Ritz (2005)</td>
<td>Los Angeles County, CA</td>
<td>(n = 136,134)</td>
<td>1-h:</td>
<td>Varying distances from monitor</td>
<td>T1: NR</td>
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<tr>
<td></td>
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<td>21.1-22.2</td>
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<td>T3: NR</td>
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<td></td>
<td>6 weeks before birth:</td>
<td>NR</td>
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<td>Brauer et al. (2008)</td>
<td>Vancouver, BC, Canada</td>
<td>(n = 70,249)</td>
<td>24-h avg:</td>
<td>Nearest Monitor (within 10 km)</td>
<td>Entire pregnancy:</td>
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<td>First 30 days of pregnancy: NR</td>
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<td>Last 30 days of pregnancy: NR</td>
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<td>T1: NR</td>
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<td>T3: NR</td>
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<tr>
<td>Dugandzic et al. (2006)</td>
<td>Nova Scotia, Canada</td>
<td>(n = 74,284)</td>
<td>24-h avg:</td>
<td>Nearest Monitor (within 25 km)</td>
<td>T1: 0.97 (0.81, 1.18)ᵃ</td>
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<td>21</td>
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<td>T2: 1.06 (0.87, 1.27)ᵃ</td>
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<td></td>
<td>T3: 1.01 (0.83-1.24)ᵃ</td>
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<tr>
<td>Mannes et al. (2005)</td>
<td>Sydney, Australia</td>
<td>(n = 138,056)</td>
<td>1-h max:</td>
<td>Citywide avg and &lt;5 km from monitor</td>
<td>T1: -0.9 g (-6.6, 4.8)</td>
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<td></td>
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<td>31.6</td>
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<td>T2: -7.5 g (-13.8, 1.2)</td>
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<td>T3: -4.5 g (-10.8, 1.8)</td>
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<td>Last 30 days</td>
<td>-1.1 g (-5.6, 3.4)</td>
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<td>Hansen et al. (2007)</td>
<td>Brisbane, Australia</td>
<td>(n = 26,617)</td>
<td>8 h max:</td>
<td>Citywide avg</td>
<td>T1: 2.8 g (-10.5, 16.0)</td>
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<td>26.7</td>
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<td>T2: 4.4 g (-11.4, 20.1)</td>
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<td>T3: 11.3 g (-4.4, 27.1)</td>
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<td>Gouveia et al. (2004)</td>
<td>Sao Paulo, Brazil</td>
<td>(n = 179,460)</td>
<td>1-h max:</td>
<td>Citywide avg</td>
<td>T1: -3.2 g (-25.6, 19)</td>
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<td>31.5</td>
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<td>T2: -0.2 g (-23.8, 23.4)</td>
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</tbody>
</table>
Table 7-5 provides a brief overview of the epidemiologic studies of birth weight. In summary, only the Children’s Health Study conducted in southern California (Salam et al., 2005) provides strong evidence for an effect of ambient O₃ on birth weight. The study by Morello-Frosch et al. (2010), also conducted in California, provides support for the results of the Children’s Health Study. Additional studies, conducted in the U.S., Canada, Australia, Latin America, and Asia, provide limited and inconsistent evidence to support the effect reported in the Children’s Health Study. The toxicological literature on the effect of O₃ on birth weight is sparse. In some studies, the reporting of birth weight may be avoided because birth weight can be confounded by decreased litter size resulting from an increased rate of pup resorption (aborted pups) in O₃ exposed dams. In one toxicological study by Haro and Paz (1993), no differences in litter size were observed and decreased birth weight in pups from dams who were exposed to 1ppm O₃ during pregnancy (short-term exposure, ~22 days) was reported. A second animal toxicology study recapitulated these findings with pregnant BALB/c mice exposed to O₃ (1.2 ppm, short-term exposure, GD9-18) that produced pups with significantly decreased birth weights (Sharkhuu et al., 2011).

7.4.4 Preterm Birth

Preterm birth (PTB) is a syndrome (Romero et al., 2006) that is characterized by multiple etiologies. It is therefore unusual to be able to identify an exact cause for each PTB. In addition, PTB is not an adverse outcome in itself, but an important determinant of health status (i.e., neonatal morbidity and mortality). Although some overlap exists for common risk factors, different etiologic entities related to distinct risk factor profiles and leading to different neonatal and postneonatal complications are attributed to PTB and measures of fetal growth. Although both restricted fetal growth and PTB can result in LBW, prematurity does not have to result in LBW or growth restricted babies.
A major issue in studying environmental exposures and PTB is selecting the relevant exposure period, since the biological mechanisms leading to PTB and the critical periods of vulnerability are poorly understood (Bobak, 2000). Short-term exposures proximate to the birth may be most relevant if exposure causes an acute effect. However, exposure occurring in early gestation might affect placentation, with results observable later in pregnancy, or cumulative exposure during pregnancy may be the most important determinant. The studies reviewed have dealt with this issue in different ways. Many have considered several exposure metrics based on different periods of exposure. Often the time periods used are the first month (or first trimester) of pregnancy and the last month (or 6 weeks) prior to delivery. Using a time interval prior to delivery introduces an additional problem since cases and controls are not in the same stage of development when they are compared. For example, a preterm infant delivered at 36 weeks is a 32-week fetus 4 weeks prior to birth, while an infant born at term (40 weeks) is a 36-week fetus 4 weeks prior to birth.

Recently, investigators have examined the association of PTB with both short-term (i.e., hours, days, or weeks) and long-term (i.e., months or years) exposure periods. Time-series studies have been used to examine the association between air pollution concentrations during the days immediately preceding birth. An advantage of these time-series studies is that this approach can remove the influence of covariates that vary across individuals over a short period of time. Retrospective cohort and case-control studies have been used to examine long-term exposure periods, often averaging air pollution concentrations over months or trimesters of pregnancy.

Studies of PTB fail to show consistency in pollutants and periods during pregnancy when an effect occurs. For example, while some studies find the strongest effects associated with exposures early in pregnancy, others report effects when the exposure is limited to the second or third trimester. However, the effect of air pollutant exposure during pregnancy on PTB has a biological basis. There is an expanding list of possible mechanisms that may explain the association between O3 exposure and PTB (see Section 5.4.2.4).

Many studies of PTB compare exposure in quartiles, using the lowest quartile as the reference (or control) group. No studies use a truly unexposed control group. If exposure in the lowest quartile confers risk, than it may be difficult to demonstrate additional risk associated with a higher quartile. Thus negative studies must be interpreted with caution.

Preterm birth occurs both naturally (idiopathic PTB), and as a result of medical intervention (iatrogenic PTB). Ritz et al. (2007); (2000) excluded all births by Cesarean section to limit their studies to idiopathic PTB. No other studies attempted to distinguish the type of PTB, although air pollution exposure maybe associated with only one type. This is a source of potential effect misclassification.

Generally, studies of air pollution and birth outcomes conducted in North America and the United Kingdom have not identified an association between PTB and maternal exposure to O3. Most recently, Darrow et al. (2009) used vital record data
to construct a retrospective cohort of 476,489 births occurring between 1994 and 2004 in 5 central counties of metropolitan Atlanta, GA. Using a time-series approach, the authors examined aggregated daily counts of PTB in relation to ambient levels of CO, NO₂, SO₂, O₃, PM₁₀, PM₂.₅ and speciated PM measurements. This study investigated 3 gestational windows of short- and long-term exposure: the final week of gestation (short-term exposure), and the first month of gestation and the final 6 weeks of gestation (long-term exposure). The authors did not observe associations of PTB with O₃ concentrations for any of the exposure periods.

A number of U.S. studies were conducted in southern California, and report somewhat inconsistent results. Ritz et al. (2000) evaluated the effect of air pollution (CO, NO₂, O₃, PM₁₀) exposure during pregnancy on the occurrence of PTB in a cohort of 97,518 neonates born in southern California between 1989 and 1993. The authors use both short- and long-term exposure windows, averaging pollutant measures taken at the closest air-monitoring station over distinct periods, such as 1, 2, 4, 6, 8, 12, and 26 weeks before birth and the whole pregnancy period. Additionally, they calculated average exposures for the first and second months of pregnancy. The authors found no consistent effects associated with O₃ concentration over any of the pregnancy periods in single or multipollutant models. Wilhelm and Ritz (2005) extended previous analyses of PTB (Ritz et al., 2000; Ritz and Yu, 1999) in California to include 1994-2000. The authors examined varying residential distances from monitoring stations to see if the distance affected risk estimation, because effect attenuation may result from local pollutant heterogeneity inadequately captured by ambient monitors. The authors analyzed the association between long-term O₃ exposure during varying periods of pregnancy and PTB, finding a positive association between O₃ levels in both the first trimester of pregnancy (RR = 1.23 [95% CI: 1.06, 1.42] per 10 ppb increase in 24-h avg O₃) and the first month of pregnancy (results for first trimester exposure were similar, but slightly smaller, quantitative results not presented) in models containing all pollutants. No association was observed between O₃ in the 6 weeks before birth and preterm delivery. Finally, Ritz et al. (2007) conducted a case-control survey nested within a birth cohort and assessed the extent to which residual confounding and exposure misclassification impacted air pollution effect estimates. The authors calculated mean long-term exposure levels for three gestational periods: the entire pregnancy, the first trimester, and the last 6 weeks before delivery. Though positive associations were observed for CO and PM₂.₅, no consistent patterns of increase in the odds of PTB for O₃ or NO₂ were observed.

A study conducted in Canada evaluated the impacts of air pollution (including CO, NO₂, NO, O₃, SO₂, PM₂.₅, and PM₁₀) on PTBs (1999-2002) using spatiotemporal residential exposure metrics by month of pregnancy (long-term exposure) in Vancouver, BC (Brauer et al., 2008). The authors did not observe consistent associations with any of the pregnancy average exposure metrics except for PM₂.₅ for PTB. The O₃ associations were largely protective, and no quantitative results were presented for O₃. Additionally, Lee et al. (2008c) used time-series techniques to investigate the associations of short-term exposure to O₃ and PTB in London, England. In addition to exposure on the day of birth, cumulative exposure up to
1 week before birth was investigated. The risk of PTB did not increase with exposure to the levels of ambient air pollution experienced by this population.

Conversely, studies conducted in Australia and China provide evidence for an association between ambient O₃ and PTB. Hansen et al. (2006) reported that long-term exposure to O₃ during the first trimester was associated with an increased risk of PTB (OR = 1.38, [95% CI: 1.14, 1.69] per 10 ppb increase). Although the test for trend was significant due to the strong effect in the highest quartile, there was not an obvious exposure-response pattern across the quartiles of O₃ during the first trimester. The effect estimate was diminished and lost statistical significance when PM₁₀ was included in the model (OR = 1.23, [95% CI: 0.97, 1.59] per 10 ppb increase). Maternal exposure to O₃ during the 90 days prior to birth showed a weak, positive association with PTB (OR = 1.09, [95% CI: 0.85, 1.39] per 10 ppb increase). Jalaludin et al. (2007) found that O₃ levels in the month and three months preceding birth had a statistically significant association with PTB. Ozone levels in the first trimester of pregnancy were associated with increased risks for PTBs (OR = 1.15 [95% CI: 1.05, 1.24] per 10 ppb increase in 1-h max O₃ concentration), and remained a significant predictor of PTB in copollutant models (ORs between 1.07 and 1.10). Jiang et al. (2007) examined the effect of short- and long-term exposure to air pollution on PTB, including risk in relation to levels of pollutants for a single day exposure window with lags from 0 to 6 days before birth. An increase of 10 ppb of the 8-week avg of O₃ corresponded to 9.47% (95% CI: 0.70, 18.7%) increase in PTBs. Increases in PTB were also observed for PM₁₀, SO₂, and NO₂. The authors did not observe any significant effect of short-term exposure to outdoor air pollution on PTB among the 1-day time windows examined in the week before birth.

Few data are available from toxicological studies; a study reported a nearly statistically significant increase in pregnancy duration (short-term exposure) in mice when exposed to 0.8 or 1.2 ppm O₃. This phenomenon was most likely due to the anorexigenic effect of relatively high O₃ concentrations (Bignami et al., 1994).

Table 7-6 provides a brief overview of the epidemiologic studies of PTB. In summary, the evidence is consistent when examining short-term exposure to O₃ during late pregnancy and reports no association with PTB. However when long-term exposure to O₃ early in pregnancy is examined the results are inconsistent. Generally, studies conducted in the U.S., Canada, and England find no association with O₃ and PTB, while studies conducted in Australia and China report an O₃ effect on PTB.
Table 7-6  Brief summary of epidemiologic studies of preterm birth (PTB).

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Sample Size</th>
<th>Mean O₃ (ppb)</th>
<th>Exposure assessment</th>
<th>Effect Estimatea (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Darrow et al. (2009)</td>
<td>Atlanta, GA</td>
<td>(n = 476,489)</td>
<td>8-h max: 44.1</td>
<td>Population-weighted spatial averages Nearest Monitor (within 4 miles)</td>
<td>First month: 0.98 (0.97, 1.00) Last week: 0.99 (0.98, 1.00) Last 6 weeks: 1.00 (0.98, 1.02)</td>
</tr>
<tr>
<td>Ritz et al. (2000)</td>
<td>California, U.S.</td>
<td>(n = 97,158)</td>
<td>8 h: 36.9</td>
<td>&lt;2 mi of monitor</td>
<td>First month: NR Last 6 weeks: NR</td>
</tr>
<tr>
<td>Wilhelm and Ritz (2005)</td>
<td>Los Angeles, CA</td>
<td>(n = 106,483)</td>
<td>1 h: 21.1-22.2</td>
<td>Varying distances to monitor</td>
<td>First month: 1.23 (1.06, 1.42) T1: NR T2: 1.38 (1.14, 1.66) Last 6 weeks: NR</td>
</tr>
<tr>
<td>Ritz et al. (2007)</td>
<td>Los Angeles, CA</td>
<td>(n = 58,316)</td>
<td>24-h avg: 22.5</td>
<td>Nearest monitor to ZIP code</td>
<td>Entire pregnancy: NR T1: 0.93 (0.82, 1.06) Last 6 weeks: NR</td>
</tr>
<tr>
<td>Brauer et al. (2008)</td>
<td>Vancouver, BC, Canada</td>
<td>(n = 70,249)</td>
<td>24-h avg: 14</td>
<td>Nearest Monitor (within 10 km) Inverse Distance Weighting (IDW)</td>
<td>Entire pregnancy: NR First 30 days of pregnancy: NR T1: NR T3: NR</td>
</tr>
<tr>
<td>Lee et al. (2008c)</td>
<td>London, UK</td>
<td>24-h avg: NR</td>
<td>1 monitor</td>
<td>Lag 0: 1.00 (1.00, 1.01)</td>
<td></td>
</tr>
<tr>
<td>Hansen et al. (2006)</td>
<td>Brisbane, Australia</td>
<td>(n = 28,200)</td>
<td>8-h max: 26.7</td>
<td>Citywide avg</td>
<td>T1: 1.39 (1.15, 1.70) T3: 1.09 (0.88, 1.39)</td>
</tr>
<tr>
<td>Jalaludin et al. (2007)</td>
<td>Sydney, Australia</td>
<td>(n = 123,840)</td>
<td>1-h max: 30.9</td>
<td>Citywide avg and &lt;5 km from monitor</td>
<td>First month: 1.04 (0.95, 1.13) T1: 1.15 (1.05, 1.24) T3: 0.98 (0.89, 1.07) Last month: 0.98 (0.88, 1.06)</td>
</tr>
<tr>
<td>Jiang et al. (2007)</td>
<td>Shanghai, China</td>
<td>(n = 3,346 preterm births)</td>
<td>8-h avg: 32.7</td>
<td>Citywide avg</td>
<td>4 wks before birth: 1.06 (1.00, 1.12) 6 wks before birth: 1.06 (0.99, 1.13) 8 wks before birth: 1.09 (1.01, 1.19) Acute effects, L0 to L6: NR* (relative risk results presented in figure 1) *NR: No quantitative results reported; however, preterm birth was not significantly associated with outdoor O₃ air pollution in any lag day (0 – 6) that was considered in the Jiang et al. (2007) study.</td>
</tr>
</tbody>
</table>

*Relative risk of PTB per 10 ppb change in O₃.
T1 = First Trimester, T2 = Second Trimester, T3 = Third Trimester.
L0 = Lag 0, L1= Lag 1, L2 = Lag 2, L3 = Lag 3, L4 = Lag 4, L5 = Lag 5, L6 = Lag 6.
NR: No quantitative results reported.
7.4.5 Fetal Growth

Low birth weight has often been used as an outcome measure because it is easily available and accurately recorded on birth certificates. However, LBW may result from either short gestation, or inadequate growth in utero. Most of the studies investigating air pollution exposure and LBW limited their analyses to term infants to focus on inadequate growth. A number of studies were identified that specifically addressed growth restriction in utero by identifying infants who failed to meet specific growth standards. Usually these infants had birth weight less than the 10th percentile for gestational age, using an external standard. Many of these studies have been previously discussed, since they also examined other reproductive outcomes (i.e., LBW or PTB).

Fetal growth is influenced by maternal, placental, and fetal factors. The biological mechanisms by which air pollutants may influence the developing fetus remain largely unknown. Several mechanisms have been proposed, and are the same as those hypothesized for birth weight (see Section 5.4.2.4). Additionally, in animal toxicology studies, O₃ causes transient anorexia in exposed pregnant dams. This may be one of many possible contributors to O₃-dependent decreased fetal growth.

A limitation of environmental studies that use birth weight as a proxy measure of fetal growth is that patterns of fetal growth during pregnancy cannot be assessed. This is particularly important when investigating pollutant exposures during early pregnancy as birth weight is recorded many months after the exposure period. The insult of air pollution may have a transient effect on fetal growth, where growth is hindered at one point in time but catches up at a later point. For example, maternal smoking during pregnancy can alter the growth rate of individual body segments of the fetus at variable developmental stages, as the fetus experiences selective growth restriction and augmentation (Lampl and Jeanty, 2003).

The terms small-for-gestational-age (SGA), which is defined as a birth weight <10th percentile for gestational age (and often sex and/or race), and intrauterine growth retardation (IUGR) are often used interchangeably. However, this definition of SGA does have limitations. For example, using it for IUGR may overestimate the percentage of “growth-restricted” neonates as it is unlikely that 10% of neonates have growth restriction (Wollmann, 1998). On the other hand, when the 10th percentile is based on the distribution of live births at a population level, the percentage of SGA among PTB is most likely underestimated (Hutcheon and Platt, 2008). Nevertheless, SGA represents a statistical description of a small neonate, whereas the term IUGR is reserved for those with clinical evidence of abnormal growth. Thus all IUGR neonates will be SGA, but not all SGA neonates with be IUGR (Wollmann, 1998). In the following section the terms SGA and IUGR are referred to as each cited study used the terms.

Over the past decade a number of studies examined various metrics of fetal growth restriction. Salam et al. (2005) assessed the effect of increasing O₃ concentrations on IUGR in a population of infants born in California from 1975-1987 as part of the
Children’s Health Study. The authors reported that maternal O₃ exposures averaged over the entire pregnancy and during the third trimester were associated with increased risk of IUGR. A 10-ppb difference in 24-h maternal O₃ exposure during the third trimester increased the risk of IUGR by 11% (95% CI: 0, 20%). **Brauer et al. (2008)** evaluated the impacts of air pollution (CO, NO₂, NO, O₃, SO₂, PM₂.₅, PM₁₀) on SGA (1999-2002) using spatiotemporal residential exposure metrics by month of pregnancy in Vancouver, BC. The O₃ associations were largely protective (OR = 0.87, [95% CI: 0.81, 0.93] for a 10 ppb increase in inverse distance weighted SGA), and no additional quantitative results were presented for O₃. **Liu et al. (2007b)** examined the association between IUGR among singleton term live births and SO₂, NO₂, CO, O₃, and PM₂.₅ in 3 Canadian cities for the period 1985-2000. No increase in the risk of IUGR in relation to exposure to O₃ averaged over each month and trimester of pregnancy was noted. 

Three studies conducted in Australia provide evidence for an association between ambient O₃ and fetal growth restriction. **Hansen et al. (2007)** examined SGA among singleton, full-term births in Brisbane, Australia in relation to ambient air pollution (bsp, PM₁₀, NO₂, O₃) during pregnancy. They also examined head circumference and crown-heel length in a subsample of term neonates. Trimester specific exposures to all pollutants were not statistically significantly associated with a reduction in head circumference or an increased risk of SGA. When monthly-specific exposures were examined, the authors observed an increased risk of SGA associated with exposure to O₃ during month 4 (OR = 1.11 [95% CI: 1.00, 1.24] per 10 ppb increase). In a subsequent study, **Hansen et al. (2008)** examined the possible associations between fetal ultrasonic measurements and ambient air pollution (PM₁₀, O₃, NO₂, SO₂) during early pregnancy. This study had two strengths: (1) fetal growth was assessed during pregnancy as opposed to at birth; and (2) there was little delay between exposures and fetal growth measurements, which reduces potential confounding and uses exposures that are concurrent with the observed growth pattern of the fetus. Fetal ultrasound biometric measurements were recorded for biparietal diameter (BPD), femur length, abdominal circumference, and head circumference. To further improve exposure assessment, the authors restricted the samples to include only scans from women for whom the centroid of their postcode was within 14 km of an air pollution monitoring site. Ozone during days 31-60 was associated with decreases in all of the fetal growth measurements, and a 1.78 mm reduction in abdomen circumference per 10 ppb increase in O₃ concentration, though this effect did not persist in copollutant models. The change in ultrasound measurements associated with O₃ during days 31-60 of gestation indicated that increasing O₃ concentration decreased the magnitude of ultrasound measurements for women living within 2 km of the monitoring site. The relationship decreased toward the null as the distance from the monitoring sites increased. When assessing effect modification due to SES, there was some evidence of effect modification for most of the associations, with the effects of air pollution stronger in the highest SES quartile. In the third study, **Mannes et al. (2005)** estimated the effects of pollutant (PM₁₀, PM₂.₅, NO₂, CO and O₃) exposure in the first, second and third trimesters of pregnancy and risk of SGA in Sydney, Australia. Citywide average air pollutant concentrations in the last month, third trimester, and first trimester of pregnancy had no effect on SGA.
Concentrations of O₃ in the second trimester of pregnancy had small but adverse effects on SGA (OR = 1.10 [95% CI: 1.00, 1.14] per 10 ppb increment). This effect disappeared when the analysis was limited to births with a maternal address within 5 km of a monitoring station (OR = 1.00 [95% CI: 0.60, 1.79] per 10 ppb increment).

Very little information from toxicological studies is available to address effects on fetal growth. However, there is evidence to suggest that prenatal (short-term) exposure to O₃ can affect postnatal growth. A few studies reported that mice or rats exposed developmentally (gestationally ± lactationally) to O₃ had deficits in body weight gain in the postpartum period (Bignami et al., 1994; Haro and Paz, 1993; Kavlock et al., 1980).

Table 7-7 provides a brief overview of the epidemiologic studies of fetal growth restriction. In summary, the evidence is inconsistent when examining exposure to O₃ and fetal growth restriction. Similar to PTB, studies conducted in Australia have reported an effect of O₃ on fetal growth, whereas studies conducted in other areas generally have not found such an effect. This may be due to the restriction of births to those within 2-14 km of a monitoring station, as was done in the Australian studies.
Table 7-7  Brief summary of epidemiologic studies of fetal growth.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Mean O₃ (ppb)</th>
<th>Exposure assessment</th>
<th>Effect Estimate⁹ (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Salam et al. (2005)</td>
<td>California, U.S. (n = 3,901)</td>
<td>24-h avg: 27.3</td>
<td>ZIP code level</td>
<td>Entire pregnancy: 1.16 (1.00, 1.32)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>8 h: 50.6</td>
<td></td>
<td>T1: 1.00 (0.94, 1.11)</td>
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<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T2: 1.06 (1.00, 1.12)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T3: 1.11 (1.00, 1.17)</td>
</tr>
<tr>
<td>Brauer et al. (2008)</td>
<td>Vancouver, BC, Canada (n = 70,249)</td>
<td>24-h avg: 14</td>
<td>Nearest Monitor (within 10 km) Inverse Distance Weighting (IDW)</td>
<td>Entire pregnancy: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>First 30 days of pregnancy: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Last 30 days of pregnancy: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T1: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T3: NR</td>
</tr>
<tr>
<td>Liu et al. (2007b)</td>
<td>Calgary, Edmonton, and Montreal, Canada (n = 16,430)</td>
<td>24-h avg: 16.5</td>
<td>Census Subdivision avg</td>
<td>Entire pregnancy: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1-h max: 31.2</td>
<td></td>
<td>(results presented in figure)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T1: NR</td>
</tr>
<tr>
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<td></td>
<td></td>
<td></td>
<td>T2: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T3: NR</td>
</tr>
<tr>
<td>Hansen et al. (2007)</td>
<td>Brisbane, Australia (n = 26,617)</td>
<td>8-h max: 26.7</td>
<td>Citywide avg</td>
<td>T1: 1.01 (0.89, 1.15)</td>
</tr>
<tr>
<td>Hansen et al. (2008)</td>
<td>Brisbane, Australia (n = 15,623)</td>
<td>8-h avg: 24.8</td>
<td>Within 2 km of monitor</td>
<td>M1: -0.32 (-1.56, 0.91)⁸</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>M2: -0.58 (-1.97, 0.80)⁸</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>M3: 0.26 (-1.07, 1.59)⁸</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>M4: 0.11 (-0.98, 1.21)⁸</td>
</tr>
<tr>
<td>Mannes et al. (2005)</td>
<td>Sydney, Australia (n = 138,056)</td>
<td>1-h max: 31.6</td>
<td>Citywide avg and &lt;5 km from monitor</td>
<td>T1: 0.90 (0.48, 1.34)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T2: 1.00 (0.60, 1.79)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>T3: 1.10 (0.66, 1.97)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Last 30 days of pregnancy: 1.10 (0.74, 1.79)</td>
</tr>
</tbody>
</table>

⁹Relative risk of fetal growth restriction per 10 ppb change in O₃, unless otherwise noted.

Mean change in fetal ultrasonic measure of head circumference recorded between 13 and 26 weeks gestation for a 10-ppb increase in maternal exposure to O₃ during early pregnancy

T1 = First Trimester, T2 = Second Trimester, T3 = Third Trimester

M1 = Month 1, M2 = Month 2, M3 = Month 3, M4 = Month 4

NR: No quantitative results reported

7.4.6 Postnatal Growth

Postnatal weight and height are routinely measured in children as indicators of growth and somatic changes. Toxicological studies often follow these endpoints to ascertain if a known exposure has an effect in the postnatal window, an effect which can be permanent. Time-pregnant BALB/c mice were exposed to O₃ (0, 0.4, 0.8, or 1.2 ppm) GD9-18 (short-term exposure) with parturition at GD20-21 (Sharkhuu et al., 2011). As the offspring aged, postnatal litter body weight continued to be significantly decreased in the highest concentration (1.2 ppm) O₃ group at PND3 and PND7. When the pups were weighed separately by sex at PND42, the males with the
highest concentration of O₃ exposure (1.2 ppm, GD9-18) had significant decrements in body weight (Sharkhuu et al., 2011).

Significant decrements in body weight at 4 weeks of age were reported in C57Bl/6 mice that were exposed to postnatal O₃ (short-term exposure, PND2-28 exposure, 1 ppm O₃, 3 hours/day, 3 days/week) (Auten et al., 2012). Animals with co-exposure to in utero DE (short-term exposure, dam GD9-GD17; inhalation 0.5 or 2.0 mg/m³ O₃; 4 h/day via inhalation; or oropharyngeal aspiration DEPs, 2×/week) + postnatal O₃ (aforementioned short-term exposure) also had significantly reduced body weight.

### 7.4.7 Birth Defects

Despite the growing body of literature evaluating the association between ambient air pollution and various adverse birth outcomes, relatively few studies have investigated the effect of temporal variations in ambient air pollution on birth defects. Heart defects and oral clefts have been the focus of the majority of these recent studies, given the higher prevalence than other birth defects and associated mortality. Mechanistically, air pollutants could be involved in the etiology of birth defects via a number of key events (see Section 5.4.2.4).

Several studies have been conducted examining the relationship between O₃ exposure during pregnancy and birth defects and reported a positive association with cardiac defects. The earliest of these studies was conducted in southern California (Ritz et al., 2002). This study evaluated the effect of air pollution on the occurrence of cardiac birth defects in neonates and fetuses delivered in southern California in 1987-1993. Maternal exposure estimates were based on data from the fixed site closest to the mother’s ZIP code area. When using a case-control design where cases were matched to 10 randomly selected controls, results showed increased risks for aortic artery and valve defects (OR = 1.56 [95% CI: 1.16, 2.09] per 10 ppb O₃), pulmonary artery and valve anomalies (OR = 1.34 [95% CI: 0.96, 1.87] per 10 ppb O₃), and conotruncal defects (OR = 1.36 [95% CI: 0.91, 2.03] per 10 ppb O₃) in a dose-response manner with second-month O₃ exposure. A study conducted in Texas (Gilboa et al., 2005) looked at a similar period of exposure but reported no association with most of the birth defects studied (O₃ concentration was studied using quartiles with the lowest representing <18 ppb and the highest representing ≥31 ppb). The authors found slightly elevated odds ratios for pulmonary artery and valve defects. They also detected an inverse association between O₃ exposure and isolated ventricular septal defects. Overall, this study provided some weak evidence that air pollution increases the risk of cardiac defects. Hansen et al. (2009) investigated the possible association between ambient air pollution concentrations averaged over weeks 3-8 of pregnancy and the risk of cardiac defects. When analyzing all births with exposure estimates for O₃ from the nearest monitor there was no indication for an association with cardiac defects. There was also no adverse association when restricting the analyses to only include births where the mother
resided within 12 km of a monitoring station. However, among births within 6 km of a monitor, a 10 ppb increase in O\textsubscript{3} was associated with an increased risk of pulmonary artery and valve defects (OR = 8.76 [95% CI: 1.80, 56.55]). As indicated by the very wide credible intervals, there were very few cases in the sensitivity analyses for births within 6 km of a monitor, and this effect could be a result of type I errors. Dadvand et al. (2011) investigated the association between maternal exposure to ambient air pollution concentrations averaged over weeks 3-8 of pregnancy and the occurrence of cardiac birth defects in England. Similar to Hansen et al. (2009), they found no associations with maternal exposure to O\textsubscript{3} except for when the analysis was limited to those subjects residing within a 16 km distance of a monitoring station (OR for malformations of pulmonary and tricuspid valves=1.64 [95% CI: 1.04, 2.60] per 10 ppb increase in O\textsubscript{3}).

Despite the association between O\textsubscript{3} and cardiac defects observed in the above studies, a recent study did not observe an increased risk of cardiac birth defects associated with ambient O\textsubscript{3} concentrations. The study, conducted in Atlanta, GA, examined O\textsubscript{3} exposure during weeks 3-7 of of pregnancy and reported no association with risk of cardiovascular malformations (Strickland et al., 2009).

Several of these studies have also examined the relationship between O\textsubscript{3} exposure during pregnancy and oral cleft defects. The study by Ritz et al. (2002) evaluated the effect of air pollution on the occurrence of orofacial birth defects and did not observe strong associations between ambient O\textsubscript{3} concentration and orofacial defects. They did report an OR of 1.13 (95% CI: 0.90, 1.40) per 10 ppb during the second trimester for cleft lip with or without cleft palate. Similarly, Gilboa et al. (2005) reported an OR of 1.09 (95% CI: 0.70, 1.69) for oral cleft defects when the fourth quartile was contrasted with the first quartile of exposure during 3-8 weeks of pregnancy. Hansen et al. (2009) reported no indication for an association with cleft defects and air pollution concentrations averaged over weeks 3-8 of pregnancy. Hwang and Jaakkola (2008) conducted a population-based case-control study to investigate exposure to ambient air pollution and the risk of cleft lip with or without cleft palate in Taiwan. The risk of cleft lip with or without cleft palate was increased in relation to O\textsubscript{3} levels in the first gestational month (OR = 1.17 [95% CI: 1.01, 1.36] per 10 ppb) and second gestational month (OR = 1.22 [95% CI: 1.03, 1.46] per 10 ppb), but was not related to any of the other pollutants. In three-pollutant models, the effect estimates for O\textsubscript{3} exposure were stable for the four different combinations of pollutants and were all statistically significant. Marshall et al. (2010) compared estimated exposure to ambient pollutants during early pregnancy (6 week period from 5 to 10 weeks into the gestational period) among mothers of children with oral cleft defects to that among mothers of controls. The authors observed no consistent elevated associations between any of the air pollutants examined and cleft malformations, though there was a weak association between cases of cleft palate only and increasing O\textsubscript{3} concentrations. This association increased when cases and controls were limited to those with residences within 10 km of the closest O\textsubscript{3} monitor (OR = 2.2 [95% CI: 1.0, 4.9], comparing highest quartile [>33 ppb] to lowest quartile [<15 ppb]).
A limited number of toxicological studies have examined birth defects in animals exposed gestationally to \( O_3 \). Kavlock et al. (1979) exposed pregnant rats to \( O_3 \) for precise periods during organogenesis. No significant teratogenic effects were found in rats exposed 8 h/day to concentrations of \( O_3 \) varying from 0.44 to 1.97 ppm during early (days 6-9), mid (days 9-12), or late (days 17 to 20) gestation, or the entire period of organogenesis (days 6-15) (short-term exposures). Earlier research found eyelid malformation following gestational and postnatal exposure to 0.2 ppm \( O_3 \) (Veninga, 1967).

Table 7-8 provides a brief overview of the epidemiologic studies of birth defects. These studies have focused on cardiac and oral cleft defects, and the results from these studies are not entirely consistent. This inconsistency could be due to the absence of true associations between \( O_3 \) and risks of cardiovascular malformations and oral cleft defects; it could also be due to differences in populations, pollution levels, outcome definitions, or analytical approaches. The lack of consistency of associations between \( O_3 \) and cardiovascular malformations or oral cleft defects might be due to issues relating to statistical power or measurement error. A recent meta-analysis of air pollution and congenital anomalies concluded that there was no statistically significant increase in risk of congenital anomalies with \( O_3 \) exposure (Vrijheid et al., 2011). These authors note that heterogeneity in the results of these studies may be due to inherent differences in study location, study design, and/or analytic methods, and comment that these studies have not employed some recent advances in exposure assessment used in other areas of air pollution research that may help refine or reduce this heterogeneity.
### Table 7-8  Brief summary of epidemiologic studies of birth defects.

<table>
<thead>
<tr>
<th>Study</th>
<th>Outcomes Examined</th>
<th>Location (Sample Size)</th>
<th>Mean O₃ (ppb)</th>
<th>Exposure Assessment</th>
<th>Exposure Window</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ritz et al. (2002)</td>
<td>Cardiac and Cleft Defects</td>
<td>Southern California (n = 3,549 cases; 10,649 controls)</td>
<td>24-h avg: NR</td>
<td>Nearest Monitor (within 10 mi)</td>
<td>Month 1,2,3 Trimester 2,3 3-mo period prior to conception</td>
</tr>
<tr>
<td>Gilboa et al. (2005)</td>
<td>Cardiac and Cleft Defects</td>
<td>7 Counties in TX (n = 5,338 cases; 4,580 controls)</td>
<td>24-h avg: NR</td>
<td>Nearest Monitor</td>
<td>Weeks 3-8 of gestation</td>
</tr>
<tr>
<td>Hwang and Jaakkola (2008)</td>
<td>Oral Cleft Defects</td>
<td>Taiwan (n = 653 cases; 6,530 controls)</td>
<td>24-h avg: 27.31</td>
<td>Inverse Distance Weighting (IDW)</td>
<td>Months 1,2,3</td>
</tr>
<tr>
<td>Strickland et al. (2009)</td>
<td>Cardiac Defects</td>
<td>Atlanta, GA (n = 3,338 cases)</td>
<td>8-h max: 39.4-43.3</td>
<td>Weighted citywide avg</td>
<td>Weeks 3-7 of gestation</td>
</tr>
<tr>
<td>Hansen et al. (2009)</td>
<td>Cardiac and Cleft Defects</td>
<td>Brisbane, Australia (n = 150,308 births)</td>
<td>8-h max: 25.8</td>
<td>Nearest Monitor</td>
<td>Weeks 3-8 of gestation</td>
</tr>
<tr>
<td>Marshall et al. (2010)</td>
<td>Oral Cleft Defects</td>
<td>New Jersey (n = 717 cases; 12,925 controls)</td>
<td>24-h avg: 25</td>
<td>Nearest Monitor (within 40 km)</td>
<td>Weeks 5-10 of gestation</td>
</tr>
<tr>
<td>Dadvand et al. (2011)</td>
<td>Cardiac Defects</td>
<td>Northeast England (n = 2,140 cases; 14,256 controls)</td>
<td>24-h avg: 18.8</td>
<td>Nearest Monitor</td>
<td>Weeks 3-8 of gestation</td>
</tr>
</tbody>
</table>

#### 7.4.8  Developmental Respiratory Effects

The issue of prenatal exposure has assumed increasing importance since ambient air pollution exposures of pregnant women have been shown to lead to adverse pregnancy outcomes, as well as to respiratory morbidity and mortality in the first year of life. Growth and development of the respiratory system take place mainly during the prenatal and early postnatal periods. This early developmental phase is thought to be very important in determining long-term lung growth. Studies have recently examined this emerging issue. Several studies were included in Section 7.2.1 and Section 7.2.3, and are included here because they reported both prenatal and post-natal exposure periods.

Mortimer et al. (2008a, b) examined the association of prenatal and lifetime exposures to air pollutants with pulmonary function and allergen sensitization in a subset of asthmatic children (ages 6-11) included in the Fresno Asthmatic Children’s Environment Study (FACES). Monthly means of pollutant levels for the years 1989-2000 were created and averaged separately across several important developmental time-periods, including the entire pregnancy, each trimester, the first 3 years of life, the first 6 years of life, and the entire lifetime. The 8-h avg O₃ concentrations were approximately 50 ppb for each of the exposure metrics.
(estimated from figure). In the first analysis (Mortimer et al., 2008a), negative effects on pulmonary function were found for exposure to PM$_{10}$, NO$_2$, and CO during key neonatal and early life developmental periods. The authors did not find a negative effect of exposure to O$_3$ among this cohort. In the second analysis (Mortimer et al., 2008b), sensitization to at least one allergen was associated, in general, with higher levels of CO and PM$_{10}$ during the entire pregnancy and second trimester and higher PM$_{10}$ during the first 2 years of life. Lower exposure to O$_3$ during the entire pregnancy or second trimester was associated with an increased risk of allergen sensitization. Although the pollutant metrics across time periods are correlated, the strongest associations with the outcomes were observed for prenatal exposures. Though it may be difficult to disentangle the effect of prenatal and postnatal exposures, the models from this group of studies suggest that each time period of exposure may contribute independently to different dimensions of school-aged children’s pulmonary function. For 4 of the 8 pulmonary-function measures (FVC, FEV$_1$, PEF, FEF$_{25-75}$), prenatal exposures were more influential on pulmonary function than early-lifetime metrics, while, in contrast, the ratio of measures (FEV$_1$/FVC and FEF$_{25-75}$/FVC) were most influenced by postnatal exposures. When lifetime metrics were considered alone, or in combination with the prenatal metrics, the lifetime measures were not associated with any of the outcomes, suggesting the timing of the exposure may be more important than the overall dose and prenatal exposures are not just markers for lifetime or current exposures.

Clark et al. (2010) investigated the effect of exposure to ambient air pollution in utero and during the first year of life on risk of subsequent asthma diagnosis (incident asthma diagnosis up to age 3-4) in a population-based nested case-control study. Air pollution exposure for each subject based on their residential address history was estimated using regulatory monitoring data, land use regression modeling, and proximity to stationary pollution sources. An average exposure was calculated for the duration of pregnancy (~15 ppb) and the first year of life (~14 ppb). In contrast to the Mortimer et al. (2008a, b) studies, the effect estimates for first-year exposure were generally larger than for in utero exposures. However, similar to the Mortimer et al. (2008a, b), the observed associations with O$_3$ were largely protective. Because of the relatively high correlation between in utero and first-year exposures for many pollutants, it was difficult to discern the relative importance of the individual exposure periods.

Latzin et al. (2009) examined whether prenatal exposure to air pollution was associated with lung function changes in the newborn. Tidal breathing, lung volume, ventilation inhomogeneity and eNO were measured in 241 unsedated, sleeping neonates (age = 5 weeks). The median of the 24-h avg O$_3$ concentrations averaged across the post-natal period was ~44 ppb. Consistent with the previous studies, no association was found for prenatal exposure to O$_3$ and lung function.

The new toxicological literature since the 2006 O$_3$ AQCD, covering respiratory changes related to developmental O$_3$ exposure, reports ultrastructural changes in bronchiole development, alterations in placental and pup cytokines, and increased pup airway hyper-reactivity. These studies are detailed below.
Fetal rat lung bronchiole development is triphasic, comprised of the glandular phase (measured at GD18), the canalicular phase (GD20), and the saccular phase (GD21). The ultrastructural lung development in fetuses of pregnant rats exposed to 1-ppm O₃ (12 h/day, out to either GD18, GD20 or GD21) was examined by electron microscopy during these three phases. In the glandular phase, bronchiolar columnar epithelial cells in fetuses of dams exposed to O₃ had cytoplasmic damage and swollen mitochondria. Bronchial epithelium at the canalicular phase in O₃ exposed pups had delayed maturation in differentiation, i.e., glycogen abundance in secretory cells had not diminished as it should with this phase of development. Congruent with this finding, delayed maturation of tracheal epithelium following early neonatal O₃ exposure (1 ppm, 4-5 h/day for first week of life) in lambs has been previously reported (Mariassy et al., 1990; Mariassy et al., 1989). Also at the canalicular phase, atypical cells were seen in the bronchiolar lumen of O₃-exposed rat fetuses. Finally, in the saccular phase, mitochondrial degradation was present in the non-ciliated bronchiolar cells of rats exposed in utero to O₃. In conclusion, O₃ exposure of pregnant rats produced ultra-structural damage to near-term fetal bronchiolar epithelium (López et al., 2008).

Exposure of laboratory animals to multiple airborne pollutants can differentially affect pup physiology. One study showed that exposure of C57BL/6 mouse dams to 0.48 mg PM intratracheally twice weekly for 3 weeks during pregnancy augmented O₃-induced airway hyper-reactivity in juvenile offspring. Maternal PM exposure also significantly increased placental cytokines above vehicle-instilled controls. Pup postnatal O₃ exposure (1 ppm 3 h/day, every other day, thrice weekly for 4 weeks) induced significantly increased cytokine levels (IL-1β, TNF-α, KC, and IL-6) in whole lung versus postnatal air exposed groups; this was further exacerbated with gestational PM exposure (Auten et al., 2009). In further studies by the same laboratory, O₃-induced AHR was studied in rodent offspring after dam gestational exposure to inhaled diesel exhaust (Auten et al., 2012). Pregnant C57BL/6 mice were exposed to diesel exhaust GD9-17 (0.5 or 2.0 mg/m³ O₃, 4h/day) via inhalation or in a separate set of animals via oropharyngeal aspiration of freshly generated DEPs (2×/week). Postnatally, the offspring were exposed to O₃ starting at PND2 (1 ppm O₃, 3 hours/day, 3 days/week for 4 weeks). Juvenile mice were then subjected to measurements of pulmonary mechanisms (at 4 weeks of age and then at 8 weeks of age). Increased inflammation of the placenta and lungs of DE exposed fetuses was reported at GD18. In animals with postnatal O₃ exposure alone, elevated inflammation was seen with significant increased levels of BAL cytokines; these O₃-related elevated levels were significantly exacerbated with prenatal DE exposure (DE+O₃). At PND28, DE+O₃ exposed offspring had significant impairment of alveolar development as measured with secondary alveolar crest development, a finding that was absent in all other exposure groups (O₃ alone, DE alone). Postnatal O₃ exposure induced AHR in methacholine challenged animals at 4 weeks of age and was exacerbated with the higher dose of DE exposure (DE+O₃). At 8 weeks of age, O₃ exposed pups had persistent AHR (+/-DE) that was significantly augmented in DE+O₃ pups. In summary, gestational DE exposure induced an inflammatory response which, when combined with postnatal O₃ exposure impaired alveolar
development, and caused an exacerbated and longer-lasting O$_3$-induced AHR in offspring.

A series of experiments using infant rhesus monkeys repeatedly exposed to 0.5 ppm O$_3$ starting at one-month of age have examined the effect of O$_3$ alone or in combination with an inhaled allergen on morphology and lung function (Plopper et al., 2007). Exposure to O$_3$ alone or allergen alone produced small but not statistically significant changes in baseline airway resistance and airway responsiveness, but the combined exposure to both O$_3$ + antigen produced statistically significant and greater than additive changes in both functional measurements. Additionally, cellular changes and significant structural changes in the respiratory tract have been observed in infant rhesus monkeys exposed to O$_3$ (Fanucchi et al., 2006). A more detailed description of these studies can be found in Section 7.2.3 (Pulmonary Structure and Function), with mechanistic information found in Section 5.4.2.4.

Lung immunological response in O$_3$ exposed pups was followed by analyzing BAL and lung tissue. Sprague Dawley (SD) pups were exposed to a single 3h exposure of air or O$_3$ (0.6 ppm) on PND 13 (Han et al., 2011). Bronchoalveolar lavage (BAL) was performed 10 hours after the end of O$_3$ exposure. BALF polymorphonuclear leukocytes (PMNs) and total BALF protein were significantly elevated in O$_3$ exposed pups. Lung tissue from O$_3$ exposed pups had significant elevations of manganese superoxide dismutase (SOD) protein and significant decrements of extracellular SOD protein.

Various immunological outcomes were followed in offspring after their pregnant dams (BALB/c mice) were exposed gestationally to O$_3$ (0, 0.4, 0.8, or 1.2 ppm, GD9-18) (Sharkhuu et al., 2011). Delayed type hypersensitivity (DTH) was initiated with initial BSA injection at 6 weeks of age and then challenge 7 days later. The normal edematous response of the exposed footpad (thickness after BSA injection) was recorded as an indicator of DTH. In female offspring, normal footpad swelling with BSA injection that was seen in air exposed animals was significantly attenuated with O$_3$ exposure (0.8 and 1.2 ppm O$_3$), implying immune suppression of O$_3$ exposure specifically in DTH. Humoral immunity was measured with the sheep red blood cell (SRBC) response. Animals received primary immunization with SRBC and then blood was drawn for SRBC IgM measurement. A SRBC booster was given 2 weeks later with blood collected 5 days after booster for IgG measurement. Maternal O$_3$ exposure had no effect on humoral immunity in the offspring as measured by IgG and IgM titers after SRBC primary and booster immunizations (Sharkhuu et al., 2011).

Toxicity assessment and allergen sensitization was also assessed in these O$_3$ exposed offspring. At PND42, animals were euthanized for analysis of immune and inflammatory markers (immune proteins, inflammatory cells, T-cell populations in the spleen). A subset of the animals was intra-nasally instilled or sensitized with ovalbumin on either PND2 and 3 or PND42 and 43. All animals were challenged with OVA on PND54, 55, and 56. One day after final OVA challenge, lung function, lung inflammation and immune response were determined. Offspring of O$_3$ exposed dams that were initially sensitized at PND3 (early) or PND42 (late) were tested to
determine the level of allergic sensitization or asthma-like inflammation after OVA challenge. Female offspring sensitized early in life developed significant eosinophilia (1.2 ppm O₃) and elevated serum OVA-specific IgE (1.2 ppm O₃), which is a marker of airway allergic inflammation. The females that were sensitized early also had significant decrements in BALF total cells, macrophages, and lymphocytes (1.2 ppm O₃). Offspring that were sensitized later (PND42) in life did not develop the aforementioned changes in BALF, but these animals did develop modest, albeit significant neutropenia (0.8 and 1.2 ppm O₃) (Sharkhuu et al., 2011).

BALF cytology in non-sensitized animals was followed. BALF of offspring born to dams exposed to O₃ was relatively unaffected (cytokines, inflammatory cell numbers/types) as were splenic T-cell subpopulations. LDH was significantly elevated in BALF of females whose mothers were exposed to 1.2 ppm during pregnancy (Sharkhuu et al., 2011). In summary, the females born to mothers exposed to O₃ developed modest immunocompromise. Males were unaffected (Sharkhuu et al., 2011).

Overall, animal toxicological studies have reported ultrastructural changes in bronchiole development, alterations in placental and pup cytokines, and increased pup airway hyper-reactivity related to exposure to O₃ during the developmental period. Epidemiologic studies have found no association between prenatal exposure to O₃ and growth and development of the respiratory system. Fetal origins of disease have received a lot of attention recently, thus additional research to further explore the inconsistencies between these two lines of evidence is warranted.

### 7.4.9 Developmental Central Nervous System Effects

The following sections describe the results of toxicological studies of O₃ and developmental central nervous system effects. No epidemiologic studies of this association have been published.

#### 7.4.9.1 Laterality

Two reports of laterality changes in mice developmentally exposed to O₃ have been reported in the literature. Mice developmentally exposed to 0.6 ppm O₃ (6 days before breeding to weaning at PND21) showed a turning preference (left turns) distinct from air exposed controls (clockwise turns) (Dell'Omo et al., 1995); in previous studies this behavior in mice has been found to correlate with specific structural asymmetries of the hippocampal mossy fiber projections (Schöpke et al., 1991). The 2006 O₃ AQCD evidence for the effect of O₃ on laterality or handedness demonstrated that rats exposed to O₃ during fetal and neonatal life showed limited, sex-specific changes in handedness after exposure to the intermediate concentration of O₃ (only seen in female mice exposed to 0.6 ppm O₃, and not in males at 0.6 ppm
or in either sex of 0.3 or 0.9 ppm O$_3$ with exposure from 6 days before breeding to PND26) (Petruzzi et al., 1999).

### 7.4.9.2 Brain Morphology and Neurochemical Changes

The nucleus tractus solitarius (NTS), a medullary area of respiratory control, of adult animals exposed prenatally to 0.5 ppm O$_3$ (12h/day, ED5-eD20) had significantly less tyrosine hydroxylase staining versus control (Boussouar et al., 2009). Tyrosine hydroxylase is the rate-limiting enzyme for dopamine synthesis and serves as a precursor for catecholamine synthesis; thus, decreased staining is used as a marker of dopaminergic or catecholaminergic cell or activity loss in these regions and thus functions in neuronal plasticity. After physical restraint stress, control animals respond at the histological level with Fos activation, a marker of neuronal activity, and tyrosine hydroxylase activation in the NTS, a response which is absent or attenuated in adult animals exposed prenatally to 0.5 ppm O$_3$ (Boussouar et al., 2009) when compared to control air exposed animals who also were restrained. The O$_3$-exposed offspring in this study were cross-fostered to control air exposed dams to avoid O$_3$-dependent dam related neonatal effects on offspring outcomes (i.e., dam behavioral or lactational contributions to pup outcomes) (Boussouar et al., 2009).

Developmental exposure to 0.3 or 0.6 ppm O$_3$ prior to mating pair formation through GD17 induced significant increased levels of BDNF in the striatum of adult (PND140) O$_3$-exposed offspring as compared to control air exposed animals; these O$_3$-exposed animals also had significantly decreased level of NGF in the hippocampus versus control (Santucci et al., 2006).

Changes in the pup cerebellum with prenatal 1 ppm O$_3$ exposure include altered morphology (Romero-Velazquez et al., 2002; Rivas-Manzano and Paz, 1999), decreased total area (Romero-Velazquez et al., 2002), decreased number of Purkinje cells (Romero-Velazquez et al., 2002), and altered monoamine neurotransmitter content with the catecholamine system affected and the indoleamine system unaffected by O$_3$ (Gonzalez-Pina et al., 2008).

### 7.4.9.3 Neurobehavioral Outcomes

Ozone administration to dams during pregnancy with or without early neonatal exposure has been shown to contribute to multiple neurobehavioral outcomes in offspring that are described in further detail below.

Ozone administration (0.4, 0.8 or 1.2 ppm O$_3$) during the majority of pregnancy (PD7-17) of CD-1 mice did not affect pup behavioral outcomes including early behavioral ultrasonic vocalizations and more permanent later measurements (PND60 or 61) including pup activity, habituation and exploration and d-
amphetamine-induced hyperactivity (Bignami et al., 1994); these pups were all cross-fostered or reared on non-O3 exposed dams.

Testing for aggressive behavior in mice continuously exposed to O3 (0.3 or 0.6 ppm from 30 days prior to mating to GD17) revealed that mice had significantly increased defensive/ submissive behavior (increased freezing posturing on the first day only of a multiple-day exam) versus air exposed controls (Santucci et al., 2006). Similarly, continuous exposure of adult animals to O3 induced significant increases in fear behavior and decreased aggression as measured by significantly decreased freezing behavior (Petruzzi et al., 1995).

Developmentally exposed animals also had significantly decreased amount of time spent nose sniffing other mice (Santucci et al., 2006); this social behavior deficit, decreased sniffing time, was not found in an earlier study with similar exposures (Petruzzi et al., 1995), but sniffing of specific body areas was measured in Santucci et al. (2006) and total number of sniffs of the entire body was measured in Petruzzi et al. (1995). The two toxicology studies exploring social behavior (sniffing) employ different study designs and find opposite effects in animals exposed to O3.

7.4.9.4 Sleep Aberrations after Developmental Ozone Exposure

The effect of gestational O3 exposure (1 ppm O3 daily for 12h/day, during dark period for the entire pregnancy) on sleep patterns in rat offspring was followed using 24 h polysomnographic recordings at 30, 60 and 90 days of age (Haro and Paz, 1993). Ozone-exposed pups manifested with inverted sleep-wake patterns or circadian rhythm phase-shift. Rat vigilance was characterized in wakefulness, slow wave sleep (SWS), and paradoxical sleep (PS) using previously characterized criteria. The O3 exposed offspring spent longer time in the wakefulness state during the light period, more time in SWS during the period of darkness, and showed significant decrements in PS. Chronic O3 inhalation significantly decreased the duration of PS during both the light and dark periods (Haro and Paz, 1993). These effects were consistent at all time periods measured (30, 60 and 90 days of age). These sleep effects reported after developmental exposures expand upon the existing literature on sleep aberrations in adult animals exposed to O3 [rodents: (Paz and Huitron-Resendiz, 1996; Arito et al., 1992); and cats: (Paz and Bazan-Perkins, 1992)]. A role for inhibition of cyclooxygenase-2 and the interleukins and prostaglandins in the O3-dependent sleep changes potentially exists with evidence from a publication on indomethacin pretreatment attenuating O3-induced sleep aberrations in adult male animals (Rubio and Paz, 2003).

7.4.10 Early Life Mortality

Infants may be particularly at risk for the effects of air pollution. Within the first year of life, infants develop rapidly; therefore their sensitivity may change within weeks
or months. During the neonatal and post-neonatal periods, the developing lung is highly sensitive to environmental toxicants. The lung is not well developed at birth, with 80% of alveoli being formed postnatally. An important question regarding the association between O3 and infant mortality is the critical window of exposure during development for which infants are at risk. Several age intervals have been explored: neonatal (<1 month); postneonatal (1 month to 1 year); and an overall interval for infants that includes both the neonatal and postneonatal periods (<1 year). Within these various age categories, multiple causes of deaths have been investigated, particularly total deaths and respiratory-related deaths. The studies reflect a variety of study designs, exposure periods, regions, and adjustment for confounders. As discussed below, a handful of studies have examined the effect of ambient air pollution on neonatal and postneonatal mortality, with the former the least studied. These studies varied somewhat with regard to the outcomes and exposure periods examined and study designs employed.

7.4.10.1 Stillbirth

Pereira et al. (1998) investigated the association among daily counts of intrauterine mortality (over 28 weeks of gestation) and air pollutant concentrations in Sao Paulo, Brazil from 1991 through 1992. The association was strong for NO2, but lesser for SO2 and CO. These associations exhibited a short lag time, less than 5 days. No significant association was detected between short-term O3 exposure and intrauterine mortality.

7.4.10.2 Infant Mortality, Less than 1 Year

Ritz et al. (2006) linked birth and death certificates for infants who died between 1989 and 2000 to evaluate the influence of outdoor air pollution on infant death in the South Coast Air Basin of California. The authors examined short- and long-term exposure periods 2 weeks, 1 month, 2 months, and 6 months before each case subject’s death and reported no association between ambient levels of O3 and infant mortality. Similarly, Diaz et al. (2004) analyzed the effects of extreme temperatures and short-term exposure to air pollutants on daily mortality in children less than 1 year of age in Madrid, Spain, from 1986 to 1997 and observed no statistically significant association between mortality and O3 concentrations. Hajat et al. (2007) analyzed time-series data of daily infant mortality counts in 10 major cities in the UK to quantify any associations with short-term changes in air pollution. When the results from the 10 cities were combined there was no relationship between O3 and infant mortality, even after restricting the analysis to just the summer months.

Conversely, a time-series study of infant mortality conducted in the southwestern part of Mexico City in the years 1993-1995 found that infant mortality was associated with short-term exposure to NO2 and O3 3-5 days before death, but not as consistently as with PM. A 10-ppb increase in 24-h avg O3 was associated with a
2.78% increase (95% CI: 0.29, 5.26%) in infant mortality (lag 3) (Loomis et al., 1999). This increase was attenuated, although still positive when evaluated in a two-pollutant model with PM$_{2.5}$. One-hour max concentrations of O$_3$ exceeded prevailing Mexican and international standards nearly every day.

### 7.4.10.3 Neonatal Mortality, Less than 1 Month

Several studies have evaluated ambient O$_3$ concentrations and neonatal mortality and observed no association. Ritz et al. (2006) linked birth and death certificates for infants who died between 1989 and 2000 to evaluate the influence of outdoor air pollution on infant death in the South Coast Air Basin of California. The authors examined short- and long-term exposure periods 2 weeks, 1 month, 2 months, and 6 months before each case subject’s death and reported no association between ambient levels of O$_3$ and neonatal mortality. Hajat et al. (2007) analyzed time-series data of daily infant mortality counts in 10 major cities in the UK to quantify any associations with short-term changes in air pollution. When the results from the 10 cities were combined there was no relationship between O$_3$ and neonatal mortality, even after restricting the analysis to just the summer months. Lin et al. (2004a) assessed the impact of short-term changes in air pollutants on the number of daily neonatal deaths in Sao Paulo, Brazil. The authors observed no association between ambient levels of O$_3$ and neonatal mortality.

### 7.4.10.4 Postneonatal Mortality, 1 Month to 1 Year

A number of studies focused on the postneonatal period when examining the effects of O$_3$ on infant mortality. Ritz et al. (2006) linked birth and death certificates for infants who died between 1989 and 2000 to evaluate the influence of outdoor air pollution on infant death in the South Coast Air Basin of California. The authors examined short- and long-term exposure periods 2 weeks, 1 month, 2 months, and 6 months before each case subject’s death and reported no association between ambient levels of O$_3$ and postneonatal mortality. Woodruff et al. (2008) evaluated the county-level relationship between cause-specific postneonatal infant mortality and long-term early-life exposure (first 2 months of life) to air pollutants across the United States. Similarly, they found no association between O$_3$ exposure and deaths from respiratory causes. In the U.K., Hajat et al. (2007) analyzed time-series data of daily infant mortality counts in 10 major cities to quantify any associations with short-term changes in air pollution. When the results from the 10 cities were combined there was no relationship between O$_3$ and postneonatal mortality, even after restricting the analysis to just the summer months. In Ciudad Juarez, Mexico, Romieu et al. (2004a) examined the daily number of deaths between 1997 and 2001, estimating the modifying effect of SES on the risk of postneonatal mortality. Ambient O$_3$ concentrations were not related to infant mortality overall, or in any of the SES groups. In a follow-up study, Carbajal-Arroyo et al. (2011) evaluated the
relationship of 1-h daily max O3 levels with postneonatal infant mortality in the
Mexico City Metropolitan Area between 1997 and 2005. Generally, short-term
exposure to O3 was not significantly related to infant mortality. However, upon
estimating the modifying effect of SES on the risk of postneonatal mortality, the
authors found that O3 was statistically significantly related to respiratory mortality
among those with low SES. In a separate analysis, the effect of PM10 was evaluated
with O3 level quartiles. PM10 alone was related to a significant increase in all-cause
mortality. The magnitude of this effect remained the same when only the days when
O3 was in the lowest quartile were included in the analyses. However, when only the
days when O3 was in the highest quartile were included in the analyses, the
magnitude of the PM10 effect increased dramatically (OR = 1.06 [95% CI: 0.909,
1.241] for PM10 on days with O3 in lowest quartile; OR = 1.26 [95% CI: 1.08, 1.47]
for PM10 on days with O3 in the highest quartile. These results suggest that while O3
alone may not have an effect on infant mortality, it may serve to potentiate the
observed effect of PM10 on infant mortality.

Tsai et al. (2006) used a case-crossover analysis to examine the relationship between
short-term exposure to air pollution and postneonatal mortality in Kaohsiung, Taiwan
during the period 1994-2000. The risk of postneonatal deaths was 1.023 (95% CI:
0.564, 1.858) per 10-ppb increase in 24-h avg O3. The confidence interval for this
effect estimate is very wide, likely due to the small number of infants that died each
day, making it difficult to interpret this result. Several other studies conducted in
Asia did not find any association between O3 concentrations and infant mortality in
the postneonatal period. Ha et al. (2003) conducted a daily time-series study in Seoul,
Korea to evaluate the effect of short-term changes in ambient 8-h O3 concentrations
on postneonatal mortality. Son et al. (2008) examined the relationship between air
pollution and postneonatal mortality from all causes among firstborn infants in Seoul,
Korea during 1999-2003. Yang et al. (2006) used a case-crossover analysis to
examine the relationship between air pollution exposure and postneonatal mortality
in Taipei, Taiwan for the period 1994-2000. The authors observed no associations
between ambient levels of O3 and postneonatal mortality.

### 7.4.10.5 Sudden Infant Death Syndrome

The strongest evidence for an association between ambient O3 concentrations and
SIDS comes from a study that evaluated the county-level relationship between SIDS
and long-term early-life exposure (first 2 months of life) to air pollutants across the
U.S. (Woodruff et al., 2008). The authors observed a 1.20 (95% CI: 1.09, 1.32) odds
ratio for a 10-ppb increase in O3 and deaths from SIDS. There was a monotonic
increase in odds of SIDS for each quartile of O3 exposure compared with the lowest
quartile (highest quartile OR = 1.51; [95% CI: 1.17, 1.96]). In a multipollutant model
including PM10 or PM2.5, CO and SO2, the OR for SIDS and O3 was not
substantially lower than that found in the single-pollutant model. When examined by
season, the relationship between SIDS deaths and O3 was generally consistent across
seasons with a slight increase for those babies born in the summer. When stratified
by birth weight, the OR for LBW babies was 1.27 (95% CI: 0.95, 1.69) per 10-ppb increase in \(O_3\) and the OR for normal weight babies was 1.16 (95% CI: 1.01, 1.32) per 10-ppb increase in \(O_3\).

Conversely, two additional studies reported no association between ambient levels of \(O_3\) and SIDS. Ritz et al. (2006) linked birth and death certificates for infants who died between 1989 and 2000 to evaluate the influence of outdoor air pollution on infant death in the South Coast Air Basin of California. The authors examined short- and long-term exposure periods 2 weeks, 1 month, 2 months, and 6 months before each case subject’s death and reported no association between ambient levels of \(O_3\) and SIDS. Dales et al. (2004) used time-series analyses to compare the daily mortality rates for SIDS and short-term air pollution concentrations in 12 Canadian cities during the period of 1984-1999. Increased daily rates of SIDS were associated with previous day increases in the levels of \(SO_2\), \(NO_2\), and \(CO\), but not \(O_3\) or PM\(_{2.5}\).

Table 7-9 provides a brief overview of the epidemiologic studies of infant mortality. These studies have focused on short-term exposure windows (e.g., 1-3 days) and long-term exposure windows (e.g., up to 6 months). Collectively, they provide no evidence for an association between ambient \(O_3\) concentrations and infant mortality.

### Table 7-9: Brief summary of infant mortality studies.

<table>
<thead>
<tr>
<th>Study</th>
<th>Location</th>
<th>Mean (O_3) (ppb)</th>
<th>Exposure Assessment</th>
<th>Effect Estimate(^a) (95% CI):</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pereira et al.</td>
<td>Sao Paulo, Brazil</td>
<td>1-h max: 33.8</td>
<td>Citywide avg</td>
<td>L0-2: 1.00 (0.99, 1.01)</td>
</tr>
<tr>
<td>(1998)</td>
<td></td>
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<tr>
<td>Diaz et al. (2004)</td>
<td>Madrid, Spain</td>
<td>24-h avg: 11.4</td>
<td>Citywide avg</td>
<td>NR</td>
</tr>
<tr>
<td>Loomis et al.</td>
<td>Mexico City, Mexico</td>
<td>24-h avg: 44.1, 1-h max: 163.5</td>
<td>1 monitor</td>
<td>L0: 0.99 (0.97, 1.02)</td>
</tr>
<tr>
<td>(1999)</td>
<td></td>
<td></td>
<td></td>
<td>L1: 0.99 (0.96, 1.01)</td>
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<tr>
<td></td>
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<td></td>
<td>L2: 1.00 (0.98, 1.03)</td>
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<td>L3: 1.03 (1.00, 1.05)</td>
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<td>L4: 1.01 (0.98, 1.03)</td>
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<td>L5: 1.02 (0.99, 1.04)</td>
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<td></td>
<td></td>
<td></td>
<td>L0-2: 1.02 (0.99, 1.05)</td>
</tr>
<tr>
<td>Ritz et al. (2006)</td>
<td>Southern California</td>
<td>24-h avg: 21.9-22.1</td>
<td>Nearest Monitor</td>
<td>2 weeks before death: 1.03 (0.93, 1.14)</td>
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<td></td>
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<td>1 mo before death: NR</td>
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<td></td>
<td>2 mo before death: 0.93 (0.89, 0.97)</td>
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<td></td>
<td></td>
<td></td>
<td>6 mo before death: NR</td>
</tr>
<tr>
<td>Hajat et al. (2007)</td>
<td>10 Cities in the UK</td>
<td>24-h avg: 20.5-42.6</td>
<td>Citywide avg</td>
<td>L0-2: 1.00 (0.96, 1.06)</td>
</tr>
<tr>
<td>Lin et al. (2004a)</td>
<td>Sao Paulo, Brazil</td>
<td>24-h avg: 38.06</td>
<td>Citywide avg</td>
<td>L0: 1.00 (0.99, 1.01)</td>
</tr>
<tr>
<td>Study</td>
<td>Location</td>
<td>Mean O₃ (ppb)</td>
<td>Exposure Assessment</td>
<td>Effect Estimateᵃ (95% CI):</td>
</tr>
<tr>
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</tr>
<tr>
<td>Ha et al. (2003)</td>
<td>Seoul, South Korea</td>
<td>8-h avg: 21.2</td>
<td>Citywide avg</td>
<td>L0: 0.93 (0.90, 0.96)</td>
</tr>
<tr>
<td>Romieu et al. (2004a)</td>
<td>Ciudad Juarez, Mexico</td>
<td>8-h avg: 43.43-55.12</td>
<td>Citywide avg</td>
<td>L1: 0.96 (0.90, 1.03)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>L2: 0.97 (0.91, 1.04)</td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td>L0-1 cum: 0.96 (0.89, 1.04)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>L0-2 cum: 0.94 (0.87, 1.02)</td>
</tr>
<tr>
<td>Carbajal-Arroyo et al. (2011)</td>
<td>Mexico City, Mexico</td>
<td>1-h max: 103.0</td>
<td>Citywide avg</td>
<td>L0: 1.00 (0.99, 1.00)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>L1: 0.99 (0.99, 0.99)</td>
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<tr>
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<td></td>
<td></td>
<td>L2: 0.99 (0.99, 1.00)</td>
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<td></td>
<td></td>
<td></td>
<td>L0-2: 0.99 (0.99, 1.00)</td>
</tr>
<tr>
<td>Son et al. (2008)</td>
<td>Seoul, South Korea</td>
<td>8-h avg: 25.61</td>
<td>Citywide avg</td>
<td>L(NR): 0.984 (0.976, 0.992)ᵇ</td>
</tr>
<tr>
<td>Tsai et al. (2006)</td>
<td>Kaohsiung, Taiwan</td>
<td>24-h avg: 23.60</td>
<td>Citywide avg</td>
<td>L0-2 cum: 1.02 (0.56, 1.86)</td>
</tr>
<tr>
<td>Woodruff et al. (2008)</td>
<td>Nationwide, U.S.</td>
<td>24-h avg: 26.6</td>
<td>County wide avg</td>
<td>First 2 mo of life: 1.04 (0.98, 1.10)</td>
</tr>
<tr>
<td>Yang et al. (2006)</td>
<td>Taipei, Taiwan</td>
<td>24-h avg: 18.14</td>
<td>Citywide avg</td>
<td>L0-2 cum: 1.00 (0.62, 1.61)</td>
</tr>
<tr>
<td>Dales et al. (2004)</td>
<td>12 Canadian cities</td>
<td>24-h: 31.77</td>
<td>Citywide avg</td>
<td>L0: NR</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>L1: NR</td>
</tr>
<tr>
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<td>L2: NR</td>
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<td>L3: NR</td>
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<td>L4: NR</td>
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<td>L5: NR</td>
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<tr>
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<td></td>
<td></td>
<td></td>
<td>Multiday lags of 2-6 days: NR</td>
</tr>
</tbody>
</table>

ᵃRelative risk of infant mortality per 10 ppb change in O₃
ᵇNo increment provided

L0 = Lag 0, L1= Lag 1, L2 = Lag 2, L3 = Lag 3, L4 = Lag 4, L5 = Lag 5, L6 = Lag 6
NR: No quantitative results reported
Table 7-10: Summary of key reproductive and developmental toxicological studies.

<table>
<thead>
<tr>
<th>Study</th>
<th>Model</th>
<th>O₃ (ppm)</th>
<th>Exposure Duration</th>
<th>Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sharkhuu et al. (2011)</td>
<td>Pregnant mice; BALB/c; F; GD9-18; effects in offspring</td>
<td>0.4, 0.8, or 1.2</td>
<td>Continuously for 10 consecutive days</td>
<td>Dams: Decreased number of dams reaching parturition. Offspring: (1)-Decreased birth weights. (2)-Decreased rate of postnatal growth (body weight). (3)-impaired delayed type hypersensitivity.(4)-No effect on humoral immunity. (5)-Significantly affected allergic airway inflammation markers (eosinophilia, IgE) in female offspring sensitized early in life. 6-BALF LDH significantly elevated in female offspring.</td>
</tr>
<tr>
<td>Bignami et al. (1994)</td>
<td>Pregnant CD-1 dams (PD7-17)</td>
<td>0.4, 0.8 or 1.2</td>
<td>Continuous</td>
<td>Reproductive success was not affected by O₃ exposure (PD7-17, proportion of successful pregnancies, litter size, ex ratio, frequency of still birth, or neonatal mortality). Ozone acted as a transient anorexigen in pregnant dams.</td>
</tr>
<tr>
<td>Haro and Paz (1993)</td>
<td>Rat dams, Exposure over the entirety of pregnancy;</td>
<td>1.0</td>
<td>12h/day during dark cycle</td>
<td>Decreased birth weight and postnatal body weight of offspring out to PND 90. Ozone-exposed pups manifested with inverted sleep-wake patterns or circadian rhythm phase-shift.</td>
</tr>
<tr>
<td>López et al. (2008)</td>
<td>Rats; Pregnant dams; GD1-GD18, GD20, or GD21</td>
<td>1.0</td>
<td>(12 h/day, out to either GD18, GD20 or GD21)</td>
<td>O₃ induced delayed maturation of near term rodent bronchioles, with ultra-structural damage to bronchiolar epithelium.</td>
</tr>
<tr>
<td>Auten et al. (2009)</td>
<td>C57BL/6 mouse pups</td>
<td>1.0</td>
<td>3 h/day, every other day, thrice weekly for 4 weeks</td>
<td>Postnatal O₃ exposure significantly increased lung inflammatory cytokine levels; this was further exacerbated with gestational PM exposure.</td>
</tr>
<tr>
<td>Plopper et al. (2007)</td>
<td>Infant rhesus monkeys</td>
<td>0.5</td>
<td>Postnatal, PND30-6month of age, 5 months of cyclic exposure, 5 days O₃ followed by 9 days of filtered air, 8h/day.</td>
<td>Non-significant increases airway resistance and airway responsiveness with O₃ or inhaled allergen alone. Allergen + O₃ produced additive changes in both measures.</td>
</tr>
<tr>
<td>Fanucchi et al. (2006)</td>
<td>Infant male Rhesus monkeys, post-natal exposure</td>
<td>0.5</td>
<td>5 months of episodic exposure, age 1 month-age 6 months, 5 days O₃ followed by 9 days of filtered air, 8h/day.</td>
<td>Cellular changes and significant structural changes in the distal respiratory tract in infant rhesus monkeys exposed to O₃ postnatally.</td>
</tr>
<tr>
<td>Dell'Omo et al. (1995)</td>
<td>CD-1 Mouse dams and pups</td>
<td>0.6</td>
<td>6 days before breeding to weaning at PND21</td>
<td>Laterality changes in offspring: Ozone exposed pups showed a turning preference (left turns) distinct from air exposed controls (clockwise turns) as adults.</td>
</tr>
<tr>
<td>Santucci et al. (2006)</td>
<td>CD-1 Mouse dams</td>
<td>0.3 or 0.6</td>
<td>Dam exposure prior to mating through GD17.</td>
<td>Developmental O₃ caused increased defensive/submissive behavior in offspring. Ozone exposed offspring also had significant elevations of striatal BDNF and hippocampal NGF v. air exposed controls.</td>
</tr>
</tbody>
</table>
### Study Model

**Han et al. (2011)**

*Model*: Rat; Sprague Dawley, M & F; PND13  
*O3 (ppm)*: 0.6  
*Exposure Duration*: 3 h, BALF examined 10h after O3 exposure  
*Effects*: BALF polymorphonuclear leukocytes and total BALF protein were significantly elevated in O3 exposed pups. Lung tissue from O3 exposed pups had significant elevations of manganese superoxide dismutase (SOD) protein and significant decrements of extra-cellular SOD protein.

**Campos-Bedolla et al. (2002)**

*Model*: Pregnant Rats; Sprague Dawley (GD5, GD10, or GD18)  
*O3 (ppm)*: 3.0  
*Exposure Duration*: 1 h on one day of gestation, uteri collected 16-18 h later  
*Effects*: Ozone inhalation modifies the contractile response of the pregnant uterus. The O3 exposed pregnant uteri had significant increases in the maximum response to acetyl choline stimulation at GD5 and 10; they also had a significant increase in maximal response to oxytocin at GD 5.

**Kavlock et al. (1980)**

*Model*: CD-1 mice; (pregnancy day 7-17)  
*O3 (ppm)*: 0.4, 0.8 and 1.2  
*Exposure Duration*: Continuous, pregnancy day 7-17  
*Effects*: O3 induced decrements in postnatal body weight gain. When O3 was co-administered with sodium salicylate, O3 synergistically increased the rate of pup resorption (1.0 ppm GD9-12).

**Jedlinska-Krakowska et al. (2006)**

*Model*: 5 month old male Wistar Hannover rats  
*O3 (ppm)*: 3.0  
*Exposure Duration*: 0.5 ppm, 5h/day for 50 days  
*Effects*: Histopathological evidence of impaired spermatogenesis (round spermatids/ 21 spermatocytes, giant spermatid cells, and focal epithelial desquamation with denudation to the 22 basement membrane). Vitamin E exposure concomitant with O3 protected against pathological changes but Vitamin C did not.

### 7.4.11 Summary and Causal Determination

The 2006 O3 AQCD concluded that the limited number of studies that investigated O3 demonstrated no associations between O3 and birth outcomes, with the possible exception of birth defects. The current review included an expanded body of evidence on the associations between O3 and reproductive and developmental effects. Recent epidemiologic and toxicological studies provide evidence for an effect of prenatal exposure to O3 on pulmonary structure and function, including lung function changes in the newborn, incident asthma, ultrastructural changes in bronchiole development, alterations in placental and pup cytokines, and increased pup airway hyper-reactivity. Also, there is limited toxicological evidence for an effect of prenatal and early life exposure on central nervous system effects, including laterality, brain morphology, neurobehavioral abnormalities, and sleep aberration. Recent epidemiologic studies have begun to explore the effects of O3 on sperm quality, and provide limited evidence for decrements in sperm concentration, while there is limited toxicological evidence for testicular degeneration associated with O3.

While the collective evidence for many of the birth outcomes examined is generally inconsistent (including birth defects), there are several well-designed, well-conducted studies that indicate an association between O3 and adverse outcomes. For example, as part of the southern California Children’s Health Study, Salam et al. (2005)
observed a concentration-response relationship of decreasing birth weight with increasing O₃ concentrations averaged over the entire pregnancy that was clearest above the 30-ppb level (see Figure 7-4). Similarly, Hansen et al. (2008) utilized fetal ultrasonic measurements and found a change in ultrasound measurements associated with O₃ during days 31-60 of gestation indicated that increasing O₃ concentration decreased an ultrasound measurement for women living within 2 km of the monitoring site.

The weight of evidence does not indicate that prenatal or early life O₃ concentrations are associated with infant mortality. Collectively, there is limited though positive toxicological evidence for O₃-induced developmental effects, including effects on pulmonary structure and function and central nervous system effects. Limited epidemiologic evidence for an effect on prenatal O₃ exposure on respiratory development provides coherence with the effects observed in toxicological studies. There is also limited epidemiologic evidence for an association with O₃ concentration and decreased sperm concentration. A recent toxicological study provides limited evidence for a possible biological mechanism (histopathology showing impaired spermatogenesis) for such an association. Additionally, though the evidence for an association between O₃ concentrations and adverse birth outcomes is generally inconsistent, there are several influential studies that indicate an association with reduced birth weight and restricted fetal growth.

Some of the key challenges to interpretation of these study results include the difficulty in assessing exposure as most studies use existing monitoring networks to estimate individual exposure to ambient air pollution (see Section 4.6); the inability to control for potential confounders such as other risk factors that affect birth outcomes (e.g., smoking); evaluating the exposure window (e.g., trimester) of importance; integrating the results from both short- and long-term exposure periods; integrating the results across a variety of reproductive and developmental outcomes; and limited evidence on the physiological mechanism of these effects.

Taking into consideration the positive evidence for developmental and reproductive outcomes from toxicological and epidemiological studies, and the few influential birth outcome studies, the evidence is suggestive of a causal relationship between exposures to O₃ and reproductive and developmental effects.

### 7.5 Central Nervous System Effects

#### 7.5.1 Effects on the Brain and Behavior

The 2006 O₃ AQCD (U.S. EPA, 2006b) included toxicological evidence that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short and long term memory, and sleep patterns. Additionally, histological signs of neurodegeneration have been observed. Reports of headache, dizziness, and irritation
of the nose with O3 exposure are common complaints in humans, and some behavioral changes in animals may be related to these symptoms rather than indicative of neurotoxicity. Research in the area of O3-induced neurotoxicity has notably increased over the past few years, and recent studies examining the effects of long-term exposure have demonstrated progressive damage in various regions of the brains of rodents in conjunction with altered behavior. Evidence from epidemiologic studies has been more limited. A recently published epidemiologic study examined the association between O3 concentration and neurobehavioral effects. Chen and Schwartz (2009) utilized data from the NHANES III cohort to study the relationship between O3 concentrations (mean annual O3 concentration 26.5 ppb) and neurobehavioral effects among adults aged 20-59 years. Annual O3 concentration was determined using inverse distance weighting for county of residence and adjacent counties (for more information on inverse distance weighting and other methods for exposure assessment, see Sections 4.5.1 and 4.6). The authors observed an association between annual O3 concentration and tests measuring coding ability (symbol-digit substitution test) and attention/short-term memory (serial-digit learning test). Each 10-ppb increase in annual O3 concentration corresponded to an aging-related cognitive performance decline of 3.5 yr for coding ability and 5.3 years for attention/short-term memory. These associations persisted in both crude and adjusted models. There was no association between O3 concentration and reaction time tests. The authors concluded that overall, there is an association between long-term O3 concentration and reduced performance on neurobehavioral tests.

A number of recent toxicological studies demonstrate various perturbations in neurologic function or histology with long-term exposure to O3, including changes similar to those observed in neurodegenerative disorders such as Parkinson’s and Alzheimer’s disease pathologies in relevant regions of the brain (Table 7-11). The central nervous system is very sensitive to oxidative stress, due in part to its high content of polyunsaturated fatty acids, high rate of oxygen consumption, and low antioxidant enzyme capacity. Oxidative stress has been identified as one of the pathophysiological mechanisms underlying neurodegenerative disease (Simonian and Coyle, 1996), and it is believed to play a role in altering hippocampal function, which causes cognitive deficits with aging (Vanguilder and Freeman, 2011). A particularly common finding in studies of O3-exposed rats is lipid peroxidation in the brain, especially in the hippocampus, which is important for higher cognitive function including contextual memory acquisition. Performance in passive avoidance learning tests is impaired when the hippocampus is injured. For example, in a subchronic study, exposure of rats to 0.25 ppm O3 (4 h/day) for 15-90 days caused a complex array of responses, including a time-dependent increase in lipid peroxidation products and immunohistochemical changes in the hippocampus that were correlated with decrements in passive avoidance behavioral tests (Rivas-Arancibia et al., 2010). Changes included increased numbers of activated microglia, a sign of inflammation, and progressive neurodegeneration. Notably, continued exposure tends to bring about progressive, cumulative damage, as shown by this study (Rivas-Arancibia et al., 2010) and others (Santiago-López et al., 2010; Guevara-Guzmán et al., 2009; Angoa-Pérez et al., 2006). The effects of O3 on passive avoidance test performance were particularly evident at 90 days for both
short- and long-term memory. The greatest extent of cell loss was also observed at this time point, whereas lipid peroxidation did not increase much beyond 60 days of exposure.

The substantia nigra is another region of the brain affected by O₃, and seems particularly sensitive to oxidative stress because the metabolism of dopamine, central to its function, is an oxidative process perturbed by redox imbalance. Oxidative stress has been implicated in the premature death of substantia nigra dopamine neurons in Parkinson’s disease. Progressive damage has been found in the substantia nigra of male rats after 15, 30, and 60 days of exposure to 0.25 ppm O₃ for 4 h/day. Santiago-López et al. (2010) observed a reduction dopaminergic neurons within the substantia nigra over time, with a complete loss of normal morphology in the remaining cells and virtually no dopamine immunoreactivity at 60 days. This was accompanied by an increase in p53 levels and nuclear translocation, a process associated with programmed cell death. Similarly, Angoa-Pérez et al. (2006) have shown progressive lipoperoxidation in the substantia nigra and a decrease in nigral neurons in ovariectomized female rats exposed to 0.25 ppm O₃, 4h/day, for 7 - 60 days. Lipid peroxidation effectively doubled between the 30 and 60 day time points. Total nigral cell number was also diminished to the greatest extent at 60 days, and cell loss was particularly evident in the tyrosine hydroxylase positive cell population (90%), indicating a selective loss of dopamine neurons or a loss of dopamine pathway functionality.

The olfactory bulb also undergoes oxidative damage in O₃-exposed animals, in some cases altering olfactory-dependent behavior. Lipid peroxidation was observed in the olfactory bulbs of ovariectomized female rats exposed to 0.25 ppm O₃ (4 h/day) for 30 or 60 days (Guevara-Guzmán et al., 2009). Ozone also induced decrements in a selective olfactory recognition memory test, which were significantly greater at 60 days compared to 30 days, and the authors note that early deficits in odor perception and memory are components of human neurodegenerative diseases. The decrements in olfactory memory did not appear to be due to damaged olfactory perception based on other tests early on, but by 60 days deficits in olfactory perception had emerged.

Memory deficits and associated morphological changes can be attenuated by administration of α-tocopherol (Guerrero et al., 1999), taurine (Rivas-Arancibia et al., 2000), and estradiol (Guevara-Guzmán et al., 2009; Angoa-Pérez et al., 2006), all of which have antioxidant properties. In the study by Angoa-Pérez et al. (2006) described above, estradiol seemed particularly effective at protecting against lipid peroxidation and nigral cell loss at 60 days compared to shorter exposure durations. The same was true for amelioration of decrements in olfactory recognition memory (Guevara-Guzmán et al., 2009), although protection against lipid peroxidation was similar for the 30 and 60 day exposures.

CNS effects have also been demonstrated in adult mice whose only exposure to O₃ occurred while in utero, a period particularly critical for brain development. Santucci et al. (2006) investigated behavioral effects and gene expression after in utero exposure of mice to 0.3 or 0.6 ppm O₃. Exposure began 30 days prior to mating and
continued throughout gestation. Testing of adult animals demonstrated increased
defensive/submissive behavior and reduced social investigation in both the 0.3 and
0.6 ppm O₃ groups. Changes in gene expression of brain-derived neurotrophic factor
(BDNF, increased in striatum) and nerve growth factor (NGF, decreased in
hippocampus) accompanied these behavioral changes. BDNF and NGF are involved
in neuronal organization and the growth, maintenance, and survival of neurons
during early development and in adulthood. This study and two others using short-
term exposures demonstrate that CNS effects can occur as a result of in utero
exposure to O₃, and although the mode of action of these effects is not known, it has
been suggested that circulating lipid peroxidation products may play a role
(Boussouar et al., 2009). Importantly, these CNS effects occurred in rodent models
after in utero only exposure to (semi-) relevant concentrations of O₃.
### Table 7-11 Central nervous system effects of long-term O₃ exposure in rats.

<table>
<thead>
<tr>
<th>Study</th>
<th>Model</th>
<th>O₃ (ppm)</th>
<th>Exposure Duration</th>
<th>Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Angoa-Pérez et al. (2006)</td>
<td>Rat; Wistar; F; Weight: 300 g; Ovariectomized</td>
<td>0.25</td>
<td>7 to 60 days, 4 h/day, 5 days/week</td>
<td>Long-term estradiol treatment protected against O₃-induced oxidative damage to nigral dopamine neurons, lipid peroxidation, and loss of tyrosine hydrolase-immunopositive cells.</td>
</tr>
<tr>
<td>Guevara-Guzmán et al. (2009)</td>
<td>Rat; Wistar; F; Weight: 264 g; Ovariectomized</td>
<td>0.25</td>
<td>30 and 60 days, 4h/day</td>
<td>Long-term estradiol treatment protected against O₃-induced oxidative stress and decreases in α and β estrogen receptors and dopamine β-hydroxylase in olfactory bulb, and deficits in olfactory social recognition memory and chocolate recognition.</td>
</tr>
<tr>
<td>Rivas-Arancibia et al. (2010)</td>
<td>Rat; Wistar; M; Weight: 250-300 g</td>
<td>0.25</td>
<td>15 to 90 days, 4h/day</td>
<td>Ozone produced significant increases in lipid peroxidation in the hippocampus, and altered the number of p53 positive immunoreactive cells, activated and phagocytic microglia, GFAP immunoreactive cells, double cortine cells, and short- and long-term memory-retention latency</td>
</tr>
<tr>
<td>Santiago-López et al. (2010)</td>
<td>Rat; Wistar; M; Weight: 250-300 g</td>
<td>0.25</td>
<td>15, 30, and 60 days, 4 h/day</td>
<td>Progressive loss of dopamine reactivity in the substantia nigra, along with morphological changes. Increased p53 levels and nuclear translocation.</td>
</tr>
<tr>
<td>Santucci et al. (2006)</td>
<td>Mice; CD-1; M; 18 weeks old</td>
<td>0.3; 0.6</td>
<td>Females continuously exposed from 30 days prior to breeding until GD17</td>
<td>Upon behavioral challenge with another male, there was a significant increase in defensive and freezing postures and decrease in the frequency of nose-sniffing. These behavioral changes were accompanied by a significant increase in BDNF in the striatum and a decrease of NGF in the hippocampus.</td>
</tr>
</tbody>
</table>

#### 7.5.2 Summary and Causal Determination

The 2006 O₃ AQCD included toxicological evidence that acute exposures to O₃ are associated with alterations in neurotransmitters, motor activity, short and long term memory, and sleep patterns. Additionally, histological signs of neurodegeneration have been observed. However, evidence regarding chronic exposure and neurobehavioral effects was not available. Recent research in the area of O₃-induced neurotoxicity has included several long-term exposure studies. Notably, the first
epidemiologic study to examine the relationship between O$_3$ exposure and neurobehavioral effects observed an association between annual O$_3$ levels and an aging-related cognitive performance decline in tests measuring coding ability and attention/short-term memory. This observation is supported by studies in rodents which demonstrate progressive oxidative stress and damage in the brain and associated decrements in behavioral tests, including those measuring memory, after subchronic exposure to 0.25 ppm O$_3$. Additionally, neurobehavioral changes are evident in animals whose only exposure to O$_3$ occurred in utero. Collectively, the limited epidemiologic and toxicological evidence is coherent and suggestive of a causal relationship between O$_3$ exposure and CNS effects.

7.6 Carcinogenic and Genotoxic Potential of Ozone

7.6.1 Introduction

The radiomimetic and clastogenic qualities of O$_3$, combined with its ability to stimulate proliferation of cells in the respiratory tract, have suggested that O$_3$ could act as a carcinogen. However, toxicological studies of tumorigenesis in the rodent lung have yielded mixed and often confusing results, and the epidemiologic evidence is equally conflicted. The 2006 O$_3$ AQCD concluded that, “the weight of evidence from recent animal toxicological studies and a very limited number of epidemiologic studies do not support ambient O$_3$ as a pulmonary carcinogen”\(^1\) (U.S. EPA, 2006b).

Multiple epidemiologic studies reported in the 2006 O$_3$ AQCD examined the association between O$_3$ concentration and cancer. The largest of these studies, by Pope et al. (2002), included 500,000 adults from the American Cancer Society’s (ACS) Cancer Prevention II study. In this study, no association was observed between O$_3$ concentration and lung cancer mortality. The Adventist Health Study of Smog (AHSMOG) also examined the association between O$_3$ concentration and lung cancer mortality (Abbey et al., 1999). There was a positive association between O$_3$ concentrations and lung cancer mortality among men. No association was reported for women. Another study using the AHSMOG cohort assessed the risk of incident lung cancer (Becson et al., 1998). Among males, an association with incidence of lung cancer was observed with increasing O$_3$ concentrations. When stratified by smoking status, the association persisted among never smokers but was null for former smokers. No association was detected for females. The Six Cities Study examined various air pollutants and mortality but did not specifically explore the association between O$_3$ concentrations and lung cancer mortality due to low variability in O$_3$ concentrations across the cities (Dockery et al., 1993). An ecologic study performed in Sao Paulo City, Brazil examined the correlations between O$_3$ concentrations in four of the city districts and incident cancer of the larynx and lung

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\(^1\) The toxicological evidence is presented in detail in Table 6-18 on page 6-116 of the 1996 O$_3$ AQCD (U.S. EPA, 1996a) and Table AX5-13 on page AX5-43 of the 2006 O$_3$ AQCD (U.S. EPA, 2006b).
reported in 1997 (Pereira et al., 2005). A correlation between the average number of days O₃ concentrations exceeded air quality standards from 1981 to 1990 and cancer incidence was present for larynx cancer but not for lung cancer.

Early toxicological research demonstrated lung adenoma¹ acceleration in mice with daily exposure to 1 ppm over 15 months (Stokinger, 1962). Later work demonstrated a significant increase in lung tumor numbers in one strain of mouse (A/J) but not another after exposure to 0.3-0.8 ppm O₃ (Last et al., 1987; Hassett et al., 1985). The A/J mouse strain is known to have a high incidence of spontaneous adenomas, and further studies using this strain found a statistically significant increase in lung tumor incidence after a 9-month exposure to 0.5 ppm and incidence and multiplicity after a 5 month exposure to 0.12 ppm with a 4-month recovery period (Witschi et al., 1999). However, these findings were discounted by the study authors due to the lack of a clear concentration-response, and results from the Hassett et al. (1985) and Last et al. (1987) studies were retrospectively deemed spurious based on what appeared to be unusually low spontaneous tumor incidences in the control groups (Witschi, 1991). A study of carcinogenicity of O₃ by the National Toxicology Program (NTP, 1994) reported increased incidences of alveolar/bronchiolar adenoma or carcinoma (combined) in female B6C3F₁ mice exposed over 2 years to 1.0 ppm O₃, but not 0.12 or .5 ppm. No effect was detected in male mice. For a lifetime exposure to 0.5 or 1.0 ppm O₃, an increase in the number of female mice with adenomas (but not carcinomas or total neoplasms) was found. The number of total neoplasms was also unaffected in male mice, but there was a marginally increased incidence of carcinoma in males exposed to 0.5 and 1.0 ppm. Thus there was equivocal evidence of carcinogenic activity in male mice and some evidence of carcinogenic activity of O₃ in females. Experimental details of the NTP mouse study are available in Table 6-19 on page 6-121 (U.S. EPA, 1996o) of the 1996 O₃ AQCD (U.S. EPA, 1996a).

In Fischer-344/N rats (50 of each sex per group), neither a 2-year nor lifetime exposure to O₃ ranging from 0.12 to 1.0 ppm was found to be carcinogenic (Boorman et al., 1994; NTP, 1994). However, a marginally significant carcinogenic effect of 0.2 ppm O₃ was reported in a study of male Sprague-Dawley rats exposed for 6 months (n = 50) (Monchaux et al., 1996). These two studies also examined co-carcinogenicity of O₃ with NNK² (Boorman et al., 1994) or a relatively high dose of radon (Monchaux et al., 1996), finding no enhancement of NNK related tumors and a slight non-significant increase in tumor incidence after combined exposure with radon, respectively. Another study exploring co-carcinogenicity was conducted in hamsters. Not only was there no enhancement of chemically induced tumors in the peripheral lung or nasal cavity, but results suggested that O₃ could potentially delay or inhibit tumor development (Witschi et al., 1993). Thus there is no concrete evidence that O₃ can act as a co-carcinogen.

¹ NOTE: Although adenomas are benign, over time they may progress to become malignant, at which point they are called adenocarcinomas. Adenocarcinoma is the predominant lung cancer subtype in most countries, and is the only lung cancer found in nonsmokers. From page 8-33 of the 1970 O₃ AQCD: “No true lung cancers have been reported, however, from experimental exposures to either O₃ alone or any other combination or ingredient of photochemical oxidants.”

² 4-(N-nitrosomethylamino)-1-(3-pyridyl)-1-butanone
Immune surveillance is an important defense against cancer, and it should be noted that natural killer (NK) cells, which destroy tumor cells in the lung, appear to be inhibited by higher concentrations of O\textsubscript{3} and either unaffected or stimulated at lower concentrations (Section 6.2.5.4, Infection and Adaptive Immunity). This aspect of tumorigenesis adds yet another layer of complexity which may be reflected by conflicting results across studies.

The following sections will examine epidemiologic studies of cancer incidence and mortality and toxicological studies that have been published since the 2006 O\textsubscript{3} AQCD. An epidemiologic study has been published with cancer as the outcome; most epidemiologic studies examine markers of exposure.

### 7.6.2 Lung Cancer Incidence and Mortality

A recent re-analysis of the full ACS CPSII cohort by the Health Effects Institute is the only epidemiologic study that has explored the association between O\textsubscript{3} concentration and cancer mortality since the last O\textsubscript{3} AQCD. Krewski et al. (2009) conducted an extended follow-up of the cohort (1982-2000). Mean O\textsubscript{3} concentration [obtained from the Aerometric Information Retrieval System (AIRS) for 1980] were 22.91 ppb for the full year and 30.15 ppb for the summer months (April-September). No association was reported between lung cancer mortality and O\textsubscript{3} concentration (HR = 1.00 [95% CI: 0.96-1.04] per 10 ppb O\textsubscript{3}). Additionally, no association was observed when the analysis was restricted to the summer months. There was also no association present in a sub-analysis of the cohort examining the relationship between O\textsubscript{3} concentration and lung cancer mortality in the Los Angeles area.

Since the 2006 O\textsubscript{3} AQCD, two toxicological studies have examined potential carcinogenicity of O\textsubscript{3} (Kim and Cho, 2009a, b). Looking across both studies, which used the same mouse strain as the National Toxicology Program study described above (NTP, 1994), 0.5 ppm O\textsubscript{3} alone or in conjunction with chemical tumor inducers did not enhance lung tumor incidence in males or females. However, a 10% incidence of oviductal carcinoma was observed in mice exposed to 0.5 ppm O\textsubscript{3} for 16 weeks. The implications of this observation are unclear, particularly in light of the lack of statistical information reported. Additionally, there is no mention of oviductal carcinoma after 32 weeks of exposure, and no oviductal carcinoma was observed after one year of exposure. The NTP study did not report any increase in tumors at extrapulmonary sites.

### 7.6.3 DNA Damage

The potential for genotoxic effects relating to O\textsubscript{3} exposure was predicted from the radiomimetic properties of O\textsubscript{3}. The decomposition of O\textsubscript{3} in water produces OH and HO\textsubscript{2} radicals, the same species that are generally considered to be the biologically active products of ionizing radiation. Ozone has been observed to cause degradation
of DNA in a number of different models and bacterial strains. The toxic effects of O₃ have been generally assumed to be confined to the tissues directly in contact with the gas, such as the respiratory epithelium. Due to the highly reactive nature of O₃, little systemic absorption is predicted. Zelac et al. (1971a, b); however, reported a significant increase in chromosome aberrations in peripheral blood lymphocytes from Chinese hamsters exposed to 0.2 ppm for 5 hours. Other in vivo exposure studies found increased DNA strand breaks in respiratory cells from guinea pigs (Ferng et al., 1997) and mice (Bornholdt et al., 2002) but only with exposure to higher concentrations of O₃ (1 ppm for 72 hours and 1 or 2 ppm for 90 minutes, respectively). In other studies there were no observations of chromosomal aberrations in germ cells, but mutagenic effects have been seen in offspring of mice exposed to 0.2 ppm during gestation (blepharophimosis or dysplasia of the eyelids). The overall evidence for mutagenic activity from in vitro studies is positive, and in the National Toxicology Program report described above, O₃ was found to be mutagenic in Salmonella, with and without S9 metabolic activation. No recent toxicological studies of DNA damage have become available since the 2006 O₃ AQCD.

A number of epidemiologic studies looked at the association between O₃ and DNA and cellular level damages. These changes may be relevant to mechanisms leading to cancers development and serve as early indicators of elevated risk of mutagenicity.

Two studies performed in California examined cytogenetic damage in relation to O₃ exposures. Huen et al. (2006) examined cytogenetic damage among African American children and their mothers in Oakland, CA. Increased O₃ (mean monthly 8-h O₃ concentrations ranged from about 30 ppb in April to 14 ppb in November) was associated with increased cytogenetic damage (micronuclei frequency among lymphocytes and buccal cells) even after adjustment for household/personal smoking status and distance-weighted traffic density. Chen et al. (2006a) recruited college students at the University of California, Berkeley who reported never smoking and compared their levels of cytogenetic damage (micronuclei frequency from buccal cells) in the spring and fall. Cytogenetic damage was greater in the fall, which the authors attributed to the increase in O₃ over the summer. However, O₃ levels over 2, 7, 10, 14, or 30 days (concentrations not given) before collection of buccal cells did not correlate with cytogenetic damage. Estimated lifetime O₃ exposure was also not correlated with cytogenetic damage. Additionally, the authors exposed a subset of the students (n = 15) to 200 ppb O₃ for 4 hours while the students exercised intermittently. Ozone was found to be associated with an increase in cytogenetic damage in degenerated cells but not in normal cells 9-10 days after exposure. Increased cytogenetic damage was also noted in peripheral blood lymphocytes collected 18 hours after exposure.

A study performed in Mexico recruited 55 male workers working indoors (n = 27) or outdoors (n = 28) in Mexico City or Puebla, Mexico in order to study the relationship between O₃ and DNA damage (detected from peripheral blood samples using the Comet assay) (Tovalin et al., 2006). The median estimated daily O₃ concentrations were estimated to be 28.5 ppb for outdoor workers and 5.1 ppb for indoor workers in
Mexico City and 36.1 ppb for outdoor workers and 19.5 ppb for indoor workers in Puebla. Overall, a positive correlation between O₃ levels and DNA damage was observed. However, when examining the relationship by city and workplace, only DNA damage in outdoor workers in Mexico City remained correlated with O₃ levels.

Three studies examining the relationship between O₃ concentration and DNA-level damage have been performed in Europe. The largest of these studies was the GenAir case-control study, which was nested within the European Prospective Investigation into Cancer and Nutrition (EPIC) study, and included individuals recruited between 1993 and 1998 from ten European countries. Only non-smokers (must not have smoked for at least 10 years prior to enrollment) were enrolled in the study. The researchers examined DNA adduct levels (DNA bonded to cancer-causing chemicals) and their relationship with O₃ concentrations (concentrations not given) (Peluso et al., 2005). A positive association was seen between DNA adduct levels and O₃ concentrations from 1990-1994 but not O₃ concentrations from 1995-1999. In adjusted analyses with DNA adduct levels dichotomized as high and low (detectable versus non-detectable), the OR was 1.97 (95% CI: 1.08, 3.58) when comparing the upper tertile of O₃ concentration to the lower two tertiles. Two other European studies were conducted in Florence, Italy. The most recent of these enrolled individuals from the EPIC study into a separate study between March and September of 1999 (Palli et al., 2009). The purpose of the study was to examine oxidative DNA damage (determined by Comet assay using blood lymphocytes) in association with varying periods of O₃ exposure. The researchers observed that longer periods of high O₃ concentrations (values not given) were more strongly correlated with oxidative DNA damage than shorter periods of time (i.e., the rho [p-value] was 0.26 [0.03] for 0-10 days and 0.35 [0.002] for 0-90 days). This correlation was stronger among men compared to women. The correlations for all time periods had p-values <0.05 for ex- and never-smokers. For current smokers, the correlation was only observed among time periods ≤25 days. When adjusted for age, sex, smoking history, traffic pollution exposure, period of blood draw, and area of residence, the association between O₃ concentrations and oxidative DNA damage was positive for O₃ concentrations 0-60 days, 0-75 days, and 0-90 days prior to blood draw. Positive, statistically significant associations were not observed among shorter time periods. The other study performed in Florence recruited healthy volunteers who reported being non-smokers or light smokers (Giovannelli et al., 2006). The estimated O₃ concentrations during the study ranged from approximately 4-40 ppb for 3-day averages, 5-35 ppb for 7-day averages, and 7.5-32.5 ppb for 30-day averages. Ozone concentrations were correlated with DNA strand breaks (measured from blood lymphocytes) over longer exposure periods (p-value: 0.002 at 30 days, p-value: 0.04 at 7 days; p-value: 0.17 at 3 days). This association was robust to control for temperature, solar radiation, sex, and age. No association was seen between O₃ concentrations and measures of oxidative DNA damage at 3, 7, or 30 days.
7.6.4 Summary and Causal Determination

The 2006 O₃ AQCD reported that evidence did not support ambient O₃ as a pulmonary carcinogen. Since the 2006 O₃ AQCD, very few epidemiologic and toxicological studies have been published that examine O₃ as a carcinogen, but collectively, study results indicate that O₃ may contribute to DNA damage. Ozone concentrations in most epidemiologic studies were measured using air monitoring data. For more information on long-term exposure assessment, see Section 4.6.3.2. Overall, the evidence is inadequate to determine if a causal relationship exists between ambient O₃ exposures and cancer.

7.7 Mortality

A limited number of epidemiologic studies have assessed the relationship between long-term exposure to O₃ and mortality in adults. The 2006 O₃ AQCD concluded that an insufficient amount of evidence existed “to suggest a causal relationship between chronic O₃ exposure and increased risk for mortality in humans” (U.S. EPA, 2006b). In addition to the infant mortality studies discussed in Section 7.4.10, additional studies have been conducted among adults since the last review; an ecologic study that finds no association between mortality and O₃, several re-analyses of the ACS cohort, one of which specifically points to a relationship between long-term O₃ exposure and an increased risk of respiratory mortality, and a study of four cohorts of persons with potentially predisposing conditions. These studies supplement the evidence from long-term cohort studies characterized in previous reviews of O₃, and are summarized here briefly.

In the Harvard Six Cities Study (Dockery et al., 1993), adjusted mortality rate ratios were examined in relation to long-term mean O₃ concentrations in six cities: Topeka, KS; St. Louis, MO; Portage, WI; Harriman, TN; Steubenville, OH; and Watertown, MA. Mean O₃ concentrations from 1977 to 1985 ranged from 19.7 ppb in Watertown to 28.0 ppb in Portage. Long-term mean O₃ concentrations were not found to be associated with mortality in the six cities. However, the authors noted that “The small differences in O₃ levels among the (six) cities limited the power of the study to detect associations between mortality and O₃ levels.” In addition, while total and cardio-pulmonary mortality were considered in this study, respiratory mortality was not specifically considered.

In a subsequent large prospective cohort study of approximately 500,000 U.S. adults, Pope et al. (2002) examined the effects of long-term exposure to air pollutants on mortality (American Cancer Society, Cancer Prevention Study II). All-cause, cardiopulmonary, lung cancer and other mortality risk estimates for long-term O₃ exposure are shown in Figure 7-5. While consistently positive associations were not observed between O₃ and mortality (effect estimates labeled “A” in Figure 7-5), the mortality risk estimates were larger in magnitude when analyses considered more accurate exposure metrics, increasing when the entire period was considered (effect...
estimates labeled “B” in Figure 7-5) and becoming marginally significant when the exposure estimate was restricted to the summer months (July to September; effect estimates labeled “C” in Figure 7-5), especially when considering cardiopulmonary deaths. In contrast, consistent positive and significant effects of PM$_{2.5}$ were observed for both lung cancer and cardio-pulmonary mortality.

<table>
<thead>
<tr>
<th>Years of Data Collection</th>
<th>Number of Metropolitan Areas</th>
<th>Number of Participants (in thousands)</th>
<th>1-h max O$_3$ Mean (SD)</th>
</tr>
</thead>
<tbody>
<tr>
<td>A 1980-1981</td>
<td>134</td>
<td>559</td>
<td>47.9 (11.0)</td>
</tr>
<tr>
<td>B 1982-1998</td>
<td>119</td>
<td>525</td>
<td>45.5 (7.3)</td>
</tr>
<tr>
<td>C 1982-1998 (July – Sept)</td>
<td>134</td>
<td>557</td>
<td>59.7 (12.8)</td>
</tr>
</tbody>
</table>

Source: Reprinted with permission of American Medical Association Pope et al. (2002).

Figure 7-5 Adjusted O$_3$-mortality relative risk estimates (95% CI) by time period of analysis per subject-weighted mean O$_3$ concentration in the Cancer Prevention Study II by the American Cancer Society.

A study by Abbey et al. (1999) examined the effects of long-term air pollution exposure, including O$_3$, on all-cause (n = 1,575), cardiopulmonary (n = 1,029), nonmalignant respiratory (n = 410), and lung cancer (n = 30) mortality in the long-term prospective Adventist Health Study of Smog (AHSMOG) of 6,338 nonsmoking, non-Hispanic white individuals living in California. A particular strength of this study was the extensive effort devoted to assessing long-term air pollution exposures, including interpolation to residential and work locations from monitoring sites over time and space. No associations with long-term O$_3$ exposure were observed for all cause, cardiopulmonary, and nonmalignant respiratory mortality. In a follow-up, Chen et al. (2005) utilized data from the AHSMOG study and reported no evidence of associations between long-term O$_3$ exposure (mean O$_3$ concentration 26.2 ppb)
and fatal coronary heart disease. Thus, no association of chronic O₃ exposure with mortality outcomes has been detected in this study.

Lipfert et al. (2003, 2000) reported positive effects on all-cause mortality for peak O₃ exposures (95th percentile levels) in the U.S. Veterans Cohort study of approximately 50,000 middle-aged men recruited with a diagnosis of hypertension. The actual analysis involved smaller subcohorts based on exposure and mortality follow-up periods. Four separate exposure periods were associated with three mortality follow-up periods. For concurrent exposure periods, peak O₃ was positively associated with all-cause mortality, with a 9.4% (95% CI: 0.4, 18.4) excess risk per mean 95th percentile O₃ less estimated background level (not stated). “Peak” refers, in this case, to the 95th percentile of the hourly measurements, averaged by year and county. In a further analysis, Lipfert et al. (2003) reported the strongest positive association for concurrent exposure to peak O₃ for the subset of subjects with low diastolic blood pressure during the 1982 to 1988 period. Two more recent studies of this cohort focused specifically on traffic density (Lipfert et al., 2006a; 2006b). Lipfert et al. (2006b) concluded that: “Traffic density is seen to be a significant and robust predictor of survival in this cohort, more so than ambient air quality, with the possible exception of O₃,” reporting a significant O₃ effect even with traffic density included in the model: RR = 1.080 per 40 ppb peak O₃ (95% CI: 1.019, 1.146). However, in Lipfert et al. (2006a), which considers only the EPA Speciation Trends Network (STN) sites, O₃ drops to non-significant predictor of total mortality for this cohort. The authors acknowledge that: “Peak O₃ has been important in analyses of this cohort for previous periods, but in the STN data set, this variable has limited range and somewhat lower values and its small coefficient of variation results in a relatively large standard error.” The restriction to subjects near STN sites likely reduced the power of this analysis, though the size of the remaining subjects considered was not reported in this paper. In addition, these various Veterans Cohort studies considered only total mortality, and did not consider mortality on a by-cause basis.

An ecological study in Brisbane, Australia used a geospatial approach to analyze the association of long-term exposure to gaseous air pollution with cardio-respiratory mortality, in the period 1996-2004 (Wang et al., 2009c). A generalized estimating equations model was employed to investigate the impact of NO₂, O₃ and SO₂, but PM was not addressed. The results indicated that long-term exposure to O₃ was not associated with cardio-respiratory mortality, but the fact that this study considered only one city, and that the range of O₃ exposure across that city (23.7-35.6 ppb) was low and slight in variation in comparison to the range of other pollutants across the city, limited study power. In addition, confounding factors (e.g., smoking) could not be addressed at the individual level in this ecological study. Respiratory mortality was not evaluated separately.

A recent study by Zanobetti and Schwartz examined whether year-to-year variations in 8-h mean daily O₃ concentrations for the summer (May-September) around their city-specific long-term trend were associated with year-to-year variations in mortality around its long-term trend. This association was examined among Medicare
participants with potentially predisposing conditions, including COPD, diabetes, CHF, and MI, defined as patients discharged alive after an emergency admission for one of these four conditions. The analyses was repeated in 105 cities using available data from 1985 through 2006, and the results were combined using methods previously employed by these authors (Zanobetti et al., 2008; Zanobetti and Schwartz, 2007). This study design eliminated potential confounding by factors that vary across city, which is a common concern in most air pollution cohort studies, and also avoided both confounding by cross-sectional factors that vary by city and the short-term factors that confound daily time-series studies, but are not present in annual analyses. The average 8-h mean daily summer O_3 concentrations ranged from 15.6 ppb (Honolulu, HI) to 71.4 ppb (Bakersfield, CA) for the 105 cities. The authors observed associations between yearly fluctuations in summer O_3 concentrations and mortality in each of the four cohorts; the hazard ratios (per 10 ppb increment) were 1.12 (95% CI: 1.06, 1.17) for the CHF cohort, 1.19 (95% CI 1.12, 1.25) for the MI cohort, 1.14 (95% CI: 1.10, 1.21) for the diabetes cohort, and 1.14 (95% CI: 1.08, 1.19) for the COPD cohort. A key advantage to this study is that fluctuations from summer to summer in O_3 concentrations around long-term level and trend in a specific city are unlikely to be correlated with most other predictors of mortality risk; except for temperature, which was controlled for in the regression. Key limitations of the study were the inability to control for PM2.5, since it was not reliably measured in these cities until 1999, and the inability to separate specific causes of death (e.g., respiratory, cardiovascular), since Medicare does not provide the underlying cause of death.

In the most recent follow-up analyses of the ACS cohort (Jerrett et al., 2009; Smith et al., 2009a), the effects of long-term exposure to O_3 were evaluated alone, as well as in copollutant models with PM_{2.5} and components of PM_{2.5}. Jerrett et al. (2009) utilized the ACS cohort with data from 1977 through 2000 (mean O_3 concentration ranged from 33.3 to 104.0 ppb) and subdivided cardiopulmonary deaths into respiratory and cardiovascular, separately, as opposed to combined into one category, as was done by Pope et al. (2002). Increases in exposure to O_3 were associated with an elevated risk of death from cardiopulmonary, cardiovascular, ischemic heart disease, and respiratory causes. Consistent with study hypotheses, inclusion of PM_{2.5} concentrations measured in 1999-2000 (the earliest years for which it was available) as a copollutant attenuated the association with O_3 for all end points except death from respiratory causes, for which a significant association persisted (Table 7-12). The association between increased O_3 concentrations and increased risk of death from respiratory causes was insensitive to the use of a random-effects survival model allowing for spatial clustering within the metropolitan area and state of residence, and adjustment for several ecologic variables considered individually. Subgroup analyses showed that temperature and region of country, but not sex, age at enrollment, body-mass index, education, or PM_{2.5} concentration, modified the effects of O_3 on the risk of death from respiratory causes (i.e., risks were higher at higher temperature, and in the Southeast, Southwest, and Upper Midwest). Ozone threshold analyses indicated that the threshold model was not a better fit to the data (p >0.05) than a linear representation of the overall O_3-mortality association. Overall, this new analysis indicates that long-term exposure to PM_{2.5} increases risk of cardiac death,
while long-term exposure to O$_3$ is specifically associated with an increased risk of respiratory death, and suggests that combining cardiovascular and respiratory causes of mortality into one category for analysis may obscure any effect that O$_3$ may have on respiratory-related causes of mortality.

### Table 7-12  Relative risk (and 95% CI) of death attributable to a 10-ppb change in the ambient O$_3$ concentration.

<table>
<thead>
<tr>
<th>Cause of Death</th>
<th>O$_3$ (96 MSAs)$^a$</th>
<th>O$_3$ (86 MSAs)$^a$</th>
<th>O$<em>3$ +PM$</em>{2.5}$ (86 MSAs)$^a$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Any Cause</td>
<td>1.001 (0.996, 1.007)</td>
<td>1.001 (0.996, 1.007)</td>
<td>0.989 (0.981, 0.996)</td>
</tr>
<tr>
<td>Cardiopulmonary</td>
<td>1.014 (1.007, 1.022)</td>
<td>1.016 (1.008, 1.024)</td>
<td>0.992 (0.982, 1.003)</td>
</tr>
<tr>
<td>Respiratory</td>
<td>1.029 (1.010, 1.048)</td>
<td>1.027 (1.007, 1.046)</td>
<td>1.040 (1.013, 1.067)</td>
</tr>
<tr>
<td>Cardiovascular</td>
<td>1.011 (1.003, 1.023)</td>
<td>1.014 (1.005, 1.023)</td>
<td>0.983 (0.971, 0.994)</td>
</tr>
<tr>
<td>Ischemic Heart Disease</td>
<td>1.015 (1.003, 1.026)</td>
<td>1.017 (1.006, 1.029)</td>
<td>0.973 (0.958, 0.988)</td>
</tr>
</tbody>
</table>

$^a$Ozone concentrations were measured from April to September during the years from 1977 to 2000, with follow-up from 1982 to 2000; changes in the concentration of PM$_{2.5}$ of 10 µg/m$^3$ were recorded for members of the cohort in 1999 and 2000.

Source: Reprinted with permission of Massachusetts Medical Society (Jerrett et al., 2009).

In a similar analysis, Smith et al. (2009a) used data from 66 Metropolitan Statistical Areas (MSAs) in the ACS cohort to examine the association of O$_3$ concentrations during the warm season and all-cause and cardiopulmonary mortality. Mortality effects were estimated in single pollutant and copollutant models, adjusting for two PM$_{2.5}$ constituents, sulfate, and EC. When all-cause mortality was investigated, there was a 0.8% (95% CI: -0.31, 1.9) increase associated with a 10 ppb increase in O$_3$ concentration. This association was diminished when sulfate or EC were included in the model. There was a 2.48% (95% CI: 0.74, 4.3) increase in cardiopulmonary mortality associated with a 10 ppb increase in O$_3$ concentration. The cardiopulmonary association was robust to adjustment for sulfate, and diminished, though still positive, after adjustment for EC (1.63% increase; 95% CI: -0.41, 3.7). Smith et al. (2009a) did not specifically separate out cardiovascular and respiratory causes of death from the cardiopulmonary category, as was done by Jerrett et al. (2009).

#### 7.7.1 Summary and Causal Determination

The 2006 O$_3$ AQCD concluded that an insufficient amount of evidence existed “to suggest a causal relationship between chronic O$_3$ exposure and increased risk for mortality in humans” (U.S. EPA, 2006b). Several additional studies have been conducted since the last review that evaluate cause-specific and total mortality. An ecologic study conducted in Australia observed no association between cardiopulmonary mortality and O$_3$ (Wang et al., 2009c). Two reanalyses of the ACS cohort were conducted; one provides weak evidence for an association with cardiopulmonary mortality (Smith et al., 2009a) while the other specifically points to
a relationship between long-term O₃ exposure and an increased risk of respiratory mortality (Jerrett et al., 2009). Most recently, a study of four cohorts of Medicare enrollees with potentially predisposing conditions observed associations between O₃ and total mortality among each of the cohorts (Zanobetti and Schwartz, 2011).

When considering the entire body of evidence, there is limited support for an association with long-term exposure to ambient O₃ and total mortality. There is inconsistent evidence for an association between long-term exposure to ambient O₃ and cardiopulmonary mortality, with several analyses from the ACS cohort reporting some positive associations (Smith et al., 2009a; Pope et al., 2002) while other studies reported no association (Wang et al., 2009c; Abbey et al., 1999; Dockery et al., 1993). The strongest evidence for an association between long-term exposure to ambient O₃ concentrations and mortality is derived from associations reported in the Jerrett et al. (2009) study for respiratory mortality that remained robust after adjusting for PM₂.₅ concentrations. Finally, a recent analysis reported associations of ambient O₃ concentrations and total mortality in potentially at-risk populations in the Medicare Cohort (Zanobetti and Schwartz, 2011), while earlier studies generally report no associations with total mortality (Lipfert et al., 2006a; Lipfert et al., 2003; Pope et al., 2002; Abbey et al., 1999; Dockery et al., 1993). Studies of cardiopulmonary and total mortality provide limited evidence for an association with long-term exposure to ambient O₃ concentrations. The study by Jerrett et al. (2009) observes an association between long-term exposure to ambient O₃ concentrations and respiratory mortality that remained robust after adjusting for PM₂.₅ concentrations. Coherence and biological plausibility for this observation is provided by evidence from epidemiologic, controlled human exposure, and animal toxicological studies for the effects of short- and long-term exposure to O₃ on respiratory effects (see Sections 6.2 and 7.2). Respiratory mortality is a relatively small portion of total mortality [about 7.6% of all deaths in 2010 were due to respiratory causes (Murphy et al., 2012)], thus it is not surprising that the respiratory mortality signal may be difficult to detect in studies of cardiopulmonary or total mortality. Based on the recent evidence for respiratory mortality along with limited evidence for total and cardiopulmonary mortality, the evidence is suggestive of a causal relationship between long-term O₃ exposures and total mortality.

### 7.8 Overall Summary

The evidence reviewed in this chapter describes the recent findings regarding the health effects of long-term exposure to ambient O₃ concentrations. Table 7-13 provides an overview of the causal determinations for each of the health categories evaluated.
Table 7-13  Summary of causal determinations for long-term exposures to O₃.

<table>
<thead>
<tr>
<th>Health Category</th>
<th>Causal Determination</th>
</tr>
</thead>
<tbody>
<tr>
<td>Respiratory Effects</td>
<td>Likely to be a causal relationship</td>
</tr>
<tr>
<td>Cardiovascular Effects</td>
<td>Suggestive of a causal relationship</td>
</tr>
<tr>
<td>Reproductive and Developmental Effects</td>
<td>Suggestive of a causal relationship</td>
</tr>
<tr>
<td>Central Nervous System Effects</td>
<td>Suggestive of a causal relationship</td>
</tr>
<tr>
<td>Carcinogenicity and Genotoxicity</td>
<td>Inadequate to infer a causal relationship</td>
</tr>
<tr>
<td>Total Mortality</td>
<td>Suggestive of a causal relationship</td>
</tr>
</tbody>
</table>
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8 POPULATIONS POTENTIALLY AT INCREASED RISK FOR OZONE-RELATED HEALTH EFFECTS

Interindividual variation in human responses to air pollution exposure can result in some groups being at increased risk for detrimental effects in response to ambient exposure to an air pollutant. The NAAQS are intended to provide an adequate margin of safety for both the population as a whole and those potentially at increased risk for health effects in response to ambient air pollution\(^1\). To facilitate the identification of populations and lifestages at greater risk for air pollutant related health effects, studies have evaluated factors that may contribute to the susceptibility and/or vulnerability of an individual to air pollutants. The definitions of susceptibility and vulnerability have been found to vary across studies, but in most instances “susceptibility” refers to biological or intrinsic factors (e.g., lifestage, sex, pre-existing disease/conditions) while “vulnerability” refers to non-biological or extrinsic factors (e.g., socioeconomic status [SES]) (\textit{U.S. EPA, 2010c, 2009d}). In some cases, the terms “at-risk” and “sensitive” populations have been used to encompass these concepts more generally. The main goal of this evaluation is to identify and understand those factors that result in a population or lifestage being at increased risk of an air pollutant-related health effect, not to categorize the factors. To this end, previous ISAs and reviews (\textit{Sacks et al., 2011; U.S. EPA, 2010c, 2009d}) have used “susceptible populations” to encompass these various factors. In this chapter, “at-risk” is the all-encompassing term used for groups with specific factors that increase the risk of an air pollutant (e.g., O\(_3\))-related health effect in a population.

Individuals, and ultimately populations, could experience increased risk for air pollutant induced health effects in multiple different ways. A group with intrinsically increased risk would have some factor(s) that increases risk for an effect through a biological mechanism. In general, people in this category would have a steeper concentration-risk relationship, compared to those not in the category. Potential factors that are often considered intrinsic include genetic background and sex. A group of people could also have extrinsically increased risk, which would be through an external, non-biological factor. Examples of extrinsic factors include SES and diet.

In addition, some groups are at increased risk due to differential exposure, which can encompass multiple forms. This includes increased risk due to increased internal dose at a given exposure concentration. For example, individuals may have a greater dose of delivered pollutant because of their breathing pattern. This group would include persons who work outdoors or exercise outdoors. Some outdoor workers could also have greater exposure (concentration x time), regardless of the delivered dose and this greater exposure may increase the risk of O\(_3\)-related health effects.

\(1\) The legislative history of section 109 indicates that a primary standard is to be set at “the maximum permissible ambient air level ... which will protect the health of any [sensitive] group of the population,” and that for this purpose “… reference should be made to a representative sample of persons comprising the sensitive group rather than to a single person in such a group.” [S. Rep. No. 91- 1196, 91st Cong., 2d Sess. 10 (1970)].
Finally, there are those who might be placed at increased risk for experiencing a greater exposure, and therefore increased risk of health effects, by being exposed at a higher concentration. For example, groups of people exposed to higher air pollutant concentrations due to less availability/use of home air conditioners (i.e., more open windows on high O₃ days) or close proximity to known sources of air pollution.

Some factors described above are multifaceted. For example, SES may affect access to medical care, which itself may contribute to the presence of pre-existing diseases and conditions considered as intrinsic factors. Additionally, children tend to spend more time outdoors at higher levels of activity than adults, which leads to increased intake dose and exposure, but they also have biological (i.e., intrinsic) differences when compared to adults.

The emphasis of this chapter is to identify and understand the factors that potentially increase the risk of O₃-related health effects, regardless of whether the increased risk is due to intrinsic factors, extrinsic factors, increased dose, increased exposure, or a combination. The following sections examine factors that potentially lead to increased risk of O₃-related health effects and characterize the overall weight of evidence for each factor. Most of the factors are related to greater health effects given a specific dose but there is also discussion of increased internal dose and/or exposure at a given concentration integrated throughout the sections (i.e., lifestage, outdoor workers, and air conditioning use).

**Approach to Classifying Potential At-Risk Factors**

To identify factors that potentially lead to some populations being at greater risk to air pollutant related health effects, the evidence across relevant scientific disciplines (i.e., exposure sciences, dosimetry, controlled human exposure, toxicology, and epidemiology) was evaluated. In this systematic approach, the collective evidence is used to examine coherence of effects across disciplines and determine biological plausibility. By first focusing on studies that conduct stratified analyses (i.e., epidemiologic or controlled human exposure) it is possible to identify factors that may result in some populations being at greater risk of an air pollutant related health effect. These types of studies allow for an evaluation of populations exposed to similar air pollutant (e.g., O₃) concentrations within the same study design. Experimental studies also provide important lines of evidence in the evaluation of factors that may lead to increased risk of an air pollutant related health effect. Toxicological studies conducted using animal models of disease and controlled human exposure studies that examine individuals with underlying disease or genetic polymorphisms may provide evidence to inform whether a population is at increased risk of an air pollutant related health effect in the absence of stratified epidemiologic analyses. Additionally these studies can provide support for coherence with the health effects observed in epidemiologic studies as well as an understanding of biological plausibility. Information on factors that may result in increased risk of O₃-related health effects can also be obtained from studies that examine exposure differences between populations. The collective results across the scientific
disciplines comprise the overall weight of evidence that is used to determine whether a specific factor results in a population being at increased risk of an air pollutant related health effect.

Building on the causal framework discussed in detail in the Preamble and used throughout the ISA, conclusions are made regarding the strength of evidence, based on the evaluation and synthesis across scientific disciplines, for each factor that may contribute to increased risk of an O\textsubscript{3}-related health effect. The conclusions were drawn while considering the “Aspects to Aid in Judging Causality” discussed in Table 1 of the Preamble. The categories considered for evaluating the potential increased risk of an air pollutant-related health effect are “adequate evidence,” “suggestive evidence,” “inadequate evidence,” and “evidence of no effect.” They are described in more detail in Table 8-1.

<table>
<thead>
<tr>
<th>Health Effects</th>
</tr>
</thead>
<tbody>
<tr>
<td>Adequate evidence</td>
</tr>
<tr>
<td>There is substantial, consistent evidence within a discipline to conclude that a factor results in a population or lifestage being at increased risk of air pollutant-related health effect(s) relative to some reference population or lifestage. Where applicable this includes coherence across disciplines. Evidence includes multiple high-quality studies.</td>
</tr>
<tr>
<td>Suggestive evidence</td>
</tr>
<tr>
<td>The collective evidence suggests that a factor results in a population or lifestage being at increased risk of an air pollutant-related health effect relative to some reference population or lifestage, but the evidence is limited due to some inconsistency within a discipline or, where applicable, a lack of coherence across disciplines.</td>
</tr>
<tr>
<td>Inadequate evidence</td>
</tr>
<tr>
<td>The collective evidence is inadequate to determine if a factor results in a population or lifestage being at increased risk of an air pollutant-related health effect relative to some reference population or lifestage. The available studies are of insufficient quantity, quality, consistency and/or statistical power to permit a conclusion to be drawn.</td>
</tr>
<tr>
<td>Evidence of no effect</td>
</tr>
<tr>
<td>There is substantial, consistent evidence within a discipline to conclude that a factor does not result in a population or lifestage being at increased risk of air pollutant-related health effect(s) relative to some reference population or lifestage. Where applicable this includes coherence across disciplines. Evidence includes multiple high-quality studies.</td>
</tr>
</tbody>
</table>

This chapter evaluates the various factors indicated in the literature that may result in a population being at increased risk of an O\textsubscript{3}-related health effect. For further detail on the epidemiologic, controlled human exposure, and toxicological studies included in this chapter, see Chapters 5, 6, and 7.

### 8.1 Genetic Factors

The potential effects of air pollution on individuals with specific genetic characteristics have been examined; studies often target polymorphisms in already identified candidate susceptibility genes or in genes whose protein products are thought to be involved in the biological mechanism underlying the health effect of an air pollutant (Sacks et al., 2011). As a result, multiple studies that examined the effect of short- and long-term O\textsubscript{3} exposure on respiratory function have focused on whether various gene profiles lead to an increased risk of O\textsubscript{3}-related health effects.
For more details on the function and mode of action of the genetic factors discussed in this section, see Section 5.4.2.1. Additionally, a limited number of toxicological studies have examined the joint effects of nutrition and genetics. Details on these toxicological studies of nutrition and genetics can be found in Section 5.4.2.3.

Multiple genes, including glutathione S-transferase Mu 1 (GSTM1) and tumor necrosis factor-α (TNF-α) were evaluated in the 2006 O3 AQCD and found to have a “potential role… in the innate susceptibility to O3” (U.S. EPA, 2006b). Epidemiologic, controlled human exposure, and toxicological studies performed since the 2006 O3 AQCD have continued to examine the roles of GSTM1 and TNF-α in modifying O3-related health effects and have examined other gene variants that may also increase risk. Due to small sample sizes, many controlled human exposure studies are limited in their ability to test genes with low frequency minor alleles and therefore, some genes important for O3-related health effects may not have been examined in these types of studies. A summary of effect measure modification findings from epidemiologic and controlled human exposure studies discussed in this section is included in Table 8-2 and from animal toxicology studies in Table 8-3.

Epidemiologic studies that examined the effects of short-term exposure to O3 on lung function included analyses of potential gene-environment interactions. Romieu et al. (2006) reported an association between O3 and respiratory symptoms that were larger among children with GSTM1 null or glutathione S-transferase P 1 (GSTP1) Val/Val genotypes compared with children with GSTM1 positive or GSTP1 Ile/Ile or Ile/Val genotypes, respectively. However, results suggested that O3-associated decreases in lung function may be greater among children with GSTP1 Ile/Ile or Ile/Val compared to GSTP1 Val/Val. Alexeeff et al. (2008) reported greater O3-related decreases in lung function among GSTP1 Val/Val adults than those with GSTP1 Ile/Ile or GSTP1 Ile/Val genotypes. In addition, they detected greater O3-associated decreases in lung function for adults with long GT dinucleotide repeats in heme-oxygenase-1 (HMOX1) promoters.
Table 8-2  Summaries of results from epidemiologic and controlled human exposures studies of modification by genetic variants.

<table>
<thead>
<tr>
<th>Gene variant</th>
<th>Comparison group</th>
<th>Health outcome /population</th>
<th>Effect modification of association for the gene variant</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>GSTM1 null</td>
<td>GSTM1 positive</td>
<td>Respiratory symptoms among asthmatic children</td>
<td>↑</td>
<td>Romieu et al. (2006)</td>
</tr>
<tr>
<td>GSTP1 Val/Val</td>
<td>GSTP1 Ile/Ile or Ile/Val</td>
<td>Respiratory symptoms among asthmatic children</td>
<td>↑</td>
<td></td>
</tr>
<tr>
<td>GSTP1 Ile/Ile or Ile/Val</td>
<td>GSTP1 Val/Val</td>
<td>Lung function among asthmatic children</td>
<td>↓</td>
<td></td>
</tr>
<tr>
<td>GSTP1 Ile/Ile or Val/Val</td>
<td>GSTP1 Ile/Ile</td>
<td>Lung function among adults</td>
<td>↓</td>
<td>Alexeiff et al. (2008)</td>
</tr>
<tr>
<td>HMOX1 S/L or L/L</td>
<td>HMOX1 S/S</td>
<td>Lung function among adults</td>
<td>↓</td>
<td></td>
</tr>
<tr>
<td>NQO1 wildtype and GSTM1 null</td>
<td>Other combinations</td>
<td>Lung function among healthy adults with exercise</td>
<td>↓</td>
<td>Bergamaschi et al. (2001)</td>
</tr>
<tr>
<td>NQO1 wildtype and GSTM1 null</td>
<td>Other combinations</td>
<td>Lung function among mild-to-moderate asthmatics with moderate exercise</td>
<td>=</td>
<td>Vagaggini et al. (2010)</td>
</tr>
<tr>
<td>NQO1 wildtype and GSTM1 null</td>
<td>Other combinations</td>
<td>Inflammatory responses among mild-to-moderate asthmatics with moderate exercise</td>
<td>=</td>
<td></td>
</tr>
<tr>
<td>GSTM1 null</td>
<td>GSTM1 positive</td>
<td>Lung function among healthy adults with intermittent moderate exercise</td>
<td>=</td>
<td>Kim et al. (2011)</td>
</tr>
<tr>
<td>GSTM1 null</td>
<td>GSTM1 positive</td>
<td>Inflammatory responses among healthy adults with intermittent moderate exercise</td>
<td>=</td>
<td></td>
</tr>
<tr>
<td>GSTM1 null</td>
<td>GSTM1 positive</td>
<td>Lung function among asthmatic children</td>
<td>↓</td>
<td>Romieu et al. (2004b)</td>
</tr>
<tr>
<td>GSTM1 null</td>
<td>GSTM1 positive</td>
<td>Lung function among healthy adults with intermittent moderate exercise</td>
<td>=</td>
<td>Alexis et al. (2009)</td>
</tr>
<tr>
<td>GSTM1 null</td>
<td>GSTM1 positive</td>
<td>Inflammatory changes among healthy adults with intermittent moderate exercise</td>
<td>↑</td>
<td></td>
</tr>
</tbody>
</table>
Several controlled human exposure studies have reported that genetic polymorphisms of antioxidant enzymes may modulate pulmonary function and inflammatory responses to O₃ challenge. Healthy carriers of NAD(P)H quinone oxidoreductase 1 (NQO1) wild type (wt) in combination with GSTM1 null genotype had greater decreases in lung function parameters with exposure to O₃ (Bergamaschi et al., 2001). Vagaggini et al. (2010) exposed mild-to-moderate asthmatics to O₃ during moderate exercise. In subjects with NQO1 wt and GSTM1 null, there was no evidence of changes in lung function or inflammatory responses to O₃. Kim et al. (2011) also recently conducted a study among young adults, about half of whom were GSTM1-null and half of whom were GSTM1-sufficient. They detected no difference in the FEV₁ responses to O₃ exposure by GSTM1 genotype and did not examine NQO1. In another study that examined GSTM1 but not NQO1, asthmatic children with GSTM1 null genotype (Romieu et al., 2004b) were reported to have greater decreases in lung function in relation to O₃ exposure. Additionally, supplementation with antioxidants (Vitamins C and E) had a slightly more beneficial effect among GSTM1 null children (for more on modification by diet, see Section 8.4.1).

In a study of healthy volunteers with GSTM1 sufficient and GSTM1 null genotypes exposed to O₃ with exercise, Alexis et al. (2009) found genotype effects on inflammatory responses but not lung function responses to O₃. At 4 hours post-O₃ exposure, individuals with either GSTM1 genotype had statistically significant increases in sputum neutrophils with a tendency for a greater increase in GSTM1 sufficient than GSTM1 nulls. At 24 hours postexposure, neutrophils had returned to baseline levels in the GSTM1 sufficient individuals. In the GSTM1 null subjects, neutrophil levels increased from 4 to 24 hours and were significantly greater than both baseline levels and levels at 24 hours in the GSTM1 sufficient individuals. In addition, O₃ exposure increased the expression of the surface marker CD14 in airway neutrophils of GSTM1 null subjects compared with GSTM1 sufficient subjects. CD14 and TLR4 are co-receptors for endotoxin, and signaling through this innate immune pathway has been shown to be important for a number of biological responses to O₃ exposure in toxicological studies (Garantziotis et al., 2010; Hollingsworth et al., 2010; Hollingsworth et al., 2004; Kleeberger et al., 2000).

Alexis et al. (2009) also demonstrated decreased numbers of airway macrophages at 4 and 24 hours following O₃ exposure in GSTM1 sufficient subjects. Airway macrophages in GSTM1 null subjects were greater in number and found to have greater oxidative burst and phagocytic capability following O₃ exposure than those of GSTM1 sufficient subjects. Airway macrophages and dendritic cells from GSTM1 null subjects exposed to O₃ expressed higher levels of the surface marker HLA-DR; again suggesting activation of the innate immune system. Since there was no FA control in the Alexis et al. (2009) study, effects of the exposure other than O₃ cannot be ruled out. In general, the findings between these studies are inconsistent. It is possible that different genes may be important for different phenotypes. Additional studies, which include appropriate controls, are needed to clarify the influence of genetic polymorphisms on O₃ responsiveness in humans.
Table 8-3  Summaries of results from animal toxicology studies of modification by genetic variants.

<table>
<thead>
<tr>
<th>Gene variant</th>
<th>Reference*</th>
<th>Exposure</th>
<th>Health outcome /population</th>
</tr>
</thead>
</table>
| Tlr4         | Hollingsworth et al. (2004);  
              | Kleeberger et al. (2000)       | 0.3 ppm, 72 hours  
              | 0.3 ppm, 72 hours  
              | 2.0 ppm, 3 hours | Decreased hyperpermeability  
              | No genotype difference in hyperpermeability. BALF cells, or AHR at 0.3ppm. Reduced AHR at 2.0ppm. |
|              | Williams et al. (2007b)        | 0.3 ppm, 3-24 hours | Decreased AHR. Reduced inflammation at 3 hours |
| Tlr2         | Williams et al. (2007b)        | 0.3 ppm, 3-24 hours | Decreased inflammation and AHR |
| MyD88        | Williams et al. (2007b)        | 0.3 ppm, 3-24 hours | Decreased inflammation, hyperpermeability, and AHR |
| Tnfr1/Tnfr2  | Cho et al. (2001)              | 0.3 ppm, 3-48 hours | Decreased BALF cells, neutrophilia and lung damage.  
              | Cho et al. (2007)              | 2.0 ppm, 3 hours | Reduced AHR |
| Nfkb         | Cho et al. (2007)              | 0.3 ppm, 6-48 hours | Decreased inflammation, hyperpermeability, and lung damage |
| Nfkb         | Cho et al. (2007)              | 0.3 ppm, 6-48 hours | Decreased inflammation, hyperpermeability, and lung damage |
| Jnk          | Cho et al. (2007)              | 0.3 ppm, 6-48 hours | Decreased inflammation, hyperpermeability, and lung damage |
| Il6          | Johnston et al. (2005b)        | 0.3 ppm, 3-72 hours | Decreased neutrophilia and hyperpermeability, reduced soluble TNFRs, no effect on AHR |
| Il10         | Backus et al. (2010)           | 0.3 ppm, 24-72 hours | Increased inflammation |
| Marco        | Dahl et al. (2007)             | 0.3 ppm, 48 hours | Increased inflammation, 8-isoprostane, and hyperpermeability |
| Nos2         | Kleeberger et al. (2001)       | 0.3 ppm, 72 hours | Decreased hyperpermeability, no effect on BALF cells |
|              | Fakhrzadeh et al. (2002)       | 0.8 ppm, 3 hours | Reduced AM nitric oxide, reactive nitrogen species, and superoxide anion, decreased PGE2, increased COX expression, decreased hyperpermeability and BALF cells |
|              | Kenyon et al. (2002)           | 1.0 ppm,  
              | 8 h/night for 3 nights | Increased hyperpermeability, neutrophilia, MMP-9 activity, and protein nitration products |
| Hsp70        | Bauer et al. (2011)            | 0.3 ppm, 6-72 hours | Decreased hyperpermeability and inflammation |
| NQO1         | Vojnov et al. (2009)           | 1.0 ppm, 3 hours | Reduced inflammation and AHR |
| Csb          | Kooter et al. (2007)           | 0.8 ppm, 8 hours | Decreased TNF-α in BALF. No genotype difference in neutrophilia or lung damage. |
| Mmp9         | Yoon et al. (2007)             | 0.3 ppm 6-72 hours | Increased hyperpermeability, neutrophilia, inflammation, and lung damage |
| CD44         | Garantziolis et al. (2009)     | 2.0 ppm, 3 hours | Decreased AHR |
| Cxcr2        | Johnston et al. (2005)         | 1.0 ppm, 3 hours | Reduced neutrophilia, lung injury, and AHR. No change in chemokine expression or hyperpermeability |
| II13         | Williams et al. (2008b)        | 3.0 ppm, 3 hours | Reduced AHR, BALF cells, and neutrophilia |

*The table includes animal toxicology studies where responses are assessed after gene deletion.
In general, toxicological studies have reported differences in cardiac and respiratory effects after O₃ exposure among different mouse strains, which alludes to differential risk among individuals due to genetic variability (Tankersley et al., 2010; Chuang et al., 2009; Hamade and Tankersley, 2009; Hamade et al., 2008). Thus strains of mice which are prone to or resistant to O₃-induced effects have been used to systematically identify candidate genes that may increase risk of O₃-related health effects. Genome wide linkage analyses have identified quantitative trait loci for O₃-induced lung inflammation and hyperpermeability on chromosome 17 (Kleeberger et al., 1997) and chromosome 4 (Kleeberger et al., 2000), respectively, using recombinant inbred strains of mice. More specifically, these studies found that TNF (protein product is the inflammatory cytokine TNF-α) and Tlr4 (protein product is TLR4, involved in endotoxin responses) were candidate susceptibility genes (Kleeberger et al., 2000; Kleeberger et al., 1997). The TNF receptors 1 and 2 have also been found to play a role in injury, inflammation, and airway hyperreactivity in studies of O₃-exposed knockout mice (Cho et al., 2007; Cho et al., 2001) through NF-κB and MAPK/AP-1 (Jnk) signaling pathways (Cho et al., 2007). In addition to Tlr4, other innate immune pattern recognition signaling pathway genes, including Tlr2 and Myd88, appear to be important in responses to O₃, as demonstrated by Williams et al. (2007b). A role for the inflammatory cytokine IL-6 has been demonstrated in gene-deficient mice with respect to inflammation and injury, but not AHR (Johnston et al., 2005b; Yu et al., 2002). Other studies have demonstrated a key role for CCR2, the chemokine receptor for the neutrophil chemokines KC and MIP-2, (Johnston et al., 2005a) and CD44, the major receptor for the extracellular matrix component hyaluronan (Garantziotis et al., 2009) in O₃-mediated AHR. Mice deficient in IL-10, an anti-inflammatory cytokine, demonstrated increased pulmonary inflammation in response to O₃ exposure (Backus et al., 2010). Thus genes related to innate immune signaling and pro- and anti-inflammatory genes are important for O₃-induced responses.

Altered O₃ responses between mouse strains could be due to genetic variability in nuclear factor erythroid 2-related factor 2 (Nrf-2), suggesting a role for genetic differences in altering the formation of ROS (Hamade et al., 2010). Additionally, some studies have reported O₃-related effects to vary by Inf-1 and Inf-2 quantitative trait loci (Tankersley and Kleeberger, 1994) and a gene coding for Clara cell secretory protein (CCSP) (Broeckaert et al., 2003; Wattieze et al., 2003). Other investigations in inbred mouse strains found that differences in expression of certain proteins, such as CCSP (Broeckaert et al., 2003) and MARCO (Dahl et al., 2007), are responsible for phenotypic characteristics, such as epithelial permeability and scavenging of oxidized lipids, respectively, which confer sensitivity to O₃.

Nitric oxide (NO), derived from activated macrophages, is produced upon exposure to O₃ and is thought to participate in lung damage. Mice deficient in the gene for inducible nitric oxide synthase (NOS2/NOSII/iNOS) are partially protected against lung injury (Kleeberger et al., 2001), and it appears that O₃-induced iNOS expression is tied to the TLR4 pathway described above. Similarly, iNOS deficient mice do not produce reactive nitrogen intermediates after O₃ exposure, in contrast to their wild-type counterparts, and also produce less PGE2 comparatively (Fakhrzadeh et
These gene-deficient mice were protected from O$_3$-induced lung injury and inflammation. In contrast, another study using a similar exposure concentration but longer duration of exposure found that iNOS deficient mice were more at risk of O$_3$-induced lung damage (Kenyon et al., 2002). Therefore, the role of iNOS in mediating the response to O$_3$ exposure is likely dependent on the exposure concentration and duration.

Voynow et al. (2009) have shown that NQO1 deficient mice, like their human counterparts, are resistant to O$_3$-induced AHR and inflammation. NQO1 catalyzes the reduction of quinones to hydroquinones, and is capable of both protective detoxification reactions and redox cycling reactions resulting in the generation of reactive oxygen species. Reduced production of inflammatory mediators and cells and blunted AHR were observed in NQO1 null mice after exposure to O$_3$. These results correlated with those from in vitro experiments in which human bronchial epithelial cells treated with an NQO1 inhibitor exhibited reduced inflammatory responses to exposure to O$_3$. This study may provide biological plausibility for the increased biomarkers of oxidative stress and increased pulmonary function decrements observed in O$_3$-exposed individuals bearing both the wild-type NQO1 gene and the null GSTM1 gene (Bergamaschi et al., 2001). Deletion of the gene for MMP9 also conferred protection against O$_3$-induced airways inflammation and injury (Yoon et al., 2007).

The role of TNF-$\alpha$ signaling in O$_3$-induced responses has been previously established through depletion experiments, but a more recent toxicological study investigated the effects of combined O$_3$ and PM exposure in transgenic TNF overexpressing mice. Kumarathasan et al. (2005) found that subtle effects of these pollutants were difficult to identify in the midst of the severe pathological changes caused by constitutive TNF-$\alpha$ overexpression. However, there was evidence that TNF transgenic mice were at increased risk of O$_3$/PM-induced oxidative stress, and they exhibited elevation of a serum creatine kinase after pollutant exposure, which may suggest potential systemic or cardiac related effects. Differential risk of O$_3$ among inbred strains of animals does not seem to be dose dependent since absorption of $^{18}$O in various strains of mice did not correlate with resistance or sensitivity (Vancza et al., 2009).

Defects in DNA repair mechanisms may also confer increased risk of O$_3$-related health effects. Cockayne syndrome, a rare autosomal recessive disorder in humans, is characterized by UV sensitivity abnormalities, neurological abnormalities, and premature aging. The same genetic defect in mice (Csb$^{-/-}$) makes them sensitive to oxidative stressors, including O$_3$. Kooter et al. (2007) demonstrated that Csb$^{-/-}$ mice produced significantly more TNF-$\alpha$ after exposure to O$_3$ than their wild-type counterparts. However, there were no statistically significant differences in other markers of inflammation or lung injury between the two strains of mice.

Overall, for variants in multiple genes there is adequate evidence for involvement in populations being more at-risk than others to the effects of O$_3$ exposure on health. Controlled human exposure and epidemiologic studies have reported evidence of O$_3$-related increases in respiratory symptoms or decreases in lung function with variants
including GSTM1, GSTP1, HMOX1, and NQO1. NQO1 deficient mice were found to be resistant to O₃-induced AHR and inflammation, providing biological plausibility for results of studies in humans. Additionally, studies of rodents have identified a number of other genes that may affect O₃-related health outcomes, including genes related to innate immune signaling and pro- and anti-inflammatory genes, which have not been investigated in human studies.

### 8.2 Pre-existing Disease/Conditions

Individuals with certain pre-existing diseases are likely to constitute an at-risk population. This may be the result of individuals with a pre-existing disease/condition having less reserve than healthy individuals, so although the absolute change may be the same, the health consequences are different. Previous O₃ AQCDs concluded that some people with pre-existing pulmonary disease, especially asthma, are among those at increased risk of an O₃-related health effect. Extensive toxicological evidence indicates that altered physiological, morphological and biochemical states typical of respiratory diseases may render people at risk of an additional oxidative burden induced by O₃ exposure. In addition, a number of epidemiologic studies found that some individuals with respiratory diseases are at increased risk of O₃-related effects. The majority of the studies identified in previous AQCDs focused on whether pre-existing respiratory diseases result in increased risk of O₃-related health effects, with a limited number of studies examining other pre-existing diseases, such as cardiovascular.

Studies identified since the completion of the 2006 O₃ AQCD that examined whether pre-existing diseases and conditions lead to increased risk of O₃-induced health effects were identified and are summarized below. Table 8-4 displays the prevalence rates of some of these conditions categorized by age and region among adults in the U.S. population; data for children, when available, are presented within the following sections. Substantial proportions of the U.S. population are affected by these conditions and therefore may represent potentially large at-risk populations. While these diseases and conditions represent biological or intrinsic factors that could lead to increased risk, the pathways to their development may have intrinsic or extrinsic origins.
Table 8-4  Prevalence of respiratory diseases, cardiovascular diseases, and diabetes among adults by age and region in the U.S.

<table>
<thead>
<tr>
<th>Chronic Disease/Condition</th>
<th>18-44</th>
<th>45-64</th>
<th>65-74</th>
<th>75+</th>
<th>North east</th>
<th>Mid west</th>
<th>South</th>
<th>West</th>
</tr>
</thead>
<tbody>
<tr>
<td>Respiratory Diseases</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Asthma*</td>
<td>16,380</td>
<td>7.2</td>
<td>7.5</td>
<td>7.8</td>
<td>6.4</td>
<td>7.7</td>
<td>8.0</td>
<td>5.9</td>
</tr>
<tr>
<td>COPD</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Chronic Bronchitis</td>
<td>9,832</td>
<td>3.2</td>
<td>5.5</td>
<td>5.9</td>
<td>5.3</td>
<td>3.4</td>
<td>4.8</td>
<td>5.2</td>
</tr>
<tr>
<td>Emphysema</td>
<td>3,789</td>
<td>0.2</td>
<td>2.0</td>
<td>5.7</td>
<td>5.0</td>
<td>1.2</td>
<td>1.9</td>
<td>1.9</td>
</tr>
<tr>
<td>Cardiovascular Diseases</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>All Heart Disease</td>
<td>26,628</td>
<td>4.6</td>
<td>12.3</td>
<td>26.7</td>
<td>39.2</td>
<td>11.3</td>
<td>12.7</td>
<td>12.2</td>
</tr>
<tr>
<td>Coronary Heart Disease</td>
<td>14,428</td>
<td>1.1</td>
<td>6.7</td>
<td>16.9</td>
<td>26.7</td>
<td>5.7</td>
<td>6.5</td>
<td>7.3</td>
</tr>
<tr>
<td>Hypertension</td>
<td>56,159</td>
<td>8.7</td>
<td>32.5</td>
<td>54.4</td>
<td>61.1</td>
<td>22.9</td>
<td>24.1</td>
<td>27.1</td>
</tr>
<tr>
<td>Diabetes</td>
<td>18,651</td>
<td>2.3</td>
<td>12.1</td>
<td>20.4</td>
<td>17.3</td>
<td>4.5</td>
<td>7.6</td>
<td>9.0</td>
</tr>
</tbody>
</table>

*aAsthma prevalence is reported for “still has asthma.”

Source: Pleis et al. (2009); National Center for Health Statistics.

8.2.1 Influenza/Infections

Recent studies have indicated that underlying infections may increase the risk of O₃-related health effects because O₃ exposure likely impairs host defenses, which may increase the body’s response to an infectious agent. However, there is little epidemiologic or experimental evidence that infection or influenza itself renders an individual at greater risk of an O₃-induced health effect. A study of hospitalizations in Hong Kong reported that increased levels of influenza intensity resulted in increased excess risk of respiratory disease hospitalizations related to O₃ exposure (Wong et al., 2009). In addition, a study of lung function in asthmatic children reported decreases in lung function with increased short-term O₃ exposure for those with upper respiratory infections but not for those without infections (Lewis et al., 2005). Toxicological studies provide biological plausibility for the increase in O₃-induced health effects observed in epidemiologic studies that examined infections by way of studies that demonstrated that exposure to 0.08 ppm O₃ increased streptococcus-induced mortality, regardless of whether O₃ exposure preceded or followed infection (Miller et al., 1978; Coffin and Gardner, 1972; Coffin et al., 1967). Overall, the epidemiologic and experimental evidence supports the potential for increased risk to be conferred by an infection but the number of studies is limited. There have only been a few epidemiologic studies and these studies examine
different outcomes (respiratory-related hospital admissions or lung function) and different modifiers (influenza or respiratory infection). In some of the toxicological studies, the O₃ exposure came before the infection. Therefore, evidence is inadequate to determine if influenza/infections increase the risk of O₃-related health effects.

8.2.2 Asthma

Previous O₃ AQCDs identified individuals with asthma as a population at increased risk of O₃-related health effects. Within the U.S., approximately 7.3% of adults have reported currently having asthma (Pleis et al., 2009), and 9.5% of children have reported currently having asthma (Bloom et al., 2008). For more detailed prevalence by age, see Table 8-5.

<table>
<thead>
<tr>
<th>Age (years)</th>
<th>N (in thousands)</th>
<th>Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>0-4</td>
<td>1,276</td>
<td>6.2</td>
</tr>
<tr>
<td>5-11</td>
<td>3,159</td>
<td>11.2</td>
</tr>
<tr>
<td>12-17</td>
<td>2,518</td>
<td>10.2</td>
</tr>
<tr>
<td>18-44</td>
<td>7,949</td>
<td>7.2</td>
</tr>
<tr>
<td>45-64</td>
<td>5,768</td>
<td>7.5</td>
</tr>
<tr>
<td>65-74</td>
<td>1,548</td>
<td>7.8</td>
</tr>
<tr>
<td>75+</td>
<td>1,116</td>
<td>6.4</td>
</tr>
</tbody>
</table>

*Note: Asthma prevalence is reported for “still has asthma.”

Source: Statistics for adults: Pleis et al. (2009); statistics for children: Bloom et al. (2008); National Center for Health Statistics.

Multiple epidemiologic studies included within this ISA have evaluated the potential for increased risk of O₃-related health effects among individuals with asthma. A study of lifeguards in Texas reported decreased lung function with short-term O₃ exposure among both individuals with and without asthma, however, the decrease was greater among those with asthma (Thaller et al., 2008). A Mexican study of children ages 6-14 detected an association between short-term O₃ exposure and wheeze, cough, and bronchodilator use among asthmatics but not non-asthmatics, although this may have been the result of a small non-asthmatic population (Escamilla-Nuñez et al., 2008). A study of modification by airway hyperresponsiveness (AHR) (a condition common among asthmatics) reported greater short-term O₃-associated decreases in lung function in elderly individuals with AHR, especially among those who were obese (Alexeeff et al., 2007). However, no evidence for increased risk was found in a study performed among children in Mexico City that examined the effect of short-term O₃ exposure on respiratory health (Barraza-Villarreal et al., 2008). In this study, a positive association was reported for airway inflammation among asthmatic children, but the observed association was
similar in magnitude to that of non-asthmatics. Similarly, a study of children in California reported an association between O₃ concentration and exhaled nitric oxide fraction (FeNO) that persisted both among children with and without asthma as well as those with and without respiratory allergy (Berhane et al., 2011). Finally, Khatri et al. (2009) found no association between short-term O₃ exposure and altered lung function for either asthmatic or non-asthmatic adults, but did note a decrease in lung function among individuals with allergies.

Evidence for difference in effects among asthmatics has been observed in studies that examined the association between O₃ exposure and altered lung function by asthma medication use. A study of children with asthma living in Detroit reported a greater association between short-term O₃ and lung function for corticosteroid users compared with noncorticosteroid users (Lewis et al., 2005). Conversely, another study of children found decreased lung function among noncorticosteroid users compared to corticosteroid users, although in this study, a large proportion of non-users were considered to be persistent asthmatics (Hernández-Cadena et al., 2009). Lung function was not related to short-term O₃ exposure among corticosteroid users and non-users in a study taking place among children during the winter months in Canada (Liu et al., 2009a). Additionally, a study of airway inflammation among individuals aged 12-65 years old reported a counterintuitive inverse association with O₃ of similar magnitude for all groups of corticosteroid users and non-users (Qian et al., 2009).

Controlled human exposure studies that have examined the effects of O₃ on individuals with asthma and healthy controls are limited. Based on studies reviewed in the 1996 and 2006 O₃ AQCDs, subjects with asthma appeared to be at least as sensitive to acute effects of O₃ in terms of FEV₁ and inflammatory responses as healthy non-asthmatic subjects. For instance, Horstman et al. (1995) observed that mild-to-moderate asthmatics, on average, experienced double the O₃-induced FEV₁ decrement of healthy subjects (19% versus 10%, respectively, p = 0.04). Moreover, a statistically significant positive correlation between FEV₁ responses to O₃ exposure and baseline lung function was observed in individuals with asthma, i.e., responses increased with severity of disease. Kreit et al. (1989) performed a short duration study in which asthmatics also showed a considerably larger average O₃-induced FEV₁ decrement than the healthy controls (25% versus 16%, respectively) following exposure to O₃ with moderate-heavy exercise. Alexis et al. (2000) and Jorres et al. (1996) also reported a tendency for slightly greater FEV₁ decrements in asthmatics than healthy subjects. Minimal evidence exists suggesting that individuals with asthma have smaller O₃-induced FEV₁ decrements than healthy subjects (3% versus 8%, respectively) (Mudway et al., 2001). However, the asthmatics in that study also tended to be older than the healthy subjects, which could partially explain their lesser response since FEV₁ responses to O₃ exposure diminish with age. Individuals with asthma also had more neutrophils in the BALF (18 hours postexposure) than similarly exposed healthy individuals (Peden et al., 1997; Scannell et al., 1996; Basha et al., 1994). Furthermore, a study examining the effects of O₃ on individuals with atopic asthma and healthy controls reported that greater numbers of neutrophils, higher levels of cytokines and hyaluronan, and greater expression of macrophage
cell-surface markers were observed in induced sputum of atopic asthmatics compared with healthy controls (Hernandez et al., 2010). Differences in O₃-induced epithelial cytokine expression were noted in bronchial biopsy samples from asthmatics and healthy controls (Bosson et al., 2003). Cell-surface marker and cytokine expression results, and the presence of hyaluronan, are consistent with O₃ having greater effects on innate and adaptive immunity in these asthmatic individuals (see Section 5.4.2.2). In addition, studies have demonstrated that O₃ exposure leads to increased bronchial reactivity to inhaled allergens in mild allergic asthmatics (Kehrl et al., 1999; Jorres et al., 1996) and to the influx of eosinophils in individuals with pre-existing allergic disease (Vagaggini et al., 2002; Peden et al., 1995). Taken together, these results point to several mechanistic pathways which could account for the increased risk of O₃-related health effects in subjects with asthma (see Section 5.4.2.2).

Toxicological studies provide biological plausibility for greater effects of O₃ among those with asthma or AHR. In animal toxicological studies, an asthmatic phenotype is modeled by allergic sensitization of the respiratory tract. Many of the studies that provide evidence that O₃ exposure is an inducer of AHR and remodeling utilize these types of animal models. For example, a series of experiments in infant rhesus monkeys have shown these effects, but only in monkeys sensitized to house dust mite allergen (Fanucchi et al., 2006; Joad et al., 2006; Schelegle et al., 2003). Similarly, Funabashi et al. (2004) demonstrated changes in pulmonary function in mice exposed to O₃, and Wagner et al. (2007) demonstrated enhanced inflammatory responses in rats exposed to O₃, but only in animals sensitized to allergen. In general, it is the combined effects of O₃ and allergic sensitization which result in measurable effects on pulmonary function. In a bleomycin induced pulmonary fibrosis model, exposure to 250 ppb O₃ for 5 days increased pulmonary inflammation and fibrosis, along with the frequency of bronchopneumonia in rats (Oyarzún et al., 2005). Thus, short-term exposure to O₃ may enhance damage in a previously injured lung.

In the 2006 O₃ AQCD, the potential for individuals with asthma to have greater risk of O₃-related health effects was supported by a number of controlled human exposure studies, evidence from toxicological studies, and a limited number of epidemiologic studies. Overall, in the recent epidemiologic literature some, but not all, studies report greater risk of health effects among individuals with asthma. Studies examining effect measure modification of the relationship between short-term O₃ exposure and altered lung function by corticosteroid use provided limited and inconsistent evidence of O₃-related health effects. Additionally, recent studies of behavioral responses have found that studies do not take into account individual behavioral adaptations to forecasted air pollution levels (such as avoidance and reduced time outdoors), which may underestimate the observed associations in studies that examined the effect of O₃ exposure on respiratory health (Neidell and Kinney, 2010). This could explain some inconsistency observed among recent epidemiologic studies. The evidence from controlled human exposure studies provides support for increased decrements in FEV₁ and greater inflammatory responses to O₃ in individuals with asthma than in healthy individuals without a history of asthma. These studies are often performed among individuals with mild asthma and therefore it is possible that individuals with severe asthma may have an
even greater risk of O$_3$-related health effects. The collective evidence for increased risk of O$_3$-related health effects among individuals with asthma from controlled human exposure studies is supported by recent toxicological studies which provide biological plausibility for heightened risk of asthmatics to respiratory effects due to O$_3$ exposure. Evidence indicating O$_3$-induced respiratory effects among individuals with asthma is further supported by additional studies of O$_3$-related respiratory effects (Section 6.2). Overall, there is adequate evidence for asthmatics to be an at-risk population based on the substantial, consistent evidence among controlled human exposure studies and coherence from epidemiologic and toxicological studies.

8.2.3 Chronic Obstructive Pulmonary Disease (COPD)

In the U.S. over 4% of adults report having chronic bronchitis and almost 2% report having emphysema, both of which are classified as COPD (Pleis et al., 2009).

A recent study reported no association between O$_3$ exposure and lung function regardless of whether the study participant had COPD or other pre-existing diseases (asthma or IHD) (Lagorio et al., 2006).

Peel et al. (2007) found that individuals with COPD were at increased risk of cardiovascular ED visits in response to short-term O$_3$ exposure compared to healthy individuals in Atlanta, GA. The authors reported that short-term O$_3$ exposure was associated with higher odds of an emergency department (ED) visit for peripheral and cerebrovascular disease among individuals with COPD compared to individuals without COPD. However, pre-existing COPD did not increase the odds of hospitalization for all CVD outcomes (i.e., IHD, dysrhythmia, or congestive heart failure). In an additional study performed in Taiwan, individuals with and without COPD had higher odds of congestive heart failure associated with O$_3$ exposure on warm days (Lee et al., 2008a). As discussed in Section 6.3, most studies reported no overall association between O$_3$ concentration and CVD morbidity.

Recent epidemiologic evidence indicates that persons with COPD may have increased risk of O$_3$-related cardiovascular effects, but little information is available on whether COPD leads to an increased risk of O$_3$-induced respiratory effects. Overall, this small number of studies provides inadequate evidence to determine whether COPD results in increased risk of O$_3$-related health effects.

8.2.4 Cardiovascular Disease (CVD)

Cardiovascular disease has become increasingly prevalent in the U.S., with about 12% of adults reporting a diagnosis of heart disease (Table 8-4). A high prevalence of other cardiovascular-related conditions has also been observed, such as hypertension which is prevalent among approximately 24% of adults. In the 2006 O$_3$ AQCD, little evidence was available regarding whether pre-existing CVD
contributed to increased risk of O₃-related health effects. Recent epidemiologic studies have examined cardiovascular-related diseases as modifiers of the O₃-outcome associations; however, no recent evidence is available from controlled human exposure studies or toxicological studies.

Peel et al. (2007) compared the associations between short-term O₃ exposure and cardiovascular ED visits in Atlanta, GA among multiple comorbid conditions. The authors found no evidence of increased risk of cardiovascular ED visits in individuals previously diagnosed with dysrhythmia, congestive heart failure, or hypertension compared to healthy individuals. Similarly, a study in France examined the association between O₃ concentrations and ischemic cerebrovascular events (ICVE) and myocardial infarction (MI) and the influence of multiple vascular risk factors on any observed associations (Henrotin et al., 2010). The association between O₃ exposure and ICVE was elevated for individuals with multiple risk factors, specifically individuals with diabetes or hypertension. For the association between O₃ and MI, increased odds were apparent only for those with hypercholesterolemia. In a study conducted in Taiwan, a positive association was observed for O₃ on warm days and congestive heart failure hospital admissions, but the association did not differ between individuals with/without hypertension or with/without dysrhythmia (Lee et al., 2008a). Another study in Taiwan reported that the association between O₃ levels and ED visits for arrhythmias were greater on warm days among those with congestive heart failure compared to those without congestive heart failure; however, the estimate and 95% CIs for those without congestive heart failure is completely contained within the 95% CI of those with congestive heart failure (Chiu and Yang, 2009).

Although not studied extensively, a study has examined the increased risk of O₃-related changes in blood markers for individuals with CVD. There was a greater association between O₃ exposure and some, but not all, blood inflammatory markers among individuals with a history of CVD (Liao et al., 2005). Liao et al. (2005) found that increased fibrinogen was positively associated with short-term O₃ exposure but this association was present only among individuals with a history of CVD.

No association was observed among those without a history of CVD. However, for another biomarker (vWF), CVD status did not modify the positive association with short-term O₃ exposure (Liao et al., 2005).

Mortality studies provide some evidence for a potential increase in O₃-induced mortality in individuals with pre-existing atrial fibrillation and atherosclerosis. In a study of 48 U.S. cities, increased risk of mortality with short-term O₃ exposure was observed only among individuals with secondary atrial fibrillation (Medina-Ramón and Schwartz, 2008). No association was observed for short-term O₃ exposure and mortality in a study of individuals with diabetes with or without CVD prior to death; however, there was some evidence of increased risk of mortality during the warm season if individuals had diabetes and atherosclerosis compared to only having diabetes (Goldberg et al., 2006).

Finally, although not extensively examined, a study explored whether a pre-existing CVD increased the risk of an O₃-induced respiratory effect. Lagorio et al. (2006)
examined the effect of O₃ exposure on lung function among participants with a variety of pre-existing diseases, including IHD. No association was observed regardless of whether the participant had IHD.

Overall, most short-term exposure studies did not report increased O₃-related cardiovascular morbidity for individuals with pre-existing CVD. However, as discussed in Section 6.3, most studies reported no overall association between O₃ concentration and CV morbidity. Thus, it is likely the association would be null regardless of the stratification. A limited number of studies examined whether cardiovascular disease modifies the association between O₃ and respiratory effects. There was some evidence that cardiovascular disease increases the risk of O₃-related mortality but again the number of studies was limited. Currently, evidence is inadequate to classify pre-existing CVD as a potential at-risk factor for O₃-related health effects. Future research among those with CVD compared to those without will increase the understanding of potential increased risk of O₃-related health effects among this group.

### 8.2.5 Diabetes

The literature has not extensively examined whether individuals with diabetes (about 8% of U.S. adults) are potentially at increased risk of O₃-related health effects. In a study of short-term O₃ exposure and cardiovascular ED visits in Atlanta, GA, no association was observed for individuals with or without diabetes (Peel et al., 2007). A similar study conducted in Taiwan reported a positive association between O₃ exposure on warm days and hospital admissions for congestive heart failure; however, no modification of the association by diabetes was observed (Lee et al., 2008a). Finally, in a study of O₃ exposure and ED visits for arrhythmia in Taiwan, there was no evidence of effect measure modification by diabetes on warm or cool days (Chiu and Yang, 2009). Currently, the limited number of epidemiologic studies as well as the lack of controlled human exposure or toxicological studies provides inadequate evidence to indicate whether diabetes results in a potentially increased risk of O₃-related health effects.

### 8.2.6 Hyperthyroidism

Hyperthyroidism has been identified in toxicological studies as a potential factor that may lead to increased risk of O₃-related health effects but has not yet been explored in epidemiologic or controlled human exposure studies. Lung damage and inflammation due to oxidative stress may be modulated by thyroid hormones. Compared to controls, hyperthyroid rats exhibited elevated levels of BAL neutrophils and albumin after a 4-hour exposure to O₃, indicating O₃-induced inflammation and damage. Hyperthyroidism did not affect production of reactive oxygen or nitrogen species, but BAL phospholipids were increased, indicating greater activation of Type II cells and surfactant protein production compared to normal rats (Huffman et al.,
Thus, this study provides some underlying evidence, which suggests that individuals with hyperthyroidism may represent an at-risk population; however, overall the lack of additional studies provides inadequate evidence to determine whether hyperthyroidism results in potentially increased risk of O₃-related health effects.

### 8.3 Sociodemographic Factors

#### 8.3.1 Lifestage

The 1996 and 2006 O₃ AQCDs identified children, especially those with asthma, and older adults as at-risk populations. These previous AQCDs reported clinical (controlled human exposure) evidence that children have greater spirometric responses to O₃ than middle-aged and older adults ([U.S. EPA, 1996a](https://www.epa.gov/)). Similar results were observed for symptomatic responses and O₃ exposure. Among older adults, most studies reported in the 2006 O₃ AQCD reported greater effects of short-term O₃ exposure and mortality compared to other age groups ([U.S. EPA, 2006b](https://www.epa.gov/)). Evidence published since the 2006 O₃ AQCD, summarized below, further supports these findings.

#### 8.3.1.1 Children

The 2000 Census reported that 28.6% of the U.S. population was under 20 years of age, with 14.1% under the age of 10 ([SSDAN CensusScope, 2010a](https://www.census.gov/)). Children’s respiratory systems are undergoing lung growth until about 18-20 years of age and are therefore thought to be intrinsically more at risk for O₃-induced damage ([U.S. EPA, 2006b](https://www.epa.gov/)). It is generally recognized that children spend more time outdoors than adults, and therefore would be expected to have higher exposure to O₃ than adults. The ventilation rates also vary between children and adults, particularly during moderate/heavy activity. Children aged 11 years and older and adults have higher absolute ventilation rates than children aged 1-11 years. However, children have higher ventilation rates relative to their lung volumes, which tends to increase dose normalized to lung surface area. Exercise intensity has a substantial effect on ventilation rate, with high intensity activities resulting in nearly double the ventilation rate during moderate activity among children and those adults less than 31 years of age. For more information on time spent outdoors and ventilation rate differences by age group, see Section 4.4.1.

The 1996 O₃ AQCD, reported clinical evidence that children, adolescents, and young adults (<18 years of age) appear, on average, to have nearly equivalent spirometric responses to O₃ exposure, but have greater responses than middle-aged and older adults ([U.S. EPA, 1996a](https://www.epa.gov/)). Symptomatic responses (e.g., cough, shortness of breath,
pain on deep inspiration) to O₃ exposure, however, appear to increase with age until early adulthood and then gradually decrease with increasing age (U.S. EPA, 1996a). For subjects aged 18-36 years, McDonnell et al. (1999b) reported that symptom responses from O₃ exposure also decrease with increasing age. Complete lung growth and development is not achieved until 18-20 years of age in women and the early 20s for men; pulmonary function is at its maximum during this time as well. Additionally, PBPK modeling reported lung regional extraction of O₃ to be higher in infants compared to adults. This is thought to be due to the smaller nasal and pulmonary regions’ surface area in children under the age of 5 years compared to the total airway surface area observed in adults (Sarangapani et al., 2003).

Recent epidemiologic studies have examined different age groups and their risk to O₃-related respiratory hospital admissions and ED visits. A study in Cyprus of short-term O₃ concentrations and respiratory hospital admissions detected possible effect measure modification by age with a larger association among individuals <15 years of age compared with those >15 years of age. However, this difference was only apparent with a 2-day lag (Middleton et al., 2008). Similarly, a Canadian study of asthma-ED visits reported the strongest O₃-related associations among 5 to 14 year-olds compared to the other age groups (ages examined 0-75+) (Villeneuve et al., 2007). Greater O₃-associated risk in asthma-related ED visits were also reported among children (<15 years) as compared to adults (15 to 64 years) in a study from Finland (Halonen et al., 2009). A study of New York City hospital admissions demonstrated an increase in the association between O₃ exposure and asthma-related hospital admissions for 6 to 18 year-olds compared to those <6 years old and those >18 years old (Silverman and Ito, 2010). When examining long-term O₃ exposure and asthma hospital admissions among children, associations were determined to be larger among children 1 to 2 years old compared to children 2 to 6 years old (Lin et al., 2008b). A few studies reported positive associations among both children and adults and no modification of the effect by age. A study performed in Hong Kong examined O₃ exposure and asthma-related hospital admissions for ages 0 to 14, 15 to 65, and >65 (Ko et al., 2007). The researchers reported that the association was greater among the 0 to 14 and 14 to 65 age groups compared to the >65 age group. Another study looking at asthma-related ED visits and O₃ exposure in Maine reported positive associations for all age groups (ages 2 to 65) (Paulu and Smith, 2008). Effects of O₃ exposure on asthma hospitalizations among both children and adults (<18 and ≥ 18 years old) were demonstrated in a study in Washington, but only children (<18 years of age) had statistically significant results at lag day 0, which the authors wrote, “suggests that children are more immediately responsive to adverse effects of O₃ exposure” (Mar and Koenig, 2009).

The evidence reported in epidemiologic studies is supported by recent toxicological studies which observed O₃-induced health effects in immature animals. Early life exposures of multiple species of laboratory animals, including infant monkeys, resulted in changes in conducting airways at the cellular, functional, ultra-structural, and morphological levels. Carey et al. (2007) conducted a study of O₃ exposure in infant rhesus macaques, whose respiratory tract closely resembles that of humans. Monkeys were exposed either acutely for 5 days to 0.5 ppm O₃, or episodically for 5 days to 0.25 ppm O₃.
biweekly cycles alternating 5 days of 0.5 ppm O₃ with 9 days of filtered air, designed to mimic human exposure (70 days total). All monkeys acutely exposed to O₃ had moderate to marked necrotizing rhinitis, with focal regions of epithelial exfoliation, numerous infiltrating neutrophils, and some eosinophils. The distribution, character, and severity of lesions in episodically exposed infant monkeys were similar to that of acutely exposed animals. Neither exposure protocol for the infant monkeys produced mucous cell metaplasia proximal to the lesions, an adaptation observed in adult monkeys exposed continuously to 0.3 ppm O₃ in another study (Harkema et al., 1987a). Functional (increased airway resistance and responsiveness with antigen + O₃ co-exposure) and cellular changes in conducting airways (increased numbers of inflammatory eosinophils) were common manifestations of exposure to O₃ among both the adult and infant monkeys (Plopper et al., 2007). In addition, the lung structure of the conducting airways in the infant monkeys was stunted by O₃ and this aberrant development was persistent 6 months postexposure. This developmental endpoint was not, of course, studied in the adult monkey experiments (Fanucchi et al., 2006). Thus, some functional and biochemical effects were similar between the infant and adult monkeys exposed to O₃, but because the study designs did not include concentration-response experiments, it is not possible to determine whether the infant monkeys were more at risk for the effects of O₃.

Similarly, rat fetuses exposed to O₃ in utero had ultrastructural changes in bronchiolar epithelium when examined near the end of gestation (López et al., 2008). In addition, exposure of mice to mixtures of air pollutants early in development affected pup lung cytokine levels (TNF, IL-1, KC, IL-6, and MCP-1) (Auten et al., 2009). In utero exposure of animals to PM augmented O₃-induced airway hyper-reactivity in these pups as juveniles.

Age may affect the inflammatory response to O₃ exposure. In comparing neonatal mice to adult mice, increased bronchoalveolar lavage (BAL) neutrophils were observed in four strains of neonates 24 hours after exposure to 0.8 ppm O₃ for 5 hours (Vancza et al., 2009). Three of these strains also exhibited increased BAL protein, although the two endpoints were not necessarily consistently correlated in a given strain. In some strains, however, adults were responsive, indicating a strain-age interaction. Measurement of ¹⁸O determined that the observed strain- and age-dependent differences were not due to absorbed O₃ dose. Using electron microscopy, Bils (1970) studied the lungs of mice of different ages (4 days or 1 to 2 months) exposed to 0.6 to 1.3 ppm O₃ for 6 to 7 h/day for 1 to 2 days and noted swelling of the alveolar epithelial lining cells without intra-alveolar edema. Swelling of endothelial cells and occasional breaks in the basement membrane were observed. These effects were most evident in younger mice exposed for 2 days. Toxicological studies reported that the difference in effects among younger lifestage test animals may be due to age-related changes in endogenous antioxidants and sensitivity to oxidative stress. A recent study demonstrated that 0.25 ppm O₃ exposure differentially altered expression of metalloproteinases in the skin of young (8 weeks old) and aged (18 months old) mice, indicating age-related variations in risk of oxidative stress (Fortino et al., 2007). Valacchi et al. (2007) found that aged mice had more Vitamin E in their plasma but less in their lungs compared to young mice.
which may affect their pulmonary antioxidant defenses. Servais et al. (2005) found higher levels of oxidative damage indicators in immature (3 weeks old) and aged (20 months old) rats compared to adult rats, the latter which were relatively resistant to an intermittent 7-day exposure to 0.5 ppm O₃. Immature rats exhibited a higher ventilation rate, which may have increased exposure. Additionally, a series of toxicological studies reported an association between O₃ exposure and bradycardia that was present among young but not older mice (Hamade et al., 2010; Tankersley et al., 2010; Hamade and Tankersley, 2009; Hamade et al., 2008). Regression analysis revealed an interaction between age and strain on heart rate, which implies that aging may affect heart rate differently among mouse strains (Tankersley et al., 2010).

The authors proposed that the genetic differences between the mice strains could be altering the formation of ROS, which tends to increase with age, thus modulating the changes in cardiopulmonary physiology after O₃ exposure.

The previous and recent human clinical (controlled human exposure) and toxicological studies reported evidence of increased risk from O₃ exposure for younger ages, which provides coherence and biological plausibility for the findings from epidemiologic studies. Although there was some inconsistency, generally, the epidemiologic studies reported larger associations for respiratory hospital admissions and ED visits for children than adults. The interpretation of these studies is limited by the lack of consistency in comparison age groups and outcomes examined. Toxicological studies observed O₃-induced health effects in immature animals, including infant monkeys, though the effects were not consistently greater in young animals than adults. However, overall, the epidemiologic, controlled human exposure, and toxicological studies provide substantial and consistent evidence within and across disciplines. Therefore, there is adequate evidence to conclude that children are at increased risk of O₃-related health effects.

### 8.3.1.2 Older Adults

Older adults may be at greater risk of health effects associated with O₃ exposure through a variety of intrinsic pathways. In addition, older adults may differ in their exposure and internal dose. Older adults spend slightly more time outdoors than adults aged 18-64 years. Older adults also have somewhat lower ventilation rates than adults aged 31 - less than 61 years. For more information on time spent outdoors and ventilation rate differences by age group, see Section 4.4.1. The gradual decline in physiological processes that occur with aging may lead to increased risk of O₃-related health effects (U.S. EPA, 2006a). Respiratory symptom responses to O₃ exposure appears to increase with age until early adulthood and then gradually decrease with increasing age (U.S. EPA, 1996a), which may put older adults at increased risk by withstanding continued O₃ exposure and thus not seeking relief and avoiding exposure. In addition, older adults, in general, have a higher prevalence of pre-existing diseases, with the exception of asthma, compared to younger age groups and this may also lead to increased risk of O₃-related health effects (see Table 8-4 that gives pre-existing rates by age). With the number of older Americans increasing
in upcoming years (estimated to increase from 12.4% of the U.S. population to 19.7% between 2000 to 2030, which is approximately 35 million and 71.5 million individuals, respectively) this group represents a large population potentially at risk of O₃-related health effects (SSDAN CensusScope, 2010a; U.S. Census Bureau, 2010).

The majority of recent studies reported greater effects of short-term O₃ exposure and mortality among older adults, which is consistent with the findings of the 2006 O₃ AQCD. A study conducted in 48 cities across the U.S. reported larger effects among adults ≥ 65 years old compared to those <65 years (Medina-Ramón and Schwartz, 2008). Further investigation of this study population revealed a trend of O₃-related mortality risk that gets larger with increasing age starting at age 50 (Zanobetti and Schwartz, 2008a). A study of 7 urban centers in Chile reported similar results, with greater effects in adults ≥ 65 years old, however the effects were smaller among those ≥ 85 years old compared to those in the 75 to 84 years old age range (Cakmak et al., 2007). More recently, a study conducted in the same area reported similar associations between O₃ exposure and mortality in adults aged <64 years old and 65 to 74 years old, but the risk was increased among older age groups (Cakmak et al., 2011). A study performed in China reported greater effects in populations ≥ 45 years old (compared to 5 to 44 year-olds), with statistically significant effects present only among those ≥ 65 years old (Kan et al., 2008). An Italian study reported higher risk of all-cause mortality associated with increased O₃ concentrations among individuals ≥ 85 years old as compared to those 35 to 84 years old. Those 65 to 74 and 75 to 84 years old did not show a greater increase in risk compared to those aged 35 to 64 years (Stafoggia et al., 2010). The Air Pollution and Health: A European and North American Approach (APHENA) project examined the association between O₃ exposure and mortality for those <75 and ≥ 75 years of age. In Canada, the associations for all-cause and cardiovascular mortality were greater among those ≥ 75 years old in the summer-only and all-year analyses. Age groups were not compared in the analysis for respiratory mortality in Canada. In the U.S., the association for all-cause mortality was slightly greater for those <75 years of age compared to those ≥ 75 years old in summer-only analyses. No consistent pattern was observed for CVD mortality. In Europe, slightly larger associations for all-cause mortality were observed in those <75 years old in all-year and summer-only analyses. Larger associations were reported among those <75 years for CVD mortality in all-year analyses, but the reverse was true for summer-only analyses (Katsouyanni et al., 2009).

Multiple epidemiologic studies of O₃ exposure and hospital admissions were stratified by age groups. A positive association was reported between short-term O₃ exposure and respiratory hospital admissions for adults ≥ 65 years old but not for those adults aged 15 to 64 years (Halonen et al., 2009). In the same study, no association was observed between O₃ concentration and respiratory mortality among those ≥ 65 years old or those 15 to 64 years old; however, an inverse association between O₃ concentration and cardiovascular mortality was present among individuals ≥ 65 years old but not among individuals <65 years old. This inverse association among those ≥ 65 years old persisted when examining hospital
admissions for coronary heart disease. A study of CVD-related hospital visits in Bangkok, Thailand reported an increase in percent change for hospital visits with previous day and cumulative 2-day O₃ levels among those ≥ 65 years old, whereas no association was present for individuals less than 65 years of age (Buadong et al., 2009). No association was observed for current day or cumulative 3-day averages in any age group. A study examining O₃ and hospital admissions for CVD-related health effects reported no association for individuals aged 15 to 64 or individuals aged ≥ 65 years, although one lag-time did show an inverse effect for coronary heart disease among elderly that was not present among 15 to 64 year-olds (Halonen et al., 2009). However, as discussed in the section on CVD hospital admissions (Section 6.3.2.7), results were inconsistent and often null so it is plausible that no association would be observed regardless of age. No modification by age (40 to 64 year-olds versus >64 years old) was observed in a study from Brazil examining O₃ levels and COPD ED visits (Arbex et al., 2009).

Biological plausibility for differences by age is provided by toxicological studies. Ozone exposure resulted in an increase in left ventricular chamber dimensions at end diastole (LVEDD) in young and old mice, whereas decreases in left ventricular posterior wall thickness at end systole (PWTES) were only observed among older mice (Tankersley et al., 2010). Other toxicological studies also indicate increased risk in older animals for additional endpoints, including neurological and immune. The hippocampus, one of the main regions affected by age-related neurodegenerative diseases, may be more sensitive to oxidative damage in aged rats. In a study of young (47 days) and aged (900 days) rats exposed to 1 ppm O₃ for 4 hours, O₃-induced lipid peroxidation occurred to a greater extent in the striatum of young rats, whereas it was highest in the hippocampus in aged rats (Rivas-Arancibia et al., 2000).

In young mice, healing of skin wounds is not significantly affected by O₃ exposure (Lim et al., 2006). However, exposure to 0.5 ppm O₃ for 6 h/day significantly delays wound closure in aged mice.

Although some outcomes reported mixed findings regarding an increase in risk for older adults, recent epidemiologic studies report consistent positive associations between short-term O₃ exposure and mortality in older adults. The evidence from mortality studies is consistent with the results reported in the 2006 O₃ AQCD and is supported by toxicological studies providing biological plausibility for increased risk of effects in older adults. Also, older adults may be experiencing increased exposure compared to younger adults due to time spend outdoors and withstanding exposures. Overall, adequate evidence is available indicating that older adults are at increased risk of O₃-related health effects based on the substantial and consistent evidence within epidemiologic studies on O₃ exposure and mortality and the coherence with toxicological studies.
8.3.2 Sex

The distribution of males and females in the U.S. is similar. In 2000, 49.1% of the U.S. population was male and 50.9% were female. However, this distribution does vary by age with a greater prevalence of females ≥ 65 years old compared to males (SSDAN CensusScope, 2010a). The 2006 O₃ AQCD did not report evidence of differences between the sexes in health responses to O₃ exposure (U.S. EPA, 2006b). Recent epidemiologic studies have evaluated the effects of short-term and long-term exposure to O₃ on multiple health endpoints stratified by sex.

A study in Maine that examined short-term O₃ concentrations and asthma ED visits detected greater effects among males ages 2 to 14 years and among females ages 15 to 34 years compared to males and females in the same age groups (no difference was detected for males and females aged 35 to 64) (Paulu and Smith, 2008). A Canadian study reported no associations between short-term O₃ and respiratory infection hospital admissions for either boys or girls under the age of 15 (Lin et al., 2005), whereas another Canadian study reported a slightly higher but non-statistically significant increase in respiratory hospital admissions for males (mean ages 47.6 to 69.0 years) (Cakmak et al., 2006b). A recent study from Hong Kong examining individuals of all ages reported no effect measure modification by sex for overall respiratory disease hospital admissions, but did detect a greater excess risk of hospital admissions for COPD among females compared to males (Wong et al., 2009). Similarly a study in Brazil found higher effect estimates for COPD ED visits among females compared to males (Arbex et al., 2009). Higher levels of respiratory hospital admissions with greater O₃ concentrations was also observed for females in a study of individuals living in Cyprus (Middleton et al., 2008). A study of lung function unrelated to hospital admissions and ED visits was conducted among lifeguards in Texas and reported decreased lung function with increased O₃ exposure among females but not males (Thaller et al., 2008). This study included individuals aged 16 to 27 years, and the majority of participants were male. A New York study found no evidence of effect measure modification of the association between long-term O₃ exposure and asthma hospital admissions among males and females between 1 and 6 years old (Lin et al., 2008b).

In addition to examining the potential modification of O₃ associations with respiratory outcomes by sex, studies also examined cardiovascular-related outcomes specifically hospital admissions and ED visits. All of these studies reported no effect modification by sex with some studies reporting null associations for both males and females (Wong et al., 2009; Middleton et al., 2008; Villeneuve et al., 2006a) and one study reporting a positive associations for both sexes (Cakmak et al., 2006a).

A French study examining the associations between O₃ concentrations and risk of ischemic strokes (not limited to ED visits or hospital admissions) reported no association for either males or females with lags of 0, 2, or 3 days (Henrotin et al., 2007). A positive association was reported for males with a lag of 1 day, but this association was null for females. The authors noted that men in the study had much higher rates of current and former smoking than women (67.4% versus 9.3%). Additionally, cardiovascular hospital admissions and ED visits overall have
demonstrated inconsistent and null results (Section 6.3.2.7). The lack of effect measure modification by sex may be indicative of the lack of association, not the lack of an effect by sex.

A biomarker study investigating the effects of O₃ concentrations on high-sensitivity C-reactive protein (hs-CRP), fibrinogen, and white blood cell (WBC) count, reported observations for various lag times ranging from 0 to 7 days (Steinvil et al., 2008). Most of the associations were null for males and females although one association between O₃ and fibrinogen was positive for males and null for females (lag day 4); however, this positive association was null or negative when other pollutants were included in the model. One study examining correlations between O₃ levels and oxidative DNA damage examined results stratified by sex. In this study Palli et al. (2009) reported stronger correlations for males than females, both during short-term exposure (less than 30 days) and long-term exposure (0-90 days). However, the authors commented that this difference could have been partially explained by different distributions of exposure to traffic pollution at work.

A few studies have examined the association between short-term O₃ concentrations and mortality stratified by sex and, in contrast with studies of other endpoints, were more consistent in reporting elevated risks among females. These studies, conducted in the U.S. (Medina-Ramón and Schwartz, 2008), Italy (Stafoggia et al., 2010), and Asia (Kan et al., 2008), reported larger effect estimates in females compared to males. In the U.S. study, the elevated risk of mortality among females was greater specifically among those ≥ 60 years old (Medina-Ramón and Schwartz, 2008). However, a recent study in Chile reported similar associations between O₃ exposure and mortality among both men and women (Cakmak et al., 2011). A long-term O₃ exposure study of respiratory mortality stratified their results by sex and reported relative risks of 1.01 (95% CI: 0.99, 1.04) for males and 1.04 (95% CIs 1.03, 1.07) for females (Jerrett et al., 2009).

Experimental research provided a further understanding of the underlying mechanisms that may explain a possible differential risk in O₃-related health effects among males and females. Several studies have suggested that physiological differences between sexes may predispose females to greater effects from O₃. In females, lower plasma and nasal lavage fluid (NLF) levels of uric acid (most prevalent antioxidant), the initial defense mechanism of O₃ neutralization, may be a contributing factor (Housley et al., 1996). Consequently, reduced absorption of O₃ in the upper airways of females may promote its deeper penetration. Dosimetric measurements have shown that the absorption distribution of O₃ is independent of sex when absorption is normalized to anatomical dead space (Bush et al., 1996). Thus, a differential removal of O₃ by uric acid seems to be minimal. In general, the physiologic response of young healthy females to O₃ exposure appears comparable to the response of young males (Hazucha et al., 2003). A few studies have examined changes in O₃ responses during various menstrual cycle phases. Lung function response to O₃ was enhanced during the follicular phase of the menstrual cycle compared to the luteal phase in a small study of women (Fox et al., 1993). However, Seal et al. (1996) later reported no effect of menstrual cycle phase in their analysis of
responses from 150 women, but conceded that the methods used by Fox et al. (1993) more precisely defined the menstrual cycle phase. Another study also reported no difference in responses among females during the follicular and luteal phases of their cycle (Weinmann et al., 1995c). Additionally, in this study the responses in women were comparable to those reported for men in the study. In a toxicological study, small differences in effects by sex were seen in adult mice with respect to pulmonary inflammation and injury after a 5-h exposure to 0.8 ppm O₃, and although adult females were generally more at risk, these differences were strain-dependent, with some strains exhibiting greater risk in males (Vancza et al., 2009). The most obvious sex difference was apparent in lactating females, which incurred the greatest lung injury or inflammation among several of the strains.

Overall, results have varied, with recent evidence for increased risk for O₃-related health effects present for females in some studies and males in other studies. Most studies examining the associations O₃ and mortality report females to be at greater risk than males, but minimal evidence is available regarding a difference between the sexes for other outcomes. Inconsistent findings were reported on whether effect measure modification exists by sex for respiratory and cardiovascular hospital admissions and ED visits, although there is some indication that females are at increased risk of O₃-related respiratory hospital admissions and ED visits. While O₃-related effects may occur in both men and women, there is suggestive evidence exists indicating that females are at potentially increased risk of O₃-related health effects as there are consistent findings among epidemiologic studies of mortality.

### 8.3.3 Socioeconomic Status

SES is often represented by personal or neighborhood SES, which is comprised of a variety of components such as educational attainment, household income, health insurance status, and other such factors. SES is often indicative of such things as access to healthcare, quality of housing, and pollution gradient to which people are exposed. One or a combination of these components could modify the risk of O₃-related health effects. Based on the 2000 Census data, 12.4% of Americans live in poverty (poverty threshold for a family of four was $17,463) (SSDAN CensusScope, 2010c). Although included below, studies stratifying by SES that are conducted outside the U.S. may not be comparable to those studies from within the United States. Having low SES in another country may be different than having low SES in the U.S. based on SES definitions, population composition, and/or conditions in that country.

Multiple epidemiologic studies have reported individuals of low SES to have increased risk for the effects of short-term O₃ exposure on respiratory hospital admissions and ED visits. In New York State, larger associations between long-term O₃ exposure and asthma hospital admissions were observed among children of mothers who did not graduate from high school, whose births were covered by Medicaid/self-paid, or who were living in poor neighborhoods compared to children.
whose mothers graduated from high school, whose births were covered by other insurance, or who were not living in poor neighborhoods, respectively (Lin et al., 2008b). In addition, a study conducted across 10 cities in Canada found the largest association between O₃ exposure and respiratory hospital admissions was among those with an educational level less than grade 9, but no consistent trend in the effect was seen across quartiles of income (Cakmak et al., 2006b). A Canadian study reported inverse effects of O₃ on respiratory hospital admissions and ED visits for all levels of SES, measured by average census tract household income (Burra et al., 2009). A study performed in Korea examined the association between O₃ concentrations and asthma hospital admissions and reported larger effect estimates in areas of moderate and low SES compared with areas of high SES (SES was based on average regional insurance rates) (Lee et al., 2006).

The examination of the potential effects of SES on O₃-related cardiovascular health effects is relatively limited. A study conducted in Canada reported the association between short-term O₃ and ED visits for cardiac disease by quartiles of neighborhood-level education and income. No effect measure modification was apparent for either measure of SES (Cakmak et al., 2006a). However, this may be due to the lack of association present between O₃ and ED visits for cardiac disease regardless of SES.

Several studies were conducted that examined the modification of the relationship between short-term O₃ concentrations and mortality by SES. A U.S. multicity study reported that communities with a higher proportion of the population unemployed had higher O₃-related mortality effect estimates (Bell and Dominici, 2008). A study in seven urban centers in Chile reported on modification of the association between O₃ exposure and mortality using multiple SES markers (Cakmak et al., 2011). Increased risk was observed among the categories of low SES for all measures (personal educational attainment, personal occupation, community income level). Additionally, the APHENA study, which examined the association between O₃ and mortality by percentage unemployed, reported a higher percent change in mortality with increased percent unemployed but this varied across the regions included in the study (U.S., Canada, Europe) (Katsouyanni et al., 2009). A Chinese study reported that the greatest effects between O₃ concentrations and mortality at lag day 0 were among individuals living in areas of high social deprivation (i.e., low SES), but this association was not consistent across lag days (at other lag times, the middle social deprivation index category had the greatest association) (Wong et al., 2008). However, another study in Asia comparing low to high educational attainment populations reported no evidence of greater mortality effects (total, CVD, or respiratory) (Kan et al., 2008). Additionally, a study in Italy reported no difference in risk of mortality among census-block level derived income levels (Stafoggia et al., 2010). A study of infant mortality in Mexico reported no association between O₃ concentrations and infant mortality among any of the three levels of SES determined using a socioeconomic index based on residential areas (Romieu et al., 2004a). Another study in Mexico reported a positive association between O₃ levels at lag 0 and respiratory-related infant mortality in only the low SES group (determined based on education, income, and household conditions across residential areas), but no
association was observed in any of the SES groups with other lags (Carbajal-Arroyo et al., 2011).

Studies of O₃ concentrations and reproductive outcomes have also examined associations by SES levels. A study in California reported greater decreases in birth weight associated with full pregnancy O₃ concentration for those with neighborhood poverty levels of at least 7% compared with those in neighborhoods with less than 7% poverty (the authors do not provide information on how categories of the SES variable were determined) (Morello-Frosch et al., 2010). No dose response was apparent and those with neighborhood poverty levels of 7-21% had greater decreases observed for the association than those living in areas with poverty rates of at least 22%. An Australian study reported an inverse association between O₃ exposure during days 31-60 of gestation and abdominal circumference during gestation (Hansen et al., 2008). The interaction with SES (area-level measured socioeconomic disadvantage) was examined and although the inverse association remained statistically significant in only the highest SES quartile, there were large confidence interval overlaps among estimates for each quartile so no difference in the association for the quartiles was apparent.

Evidence from a controlled human exposure study that examined O₃ effects on lung function does not provide support for greater O₃-related health effects in individuals of lower SES. In a follow-up study on modification by race, Seal et al. (1996) reported that, of three SES categories, individuals in the middle SES category showed greater concentration-dependent decline in percent-predicted FEV₁ (4-5% at 400 ppb O₃) than in low and high SES groups. The authors did not have an “immediately clear” explanation for this finding and controlled human exposure studies are typically not designed to answer questions about SES.

Overall, most studies of individuals have reported that individuals with low SES and those living in neighborhoods with low SES are more at risk for O₃-related health effects, resulting in increased risk of respiratory hospital admissions and ED visits. Inconsistent results have been observed in the few studies examining effect modification of associations between O₃ exposure and mortality and reproductive outcomes. Also, a controlled human exposure study does not support evidence of increased risk of respiratory morbidity among individuals with lower SES. Overall, evidence is suggestive of SES as a factor affecting risk of O₃-related health outcomes based on collective evidence from epidemiologic studies of respiratory hospital admissions but inconsistency among epidemiologic studies of mortality and reproductive outcomes. Further studies are needed to confirm this relationship, especially in populations within the U.S.

### 8.3.4 Race/Ethnicity

Based on the 2000 Census, 69.1% of the U.S. population identified as non-Hispanic whites. Approximately 12.1% of people reported their race/ethnicity as non-Hispanic black and 12.6% reported being Hispanic (SSDAN CensusScope, 2010b).
Only a few studies examined the associations between short-term O₃ concentrations and mortality and reported higher effect estimates among blacks (Medina-Ramón and Schwartz, 2008) and among communities with larger proportions of blacks (Bell and Dominici, 2008). Another study examined long-term exposure to O₃ concentrations and asthma hospital admissions among children in New York State. These authors reported no statistically significant difference in the odds of asthma hospital admissions for blacks compared to other races but did detect higher odds for Hispanics compared to non-Hispanics (Lin et al., 2008b).

Additionally, recent epidemiologic studies have stratified by race when examining the association between O₃ concentration and birth outcomes. A study conducted in Atlanta, GA reported decreases in birth weight with increased third trimester O₃ concentrations among Hispanics but not among non-Hispanic whites (Darrow et al., 2011b). A California study reported that the greatest decrease in birth weight associated with full pregnancy O₃ concentration was among non-Hispanic whites (Morello-Frosch et al., 2010). This inverse association was also apparent, although not as strong, for Hispanics and non-Hispanic blacks. Increased birth weight was associated with higher O₃ exposure among non-Hispanic Asians and Pacific Islanders but these results were not statistically significant.

Similar to the epidemiologic studies, a controlled human exposure study suggested differences in lung function responses by race (Seal et al., 1993). The independent effects of sex-race group and O₃ concentration on lung function were positive, but the interaction between sex-race group and O₃ concentration was not statistically significant. The findings indicated some overall difference between the sex-race groups that was independent of O₃ concentration (the concentration-response curves for the four sex-race groups are parallel). In a multiple comparison procedure on data collapsed across all O₃ concentrations for each sex-race group, both black men and black women had larger decrements in FEV₁ than did white men. The authors noted that the O₃ dose per unit of lung tissue would be greater in blacks and females than whites and males, respectively. That this difference in tissue dose might have affected responses to O₃ cannot be ruled out. The college students recruited for the Seal et al. (1993) study were probably from better educated and more SES advantaged families, thus reducing potential for these variables to be confounding factors. Que et al. (2011) also examined pulmonary responses to O₃ exposure in blacks of African American ancestry and in whites. On average, the black males experienced the greatest decrements in FEV₁ following O₃ exposure. This decrease was larger than the decrement observed among black females, white males, and white females.

Overall, the results of recent studies indicate that there may be race-related increase in risk of O₃-related health effects for some outcomes, although the overall understanding of potential effect measure modification by race is limited by the small number of studies. Additionally, these results may be confounded by other factors, such as SES. Overall, evidence is inadequate to determine if O₃-related health effects vary by race because of the insufficient quantity of studies and lack of consistency within disciplines.
8.4 Behavioral and Other Factors

8.4.1 Diet

Diet was not examined as a factor potentially affecting risk in previous O₃ AQCDs, but recent studies have examined modification of the association between O₃ and health effects by dietary factors. Because O₃ mediates some of its toxic effects through oxidative stress, the antioxidant status of an individual is an important factor that may contribute to increased risk of O₃-related health effects. Supplementation with Vitamins C and E has been investigated in a number of studies as a means of inhibiting O₃-mediated damage.

Epidemiologic studies have examined effect measure modification by diet and found evidence that certain dietary components are related to the effect O₃ has on respiratory outcomes. In a recent study the effects of fruit/vegetable intake and Mediterranean diet was examined (Romieu et al., 2009). Increases in these food patterns, which have been noted for their high Vitamins C and E and omega-3 fatty acid content, protected against O₃-related decreases in lung function among children living in Mexico City. Another study examined supplementation of the diets of asthmatic children in Mexico with Vitamins C and E (Sienra-Monge et al., 2004). Associations were detected between short-term O₃ exposure and nasal airway inflammation among children in the placebo group but not in those receiving the supplementation. The authors concluded that “Vitamin C and E supplementation above the minimum dietary requirement in asthmatic children with a low intake of Vitamin E might provide some protection against the nasal acute inflammatory response to ozone.”

The epidemiologic evidence is supported by controlled human exposure studies, which have shown that the first line of defense against oxidative stress is antioxidants-rich extracellular lining fluid (ELF) which scavenge free radicals and limit lipid peroxidation. Exposure to O₃ depletes the antioxidant level in nasal ELF probably due to scrubbing of O₃ (Mudway et al., 1999a); however, the concentration and the activity of antioxidant enzymes either in ELF or plasma do not appear to be related to O₃ responsiveness (e.g., pulmonary function and inflammation) (Samet et al., 2001; Avissar et al., 2000; Blomberg et al., 1999). Carefully controlled studies of dietary antioxidant supplementation have demonstrated some protective effects of α-tocopherol (a form of Vitamin E) and ascorbate (Vitamin C) on spirometric measures of lung function after O₃ exposure but not on the intensity of subjective symptoms and inflammatory response including cell recruitment, activation and a release of mediators (Samet et al., 2001; Trenga et al., 2001). Dietary antioxidants have also afforded partial protection to asthmatics by attenuating postexposure bronchial hyperresponsiveness (Trenga et al., 2001).

Toxicological studies provide evidence of biological plausibility to the epidemiologic and controlled human exposure studies. Wagner et al. (2009); (2007) found reductions in O₃-exacerbated nasal allergy responses in rats with γ-tocopherol
treatment (a form of Vitamin E). O₃-induced inflammation and mucus production were also inhibited by γ-tocopherol. Supplementation with Vitamins C and E partially ameliorated inflammation, oxidative stress, and airway hyperresponsiveness in guinea pigs exposed subchronically to 0.12 ppm O₃ ppm (Chhabra et al., 2010). Inconsistent results were observed in other toxicological studies of Vitamin C deficiency and O₃-induced responses. Guinea pigs deficient in Vitamin C displayed only minimal injury and inflammation after exposure to O₃ (Kodavanti et al., 1995). A recent study in mice demonstrated a protective effect of β-carotene in the skin, where it limited the production of proinflammatory markers and indicators of oxidative stress induced by O₃ exposure (Valacchi et al., 2009). Deficiency of Vitamin A, which has a role in regulating the maintenance and repair of the epithelial layer, particularly in the lung, appears to enhance the risk of O₃-induced lung injury (Paquette et al., 1996). Differentially susceptible mouse strains that were fed a Vitamin A sufficient diet were observed to have different tissue concentrations of the vitamin, potentially contributing to their respective differences in O₃-related outcomes. In addition to the studies of antioxidants, one toxicological study examined protein deficiency. Protein deficiency alters the levels of enzymes and chemicals in the brain of rats involved with redox status; exposure to 0.75 ppm O₃ has been shown to differentially affect Na⁺/K⁺ ATPase, glutathione, and lipid peroxidation, depending on the nutritional status of the animal, but the significance of these changes is unclear (Calderón Guzmán et al., 2006). There may be a protective effect of overall dietary restriction with respect to lung injury, possibly related to increased Vitamin C in the lung surface fluid (Kari et al., 1997).

There is adequate evidence that individuals with reduced intake of Vitamins E and C are at risk for O₃-related health effects based on substantial, consistent evidence both within and among disciplines. The evidence from epidemiologic studies is supported by controlled human exposure and toxicological studies.

8.4.2 Obesity

Obesity, defined as a BMI of 30 kg/m² or greater, is an issue of increasing importance in the U.S., with self-reported rates of obesity of 26.7% in 2009, up from 19.8% in 2000 (Sherry et al., 2010). BMI may affect O₃-related health effects through multiple avenues, such as, inflammation in the body, increased pre-existing disease, and poor diet. Increased risk of PM-related health effects have been observed among obese individuals compared with non-obese individuals [2009 PM ISA (U.S. EPA, 2009d)].

A few studies have been performed examining the association between BMI and O₃-related changes in lung function. An epidemiologic study reported decreased lung function with increased short-term O₃ exposure for both obese and non-obese subjects; however, the magnitude of the reduction in lung function was greater for those subjects who were obese (Alexeiff et al., 2007). Further decrements in lung function were noted for obese individuals with AHR. Controlled human exposure
studies have also detected differential effects of O₃ exposure on lung function for individuals with varying BMIs. In a retrospective analysis of data from 541 healthy, nonsmoking, white males between the ages of 18-35 years from 15 studies conducted at the U.S. EPA Human Studies Facility in Chapel Hill, North Carolina, McDonnell et al. (2010) found that increased body mass index (BMI) was found to be associated with enhanced FEV₁ responses. The BMI effect was of the same order of magnitude but in the opposite direction of the age effect whereby FEV₁ responses diminish with increasing age. In a similar analysis, Bennett et al. (2007) found enhanced FEV₁ decrements following O₃ exposure with increasing BMI in a group of healthy, nonsmoking, women (BMI range 15.7 to 33.4), but not among healthy, nonsmoking men (BMI range 19.1 to 32.9). In the women, greater O₃-induced FEV₁ decrements were seen in individuals that were overweight/obese (BMI >25) compared to normal weight (BMI from 18.5 to 25), and in normal weight compared to underweight (BMI <18.5). Even disregarding the five underweight women, a greater O₃ response in the overweight/obese category (BMI >25) was observed compared with the normal weight group (BMI from 18.5 to 24.9).

Studies in genetically and dietarily obese mice have shown enhanced pulmonary inflammation and injury with acute O₃ exposure (Johnston et al., 2008; Shore, 2007). However, a recent study found that obese mice are actually resistant to O₃-induced pulmonary injury and inflammation and reduced lung compliance following longer exposures (72 hours) at lower concentrations (0.3 ppm O₃), regardless of whether obesity was genetic- or diet-induced (Shore et al., 2009).

Multiple epidemiologic, controlled human exposure, and toxicological studies have reported suggestive evidence for increased O₃-related respiratory health effects among obese individuals. Future research of the effect modification of the relationship between O₃ and other health-related outcomes besides respiratory health effects by BMI and studies examining the role of physical conditioning will advance understanding of obesity as a factor potentially increasing an individual’s risk.

### 8.4.3 Smoking

Previous O₃ AQCDs have concluded that smoking does not increase the risk of O₃-related health effects; in fact, in controlled human exposure studies, smokers have been found to be less responsive to O₃ than non-smokers. Data from recent interviews conducted as part of the 2008 National Health Interview Survey (NHIS) (Pleis et al., 2009) have shown the rate of smoking among adults ≥ 18 years old to be approximately 20% in the United States. Approximately 21% of individuals surveyed were identified as former smokers.

Baccarelli et al. (2007) performed a study of O₃ concentrations and plasma homocysteine levels (a risk factor for vascular disease). They found no interaction of smoking (smokers versus non-smokers) for the associations between O₃ concentrations and plasma homocysteine levels. Another study examined the association between O₃ and resting heart rate and also reported no interaction with
smoking status (current smokers versus current non-smokers) (Ruidavets et al., 2005a).

A study examining correlations between O₃ levels and oxidative DNA damage examined results stratified by current versus never and former smokers (Palli et al., 2009). Ozone was positively associated with DNA damage for short-term and long-term exposures among never/former smokers. For current smokers, short-term O₃ concentrations were inversely associated with DNA damage; however, the number of current smokers in the study was small (n = 12).

The findings of Palli et al. (2009) were consistent with those from controlled human exposure studies that have confirmed that smokers are less responsive to O₃ exposure than non-smokers. Spirometric and plethysmographic pulmonary function decline, nonspecific AHR, and inflammatory responses of smokers to O₃ exposure were all weaker than those reported for non-smokers. Similarly, the time course of development and recovery from these effects, as well as their reproducibility, was not different from non-smokers. Chronic airway inflammation with desensitization of bronchial nerve endings and an increased production of mucus may plausibly explain the pseudo-protective effect of smoking (Frampton et al., 1997a; Torres et al., 1997).

These findings for smoking are consistent with the conclusions from previous AQCDs. An epidemiologic study of O₃-associated DNA damage reported smokers to be less at risk for O₃-related health effects. In addition, both epidemiologic studies of short-term exposure and CVD outcomes found no evidence of effect measure modification by smoking. No toxicological studies provide biological support for O₃-related effects. Overall, evidence of potential differences in O₃-related health effects by smoking status is inadequate due to insufficient coherence and a limited number of studies.

8.4.4 Outdoor Workers

Studies included in the 2006 O₃ AQCD reported that individuals who participate in outdoor activities or work outside to be a population at increased risk based on consistently reported associations between O₃ exposure and respiratory health outcomes in these groups (U.S. EPA, 2006b). Outdoor workers are exposed to ambient O₃ concentrations for a greater period of time than individuals who spend their days indoors. As discussed in Section 4.3.3 of this ISA, outdoor workers sampled during the work shift had a higher ratio of personal exposure to fixed-site monitor concentrations than health clinic workers who spent most of their time indoors. Additionally, an increase in dose to the lower airways is possible during outdoor exercise due to both increases in the amount of air breathed (i.e., minute ventilation) and a shift from nasal to oronasal breathing (Sawyer et al., 2007; Nodelman and Ultman, 1999; Hu et al., 1994). For further discussion of the association between FEV₁ responses to O₃ exposure and minute ventilation, refer to Section 6.2.3.1 of the 2006 O₃ AQCD. A recent study has explored the potential effect measure modification of O₃ exposure and DNA damage by indoor/outdoor
workplace (Tovalin et al., 2006). In a study of indoor and outdoor workers in Mexico, individuals who worked outdoors in Mexico City had a slight association between O₃ exposure and DNA damage (measured by comet tail length assay), whereas no association was observed for indoor workers. However, workers in another Mexican city, Puebla, demonstrated no association between O₃ levels and DNA damage, regardless of whether they worked indoors or outdoors.

Previous studies have shown that increased exposure to O₃ due to outdoor work leads to increased risk of O₃-related health effects, specifically decrements in lung function (U.S. EPA, 2006b). Additionally, outdoor workers may be an at-risk population due to their increased dose and exposure to O₃. Recent evidence from a stratified analysis does not indicate that increased O₃ exposure due to outdoor work leads to DNA damage. However, the strong evidence from the 2006 O₃ AQCD which demonstrated increased exposure, dose, and ultimately risk of O₃-related health effects in this population supports that there is adequate evidence available to indicate that increased exposure to O₃ through outdoor work increases the risk of O₃-related health effects.

### 8.4.5 Air Conditioning Use

Air conditioning use is an important indicator of O₃ exposure, as use of central air conditioning will limit exposure to O₃ by blocking the penetration of O₃ into the indoor environment and lack of air conditioning may be linked to increased exposure by use of open windows (see Section 4.3.2). Air conditioning use is a difficult effect measure modifier to examine in epidemiologic studies because it is often estimated using regional prevalence data and may not reflect individual-level use. More generally, air conditioning prevalence is associated with temperature of a region; those areas with higher temperatures have a greater prevalence of households with air conditioning. Therefore, not having air conditioning is not necessarily indicative of higher O₃ exposure. Despite these limitations, a few studies have examined effect measure modification by prevalence of air conditioning use in an area. Studies examining multiple cities across the U.S. have assessed whether associations between O₃ concentrations and hospital admissions and mortality varied among areas with high and low prevalence of air conditioning. Medina-Ramon et al. (2006) conducted a study during the warm season and observed a greater association between O₃ levels and pneumonia-hospital admissions among areas with a lower proportion of households having central air conditioning compared to areas with a larger proportion of households with air conditioning. However, a similar observation was not observed when examining COPD hospital admissions complicating the interpretation of the results from this study. Bell and Dominici (2008) found evidence of increased risk of O₃-related mortality in areas with a lower prevalence of central air conditioning in a study of 98 U.S. communities. Conversely, Medina-Ramón and Schwartz (2008) found that among individuals with atrial fibrillation, a lower risk of mortality was observed for areas with a lower prevalence of central air conditioning.
The limited number of studies that examined whether air conditioning use modifies the association between \( \text{O}_3 \) exposure and health has not provided consistent evidence across health endpoints. Therefore, the limited and inconsistent results across epidemiologic studies and the regional variation of air conditioning use has provided inadequate evidence to determine whether a lower prevalence of air conditioning use leads to a potential increased risk of \( \text{O}_3 \)-related health effects.

### 8.5 Summary

In this section, epidemiologic, controlled human exposure, and toxicological studies have been evaluated and indicate that various factors may lead to increased risk of \( \text{O}_3 \)-related health effects (Table 8-6).

The populations and lifestages identified in this section that have “adequate” evidence for increased \( \text{O}_3 \)-related health effects are individuals with certain genotypes, individuals with asthma, younger and older age groups, individuals with reduced intake of certain nutrients, and outdoor workers, based on consistency in findings across studies and evidence of coherence in results from different scientific disciplines. Multiple genetic variants have been observed in epidemiologic and controlled human exposure studies to affect the risk of \( \text{O}_3 \)-related respiratory outcomes and support is provided by toxicological studies of genetic factors. Asthma as a factor affecting risk was supported by controlled human exposure and toxicological studies, as well as some evidence from epidemiologic studies. Generally, studies of age groups reported positive associations for respiratory hospital admissions and ED visits among children. Biological plausibility for this increased risk is supported by toxicological and controlled human exposure research. Also, children have higher exposure and dose due to increased time spent outdoors and ventilation rate. Most studies comparing age groups reported greater effects of short-term \( \text{O}_3 \) exposure on mortality among older adults, although studies of other health outcomes had inconsistent findings regarding whether older adults were at increased risk. Older adults may also withstand greater \( \text{O}_3 \) exposure and not seek relief as quickly as younger adults. Multiple epidemiologic, controlled human exposure, and toxicological studies reported that reduced Vitamins E and C intake are associated with risk of \( \text{O}_3 \)-related health effects. Previous studies have shown that increased exposure to \( \text{O}_3 \) due to outdoor work leads to an increased risk of \( \text{O}_3 \)-related health effects and it is clear that outdoor workers have higher exposures, and possibly greater internal doses, of \( \text{O}_3 \), which may lead to increased risk of \( \text{O}_3 \)-related health effects.
Table 8-6  Summary of evidence for potential increased risk of O₃-related health effects.

<table>
<thead>
<tr>
<th>Evidence Classification</th>
<th>Potential At Risk Factor</th>
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<tbody>
<tr>
<td>Adequate evidence</td>
<td>Genetic factors (Section 8.1)</td>
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<td></td>
<td>Asthma (Section 8.2.2)</td>
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<td></td>
<td>Children (Section 8.3.1.1)</td>
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<td></td>
<td>Older adults (Section 8.3.1.2)</td>
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<td></td>
<td>Diet (Section 8.4.1)</td>
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<td></td>
<td>Outdoor workers (Section 8.4.4)</td>
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<tr>
<td>Suggestive evidence</td>
<td>Sex (Section 8.3.2)</td>
</tr>
<tr>
<td></td>
<td>SES (Section 8.3.3)</td>
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<tr>
<td></td>
<td>Obesity (Section 8.4.2)</td>
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<tr>
<td>Inadequate evidence</td>
<td>Influenza/Infection (Section 8.2.1)</td>
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<td>COPD (Section 8.2.3)</td>
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<td></td>
<td>CVD (Section 8.2.4)</td>
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<td>Diabetes (Section 8.2.5)</td>
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<td></td>
<td>Hyperthyroidism (Section 8.2.6)</td>
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<td></td>
<td>Race/ethnicity (Section 8.3.4)</td>
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<td></td>
<td>Smoking (Section 8.4.3)</td>
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<tr>
<td></td>
<td>Air conditioning use (Section 8.4.5)</td>
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<tr>
<td>Evidence of no effect</td>
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</tbody>
</table>

In some cases, it is difficult to determine a factor that results in potentially increased risk of effects. For example, previous assessments have included controlled human exposure studies in which some healthy individuals demonstrate greater O₃-related health effects compared to other healthy individuals. Intersubject variability has been observed for lung function decrements, symptomatic responses, pulmonary inflammation, AHR, and altered epithelial permeability in healthy adults exposed to O₃ (Que et al., 2011; Holz et al., 2005; McDonnell, 1996). These responses to O₃ exposure in healthy individuals tend to be reproducible within a given individual over a period of several months indicating differences in the intrinsic responsiveness (Holz et al., 2005; Hazucha et al., 2003; Holz et al., 1999; McDonnell et al., 1985c). In addition, there is the possibility of attenuation. In controlled human exposure and toxicological studies, pre-exposure to O₃ was observed to lead to a dampening of some responses following subsequent exposure to O₃ (for more details see Sections 5.4.2.5 and 6.2.1.1).

Limitations include the challenge of evaluating effect measure modification in epidemiologic studies with widespread populations with variation in numerous factors. For a number of the factors described below, there are few available studies. Also, some factors are inconsistent across studies, both in regards to the categorization of the variable and its measurement in the studies. Many toxicological and controlled human exposure studies are the only ones that have examined certain factors and therefore have not been replicated. In considering epidemiologic studies conducted in other countries, it is possible that those populations may differ in SES.
or other demographic indicators, thus limiting generalizability to a U.S. population. Additionally, many epidemiologic studies that stratify by factors of interest have small sample sizes, which can decrease precision of effect estimates and make drawing conclusions difficult.

These challenges and limitations in evaluating the factors that can increase risk for experiencing O₃-related health effects may contribute to conclusions that evidence for some factors, such as sex, SES, and obesity provided “suggestive” evidence of potentially increased risk. In addition, for a number of factors listed in Table 8-6 the evidence was inadequate to draw conclusions about potential increase in risk of effects. Overall, the factors most strongly supported as contributing to increased risk of O₃-related effects among various populations and lifestages were related to genetic factors, asthma, age group (children and older adults), dietary factors, and working outdoors.


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9 ENVIRONMENTAL EFFECTS: OZONE EFFECTS ON VEGETATION AND ECOSYSTEMS

9.1 Introduction

This chapter synthesizes and evaluates the relevant science to help form the scientific foundation for the review of a vegetation- and ecologically-based secondary NAAQS for O₃. The secondary NAAQS are based on welfare effects. The Clean Air Act (CAA) definition of welfare effects includes, but is not limited to, effects on soils, water, wildlife, vegetation, visibility, weather, and climate, as well as effects on materials, economic values, and personal comfort and well-being. The effects of O₃ as a greenhouse gas and its direct effects on climate are discussed in Chapter 10 of this document.

The intent of the ISA, according to the CAA, is to “accurately reflect the latest scientific knowledge expected from the presence of [a] pollutant in ambient air” (42 U.S.C.7408 and 42 U.S.C.7409. This chapter of the ISA includes scientific research from biogeochemistry, soil science, plant physiology, and ecology conducted at multiple levels of biological organization (e.g., molecular, organ, organism, population, community, ecosystem). Key information and judgments formerly found in the AQCDs regarding O₃ effects on vegetation and ecosystems are found in this chapter. This chapter of the O₃ ISA serves to update and revise Chapter 9 and AX9 of the 2006 O₃ AQCD (U.S. EPA, 2006b).

Numerous studies of the effects of O₃ on vegetation and ecosystems were reviewed in the 2006 O₃ AQCD. That document concluded that the effects of ambient O₃ on vegetation and ecosystems appear to be widespread across the U.S., and experimental studies demonstrated plausible mechanisms for these effects. Ozone effect studies published from 2005 to July 2011 are reviewed in this document in the context of the previous O₃ AQCDs. From 2005 to 2011, some areas have had very little new research published and the reader is referred back to sections of the 2006 O₃ AQCD for a more comprehensive discussion of those subjects. This chapter is focused on studies of vegetation and ecosystems that occur in the U.S. and that provide information on endpoints or processes most relevant to the review of the secondary standard. Many studies have been published about vegetation and ecosystems outside of the U.S. and North America, largely in Europe and Asia. This document includes discussion of studies of vegetation and ecosystems outside of North America only if those studies contribute to the general understanding of O₃ effects across species and ecosystems. For example, studies outside North America are discussed that consider physiological and biochemical processes that contribute to the understanding of effects of O₃ across species. Also, ecosystem studies outside of North America that contribute to the understanding of O₃ effects on general ecosystem processes are discussed in the chapter.
Sections of this chapter first discuss exposure methods, followed by effects on vegetation and ecosystems at various levels of biological organization and ends with policy-relevant discussions of exposure indices and exposure-response. Figure 9-1 is a simplified illustrative diagram of the major endpoints O₃ may affect. First, Section 9.2 presents a brief overview of various methodologies that have been, and continue to be, central to quantifying O₃ effects on vegetation (see AX9.1 of the 2006 O₃ AQCD for more detailed discussion) (U.S. EPA, 2006b). Section 9.3 through Section 9.4 begin with a discussion of effects at the cellular and subcellular level followed by consideration of the O₃ effects on plant and ecosystem processes (Figure 9-1). In Section 9.3 research is reviewed from the molecular to the biochemical and physiological levels in impacted plants, offering insight into the mode of action of O₃. Section 9.4 provides a review of the effects of O₃ exposure on major endpoints at the whole plant scale including growth, reproduction, visible foliar injury and leaf gas exchange in woody and herbaceous plants in the U.S., as well as a brief discussion of O₃ effects on agricultural crop yield and quality. Section 9.4 also integrates the effects of O₃ on individual plants in a discussion of available research for assessing the effect of O₃ on ecosystems, along with available studies that could inform assessments of various ecosystem services (see Section 9.4.1.2). The development of indices of O₃ exposure and dose modeling is discussed in Section 9.5. Finally, exposure-response relationships for a number of tree species, native vegetation, and crop species and cultivars are reviewed, tabulated, and compared in Section 9.6 to form the basis for an assessment of the potential risk to vegetation from current ambient levels of O₃.
9.2 Experimental Exposure Methodologies

9.2.1 Introduction

A variety of methods for studying plant response to O₃ exposures have been developed over the last several decades. The majority of methodologies currently used have been discussed in detail in the 1996 O₃ AQCD (U.S. EPA, 1996a) and 2006 O₃ AQCD (U.S. EPA, 2006b). This section will serve as a short overview of the methodologies and the reader is referred to the previous O₃ AQCDs for more in-depth discussion.
9.2.2 “Indoor,” Controlled Environment, and Greenhouse Chambers

The earliest experimental investigations of the effects of O₃ on plants utilized simple glass or plastic-covered chambers, often located within greenhouses, into which a flow of O₃-enriched air or oxygen could be passed to provide the exposure. The types, shapes, styles, materials of construction, and locations of these chambers have been numerous. Hogsett et al. (1987a) have summarized the construction and performance of more elaborate and better instrumented chambers since the 1960s, including those installed in greenhouses (with or without some control of temperature and light intensity).

One greenhouse chamber approach that continues to yield useful information on the relationships of O₃ uptake to both physiological and growth effects employs continuous stirred tank reactors (CSTRs) first described by Heck et al. (1978). Although originally developed to permit mass-balance studies of O₃ flux to plants, their use has more recently widened to include short-term physiological and growth studies of O₃ × CO₂ interactions (Loats and Rebbeck, 1999; Reinert et al., 1997; Rao et al., 1995; Reinert and Ho, 1995; Heagle et al., 1994a), and validation of visible foliar injury on a variety of plant species (Kline et al., 2009; Orendovici et al., 2003). In many cases, supplementary lighting and temperature control of the surrounding structure have been used to control or modify the environmental conditions (Heagle et al., 1994a).

Many investigations have utilized commercially available controlled environment chambers and walk-in rooms adapted to permit the introduction of a flow of O₃ into the controlled air-volume. Such chambers continue to find use in genetic screening and in physiological and biochemical studies aimed primarily at improving the understanding of modes of action. For example, some of the studies of the O₃ responses of common plantain (Plantago major) populations have been conducted in controlled environment chambers (Whitfield et al., 1996; Reiling and Davison, 1994).

More recently, some researchers have been interested in attempting to investigate direct O₃ effects on reproductive processes, separate from the effects on vegetative processes (Black et al., 2010). For this purpose, controlled exposure systems have been employed to expose the reproductive structures of annual plants to gaseous pollutants independently of the vegetative component (Black et al., 2010; Stewart et al., 1996).

9.2.3 Field Chambers

In general, field chamber studies are dominated by the use of various versions of the open top chamber (OTC) design, first described by Heagle et al. (1973) and Mandl et al. (1973). The OTC method continues to be a widely used technique in the U.S. and Europe for exposing plants to varying levels of O₃. Most of the new information confirms earlier conclusions and provides additional support for OTC use in
assessing plant species and in developing exposure-response relationships. Chambers are generally ~3 meters in diameter with 2.5 meter-high walls. Hogsett et al. (1987b) described in detail many of the various modifications to the original OTC designs that appeared subsequently, e.g., the use of larger chambers for exposing small trees (Kats et al., 1985) or grapevines (Mandl et al., 1989), the addition of a conical baffle at the top to improve ventilation (Kats et al., 1976), a frustum at the top to reduce ambient air incursions, and a plastic rain-cap to exclude precipitation (Hogsett et al., 1985). All versions of OTCs included the discharge of air via ports in annular ducting or interiorly perforated double-layered walls at the base of the chambers to provide turbulent mixing and the upward mass flow of air.

Chambered systems, including OTCs, have several advantages. For instance, they can provide a range of treatment levels including charcoal-filtered (CF), clean-air control, and several above ambient concentrations for O3 experiments. Depending on experimental intent, a replicated, clean-air control treatment is an essential component in many experimental designs. The OTC can provide a consistent, definable exposure because of the constant wind speed and delivery systems. Statistically robust concentration-response (C-R) functions can be developed using such systems for evaluating the implications of various alternative air quality scenarios on vegetation response. Nonetheless, there are several characteristics of the OTC design and operation that can lead to exposures that might differ from those experienced by plants in the field. First, the OTC plants are subjected to constant air flow turbulence, which, by lowering the boundary layer resistance to diffusion, may result in increased uptake. This may lead to an overestimation of effects relative to areas with less turbulence (Krupa et al., 1995; Legge et al., 1995). However, other research has found that OTC’s may slightly change vapor pressure deficit (VPD) in a way that may decrease the uptake of O3 into leaves (Piikki et al., 2008a). As with all methods that expose vegetation to modified O3 concentrations in chambers, OTCs create internal environments that differ from ambient air. This so-called “chamber effect” refers to the modification of microclimatic variables, including reduced and uneven light intensity, uneven rainfall, constant wind speed, reduced dew formation, and increased air temperatures (Fuhrer, 1994; Manning and Krupa, 1992). However, in at least one case where canopy resistance was quantified in OTCs and in the field, it was determined that gaseous pollutant exposure to crops in OTCs was similar to that which would have occurred at the same concentration in the field (Unsworth et al., 1984a, b). Because of the standardized methodology and protocols used in National Crop Loss Assessment Network (NCLAN) and other programs, the database can be assumed to be internally consistent.

While it is clear that OTCs can alter some aspects of the microenvironment and plant growth, it is important to establish whether or not these differences affect the relative response of a plant to O3. As noted in the 1996 O3 AQCD (U.S. EPA, 1996a), evidence from a number of comparative studies of OTCs and other exposure systems suggested that responses were essentially the same regardless of exposure system used and chamber effects did not significantly affect response. In studies that included exposure to ambient concentrations of O3 in both OTCs, and open-air, chamberless control plots, responses in the OTCs were the same as in open-air plots.
Examples include studies of tolerant and sensitive white clover clones (*Trifolium repens*) to ambient O\textsubscript{3} in greenhouse, open top, and ambient plots ([Heagle et al., 1996](#)), Black Cherry (*Prunus serotina*) ([Neufeld et al., 1995](#)), and three species of conifers ([Neufeld et al., 2000](#)). Experimental comparisons between exposure methodologies are reviewed in Section 9.2.6.

Another type of field chamber called a “terracosm” has been developed and used in recent studies ([Lee et al., 2009a](#)). Concern over the need to establish realistic plant-litter-soil relationships as a prerequisite to studies of the effects of O\textsubscript{3} and CO\textsubscript{2} enrichment on ponderosa pine (*Pinus ponderosa*) seedlings led Tingey et al. (1996) to develop closed, partially environmentally controlled, sun-lit chambers (“terracosms”) incorporating lysimeters (1 meter deep) containing forest soil in which the appropriate horizon structure was retained.

Other researchers have recently published studies using another type of out-door chamber called recirculating Outdoor Plant Environment Chambers (OPECs) ([Flowers et al., 2007](#)). These closed chambers are approximately 2.44 meters × 1.52 meters with a growth volume of approximately 3.7 m\textsuperscript{3} in each chamber. These chambers admit 90% of full sunlight and control temperature, humidity and vapor pressure ([Fiscus et al., 1999](#)).

### 9.2.4 Plume and FACE-Type Systems

Plume systems are chamberless exposure facilities in which the atmosphere surrounding plants in the field is modified by the injection of pollutant gas into the air above or around them from multiple orifices spaced to permit diffusion and turbulence, so as to establish relatively homogeneous conditions as the individual plumes disperse and mix with the ambient air. They can only be used to increase the O\textsubscript{3} levels in the ambient air.

The most common plume system used in the U.S. is a modification of the free-air carbon dioxide/ozone enrichment (FACE) system ([Hendrey et al., 1999; Hendrey and Kimball, 1994](#)). Although originally designed to provide chamberless field facilities for studying the CO\textsubscript{2} effects of climate change, FACE systems have been adapted to include the dispensing of O\textsubscript{3} ([Karnosky et al., 1999](#)). This method has been employed in Illinois (SoyFACE) to study soybeans ([Morgan et al., 2004; Rogers et al., 2004](#)) and in Wisconsin (Aspen FACE) to study trembling aspen (*Populus tremuloides*), birch (*Betula papyrifera*) and maple (*Acer saccharum*) ([Karnosky et al., 1999](#)). Volk et al. (2003) described a similar system for exposing grasslands that uses 7-m diameter plots. Another similar FACE system has been used in Finland ([Saviranta et al., 2010; Oksanen, 2003](#)).

The FACE systems in the U.S. discharge the pollutant gas (O\textsubscript{3} and/or CO\textsubscript{2}) through orifices spaced along an annular ring (or torus) or at different heights on a ring of vertical pipes. Computer-controlled feedback from the monitoring of gas concentration regulates the feed rate of enriched air to the dispersion pipes. Feedback
of wind speed and directional information ensures that the discharges only occur upwind of the treatment plots, and that discharge is restricted or closed down during periods of low wind speed or calm conditions. The diameter of the arrays and their height (25-30 meters) in some FACE systems requires large throughputs of enriched air per plot, particularly in forest tree systems. The cost of the throughputs tends to limit the number of enrichment treatments, although Hendrey et al. (1999) argued that the cost on an enriched volume basis is comparable to that of chamber systems.

A different FACE-type facility has been developed for the Kranzberg Ozone Fumigation Experiment (KROFEX) in Germany beginning in 2000 (Nunn et al., 2002; Werner and Fabian, 2002). The experiment aims to study the effects of O₃ on mature stands of beech (Fagus sylvatica) and spruce (Picea abies) trees in a system that functions independently of wind direction. The enrichment of a large volume of the ambient air immediately above the canopy takes place via orifices in vertical tubes suspended from a horizontal grid supported above the canopy.

Although plume systems make virtually none of the modifications to the physical environment that are inevitable with chambers, their successful use depends on selecting the appropriate numbers, sizes, and orientations of the discharge orifices to avoid “hot-spots” resulting from the direct impingement of jets of pollutant-enriched air on plant foliage (Werner and Fabian, 2002). Because mixing is unassisted and completely dependent on wind turbulence and diffusion, local gradients are inevitable especially in large-scale systems. FACE systems have provisions for shutting down under low wind speed or calm conditions and for an experimental area that is usually defined within a generous border in order to strive for homogeneity of the exposure concentrations within the treatment area. They are also dependent upon continuous computer-controlled feedback of the O₃ concentrations in the mixed treated air and of the meteorological conditions. Plume and FACE systems also are unable to reduce O₃ levels below ambient in areas where O₃ concentrations are phytotoxic.

### 9.2.5 Ambient Gradients

Ambient O₃ gradients that occur in the U.S. hold potential for the examination of plant responses over multiple levels of exposure. However, few such gradients can be found that meet the rigorous statistical requirements for comparable site characteristics such as soil type, temperature, rainfall, radiation, and aspect (Manning and Krupa, 1992); although with small plants, soil variability can be avoided by the use of plants in large pots. The use of soil monoliths transported to various locations along natural O₃ gradients is another possible approach to overcome differences in soils; however, this approach is also limited to small plants.

Studies in the 1970s used the natural gradients occurring in southern California to assess yield losses of alfalfa and tomato (Oshima et al., 1977; Oshima et al., 1976). A transect study of the impact of O₃ on the growth of white clover and barley in the U.K. was confounded by differences in the concurrent gradients of SO₂ and NO₂.

Several studies have used the San Bernardino Mountains Gradient Study in southern California to study the effects of O$_3$ and N deposition on forests dominated by ponderosa and Jeffrey pine (Jones and Paine, 2006; Arbaugh et al., 2003; Gruik, 1999; U.S. EPA, 1977). However, it is difficult to separate the effects of N and O$_3$ in some instances in these studies (Arbaugh et al., 2003). An O$_3$ gradient in Wisconsin has been used to study foliar injury in a series of trembling aspen clones (Populus tremuloides) differing in O$_3$ sensitivity (Maňkovská et al., 2005; Karnosky et al., 1999). Also in the Midwest, an east-west O$_3$ gradient around southern Lake Michigan was used to look at growth and visible foliar injury in (P. serotina) and common milkweed (Asclepias syriaca) (Bennett et al., 2006).

More recently, studies have been published that have used natural gradients to study a variety of endpoints and species. For example, Gregg et al. (2003) studied cottonwood (Populus deltoides) saplings grown in an urban to rural gradient of O$_3$ by using seven locations in the New York City area. The secondary nature of the reactions of O$_3$ formation and NO$_X$ titration reactions within the city center resulted in significantly higher cumulative O$_3$ exposures in more rural sites. Potential modifying factors such as soil composition, moisture, or temperature were either controlled or accounted for in analysis. As shown in Section 9.6.3.3, the response of this species to O$_3$ exposure was much stronger than most species. The natural gradient exposures were reproduced in parallel using OTCs, and yielded similar results. Also, the U.S. Forest Service - Forest Inventory and Analysis (FIA) program uses large-scale O$_3$ exposure patterns across the continental U.S. to study occurrences of foliar injury due to O$_3$ exposure (Smith et al., 2003) (Section 9.4.2). Finally, McLaughlin et al. (2007a; 2007b) used spatial and temporal O$_3$ gradients to study forest growth and water use in the southern Appalachians. These studies found varying O$_3$ exposures between years and between sites.

### 9.2.6 Comparative Studies

All experimental approaches used to expose plants to O$_3$ have strengths and weaknesses. One potential weakness of laboratory, greenhouse, or field chamber studies is the potential effect of the chamber on micrometeorology. In contrast, plume, FACE and gradient systems are limited by the very small number of possible exposure levels (almost always no more than two), small replication and the inability to reduce O$_3$ levels below ambient. In general, experiments that aim at characterizing the effect of a single variable, e.g., exposure to O$_3$, must not only manipulate the levels of that variable, but also control potentially interacting variables and confounders, or else account for them. However, while increasing control of environmental variables makes it easier to discern the effect of the variable of interest, it must be balanced with the ability to extend conclusions to natural,
non-experimental settings. More naturalistic exposure systems, on the other hand, let interacting factors vary freely, resulting in greater unexplainable variability. The various exposure methodologies used with O₃ vary in the balance each strikes between control of environmental inputs, closeness to the natural environment, noisiness of the response data, and ability to make general inferences. Studies have examined the comparability of results obtained though the various exposure methodologies. As noted in the 1996 O₃ AQCD, evidence from the comparative studies of OTCs and from closed chamber and O₃-exclusion exposure systems on the growth of alfalfa (*Medicago sativa*) by Olszyk et al. (1986) suggested that, since significant differences were found for fewer than 10% of the growth parameters measured, the responses were, in general, essentially the same regardless of exposure system used, and chamber effects did not significantly affect response. Heagle et al. (1988) concluded: “Although chamber effects on yield are common, there are no results showing that this will result in a changed yield response to O₃.” A study of the effects of an enclosure examined the responses of tolerant and sensitive white clover clones (*Trifolium repens*) to ambient O₃ in a greenhouse, open-top chamber, and ambient (no chamber) plots (Heagle et al., 1996). For individual harvests, greenhouse O₃ exposure reduced the forage weight of the sensitive clone 7 to 23% more than in OTCs. However, the response in OTCs was the same as in ambient plots. Several studies have shown very similar response of yield to O₃ for plants grown in pots or in the ground, suggesting that even such a significant change in environment does not alter the proportional response to O₃, providing that the plants are well watered (Heagle et al., 1983; Heagle, 1979).

A few recent studies have compared results of O₃ experiments between OTCs, FACE experiments, and gradient studies. For example, a series of studies undertaken at Aspen FACE (Isebrands et al., 2001; Isebrands et al., 2000) showed that O₃ symptom expression was generally similar in OTCs, FACE, and ambient O₃ gradient sites, and supported the previously observed variation among trembling aspen clones using OTCs (Maňkovská et al., 2005; Karnosky et al., 1999). In the SoyFACE experiment in Illinois, soybean (Pioneer 93B15 cultivar) yield loss data from a two-year study was published (Morgan et al., 2006). This cultivar is a recent selection and, like most modern cultivars, has been selected under an already high current O₃ exposure. It was found to have average sensitivity to O₃ compared to 22 other cultivars tested at SoyFACE. In this experiment, ambient hourly O₃ concentrations were increased by approximately 20% and measured yields were decreased by 15% in 2002 as a result of the increased O₃ exposure (Morgan et al., 2006). To compare these results to chamber studies, Morgan et al. (2006) calculated the expected yield loss from a linear relationship constructed from chamber data using seven-hour seasonal averages (Ashmore, 2002). They calculated an 8% expected yield loss from the 2002 O₃ exposure using that linear relationship. As reported in Section 9.2.5, Gregg et al. (2006, 2003) found similar O₃ effects on cottonwood sapling biomass growth along an ambient O₃ gradient in the New York City area and a parallel OTC study.

Finally, EPA conducted comparisons of exposure-response model predictions based on OTC studies, and more recent FACE observations. These comparisons include
yield of annual crops, and biomass growth of trees. They are presented in Section 9.6.3 of this document.

9.3 Mechanisms Governing Vegetation Response to Ozone

9.3.1 Introduction

This section focuses on the effects of O₃ stress on plants and their responses to that stress on the molecular, biochemical and physiological levels. First, the pathway of O₃ uptake into the leaf and the initial chemical reactions occurring in the substomatal cavity and apoplast will be described (Section 9.3.2); additionally, direct effects of O₃ on the stomatal apparatus will be discussed. Once O₃ has entered the substomatal cavity and apoplast, the cell initiates rapid changes in signaling pathways and gene expression that have been measured in O₃-treated plants. The next section focuses on changes in gene and protein expression measured in plants exposed to O₃, with particular emphasis on results from transcriptome (all RNA molecules produced in a cell) and proteome (all proteins produced in a cell) analyses (Section 9.3.3.2). Subsequently, the role of phytohormones such as salicylic acid (SA), ethylene (ET), jasmonic acid (JA), and abscisic acid (ABA) and their interactions in both signal transduction processes and in determining plant response to O₃ is discussed in Section 9.3.3.3. After O₃ uptake, some plants can respond to the oxidative stress with detoxification to minimize damage. These mechanisms of detoxification, with particular emphasis on antioxidant enzymes and metabolites, are reviewed in Section 9.3.4. The next section focuses on changes in primary and secondary metabolism in plants exposed to O₃, looking at photosynthesis, respiration and several secondary metabolites, some of which may also act as antioxidants and protect the plant from oxidative stress (Section 9.3.5). For many of these topics, information from the 2006 O₃ AQCD (U.S. EPA, 2006b) has been summarized, as this information is still valid and supported by more recent findings. For other topics, such as genomics and proteomics, which have arisen due to the availability of new technologies, the information is based solely on new publications with no reference to the 2006 O₃ AQCD.

As Section 9.3 focuses on mechanisms underlying effects of O₃ on plants and their response to it, the conditions that are used to study these mechanisms do not always reflect conditions that a plant may be exposed to in an agricultural setting or natural ecosystem. The goal of many of these studies is to generate an O₃ effect in a relatively short period of time and not always to simulate ambient O₃ exposures. Therefore, plants are often exposed to unrealistically high O₃ concentrations for several hours or days (acute exposure), and only in some cases to ambient or slightly elevated O₃ concentrations for longer time periods (chronic exposure). Additionally, the plant species utilized in these studies are often not agriculturally important or commonly found as part of natural ecosystems. Model organisms such as Arabidopsis thaliana are used frequently as they are easy to work with, and mutants
or transgenic plants are easy to develop or have already been developed. Furthermore, the Arabidopsis genome has been sequenced, and much is known about the molecular basis of many biochemical and cellular processes.

Many of the studies described in this section focus on changes in the expression of genes in O₃-treated plants. Some very recent studies utilizing proteomics techniques have evaluated changes in protein expression for large numbers of proteins in O₃ treated plants, and the findings from these studies support the previous results regarding changes in gene expression studies as a result of O₃ exposure. The next step in the process is to determine the implications of the measured changes occurring at the cellular level to whole plants and ecosystems, which is an important topic of study which has not been widely addressed.

The most noteworthy new body of research since the 2006 O₃ AQCD is on the understanding of molecular mechanisms underlying how plants are affected by O₃; many of the recent studies reviewed here focus on changes in gene expression in plants exposed to elevated O₃. The findings summarized in the 2006 O₃ AQCD included decreases in transcript levels of photosynthesis associated genes, and increases in transcript levels of genes encoding for pathogenesis-related proteins, enzymes needed for ethylene synthesis, antioxidant enzymes and defense genes such as phenylalanine ammonia lyase in plants exposed to O₃. These findings have been supported by the new studies, and the advent of new technologies has allowed for a more comprehensive understanding of the mechanisms governing how plants are affected by O₃.

In summary, these new studies have increased knowledge of the molecular, biochemical and cellular mechanisms occurring in plants in response to O₃ by often using artificial exposure conditions and model organisms. This information adds to the understanding of the basic biology of how plants are affected by oxidative stress in the absence of any other potential stressors. The results of these studies provide important insights, even though they may not always directly translate into effects observed in other plants under more realistic exposure conditions.

### 9.3.2 Ozone Uptake into the Leaf

Appendix AX9.2.3 of the 2006 O₃ AQCD clearly described the process by which O₃ enters plant leaves through open stomata (U.S. EPA, 2006b). This information continues to be valid and is only summarized here.

Ozone moves from the atmosphere above the canopy boundary layer into the canopy primarily by turbulent air flow. Canopy conductance is controlled by the complexity of the canopy architecture. Within the canopy, O₃ is adsorbed onto surfaces as well as being absorbed into the leaves. Absorption into leaves is controlled by leaf boundary layer and stomatal conductance, which together determine leaf conductance (Figure 9-2, Panel A). Other factors that may also limit uptake include the size of the stomatal aperture and the reactions of O₃ with biogenically-emitted
hydrocarbons (U.S. EPA, 2006b; Kurpius and Goldstein, 2003). Stomata provide the principal pathway for O₃ to enter and affect plants (Massman and Grantz, 1995; Fuentes et al., 1992; Reich, 1987; Leuning et al., 1979). Ozone moves into the leaf interior by diffusing through open stomata, and environmental conditions which promote high rates of gas exchange will favor the uptake of the pollutant by the leaf (Figure 9-2, Panel B). Once inside the substomatal cavity, O₃ is thought to rapidly react with the aqueous apoplast to form breakdown products known as reactive oxygen species (ROS), such as hydrogen peroxide (H₂O₂), superoxide (O₂⁻), hydroxyl radicals (HO⁻) and peroxy radicals (HO₂⁻) (Figure 9-3). Hydrogen peroxide is not only a toxic breakdown product of O₃, but has been shown to function as a signaling molecule, which is activated in response to both biotic and abiotic stressors. The role of H₂O₂ in signaling was described in detail in the 2006 O₃ AQCD. Additional organic molecules present in the apoplast or cell wall, such as those containing double bonds or sulfhydryls that are sensitive to oxidation, could also be converted to oxygenated molecules after interacting with O₃ (Figure 9-4). These reactions are not only pH dependent, but are also influenced by the presence of other molecules in the apoplast (U.S. EPA, 2006b). The 2006 O₃ AQCD provided a comprehensive summary of these possible interactions of O₃ with other biomolecules (U.S. EPA, 2006b). It is in the apoplast that initial detoxification reactions by antioxidant metabolites and enzymes take place, and these initial reactions are critical to reduce concentrations of the oxidative breakdown products of O₃; these reactions are described in more detail in Section 9.3.4 of this document.

9.3.2.1 Changes in Stomatal Function

Ozone-induced changes in stomatal conductance have been reviewed in detail in previous O₃ AQCDs. The findings summarized in these documents demonstrate that stomatal conductance is often reduced in plants exposed to O₃, resulting either from a direct impact of O₃ on the stomatal complex which causes closure, or as a response to increasing CO₂ concentrations in the substomatal cavity as carbon fixation is reduced. Although the nature of these effects depends upon many different factors, including the plant species, concentration and duration of the O₃ exposure, and prevailing meteorological conditions, stomatal conductance is often negatively affected by plant exposure to O₃ (Wittig et al., 2007). Decreases in conductance have been shown to result from direct as well as indirect effects on stomata (Wittig et al., 2007). However, some recent studies have reported increased conductance in response to O₃ exposure, suggesting partial stomatal dysfunction (Paoletti and Grulke, 2010).

Results from the use of Arabidopsis mutants and new technologies, which allow for analysis of guard cell function in whole plants rather than in isolated guard cells or epidermal peels, suggest that O₃ may also have a direct impact on stomatal guard cells, leading to alterations in stomatal conductance. The use of a new simultaneous O₃ exposure/gas exchange device has demonstrated that exposure of Arabidopsis ecotypes Col-0 and Ler to 150 ppb O₃ resulted in a 60-70% decline in stomatal

9-12
conductance within 9-12 minutes of beginning the exposure. Twenty to thirty minutes later, stomatal conductance had returned to its initial value, even with continuing exposure to O₃, indicating a rapid direct effect of O₃ on stomatal function (Kollist et al., 2007). This transient decrease in stomatal conductance was not observed in the abscisic acid insensitive (ABI2) Arabidopsis mutant. As the ABI2 protein is thought to regulate the signal transduction process involved in stomatal response downstream of ROS production, the authors suggest that the transient decrease in stomatal conductance in the Col-0 and Ler ecotypes results from the biological action of ROS in transducing signals, rather than direct physical damage to guard cells by ROS (Kollist et al., 2007). This rapid transient decrease in stomatal conductance was also not observed when exposing the Arabidopsis mutant slac1 (slow anion channel-associated 1) to 200 ppb O₃ (Vahisalu et al., 2008). The SLAC1 protein was shown to be essential for guard cell slow anion channel functioning and for stomatal closure in response to O₃. Based on additional studies using a variety of Arabidopsis mutants impaired in various aspects of stomatal function, Vahisalu et al. (2008) suggest that the presence of ROS in the guard cell apoplast (formed either by O₃ breakdown or through ROS production from NADPH oxidase activity) leads to the activation of a signaling pathway in the guard cells, which includes SLAC1, and results in stomatal closure.

A review by McAinsh et al. (2002) discusses the role of calcium as a part of the signal transduction pathway involved in regulating stomatal responses to pollutant stress. A number of studies in this review provide some evidence that exposure to O₃ increases the cytosolic free calcium concentration ([Ca²⁺]cyt) in guard cells, which may result in an inhibition of the plasma membrane inward-rectifying K⁺ channels in guard cells, which allow for the K⁺ uptake needed for stomatal opening (McAinsh et al., 2002; Torsethaugen et al., 1999). This would compromise the ability of the stomata to respond to various stimuli, including light, CO₂ concentration and drought. Pei et al. (2000) reported that the presence of H₂O₂ activated Ca²⁺-permeable channels, which mediate increases in [Ca²⁺]cyt in guard cell plasma membranes of Arabidopsis. They also determined that abscisic acid (ABA) induced H₂O₂ production in guard cells, leading to ABA-induced stomatal closure via activation of the membrane Ca²⁺ channels. Therefore, it is possible that H₂O₂, a byproduct of O₃ breakdown in the apoplast, could disrupt the Ca²⁺-ABA signaling pathway that is involved in regulating stomatal responses (McAinsh et al., 2002). The studies described here provide some evidence to suggest that O₃ and its breakdown products can directly affect stomatal functioning by impacting the signal transduction pathways which regulate guard cells. Stomatal sluggishness has been described as a delay in stomatal response to changing environmental conditions in sensitive species exposed to higher concentrations and/or longer-term O₃ exposures (Paoletti and Grulke, 2010, 2005; McAinsh et al., 2002). It is possible that the signaling pathways described above could be involved in mediating this stomatal sluggishness in some plant species under certain O₃ exposure conditions (Paoletti and Grulke, 2005; McAinsh et al., 2002).
Note: While details among species vary, the general overview remains the same. Light that drives photosynthesis generally falls upon the upper (adaxial) leaf surface. Carbon dioxide (CO₂) and O₃ enter through the stomata, while water vapor exits through the stomata (transpiration). Stomata are usually on the lower (abaxial) leaf surface, but may occur on the upper leaf surface in some species.

**Figure 9-2** Ozone uptake from the atmosphere (A), and The anatomy of a dicot leaf (B).
Note: (a) Ozone reacts at the double bonds to form carbonyl groups. (b) Under certain circumstances, peroxides are generated.

**Figure 9-3** Possible reactions of O$_3$ within water.

Note: (a) The typical Crigee mechanism is shown in which several reaction paths from the initial product are shown. (b) Typical reaction of ascorbic acid with O$_3$.
Source: Adapted from Mudd et al. (1996).

**Figure 9-4** The Crigee mechanism of O$_3$ attack of a double bond.
9.3.3 Cellular to Systemic Responses

9.3.3.1 Ozone Signal Transduction

New technologies allowing for large-scale analysis of oxidative stress-induced changes in gene expression have facilitated the study of signal transduction processes associated with plant response to \( \text{O}_3 \) exposure. Many of these studies have been conducted using Arabidopsis or tobacco plants, for which a variety of mutants are available and/or which can be easily genetically modified to generate either loss-of-function or over-expressing genotypes. Several comprehensive review articles provide an overview of what is known of \( \text{O}_3 \)-induced signal transduction processes and how they may help to explain differential sensitivity of plants to the pollutant (Ludwikow and Sadowski, 2008; Baier et al., 2005; Kangasjarvi et al., 2005). Additionally, analysis of several studies of transcriptome changes has also allowed for the compilation of these data to determine an initial time-course for \( \text{O}_3 \)-induced activation of various signaling compounds (Kangasjarvi et al., 2005).

Some of the earliest events that occur in plants exposed to \( \text{O}_3 \) have been described in the guard cells of stomata. Reactive oxygen species were observed in the chloroplasts of guard cells in the \( \text{O}_3 \) tolerant Col-0 Arabidopsis thaliana ecotype plants within 5 minutes of plant exposure to 350 ppb \( \text{O}_3 \) (Joo et al., 2005). Reactive oxygen species from the breakdown of \( \text{O}_3 \) in the apoplast are believed to activate GTPases (G-proteins), which, in turn, activate several intracellular sources of ROS, including ROS derived from the chloroplasts. G-proteins are also believed to play a role in activating membrane-bound NADPH oxidases to produce ROS and, as a result, propagate the oxidative burst to neighboring cells (Joo et al., 2005). Therefore, G-proteins are recognized as important molecules involved in plant responses to \( \text{O}_3 \) and may play a role in the initiation of signal transduction mechanisms resulting from the presence of ROS in the apoplast (Kangasjarvi et al., 2005; Booker et al., 2004a).

A change in the redox state of the plant may contribute to initiating the process of signaling of oxidative stress in plants. Disulfide-thiol conversions in proteins and the redox state of the glutathione pool may be important components of redox and signal transduction (Foyer and Noctor, 2005a, b).

Calcium (\( \text{Ca}^{2+} \)) has also been implicated in the transduction of signals to the nucleus in response to oxidative stress. The influx of \( \text{Ca}^{2+} \) from the apoplast into the cell occurs early during plant exposure to \( \text{O}_3 \), and it is thought to play a role in regulating the activity of protein kinases, which are discussed below (Baier et al., 2005; Hamel et al., 2005). Calcium channel blockers inhibited \( \text{O}_3 \)-induced activation of protein kinases in tobacco suspension cells exposed to 500 ppb \( \text{O}_3 \) for 10 minutes, indicating that the opening of \( \text{Ca}^{2+} \) channels is an important upstream signaling event or that the (as yet unknown) upstream process has a requirement for \( \text{Ca}^{2+} \) (Samuel et al., 2000).

Further transmission of information regarding the presence of ROS to the nucleus involves mitogen-activated protein kinases (MAPKs), which phosphorylate proteins.
and activate various cellular responses (Hamel et al., 2005). Mitogen-activated protein kinases are induced in several different plant species in response to O₃ exposure, including tobacco (Samuel et al., 2005), Arabidopsis (Ludwikow et al., 2004), the shrub Phillyrea latifolia (Paolacci et al., 2007) and poplar (Hamel et al., 2005). Disruption of these signal transduction pathways by over-expressing or suppressing MAPK activity in different Arabidopsis and tobacco lines resulted in increased plant sensitivity to O₃ (Miles et al., 2005; Samuel and Ellis, 2002). Additionally, greater O₃ tolerance of several Arabidopsis ecotypes was correlated with greater upregulation of MAPK signaling pathways upon O₃ exposure than in more sensitive Arabidopsis ecotypes (Li et al., 2006b; Mahalingam et al., 2006; Overmyer et al., 2005), indicating that determination of plant sensitivity and plant response to O₃ may, in part, be determined not only by whether these pathways are turned on, but also by the magnitude of the signals moving through these communication channels.

In conclusion, experimental evidence suggests that there are likely several different mechanisms by which signaling as part of plant response to O₃ or its breakdown products is initiated. These mechanisms may vary by species or developmental stage of the plant, or may co-exist and be activated by different exposure conditions. Calcium and protein kinases are likely involved in relaying information to the nucleus and other cellular compartments as a first step in determining whether and how the plant will respond to the stress.

9.3.3.2 Gene and Protein Expression Changes in Response to Ozone

The advent of DNA microarray technology has allowed for the study of gene expression in cells on a large scale. Rather than assessing changes in gene expression of individual genes, DNA microarrays facilitate the evaluation of entire transcriptomes, providing a comprehensive picture of simultaneous alterations in gene expression. In addition, these studies have provided more insight into the complex interactions between molecules, how those interactions lead to the communication of information in the cell (or between neighboring cells), and which role these interactions play in determining tolerance or sensitivity and how a plant may respond to stresses such as O₃ (Ludwikow and Sadowski, 2008). Transcriptome analysis of O₃-treated plants has been performed in several species, including Arabidopsis thaliana (Li et al., 2006b; Tosti et al., 2006; Heidenreich et al., 2005; Mahalingam et al., 2005; Tamaoki et al., 2003), pepper (Capsicum annuum) (Lee and Yun, 2006), clover (Medicago truncatula) (Puckette et al., 2008), Phillyrea latifolia (Paolacci et al., 2007), poplar (Street et al., 2011), and European beech (Fagus sylvatica) (Olbrich et al., 2010; Olbrich et al., 2009; Olbrich et al., 2005). In some cases, researchers compared transcriptomes of two or more cultivars, ecotypes or mutants that differed in their sensitivity to O₃ (Puckette et al., 2008; Rizzo et al., 2007; Lee and Yun, 2006; Li et al., 2006b; Tamaoki et al., 2003). Species, O₃ exposure conditions (concentration, duration of exposure) and sampling times varied
considerably in these studies. However, functional classification of the genes that were either upregulated or downregulated by plant exposure to O₃ exhibited common trends. Genes involved in plant defense, signaling and those associated with the synthesis of plant hormones and secondary metabolism were generally upregulated, while those related to photosynthesis and general metabolism were typically downregulated in O₃-treated plants (Puckette et al., 2008; Lee and Yun, 2006; Li et al., 2006b; Tosti et al., 2006; Obriech et al., 2005; Tamaoki et al., 2003).

Analysis of the transcriptome has been used to evaluate differences in gene expression between sensitive and tolerant plants in response to O₃ exposure. In pepper, 67% of the 180 genes studied that were affected by O₃ were differentially regulated in the sensitive and tolerant cultivars. At both 0 hours and 48 hours after a 3-day exposure at 150 ppb, O₃ responsive genes were either upregulated or downregulated more markedly in the sensitive than in the tolerant cultivar (Lee and Yun, 2006). Transcriptome analysis also revealed differences in timing and magnitude of changes in gene expression between sensitive and tolerant clovers. Acute exposure (300 ppb O₃ for 6 hours) led to the production of an oxidative burst in both clovers (Puckette et al., 2008). However, the sensitive-Jemalong cultivar exhibited a sustained ROS burst and a concomitant downregulation of defense response genes at 12 hours after the onset of exposure, while the tolerant JE 154 accession showed much more rapid and large-scale transcriptome changes than the Jemalong cultivar (Puckette et al., 2008).

Arabidopsis ecotypes WS and Col-0 were exposed to 1.2 × ambient O₃ concentrations for 8-12 days at the SoyFACE site (Li et al., 2006b). The sensitive WS ecotype showed a far greater number of changes in gene expression in response to this low-level O₃ exposure than the tolerant Col-0 ecotype. In a different study, exposure of the WS ecotype to 300 ppb O₃ for 6 hours showed a rapid induction of genes leading to cell death, such as proteases, and downregulation or inactivation of cell signaling genes, demonstrating an ineffective defense response in this O₃ sensitive ecotype (Mahalingam et al., 2006).

The temporal response of plants to O₃ exposure was evaluated in the Arabidopsis Col-0 ecotype during a 6-hour exposure at 350 ppb O₃ and for 6 hours after the exposure was completed. Results of this study, shown in Figure 9-5 indicate that genes associated with signal transduction and regulation of transcription were in the class of early upregulated genes, while genes associated with redox homeostasis and defense/stress response were in the class of late upregulated genes (Mahalingam et al., 2005).

A few studies have been conducted to evaluate transcriptome changes in response to longer term chronic O₃ exposures in woody plant species. Longer term exposures resulted in the upregulation of genes associated with secondary metabolites, including isoprenoids, polyamines and phenylpropanoids in 2-year-old seedlings of the Mediterranean shrub Phillyrea latifolia exposed to 110 ppb O₃ for 90 days (Paolacci et al., 2007). In 3-year-old European beech saplings exposed to O₃ for 20 months (with monthly average twice ambient O₃ concentrations ranging from 11 to 80 ppb), O₃-induced changes in gene transcription were similar to those observed
for herbaceous species (Olbrich et al., 2009). Genes encoding proteins associated with plant stress response, including ethylene biosynthesis, pathogenesis-related proteins and enzymes detoxifying ROS, were upregulated. Some genes associated with primary metabolism, cell structure, cell division and cell growth were reduced (Olbrich et al., 2009). In a similar study using adult European beech trees, it was determined that the magnitude of the transcriptional changes described above was far greater in the saplings than in the adult trees exposed to the same O₃ concentrations for the same time period (Olbrich et al., 2010).

The results from transcriptome studies described above have been substantiated by results from proteome analysis in rice, poplar, European beech, wheat, and soybean. Exposure of soybean to 120 ppb O₃ for 12 hours/day for 3 days in growth chambers resulted in decreases in the quantity of proteins associated with photosynthesis, while proteins involved with antioxidant defense and carbon metabolism increased (Ahsan et al., 2010). Young poplar plants exposed to 120 ppb O₃ in a growth chamber for 35 days also showed significant changes in proteins involved in carbon metabolism (Bohler et al., 2007). Declines in enzymes associated with carbon fixation, the Calvin cycle and photosystem II were measured, while ascorbate peroxidase and enzymes associated with glucose catabolism increased in abundance. In another study to determine the impacts of O₃ on both developing and fully expanded poplar leaves, young poplars were exposed to 120 ppb O₃ for 13 hours/day for up to 28 days (Bohler et al., 2010). Impacts on protein quantity only occurred after the plants had been exposed to O₃ for 14 days, and at this point in time, several Calvin cycle enzymes were reduced in quantity, while the effects on the light reactions appeared later, at 21 days after beginning treatment. Some of the antioxidant enzymes increased in abundance with O₃ treatment, while others (ascorbate peroxidase) did not. In relationship to leaf expansion, it was shown that O₃ did not affect protein quantity until leaves had reached full expansion, after about 7 days (Bohler et al., 2010).

Two-week-old rice seedlings exposed to varying levels of O₃ (4, 40, 80, 120 ppb) in a growth chamber for 9 days showed reductions in quantities of proteins associated with photosynthesis and energy metabolism, and increases in some antioxidant and defense related proteins (Feng et al., 2008a). A subsequent study of O₃-treated rice seedlings (exposed to 200 ppb O₃ for 24 hours) focusing on the integration of transcriptomics and proteomics, supported and further enhanced these results (Cho et al., 2008). The authors found that of the 22,000 genes analyzed from the rice genome, 1,535 were differentially regulated by O₃. Those differentially regulated genes were functionally categorized as transcription factors, MAPK cascades, those encoding for enzymes involved in the synthesis of jasmonic acid (JA), ethylene (ET), shikimate, tryptophan and lignin, and those involved in glycolysis, the citric acid cycle, oxidative respiration and photosynthesis. The authors determined that the proteome and metabolome (all small molecule metabolites in a cell) analysis supported the results of the transcriptome changes described above (Cho et al., 2008). This type of study, which ties together results from changes in gene expression, protein quantity and activity, and metabolite levels, provides the most
complete picture of the molecular and biochemical changes occurring in plants exposed to a stressor such as O₃.

Sarkar et al. (2010) compared proteomes of two cultivars of wheat grown in OTCs at several O₃ concentrations, including filtered air, ambient O₃ (mean concentration 47 ppb), ambient + 10 ppb and ambient + 20 ppb for 5 hours/day for 50 days. Declines in the rate of photosynthesis and stomatal conductance were related to decreases in proteins involved in carbon fixation and electron transport and increased proteolysis of photosynthetic proteins such as the large subunit of ribulose-1,6-bisphosphate carboxylase/oxygenase (Rubisco). Enzymes that take part in energy metabolism, such as ATP synthesis, were also downregulated, while defense/stress related proteins were upregulated in O₃-treated plants. In comparing the two wheat cultivars, Sarkar et al. (2010) found that while the qualitative changes in protein expression between the two cultivars were similar, the magnitude of these changes differed between the sensitive and tolerant wheat cultivars. Greater foliar injury and a smaller decline in stomatal conductance was observed in the sensitive cultivar as compared to the more tolerant cultivar, along with greater losses in photosynthetic enzymes and higher quantities of antioxidant enzymes. Results from a three-year exposure of European beech saplings to elevated O₃ (AOT40 value was 52.6 ppm-h for 2006, when trees were sampled) supported the results from the short-term exposure studies described above (Kerner et al., 2011). The O₃ treatment of the saplings resulted in reductions in enzymes associated with the Calvin cycle, which could lead to reduced carbon fixation. Enzymes associated with carbon metabolism/catabolism were increased, and quantities of starch and sucrose were reduced in response to the O₃ treatment in these trees, indicating a potential impact of O₃ on overall carbon metabolism in long-term exposure conditions (Kerner et al., 2011).

Transcriptome and proteome studies have provided valuable information about O₃ effects on plants. These studies allow for simultaneous analysis of changes in the expression patterns of many different genes and proteins, and also provide information on how these molecules might interact with one another as a result of plant exposure to oxidative stress. Gene and protein expression patterns generally differ between O₃-sensitive and tolerant plants, which could result from differential uptake or detoxification of O₃ or from differential regulation of the transcriptome and proteome.
Note: (A) Temporal profile of the oxidative stress response to O₃. The biphasic O₃-induced oxidative burst is represented in black, with the ROS (reactive oxygen species) control measurements shown as a broken line. Average transcript profiles are shown for early upregulated genes (yellow, peaks at 0.5-1 hours), and the 3 hours (blue), 4.5 hours (red) and 9-12 hours (green) late upregulated genes and for the downregulated genes coding for photosynthesis proteins (brown). PR = pathogenesis related. (B) Diagrammatic representation of redox regulation of the oxidative stress response. TF = transcription factor; SA = salicylic acid. Source: Reprinted with permission of Springer (Mahalingam et al., 2005).

**Figure 9-5**  
Composite diagram of major themes in the temporal evolution of the genetic response to O₃ stress.

All of these studies describe common trends for changes in gene and protein expression which occur in a variety of plant species exposed to O₃. While genes associated with carbon assimilation and general metabolism are typically downregulated, genes associated with signaling, catabolism, and defense are upregulated. The magnitude of these changes in gene and protein expression appears to be related to plant species, age and their sensitivity or tolerance to O₃.
9.3.3.3 Role of Phytohormones in Plant Response to Ozone

Many studies of O₃ effects on plants have analyzed the importance of plant hormones such as SA, ET and JA in determining plant response to O₃. The 2006 O₃ AQCD documents the O₃-induced production of ET and its role in promoting the formation of leaf lesions. Transcriptome analysis and the use of a variety of mutants have allowed for further elucidation of the complex interactions between SA, ET, JA and the role of abscisic acid (ABA) in mediating plant response to O₃ (Ludwikow and Sadowski, 2008). In addition to their roles in signaling pathways, phytohormones also appear to regulate, and be regulated by, the MAPK signaling cascades described previously. Most evidence suggests that while ET and SA are needed to develop O₃-induced leaf lesions, JA acts antagonistically to SA and ET to limit the lesions (Figure 9-6) (Kangasjarvi et al., 2005).

The rapid production of ET in O₃ treated plants has been described in many plant species and has been further characterized through the use of a variety of mutants that either over-produce or are insensitive to ET. Production of stress ET in O₃-treated plants, which is thought to be part of a wounding response, was found to be correlated to the degree of injury development in leaves (U.S. EPA, 2006b). More recent studies have supported these conclusions and have also focused on the interactions occurring between several oxidative-stress induced phytohormones. Yoshida et al. (2009) determined that ET likely amplifies the oxidative signal generated by ROS, thereby promoting lesion formation. By analyzing the O₃-induced transcriptome of several Arabidopsis mutants of the Col-0 ecotype, Tamaoki et al. (2003) determined that at 12 hours after initiating the O₃ exposure (200 ppb for 12 hours), the ET and JA signaling pathways were the main pathways used to activate plant defense responses, with a lesser role for SA. The authors also demonstrated that low levels of ET production could stimulate the expression of defense genes, rather than promoting cell death which occurs when ET production is high. Tosti et al. (2006) supported these findings by showing that plant exposure to O₃ not only results in activation of the biosynthetic pathways of ET, JA and SA, but also increases the expression of genes related to the signal transduction pathways of these phytohormones in O₃-treated Arabidopsis plants (300 ppb O₃ for 6 hours). Conversely, in the O₃ sensitive Ws ecotype, its sensitivity may, in part, be due to intrinsically high ET levels leading to SA accumulation, and the high ET and SA may act to repress JA-associated genes, which would serve to inhibit the spread of lesions (Mahalingam et al., 2006). Ogawa et al. (2005) found that increases in SA in O₃-treated plants leads to the formation of leaf lesions in tobacco plants exposed to 200 ppb O₃ for 6 hours. Furthermore, in transgenic tobacco plants with reduced levels of ET production in response to O₃ exposure, several genes encoding for enzymes in the biosynthetic pathway of SA were suppressed, suggesting that SA levels are, in part, controlled by ET in the presence of O₃.

Exposure of the Arabidopsis mutant rcd1 to acute doses of O₃ (250 ppb O₃ for 8 hours/day for 3 days) resulted in programmed cell death (PCD) and the formation of leaf lesions (Overmyer et al., 2000). They determined that the observed induction of ET synthesis promotes cell death, and that ET perception and signaling are
required for the accumulation of superoxide, which leads to cell death and propagation of lesions. Jasmonic acid, conversely, contains the spread of leaf lesions (Overmyer et al., 2000). Transcriptome analysis of several Arabidopsis mutants, which are insensitive to SA, ET and JA, exposed to 12-h of 200 ppb O3 showed that approximately 78 of the upregulated genes measured in this study were controlled by ET and JA signaling pathways, while SA signaling pathways were suggested to antagonize ET and JA pathways (Tamaoki et al., 2003). In a subsequent transcriptome study on the Col-0 ecotype exposed to 150 ppb O3 for 48-h, JA and ET synthesis were downregulated, while SA was upregulated in O3-treated plants. In cotton plants exposed to a range of O3 concentrations (0-120 ppb) and methyl jasmonate (MeJA), Grantz et al. (2010b) determined that exogenous applications of MeJA did not protect plants from chronic O3 exposure.

Abscisic acid has been investigated for its role in regulating stomatal aperture and also for its contribution to signaling pathways in the plant. The role of ABA and the interaction between ABA and H2O2 in O3-induced stomatal closure was described in the 2006 O3 AQCD. It was determined that the presence of H2O2, which is formed from O3 degradation, increases the sensitivity of guard cells to ABA and, therefore, more readily results in stomatal closure. More recently, it was determined that synthesis of ABA was induced in O3-treated Arabidopsis plants (250-350 ppb O3 for 6 hours), with a more pronounced induction in the O3 sensitive rcd3 mutant as compared to the wildtype Col-0 (Overmyer et al., 2008). The rcd3 mutant also exhibited a lack of O3-induced stomatal closure, and the RCD3 protein has been shown to be required for slow anion channels (Overmyer et al., 2008). Ludwikow et al. (2009) used Arabidopsis ABI1td mutants, in which a key negative regulator of ABA action (abscisic acid insensitive1 protein phosphatase 2C) has been knocked out, to examine O3 responsive genes in this mutant compared to the Arabidopsis Col-0. Results of this study indicate a role for ABI1 in negatively regulating the synthesis of both ABA and ET in O3-treated plants (350 ppb O3 for 9 hours). Additionally, ABI1 may stimulate JA-related gene expression, providing evidence for an antagonistic interaction between ABA and JA signaling pathways (Ludwikow et al., 2009).

Nitric oxide (NO) has also been shown to play a role in regulating gene expression in plants in response to O3 exposure. However, little is known to date about NO and its role in the complex interactions of molecules in response to O3. Exposure of tobacco to O3 (150 ppb for 5 hours) stimulated NO and NO-dependent ET production, while NO production itself did not depend on the presence of ET (Ederli et al., 2006). Analysis of O3-treated Arabidopsis indicated the possibility of a dual role for NO in the initiation of cell death and later lesion containment (Ahlfors et al., 2009).

While much work remains to be done to better elucidate what determines plant sensitivity and response to O3, it is clear that the mechanism for response to O3 and signal transduction is very complex. Many of the phytohormones and other signaling molecules thought to be involved in these processes are interactive and depend upon a variety of other factors, which could be either internal or external to the plant. This results in a highly dynamic and complex system, capable of resulting in a spectrum
of plant sensitivity to oxidative stress and generating a variety of plant responses to that stress.

Note: Ozone-derived radicals induce endogenous ROS production (1) which results in salicylic acid (SA) accumulation and programmed cell death; (2) Cell death triggers ethylene (ET) production, which is required for the continuing ROS production responsible for the propagation of cell death; (3) Jasmonates counteract the progression of the cycle by antagonizing the cell death promoting function of SA and ET; (4) Abscisic acid (ABA) antagonizes ET function in many situations and might also have this role in O₃-induced cell death; (5) Mutually antagonistic interactions between ET, SA and jasmonic acid (JA) are indicated with red bars.

Source: Reprinted with permission of Blackwell Publishing Ltd. (Kangasjarvi et al., 2005).

Figure 9-6 The oxidative cell death cycle.

9.3.4 Detoxification

9.3.4.1 Overview of Ozone-induced Defense Mechanisms

Plants are exposed to an oxidizing environment on a continual basis, and many reactions that are part of the basic metabolic processes, such as photosynthesis and respiration, generate ROS. As a result, there is an extensive and complex mechanism in place to detoxify these oxidizing radicals, including both enzymes and metabolites, which are located in several locations in the cell and also in the apoplast of the cell. As O₃ enters the leaf through open stomata, the first point of contact of
O₃ with the plant is likely in the apoplast, where it breaks down to form oxidizing radicals such as H₂O₂, O₂⁻, HO⁻ and HO₂. Another source of oxidizing radicals is an oxidative burst, generated by a membrane-bound NADPH oxidase enzyme, which is recognized as an integral component of the plant's defense system against pathogens (Schraudner et al., 1998). Antioxidant metabolites and enzymes located in the apoplast are thought to form a first line of defense by detoxifying O₃ and/or the ROS that are formed as breakdown products of O₃ (Section 9.3.2). However, even with the presence of several antioxidants, including ascorbate, the redox buffering capacity of the apoplast is far less than that of the cytoplasm, as it lacks the regeneration systems necessary to retain a reduced pool of antioxidants (Foyer and Noctor, 2005b).

Redox homeostasis is regulated by the presence of a pool of antioxidants, which are typically found in a reduced state and detoxify ROS produced by oxidases or electron transport components. As ROS increase due to environmental stress such as O₃, it is unclear whether the antioxidant pool can maintain its reduced state (Foyer and Noctor, 2005b). As such, not only the quantity and types of antioxidant enzymes and metabolites present, but also the cellular ability to regenerate those antioxidants are important considerations in mechanisms of plant tolerance to oxidative stress (Dizengremel et al., 2008). Molecules such as glutathione (GSH), thioredoxins and NADPH play very important roles in this regeneration process; additionally, it has been hypothesized that alterations in carbon metabolism would be necessary to supply the needed reducing power for antioxidant regeneration (Dizengremel et al., 2008).

### 9.3.4.2 Role of Antioxidants in Plant Defense Responses

Ascorbate has been the focus of many different studies as an antioxidant metabolite that protects plants from exposure to O₃. It is found in several cellular locations, including the chloroplast, the cytosol and the apoplast (Noctor and Foyer, 1998). Ascorbate is synthesized in the cell and transported to the apoplast. Apoplastic ascorbate can be oxidized to dehydroascorbate (DHA) with exposure to O₃ and is then transported back to the cytoplasm. Here, DHA is reduced to ascorbate by the enzyme dehydroascorbate reductase (DHAR) and reduced GSH, which is part of the ascorbate-glutathione cycle (Noctor and Foyer, 1998). Many studies have focused on evaluating whether ascorbate is the primary determining factor in differential sensitivity of plants to O₃. An evaluation of several species of wildflowers in Great Smoky Mountains National Park showed a correlation between higher quantities of reduced apoplastic ascorbate and lower levels of foliar injury from O₃ exposure in a field study on tall milkweed plants (Asclepias exaltata L.) (Burkey et al., 2006; Souza et al., 2006). Cheng et al. (2007) exposed two soybean cultivars to elevated O₃ (77 ppb) and filtered air for 7 hours/day for 6 days. The differences in sensitivity between the two cultivars could not be explained by differential O₃ uptake or by the fraction of reduced ascorbate present in the apoplast. However, total antioxidant capacity of the apoplast was 2-fold higher in the tolerant Essex cultivar as compared to the sensitive Forrest cultivar, indicating that there may be other compounds in the
leaf apoplast that scavenge ROS. D’Haese et al. (2005) exposed the NC-S (sensitive) and NC-R (resistant) clones of white clover (Trifolium repens) to 60 ppb O₃ for 7 hours/day for 5 days in environmental chambers. Surprisingly, the NC-S clone had a higher constitutive concentration of apoplastic ascorbate with a higher redox status than the NC-R clone. However, the redox status of symplastic GSH was higher in NC-R, even though the concentration of GSH was not higher than in NC-S. In addition, total symplastic antioxidative capacity was not a determining factor in differential sensitivity between these two clones. Severino et al. (2007) also examined the role of antioxidants in the differential sensitivity of the two white clover clones by growing them in the field for a growing season and then exposing them to elevated O₃ (100 ppb for 8 hours/day for 10 days) in OTC at the end of the field season. The NC-R clone had greater quantities of total ascorbate and total antioxidants than the NC-S clone at the end of the experiment. In snap bean, plants of the O₃ tolerant Provider cultivar had greater total ascorbate and more ascorbate in the apoplast than the sensitive S156 cultivar after exposure to 71 ppb O₃ for 10 days in OTC (Burkey et al., 2003). While most of the apoplastic ascorbate was in the oxidized form, the ratio of reduced ascorbate to total ascorbate was higher in Provider than S156, indicating that Provider is better able to maintain this ratio to maximize plant protection from oxidative stress. Exposure of two wheat varieties to ambient (7-h average 44 ppb O₃) and elevated (7-h average 56 ppb O₃) O₃ for 60 days in open-air field conditions showed higher concentrations of reduced ascorbate in the apoplast in the tolerant Y16 variety than the more sensitive Y2 variety, however no varietal differences were seen in the decrease in reduced ascorbate quantity in response to O₃ exposure (Feng et al., 2010). To evaluate whether O₃ affected apoplastic concentrations of ascorbic acid and phenolic compounds, wildtype Arabidopsis thaliana (Col-0, Ler-0) and null mutants lacking sinapoyl and flavonol glycosides were exposed to either 125 or 175 ppb O₃ for up to 2 days. The authors determined that ascorbic acid, which was found in very low quantities in the reduced form, and the phenolic compounds did not play an important role in protecting plants from O₃ injury (Booker et al., 2012). While there is much evidence that supports an important role for ascorbate, particularly apoplastic ascorbate, in protecting plants from oxidative stressors such as O₃, it is also clear that there is much variation in the importance of ascorbate for different plant species and differing exposure conditions. Additionally, the work of several authors suggests that there may be other compounds in the apoplast which have the capacity to act as antioxidants.

While the quantities of antioxidant metabolites such as ascorbate are an important indicator of plant tolerance to O₃, the ability of the plant to recycle oxidized ascorbate efficiently also plays a large role in determining the plant’s ability to effectively protect itself from sustained exposure to oxidative stress. Tobacco plants over-expressing DHAR were better protected from exposure to either chronic (100 ppb O₃ 4 hours/day for 30 days) or acute (200 ppb O₃ for 2 hours) O₃ conditions than control plants and those with reduced expression of DHAR (Chen and Gallie, 2005). The DHAR over-expressing plants exhibited an increase in guard cell ascorbic acid, leading to a decrease in stomatal responsiveness to O₃ and an increase in stomatal conductance and O₃ uptake. Despite this, the presence of higher...
levels of ascorbic acid led to a lower oxidative load and a higher level of photosynthetic activity in the DHAR over-expressing plants (Chen and Gallie, 2005). A subsequent study with tobacco plants over-expressing DHAR confirmed some of these results. Levels of ascorbic acid were higher in the transgenic tobacco plants, and they exhibited greater tolerance to O₃ exposure (200 ppb O₃) as demonstrated by higher photosynthetic rates in the transgenic plants as compared to the control plants (Eltayeb et al., 2006). Over-expression of monodehydroascorbate reductase (MDAR) in tobacco plants also showed enhanced stress tolerance in response to O₃ exposure (200 ppb O₃), with higher rates of photosynthesis and higher levels of reduced ascorbic acid as compared to controls (Eltayeb et al., 2007). Results of these studies demonstrate the importance of ascorbic acid as a detoxification mechanism in some plant species, and also emphasize that the recycling of oxidized ascorbate to maintain a reduced pool of ascorbate is a factor in determining plant tolerance to oxidative stress.

The roles of other antioxidant metabolites and enzymes, including GSH, catalase (CAT), peroxidase (POD) and superoxide dismutase (SOD), were comprehensively reviewed in the 2006 O₃ AQCD. Based on the review of the literature, no conclusive and consistent effects of O₃ on the quantity of GSH and CAT could be identified. Both apoplastic and cytosolic POD activity increased in response to O₃ exposure, while various isoforms of SOD showed inconsistent changes in quantity in response to O₃. Additional studies have been conducted to further elucidate the roles of these antioxidant enzymes and metabolites in protecting plants from oxidative stress. Superoxide dismutase and POD activities were measured in both the tolerant Bel B and sensitive Bel W3 tobacco cultivars exposed to ambient O₃ concentrations for 2 weeks 3 times throughout a growing season (Borowiak et al., 2009). In this study, SOD and POD activity, including that of several different isoforms, increased in both the sensitive and tolerant tobacco cultivars with exposure to O₃, however the isoenzyme composition for POD differed between the sensitive and tolerant tobacco cultivars (Borowiak et al., 2009). Tulip poplar (Liriodendron tulipifera) trees exposed to increasing O₃ concentrations (from 100 to 300 ppb O₃ during a 2-week period) showed increases in activities of SOD, ascorbate peroxidase (APX), glutathione reductase (GR), MDAR, DHAR, CAT and POD in the 2-week period, although individual enzyme activities increased at different times during the 2-week period (Ryang et al., 2009).

Longer, chronic O₃ exposures in trees revealed increases in SOD and APX activity in Quercus mongolica after 45 days of plant exposure to 80 ppb O₃, which were followed by declines in the activities and quantities of these enzymes after 75 days of exposure (Yan et al., 2010). Similarly, activities of SOD, APX, DHAR, MDAR, and GR increased in Gingko biloba trees during the first 50 days of exposure to 80 ppb O₃, followed by decreases in activity below control values after 50 days of exposure (He et al., 2006). Soybean plants exposed to 70 or 100 ppb O₃ for 4 hours/day over the course of a growing season showed elevated POD activity and a decrease in CAT activity at 40 and 60 days after germination (Singh et al., 2010a).
Antioxidant enzymes and metabolites have been shown to play an important role in determining plant tolerance to O3 and mediating plant responses to O3. However, there is also some evidence to suggest that the direct reaction of ascorbate with O3 could lead to the formation of secondary toxicants, such as peroxo compounds, which may act upon signal transduction pathways and modulate plant response to O3 (Sandermann, 2008). Therefore, the role of ascorbate and other antioxidants and their interaction with other plant responses to O3, such as the activation of signal transduction pathways, is likely far more complex than is currently understood.

9.3.5 Effects on Primary and Secondary Metabolism

9.3.5.1 Light and Dark Reactions of Photosynthesis

Declines in the rate of photosynthesis in O3-treated plants have been documented for many different plant species (Booker et al., 2009; Wittig et al., 2007; U.S. EPA, 2006b). The 2006 O3 AQCD described the mechanism by which plant exposure to O3 reduces carboxylation capacity, and the more recent scientific literature confirms these findings. While several measures of the light reactions of photosynthesis are sensitive to exposure to O3 (see below), photosynthetic carbon assimilation is generally considered to be more affected by pollutant exposure, resulting in an overall decline in photosynthesis (Guidi and Degl'Innocenti, 2008; Heath, 2008; Fiscus et al., 2005). Loss of carbon assimilation capacity has been shown to result from declines in the quantity and activity of Rubisco (Calatayud et al., 2010; Goumenaki et al., 2010; Singh et al., 2009; Bagard et al., 2008; Calatayud et al., 2007a; Crous et al., 2006). Experimental evidence suggests that both decreases in Rubisco synthesis and enhanced degradation of the protein contribute to the measured reduction in its quantity. Additionally, the reduction in Rubisco quantity has been associated with the O3-induced oxidative modification of the enzyme, which is evidenced by the increases in carbonyl groups on the protein after plant exposure to O3 (U.S. EPA, 2006b). Reduced carbon assimilation has been linked to reductions in biomass and yield (Wang et al., 2009b; He et al., 2007; Novak et al., 2007; Gregg et al., 2006; Keutgen et al., 2005). Recent studies evaluating O3 induced changes in the transcriptome and proteome of several different species confirm these findings. Levels of mRNA for rbcS (the gene that encodes the small subunit [SSU] of the RuBisCO protein [ribulose-1,5-bisphosphate carboxylase/oxygenase, a major stromal enzyme involved in carbon fixation by plants]) declined in European beech saplings exposed to 300 ppb O3 for 8 hours/day for up to 26 days (Olbrich et al., 2005). Similar declines in rbcS mRNA were also measured in the beech saplings in a free air exposure system over a course of two growing seasons (Olbrich et al., 2009). Proteomics studies have also confirmed the effects of O3 on proteins involved in carbon assimilation. Reductions in quantities of the small and large subunit (rbcL) of Rubisco and Rubisco activase were measured in soybean plants exposed to 120 ppb O3 for 3 days in growth chambers (Ahsan et al., 2010). Exposure of young poplar trees to 120 ppb O3 for 35 days in exposure chambers resulted in reductions of
Rubisco, Rubisco activase, and up to 24 isoforms of Calvin cycle enzymes, most of which play a role in regenerating the CO₂ acceptor molecule, ribulose-1,5-bisphosphate (Bohler et al., 2007). Reductions in protein quantity of both the small and large subunit of Rubisco were seen in wheat plants exposed to ambient (average concentration 47.3 ppb O₃) and elevated O₃ (ambient + 10 or 20 ppb O₃) in open-top chambers for 5 hours/day for 50 days (Sarkar et al., 2010). Lettuce plants exposed to 100 ppb O₃ in growth chambers for 8 hours/day for 3 weeks also showed reductions in transcript and protein levels of the small and large subunits of Rubisco and Rubisco activase (Goumenaki et al., 2010).

Reductions in photosynthesis are not only related to declines in the quantity of Rubisco, but also of its activity level. The maximum carboxylation rate ($V_{\text{cmax}}$) has been shown to decline in plants species exposed to O₃, including lettuce (Goumenaki et al., 2010), white clover (Crous et al., 2006), young poplar trees (Bagard et al., 2008) and evergreen deciduous shrubs (Calatayud et al., 2010). While a significant proportion of the reduction in $V_{\text{cmax}}$ is caused by declines in the quantity of Rubisco, other contributors to changes in $V_{\text{cmax}}$ result from reductions in the quantity and activity of Rubisco activase, an enzyme which prepares Rubisco for carbamylation by accelerating the release of bound sugar phosphates. Reductions in Rubisco activase quantity have been observed in several studies evaluating the effects of O₃ on the proteomes of poplar (Bohler et al., 2007), European beech (Kerner et al., 2011) and soybean (Ahsan et al., 2010).

In addition to impacts on carbon assimilation, the deleterious effects of O₃ on the photosynthetic light reactions have received more attention in recent years. Chlorophyll fluorescence provides a useful measure of changes to the photosynthetic process from exposure to oxidative stress. Decreases in the Fv/Fm ratio (a measure of the maximum efficiency of Photosystem II) in dark adapted leaves indicate a decline in the efficiency of the PSII photosystems and a concomitant increase in non-photochemical quenching (Guidi and Degl'Innocenti, 2008; Scebba et al., 2006). Changes in these parameters have been correlated to differential sensitivity of plants to the pollutant. In a study to evaluate the response of 4 maple species to O₃ (exposed to an 8-h avg of 51 ppb for ambient and 79 ppb for elevated treatment in OTC), the 2 species which were most sensitive based on visible injury and declines in CO₂ assimilation also showed the greatest decreases in Fv/Fm in symptomatic leaves. In asymptomatic leaves, CO₂ assimilation decreased significantly but there was no significant decline in Fv/Fm (Calatayud et al., 2007). Degl'Innocenti et al. (2007) measured significant decreases in Fv/Fm in young and symptomatic leaves of a resistant tomato genotype (line 93.1033/1) in response to O₃ exposure (150 ppb O₃ for 3 hours in a growth chamber), but only minor decreases in asymptomatic leaves with no associated changes in net photosynthetic rate. In the O₃ sensitive tomato cultivar Cuor Di Bue, the Fv/Fm ratio did not change, while the photosynthetic rate declined significantly in asymptomatic leaves (Degl'Innocenti et al., 2007). In two soybean cultivars, Fv/Fm also declined significantly with plant exposure to O₃ (Singh et al., 2009). It appears that in asymptomatic leaves, photoinhibition, as indicated by a decrease in Fv/Fm, is not the main reason for a decline in photosynthesis.
An evaluation of photosynthetic parameters of two white clover (*Trifolium repens* cv. Regal) clones that differ in their O$_3$ sensitivity revealed that O$_3$ (40-110 ppb O$_3$ for 7 hours/day for 5 days) increased the coefficient of non-photochemical quenching ($q_{NP}$) in both the resistant (NC-R) and sensitive (NC-S) clones, however $q_{NP}$ was significantly lower for the sensitive clone ([Crous et al., 2006](#)). Sensitive *Acer* clones had a lower coefficient of non-photochemical quenching, while exposure to O$_3$ increased $q_{NP}$ in both sensitive and tolerant clones ([Calatayud et al., 2007a](#)). While exposure to O$_3$ also increased $q_{NP}$ in tomato, there were no differences in the coefficient of photochemical quenching between cultivars thought to be differentially sensitive to O$_3$ ([Degl'Innocenti et al., 2007](#)). Higher $q_{NP}$ as a result of exposure to O$_3$ indicates a reduction in the proportion of absorbed light energy being used to drive photochemistry. A lower coefficient of non-photochemical quenching in O$_3$ sensitive plants could indicate increased vulnerability to ROS generated during exposure to oxidative stress ([Crous et al., 2006](#)).

Most of the research on O$_3$ effects on photosynthesis has focused on C3 (Calvin cycle) plants because C4 (Hatch-Slack) plants have lower stomatal conductance and are, therefore, thought to be less sensitive to O$_3$ stress. However, some studies have been conducted to evaluate the effects of O$_3$ on C4 photosynthesis. In older maize leaves, Leitao et al. ([2007c](#); [2007a](#)) found that the activity, quantity and transcript levels of both Rubisco and phosphoenolpyruvate carboxylase (PEPc) decreased as a function of rising O$_3$ concentration. In younger maize leaves, the quantity, activity, and transcript levels of the carboxylases were either increased or unaffected in plants exposed to 40 ppb O$_3$ for 7 hours/day for 28-33 days, but decreased at 80 ppb ([Leitao et al., 2007b](#); [Leitao et al., 2007c](#)). In another study, Grantz et al. ([2009](#)) reported that O$_3$ exposures (4, 58, and 114 ppb, 12-h mean) decreased sugarcane biomass production by more than one third and allocation to roots by more than two thirds.

### 9.3.5.2 Respiration and Dark Respiration

While much research emphasis regarding O$_3$ effects on plants has focused on the negative impacts on carbon assimilation, other studies have measured impacts on catabolic pathways such as shoot respiration and photorespiration. Generally, shoot respiration has been found to increase in plants exposed to O$_3$. Bean plants exposed to ambient (average 12-h mean 43 ppb) and twice ambient (average 12-h mean 80 ppb) O$_3$ showed increases in respiration. When mathematically partitioned, the maintenance coefficient of respiration was significantly increased in O$_3$ treated plants, while the growth coefficient of respiration was not affected ([Amthor, 1988](#)). Loblolly pines were exposed to ambient (12-h daily mean was 45 ppb) and twice ambient (12-h daily mean was 86 ppb) O$_3$ for 12 hours/day for approximately seven months per year for 3 and 4 years. While photosynthetic activity declined with the age of the needles and increasing O$_3$ concentration, enzymes associated with respiration showed higher levels of activity with increasing O$_3$ concentration ([Dizengremel et al., 1994](#)). In their review on the role of metabolic changes in plant redox status after O$_3$ exposure, Dizengremel et al. ([2009](#)) summarized multiple
studies in which several different tree species were exposed to O$_3$ concentrations ranging from ambient to 200 ppb O$_3$ for at least several weeks. In all cases, the activity of enzymes, including phosphofructokinase, pyruvate kinase and fumarase, which are part of several catabolic pathways, were increased in O$_3$ treated plants.

Photorespiration is a light-stimulated process which consumes O$_2$ and releases CO$_2$. While it has been regarded as a wasteful process, more recent evidence suggests that it may play a role in photoprotection during photosynthesis (Bagard et al., 2008). The few studies that have been conducted on O$_3$ effects on photorespiration suggest that rates of photorespiration decline concomitantly with rates of photosynthesis. Soybean plants were exposed to ambient (daily averages 43-58 ppb) and 1.5 ambient O$_3$ (daily averages 63-83 ppb) O$_3$ in OTCs for 12 hours/day for 4 months. Rates of photosynthesis and photorespiration and photorespiratory enzyme activity declined only at the end of the growing season and did not appear to be very sensitive to O$_3$ exposure (Booker et al., 1997). Young hybrid poplars exposed to 120 ppb O$_3$ for 13 hours/day for 35 days in phytotron chambers showed that effects on photorespiration and photosynthesis were dependent upon the developmental stage of the leaf. While young leaves were not impacted, reductions in photosynthesis and photorespiration were measured in fully expanded leaves (Bagard et al., 2008).

9.3.5.3 Secondary Metabolism

Transcriptome analysis of Arabidopsis plants has revealed modulation of several genes involved in plant secondary metabolism (Ludwikow and Sadowski, 2008). Phenylalanine ammonia lyase (PAL) has been the focus of many studies involving plant exposure to O$_3$ due to its importance in linking the phenylpropanoid pathway of plant secondary metabolism to primary metabolism in the form of the shikimate pathway. Genes encoding several enzymes of the phenylpropanoid pathway and lignin biosynthesis were upregulated in transcriptome analysis of Arabidopsis plants (Col-0) exposed to 350 ppb O$_3$ for 6 hours, while 2 genes involved in flavonoid biosynthesis were downregulated (Ludwikow et al., 2004). Exposure of Arabidopsis (Col-0) to lower O$_3$ concentrations (150 ppb for 8 hours/day for 2 days) resulted in the induction of 11 transcripts involved in flavonoid synthesis. In their exposure of 2-year-old Mediterranean shrub Phillyrea latifolia to 110 ppb O$_3$ for 90 days, Paolacci et al. (2007) identified four clones that were upregulated and corresponded to genes involved in the synthesis of secondary metabolites, such as isoprenoids, polyamines and phenylpropanoids. Upregulation of genes involved in isoprene synthesis was also observed in Medicago trunculata exposed to 300 ppb O$_3$ for 6 hours, while genes encoding enzymes of the flavonoid synthesis pathway were either upregulated or downregulated (Puckette et al., 2008). Exposure of red clover to 1.5 × ambient O$_3$ (average concentrations of 32.4 ppb) for up to 9 weeks in an open field exposure system resulted in increases in leaf total phenolic content. However, the types of phenolics that were increased in response to O$_3$ exposure differed depending upon the developmental stage of the plant. While almost all of the 31 different phenolic compounds measured increased in quantity initially during the
exposure, after 3 weeks the quantity of isoflavones decreased while other phenolics increased (Saviranta et al., 2010). Exposure of beech saplings to ambient and 2 × ambient O₃ concentrations over 2 growing seasons resulted in the induction of several enzymes which contribute to lignin formation, while enzymes involved in flavonoid biosynthesis were downregulated (Olbrich et al., 2009). Exposure of tobacco Bel W3 to 160 ppb O₃ for 5 hours showed upregulation of almost all genes encoding for enzymes which are part of the prechorismate pathway (Janzik et al., 2005). Isoprenoids can serve as antioxidant compounds in plants exposed to oxidative stress (Paolacci et al., 2007).

The prechorismate pathway is the pathway leading to the formation of chorismate, a precursor to the formation of the aromatic amino acids tryptophan, tyrosine and phenylalanine. These amino acids are precursors for the formation of many secondary aromatic compounds, and, therefore, the prechorismate pathway represents a branch-point in the regulation of metabolites into either primary or secondary metabolism (Janzik et al., 2005). Exposure of the O₃ sensitive Bel W3 tobacco cultivar at 160 ppb for 5 hours showed an increase in transcript levels of most of the genes encoding enzymes of the prechorismate pathway. However, shikimate kinase (SK) did not show any change in transcript levels and only one of three isoforms of DAHPS (3-deoxy-D-arabino-heptulosonat-7-phosphate synthase), the first enzyme in this pathway, was induced by O₃ exposure (Janzik et al., 2005). Differential induction of DAHPS isoforms was also observed in European beech after 40 days of exposure to 150-190 ppb O₃. At this time point in the beech experiment, transcript levels of shikimate pathway enzymes, including SK, were generally strongly induced after an only weak initial induction after the first 40 days of exposure. Both soluble and cell-wall bound phenolic metabolites showed only minimal increases in response to O₃ for the duration of the exposure period (Alonso et al., 2007). Total leaf phenolics decreased with leaf age in Populus nigra exposed to 80 ppb O₃ for 12 hours/day for 14 days. Ozone increased the concentration of total leaf phenolics in newly expanded leaves, with the greatest increases occurring in compounds such as quercitin glycoside, which has a high antioxidant capacity (Fares et al., 2010b). While several phenylpropanoid pathway enzymes were induced in two poplar clones exposed to 60 ppb O₃ for 5 hours/day for 15 days, the degree of induction differed between the two clones. In the tolerant I-214 clone, PAL activity increased 9-fold in O₃-treated plants as compared to controls, while there was no significant difference in PAL activity in the sensitive Eridano clone (Di Baccio et al., 2008).

Polyamines such as putrescine, spermidine and spermine play a variety of roles in plants and have been implicated in plant defense responses to both abiotic and biotic stresses. They exist in both a free form and conjugated to hydroxycinnamic acids. Investigations on the role of polyamines have found that levels of putrescine increase in response to oxidative stress. This increase stems largely from the increase in the activity of arginine decarboxylase (ADC), a key enzyme in the synthesis of putrescine (Groppa and Benavides, 2008). Langebartels et al. (1991) described differences in putrescine accumulation in O₃-treated tobacco plants exposed to several O₃ concentrations, ranging from 0-400 ppb for 5-7 hours. A large and rapid
increase in putrescine occurred in the tolerant Bel B cultivar and only a small increase in the sensitive Bel W3 cultivar, which occurred only after the formation of necrotic leaf lesions. Van Buuren et al. (2002) further examined the role of polyamines in these two tobacco cultivars during an acute (130 ppb O3 for 7-h in a growth chamber) exposure. They found that while free putrescine accumulated in undamaged tissue of both cultivars, conjugated putrescine predominantly accumulated in tissues undergoing cell death after plant exposure to O3 (van Buuren et al., 2002). The authors suggest that while free putrescine may not play a role in conferring tolerance in the Bel B cultivar, conjugated putrescine may play a role in O3-induced programmed cell death in Bel W3 plants.

Isoprene is emitted by some plant species and represents the predominant biogenic source of hydrocarbon emissions in the atmosphere (Guenther et al., 2006). In the atmosphere, the oxidation of isoprene by hydroxyl radicals can enhance O3 formation in the presence of NOX, thereby impacting the O3 concentration that plants are exposed to. While isoprene emission varies widely between species, it has been proposed to stabilize membranes and provide those plant species that produce it with a mechanism of thermotolerance (Sharkey et al., 2008). It has also been suggested that isoprene may act as an antioxidant compound to scavenge O3 (Loreto and Velikova, 2001). Recent studies using a variety of plant species have shown conflicting results in trying to understand the effects of O3 on isoprene emission. Exposure to acute doses of O3 (300 ppb for 3-h) in detached leaves of Phragmites australis resulted in stimulation of isoprene emissions (Velikova et al., 2005). Similar increases in isoprene emissions were measured in Populus nigra after exposure to 100 ppb O3 for 5 days continuously (Fares et al., 2008). Isoprene emission in attached leaves of Populus alba, which were exposed to 150 ppb O3 for 11 hours/day for 30 days inside cuvettes, was inhibited, while isoprene emission and transcript levels of isoprene synthase mRNA were increased in the leaves exposed to ambient O3 (40 ppb), which were located above the leaves enclosed in the exposure cuvettes (Fares et al., 2006). Exposure of 2 genotypes of hybrid poplar to 120 ppb O3 for 6 hours/day for 8 days resulted in a significant reduction in isoprene emission in the O3-sensitive but not the tolerant genotype (Ryan et al., 2009). Similarly, O3 treatment (80 ppb 12 hours/day for 14 days) of Populus nigra showed that isoprene emission was reduced in the treated plants relative to the control plants (Fares et al., 2010b). Based on results of this and other studies, Fares et al. (2010b) concluded that the isoprenoid pathway may be induced in plants exposed to acute O3 doses, while at lower doses isoprene emission may be inhibited. Vickers et al. (2009) developed transgenic tobacco plants with the isoprene synthase gene from Populus alba and exposed them to 120 ppb O3 for 6 hours/day for 2 days. They determined that the wildtype plants showed significantly more O3 damage, including the development of leaf lesions and a decline in photosynthetic rates, than the transgenic, isoprene-emitting plants. Transgenic plants also accumulated less H2O2 and had lower levels of lipid peroxidation following exposure to O3 than the wildtype plants (Vickers et al., 2009). These results indicate that isoprene may have a protective role for plants exposed to oxidative stress.
Summary

The results of recent studies on the effects of O$_3$ stress on plants support and strengthen those reported in the 2006 O$_3$ AQCD. The most significant new body of evidence since the 2006 O$_3$ AQCD comes from research on molecular mechanisms of the biochemical and physiological changes observed in many plant species in response to O$_3$ exposure. Recent studies have employed new techniques, such as those used in evaluating transcriptomes and proteomes to perform very comprehensive analyses of changes in gene transcription and protein expression in plants exposed to O$_3$. These newer molecular studies not only provide very important information regarding the many mechanisms of plant responses to O$_3$, they also allow for the analysis of interactions between various biochemical pathways which are induced in response to O$_3$. However, many of these studies have been conducted in artificial conditions with model plants, which are typically exposed to very high, short doses of O$_3$. Therefore, additional work remains to elucidate whether these plant responses are transferable to other plant species exposed to more realistic ambient conditions.

Ozone is taken up into leaves through open stomata. Once inside the substomatal cavity, O$_3$ is thought to rapidly react with the aqueous layer surrounding the cell (apoplast) to form breakdown products such as hydrogen peroxide (H$_2$O$_2$), superoxide (O$_2^-$), hydroxyl radicals (HO·) and peroxy radicals (HO$_2^-$). Experimental evidence suggests that mitogen-activated protein kinases and calcium are important components of the signal transduction pathways, which communicate signals to the nucleus and lead to changes in gene expression in response to O$_3$. It is probable that there are multiple signal transduction pathways, and their activation may depend upon the plant species, its developmental stage and/or O$_3$ exposure conditions. Initiation of signal transduction pathways in O$_3$ treated plants has also been observed in stomatal guard cells. Reductions in stomatal conductance have been described for many plant species exposed to O$_3$. Some recent studies have also reported sluggish stomatal responses and increased stomatal conductance in some situations. New experimental evidence suggests that these effects on stomates may be due not only to a decrease in carboxylation efficiency, but also to a direct impact of O$_3$ on stomatal guard cell function, leading to a changes in stomatal conductance.

Alterations in gene transcription that have been observed in O$_3$-treated plants are now evaluated more comprehensively using DNA microarray studies, which measure changes in the entire transcriptome rather than measuring the transcript levels of individual genes. These studies have demonstrated very consistent trends, even though O$_3$ exposure conditions (concentration, duration of exposure), plant species and sampling times vary significantly. Genes involved in plant defense, signaling, and those associated with the synthesis of plant hormones and secondary metabolism are generally upregulated in plants exposed to O$_3$, while those related to photosynthesis and general metabolism are typically downregulated. Proteome studies support these results by demonstrating concomitant increases or decreases in the proteins encoded by these genes. Transcriptome analysis has also illuminated the complex interactions that exist between several different phytohormones and how
they modulate plant sensitivity and response to O₃. Experimental evidence suggests
that while ethylene and salicylic acid are needed to develop O₃-induced leaf lesions,
jasmonic acid acts antagonistically to ethylene and salicylic acid to limit the spread
of the lesions. Abscisic acid, in addition to its role in regulating stomatal aperture,
may also act antagonistically to the jasmonic acid signaling pathway. Changes in the
quantity and activity of these phytohormones and the interactions between them
reveal some of the complexity of plant responses to an oxidativestressor such as O₃.

Another critical area of interest is to better understand and quantify the capacity of
the plant to detoxify oxygen radicals using antioxidant metabolites, such as ascorbate
and glutathione, and the enzymes that regenerate them. Ascorbate remains an
important focus of research, and, due to its location in the apoplast in addition to
other cellular compartments, it is regarded as a first line of defense against oxygen
radicals formed in the apoplast. Most studies demonstrate that antioxidant
metabolites and enzymes increase in quantity and activity in plants exposed to O₃,
indicating that they play an important role in protecting plants from oxidative stress.
However, attempts to quantify the detoxification capacity of plants have remained
unsuccessful, as high quantities of antioxidant metabolites and enzymes do not
always translate into greater protection of the plant. Considerable variation exists
between plant species, different developmental stages, and the environmental and O₃
exposure conditions which plants are exposed to.

As indicated earlier, the described alterations in transcript levels of genes correlate
with observed changes quantity and activity of the enzymes and metabolites involved
in primary and secondary metabolism. In addition to the generalized upregulation of
the antioxidant defense system, photosynthesis typically declines in O₃ treated
plants. Declines in C fixation due to reductions in quantity and activity of Rubisco
were extensively described in the 2006 O₃ AQCD. More recent studies support these
results and indicate that declines in Rubisco activity may also result from reductions
in Rubisco activase enzyme quantity. Other studies, which have focused on the light
reactions of photosynthesis, demonstrate that plant exposure to O₃ results in declines
in electron transport efficiency and a decreased capacity to quench oxidizing radicals.
Therefore, the overall declines in photosynthesis observed in O₃-treated plants likely
result from combined impacts on stomatal conductance, carbon fixation and the light
reactions. While photosynthesis generally declines in plants exposed to O₃, catabolic
pathways such as respiration have been shown to increase. It has been hypothesized
that increased respiration may result from greater energy needs for defense and
repair. Secondary metabolism is generally upregulated in a variety of species
exposed to O₃ as a part of a generalized plant defense mechanism. Some secondary
metabolites, such as flavonoids and polyamines, are of particular interest as they are
known to have antioxidant properties. The combination of decreases in
C assimilation and increases in catabolism and the production of secondary
metabolites would negatively impact plants by decreasing the energy available for
growth and reproduction.
9.4 Nature of Effects on Vegetation and Ecosystems

9.4.1 Introduction

Ambient O₃ concentrations have long been known to cause visible symptoms, decreases in photosynthetic rates, decreases in growth and yield of plants as well as many other effects on ecosystems (U.S. EPA, 2006b, 1996c, 1986, 1978a). Numerous studies have related O₃ exposure to plant responses, with most effort focused on the yield of crops and the growth of tree seedlings. Many experiments exposed individual plants grown in pots or soil under controlled conditions to known concentrations of O₃ for a segment of daylight hours for some portion of the plant’s life span. Information in this section also goes beyond individual plant-scale responses to consider effects at the broader ecosystem scale, including effects related to ecosystem services.

This section will focus mainly on studies published since the release of the 2006 O₃ AQCD. However, because much O₃ research was conducted prior to the 2006 O₃ AQCD, the present discussion of vegetation and ecosystem response to O₃ exposure is largely based on the conclusions of the 1978, 1986, 1996, and 2006 O₃ AQCDs.

9.4.1.1 Ecosystem Scale, Function, and Structure

Information presented in this section was collected at multiple spatial scales or levels of biological organization, ranging from the physiology of a given species to population, community, and ecosystem investigations. An ecological population is a group of individuals of the same species and a community is an assemblage of populations of different species interacting with one another that inhabit an area. For this assessment, “ecosystem” is defined as the interactive system formed from all living organisms and their abiotic (physical and chemical) environment within a given area (IPCC, 2007a). The boundaries of what could be called an ecosystem are somewhat arbitrary, depending on the focus of interest or study. Thus, the extent of an ecosystem may range from very small spatial scales or levels of biological organization to, ultimately, the entire Earth (IPCC, 2007a). All ecosystems, regardless of size or complexity, have interactions and physical exchanges between biota and abiotic factors, this includes both structural (e.g., soil type and food web trophic levels) and functional (e.g., energy flow, decomposition, nitrification) attributes.

Ecosystems can be described, in part, by their structure, i.e., the number and type of species present. Structure may refer to a variety of measurements including the species richness, abundance, community composition and biodiversity as well as landscape attributes. Competition among and within species and their tolerance to environmental stressors are key elements of survivorship. When environmental conditions are shifted, for example, by the presence of anthropogenic air pollution,
these competitive relationships may change and tolerance to stress may be exceeded. Ecosystems may also be defined on a functional basis. “Function” refers to the suite of processes and interactions among the ecosystem components and their environment that involve nutrient and energy flow as well as other attributes including water dynamics and the flux of trace gases. Plants, via such processes as photosynthesis, respiration, C allocation, nutrient uptake and evaporation, affect energy flow, C, nutrient cycling and water cycling. The energy accumulated and stored by vegetation (via photosynthetic C capture) is available to other organisms. Energy moves from one organism to another through food webs, until it is ultimately released as heat. Nutrients and water can be recycled. Air pollution alters the function of ecosystems when elemental cycles or the energy flow are altered. This alteration can also be manifested in changes in the biotic composition of ecosystems.

There are at least three levels of ecosystem response to pollutants: (1) the individual organism and its environment; (2) the population and its environment; and (3) the biological community composed of many species and their environment (Billings, 1978). Individual organisms within a population vary in their ability to withstand the stress of environmental change. The response of individual organisms within a population is based on their genetic constitution, stage of growth at time of exposure to stress, and the microhabitat in which they are growing (Levine and Pinto, 1998).

The stress range within which organisms can exist and function determines the ability of the population to survive.

9.4.1.2 Ecosystem Services

Ecosystem structure and function may be translated into ecosystem services. Ecosystem services are the benefits people obtain from ecosystems (UNEP, 2003). Ecosystems provide many goods and services that are of vital importance for the functioning of the biosphere and provide the basis for the delivery of tangible benefits to human society. Hassan et al. (2005) define these benefits to include supporting, provisioning, regulating, and cultural services:

- Supporting services are necessary for the production of all other ecosystem services. Some examples include biomass production, production of atmospheric O₂, soil formation and retention, nutrient cycling, water cycling, and provisioning of habitat. Biodiversity is a supporting service that is increasingly recognized to sustain many of the goods and services that humans enjoy from ecosystems. These provide a basis for three higher-level categories of services.

- Provisioning services, such as products (Gitay et al., 2001), i.e., food (including game, roots, seeds, nuts and other fruit, spices, fodder), water, fiber (including wood, textiles), and medicinal and cosmetic products (such as aromatic plants, pigments).

- Regulating services that are of paramount importance for human society such as (1) C sequestration, (2) climate and water regulation, (3) protection from
natural hazards such as floods, avalanches, or rock-fall, (4) water and air purification, and (5) disease and pest regulation.

- Cultural services that satisfy human spiritual and aesthetic appreciation of ecosystems and their components including recreational and other nonmaterial benefits.

In the sections that follow, available information on individual, population and community response to O₃ will be discussed. Effects of O₃ on productivity and C sequestration, water cycling, below-ground processes, competition and biodiversity, and insects and wildlife are considered below and in the context of ecosystem services where appropriate.

### 9.4.2 Visible Foliar Injury and Biomonitoring

Visible foliar injury resulting from exposure to O₃ has been well characterized and documented over several decades on many tree, shrub, herbaceous, and crop species (U.S. EPA, 2006b, 1996b, 1984, 1978a). Visible foliar injury symptoms are considered diagnostic as they have been verified experimentally in exposure-response studies, using exposure methodologies such as CSTRs, OTCs, and free-air fumigation (see Section 9.2 for more detail on exposure methodologies). Several pictorial atlases and guides have been published, providing details on diagnosis and identification of O₃-induced visible foliar injury on many plant species throughout North America (Flagler, 1998; NAPAP, 1987) and Europe (Innes et al., 2001; Sánchez et al., 2001). Typical visible injury symptoms on broad-leaved plants include: stippling, flecking, surface bleaching, bifacial necrosis, pigmentation (e.g., bronzing), chlorosis, and/or premature senescence. Typical visible injury symptoms for conifers include: chlorotic banding, tip burn, flecking, chlorotic mottling, and/or premature senescence of needles. Although common patterns of injury develop within a species, these foliar lesions can vary considerably between and within taxonomic groups. Furthermore, the degree and extent of visible foliar injury development varies from year to year and site to site (Smith, 2012; Orendovici-Best et al., 2008; Chappelka et al., 2007; Smith et al., 2003), even among co-members of a population exposed to similar O₃ levels, due to the influence of co-occurring environmental and genetic factors (Souza et al., 2006; Chappelka et al., 2003; Somers et al., 1998). Nevertheless, Chappelka et al. (2007) reported that the average incidence of O₃-induced foliar injury was 73% on milkweed observed in the Great Smoky Mountains National Park in the years 1992-1996.

Although the majority of O₃-induced visible foliar injury occurrence has been observed on seedlings and small plants, many studies have reported visible injury of mature coniferous trees, primarily in the western U.S. (Arbaugh et al., 1998) and to mature deciduous trees in eastern North America (Schaub et al., 2005; Vollenweider et al., 2003; Chappelka et al., 1999a; Chappelka et al., 1999b; Somers et al., 1998; Hildebrand et al., 1996).
It is important to note that visible foliar injury occurs only when sensitive plants are exposed to elevated O₃ concentrations in a predisposing environment. A major modifying factor for O₃-induced visible foliar injury is the amount of soil moisture available to a plant during the year that the visible foliar injury is being assessed. This is because lack of soil moisture generally decreases stomatal conductance of plants and, therefore, limits the amount of O₃ entering the leaf that can cause injury (Matyssek et al., 2006; Panek, 2004; Grulke et al., 2003a; Panek and Goldstein, 2001; Temple et al., 1992; Temple et al., 1988). Consequently, many studies have shown that dry periods in local areas tend to decrease the incidence and severity of O₃-induced visible foliar injury; therefore, the incidence of visible foliar injury is not always higher in years and areas with higher O₃, especially with co-occurring drought (Smith, 2012; Smith et al., 2003). Other factors such as leaf age influence the severity of symptom expression with older leaves showing greater injury severity as a result of greater seasonal exposure (Zhang et al., 2010a).

Although visible injury is a valuable indicator of the presence of phytotoxic concentrations of O₃ in ambient air, it is not always a reliable indicator of other negative effects on vegetation. The significance of O₃ injury at the leaf and whole plant levels depends on how much of the total leaf area of the plant has been affected, as well as the plant’s age, size, developmental stage, and degree of functional redundancy among the existing leaf area. Previous O₃ AQCDs have noted the difficulty in relating visible foliar injury symptoms to other vegetation effects such as individual plant growth, stand growth, or ecosystem characteristics (U.S. EPA, 2006b, 1996b). As a result, it is not presently possible to determine, with consistency across species and environments, what degree of injury at the leaf level has significance to the vigor of the whole plant. However, in some cases, visible foliar symptoms have been correlated with decreased vegetative growth (Somers et al., 1998; Karnosky et al., 1996; Peterson et al., 1987; Benoit et al., 1982) and with impaired reproductive function (Chappelka, 2002; Black et al., 2000). Conversely, the lack of visible injury does not always indicate a lack of phytotoxic concentrations of O₃ or a lack of non-visible O₃ effects (Gregg et al., 2006, 2003).

### 9.4.2.1 Biomonitoring

The use of biological indicators to detect phytotoxic levels of O₃ is a longstanding and effective methodology (Chappelka and Samuelson, 1998; Manning and Krupa, 1992). A plant bioindicator can be defined as a vascular or nonvascular plant exhibiting a typical and verifiable response when exposed to a plant stress such as an air pollutant (Manning, 2003). To be considered a good indicator species, plants must (1) exhibit a distinct, verified response; (2) have few or no confounding disease or pest problems; and (3) exhibit genetic stability (U.S. EPA, 2006b). Such sensitive plants can be used to detect the presence of a specific air pollutant such as O₃ in the ambient air at a specific location or region and, as a result of the magnitude of their response, provide unique information regarding specific ambient air quality. Bioindicators can be either introduced sentinels, such as the widely used tobacco...
(Nicotiana tabacum) variety Bel W3 (Calatayud et al., 2007b; Laffray et al., 2007; Nali et al., 2007; Gombert et al., 2006; Kostka-Rick and Hahn, 2005; Heggestad, 1991) or detectors, which are sensitive native plant species (Chappelka et al., 2007; Souza et al., 2006). The approach is especially useful in areas where O$_3$ monitors are not operated (Manning, 2003). For example, in remote wilderness areas where instrument monitoring is generally not available, the use of bioindicator surveys in conjunction with the use of passive samplers (Krupa et al., 2001) may be a useful methodology (Manning, 2003). However, it requires expertise in recognizing those signs and symptoms uniquely attributable to exposure to O$_3$ as well as in their quantitative assessment.

Since the 2006 O$_3$ AQCD, new sensitive plant species have been identified from field surveys and verified in controlled exposure studies (Kline et al., 2009; Kline et al., 2008). Several multiple-year field surveys have also been conducted at National Wildlife Refuges in Maine, Michigan, New Jersey, and South Carolina (Davis, 2009, 2007a, b; Davis and Orendovici, 2006).

The USDA Forest Service through the Forest Health Monitoring Program (FHM) (1990 - 2001) and currently the Forest Inventory and Analysis (FIA) Program has been collecting data regarding the incidence and severity of visible foliar injury on a variety of O$_3$ sensitive plant species throughout the U.S. (Smith, 2012; Coulston et al., 2003; Smith et al., 2003). The plots where these data are taken are known as biosites. These biosites are located throughout the country and analysis of visible foliar injury within these sites follows a set of established protocols. For more details, see http://www.nrs.fs.fed.us/fia/topics/ozone/ (USDA, 2011). The network has provided evidence of O$_3$ concentrations high enough to induce visible symptoms on sensitive vegetation. From repeated observations and measurements made over a number of years, specific patterns of areas experiencing visible O$_3$ injury symptoms can be identified. (Coulston et al., 2003) used information gathered over a 6-year period (1994-1999) from the network to identify several species that were sensitive to O$_3$ over entire regions, including sweetgum (Liquidambar styraciflua), loblolly pine (Pinus taeda), and black cherry (P. serotina). A recent paper by Smith et al. (2012) reported trends in foliar O$_3$ injury in the northeast and north central U.S. within the biomonitoring network over a 16-year period (1994-2009). The results showed that incidence and severity of foliar injury were dependent upon local site conditions (i.e., soil moisture availability) and O$_3$ exposure. Overall, there was a declining trend in the incidence of foliar injury as peak O$_3$ concentrations declined. Nevertheless, moderate O$_3$ exposures continued to cause foliar injury at sites throughout the region.


A study by Kohut et al. (2007) assessed the estimated risk of O$_3$-induced visible foliar injury on bioindicator plants (NPS, 2006) in 244 national parks in support of the National Park Service’s Vital Signs Monitoring Network (NPS, 2007). The risk assessment was based on a simple model relating response to the interaction of species, level of O$_3$ exposure, and exposure environment. Kohut et al. (2007)
concluded that the estimated risk of visible foliar injury was high in 65 parks (27%), moderate in 46 parks (19%), and low in 131 parks (54%). Some of the well-known parks with a high risk of O$_3$-induced visible foliar injury include Gettysburg, Valley Forge, Delaware Water Gap, Cape Cod, Fire Island, Antietam, Harpers Ferry, Manassas, Wolf Trap Farm Park, Mammoth Cave, Shiloh, Sleeping Bear Dunes, Great Smoky Mountains, Joshua Tree, Sequoia and Kings Canyon, and Yosemite.

Lichens have also long been used as biomonitors of air pollution effects on forest health (Nash, 2008). It has been suspected, based on field surveys in the San Bernardino Mountains surrounding the Los Angeles air basin, that declines in lichen diversity and abundance were correlated with measured O$_3$ gradients (Gül et al., 2011). Several recent studies in North America (Geiser and Neitlich, 2007; Gombert et al., 2006; Jovan and McCune, 2006) and Europe (Nali et al., 2007; Gombert et al., 2006) have used lichens as biomonitors of atmospheric deposition (e.g., N and S) and O$_3$ exposure. Nali et al. (2007) found that epiphytic lichen biodiversity was not related to O$_3$ geographical distribution. In addition, a recent study by Riddell et al. (2010) found that lichen species, Ramalina menziesii, showed no decline in physiological response to low and moderate concentrations of O$_3$ and may not be a good indicator for O$_3$ pollution. Mosses have also been used as biomonitors of air pollution; however, there remains a knowledge gap in the understanding of the effects of O$_3$ on mosses as there has been very little information available on this topic in recent years.

9.4.2.2 Summary

Visible foliar injury resulting from exposure to O$_3$ has been well characterized and documented over several decades of research on many tree, shrub, herbaceous, and crop species (U.S. EPA, 2006b, 1996b, 1984, 1978a). Ozone-induced visible foliar injury symptoms on certain bioindicator plant species are considered diagnostic as they have been verified experimentally in exposure-response studies, using exposure methodologies such as continuous stirred tank reactors (CSTRs), OTCs, and free-air fumigation. Experimental evidence has clearly established a consistent association of visible injury with O$_3$ exposure, with greater exposure often resulting in greater and more prevalent injury. Since the 2006 O$_3$ AQCD, results of several multi-year field surveys of O$_3$-induced visible foliar injury at National Wildlife Refuges in Maine, Michigan, New Jersey, and South Carolina have been published. New sensitive species showing visible foliar injury continue to be identified from field surveys and verified in controlled exposure studies.

The use of biological indicators in field surveys to detect phytotoxic levels of O$_3$ is a longstanding and effective methodology. The USDA Forest Service through the Forest Health Monitoring (FHM) Program (1990-2001) and currently the Forest Inventory and Analysis (FIA) Program has been collecting data regarding the incidence and severity of visible foliar injury on a variety of O$_3$ sensitive plant species throughout the United States. The network has provided evidence that O$_3$
concentrations were high enough to induce visible symptoms on sensitive vegetation. From repeated observations and measurements made over a number of years, specific patterns of areas experiencing visible O₃ injury symptoms can be identified. As noted in the preceding section, a study of 244 national parks indicated that the estimated risk of visible foliar injury was high in 65 parks (27%), moderate in 46 parks (19%), and low in 131 parks (54%).

Evidence is sufficient to conclude that there is a causal relationship between ambient O₃ exposure and the occurrence of O₃-induced visible foliar injury on sensitive vegetation across the U.S.

9.4.3 Growth, Productivity and Carbon Storage in Natural Ecosystems

Ambient O₃ concentrations have long been known to cause decreases in photosynthetic rates, decreases in growth, and decreases in yield (U.S. EPA, 2006b, 1996c, 1986, 1978a). The O₃-induced damages at the plant scale may translate to damages at the stand, then ecosystem scales, and cause changes in productivity and C storage. This section focuses on the responses of C cycling to seasonal or multi-year exposures to O₃ at levels of organization ranging from individual plants to ecosystems. Quantitative responses include changes in plant growth, plant biomass allocation, ecosystem production and ecosystem C sequestration. Most information available on plant-scale responses was obtained from studies that used a single species, especially tree seedlings and crops, while some used mixtures of herbaceous species. Ecosystem changes are difficult to evaluate in natural settings, due to the complexity of interactions, the number of potential confounders, and the large spatial and temporal scales. The discussion of ecosystem effects focuses on new studies at the large-scale FACE experiments and on ecological model simulations.

9.4.3.1 Plant Growth and Biomass Allocation

The previous O₃ AQCDs concluded that there is strong evidence that exposure to O₃ decreases photosynthesis and growth in numerous plant species (U.S. EPA, 2006b, 1996b, 1984, 1978a). Studies published since the last review support those conclusions and are summarized below.

In general, research conducted over several decades has indicated that exposure to O₃ alters stomatal conductance and reduces photosynthesis in a wide variety of plant species. In a review of more than 55 studies, Wittig et al. (2007) reported that current O₃ concentrations in the northern hemisphere are decreasing stomatal conductance (13%) and photosynthesis (11%) across tree species. It was also found that younger trees (<4 years) were affected less by O₃ than older trees. Further, the authors also found that decreases in photosynthesis are consistent with the cumulative uptake of O₃ into the leaf. In contrast, several studies reported that O₃ exposure may result in loss of stomatal control, incomplete stomatal closure at night and a decoupling of
photosynthesis and stomatal conductance, which may have implications for whole-
plant water use (Section 9.4.5).

In a recently published meta-analysis, Wittig et al. (2009) quantitatively compiled
peer reviewed studies from the past 40 years on the effect of current and future O₃
exposures on the physiology and growth of forest species. They found that current
ambient O₃ concentrations as reported in those studies significantly decreased annual
total biomass growth (7%) across 263 studies. The authors calculated the ambient O₃
concentrations across these studies to average 40 ppb. This average was calculated
across the duration of each study and there were therefore many hourly exposures
well above 40 ppb. The decreased growth effect was reported to be greater (11 to
17%) in elevated O₃ exposures (97 ppb) (Wittig et al., 2009). This meta-analysis
demonstrates the coherence of O₃ effects across numerous studies and species that
used a variety of experimental techniques, and these results support the conclusion of
the previous AQCD that exposure to O₃ decreases plant growth.

In two companion papers, McLaughlin et al. (2007a; 2007b) investigated the effects of
ambient O₃ on tree growth and hydrology at forest sites in the southern Appalachian
Mountains. The authors reported that the cumulative effects of ambient levels of O₃
decreased seasonal stem growth by 30-50% for most tree species in a high O₃ year in
comparison to a low O₃ year (McLaughlin et al., 2007a). The authors also reported
that high ambient O₃ concentrations can increase whole-tree water use and in turn
reduce late-season streamflow (McLaughlin et al., 2007b); see Section 9.4.5 for more
on water cycling.

Since the 2006 O₃ AQCD, several recent studies have reported results from the
Aspen FACE “free air” O₃ and CO₂ exposure experiment in Wisconsin (Darbah et
al., 2008; Riikonen et al., 2008; Darbah et al., 2007; Kubiske et al., 2007; Kubiske et
al., 2006; King et al., 2005). At the Aspen FACE site, single-species and two-species
stands of trees were grown in 12, 30-m diameter rings corresponding to three
replications of a full factorial arrangement of two levels each of CO₂ and O₃
exposure. Over the first seven years of stand development, Kubiske et al. (2006)
observed that elevated O₃ decreased tree heights, diameters, and main stem volumes
in the aspen community by 11, 16, and 20%, respectively. In addition, Kubiske et al.
(2007) reported that elevated O₃ may change intra- and inter-species competition.
For example, O₃ treatments increased the rate of conversion from a mixed aspen-
birch community to a birch dominated community. In a comparison presented in
Section 9.6.3 of this document, EPA found that effects on biomass accumulation in
aspen during the first seven years closely agreed with the exposure-response function
based on data from earlier OTC experiments.

Several studies at the Aspen FACE site also considered other growth-related effects
of elevated O₃. Darbah et al. (2008; 2007) reported that O₃ treatments decreased
paper birch seed weight and seed germination and that this would likely lead to a
negative impact of regeneration for that species. Riikonen et al. (2008) found that
elevated O₃ decreased the amount of starch in birch buds by 16%, and reduced aspen
bud size, which may have been related to the observed delay in spring leaf
development. The results suggest that elevated O₃ concentrations have the potential
to alter C metabolism of overwintering buds, which may have carry-over effects in
the subsequent growing season (Riikonen et al., 2008).

Effects on growth of understory vegetation were also investigated at Aspen FACE.
Bandeff et al. (2006) found that the effects of elevated CO₂ and O₃ on understory
species composition, total and individual species biomass, N content, and ¹⁵N
recovery were a result of overstory community responses to those treatments;
however, the lack of apparent direct O₃ treatment effects may have been due to high
variability in the data. Total understory biomass increased with increasing light and
was greatest under the open canopy of the aspen/maple community, as well as the
more open canopy of the elevated O₃ treatments (Bandeff et al., 2006). Similarly,
data from a study by Awmack et al. (2007) suggest that elevated CO₂ and O₃ may
have indirect growth effects on red (Trifolium pratense) and white (Trifolium repens)
clover in the understory via overstory community effects; however, no direct effects
of elevated O₃ were observed.

Overall, the studies at the Aspen FACE experiment are consistent with many of the
OTC studies that were evaluated in previous O₃ AQCDs demonstrating that O₃
exposure decreases growth in numerous plant species. These results strengthen the
understanding of O₃ effects on forests and demonstrate the relevance of the
knowledge gained from trees grown in open-top chamber studies.

For some annual species, particularly crops, the relevant measurement for an
assessment of the risk of O₃ exposure is yield or growth, e.g., production of grain or
biomass. For plants grown in mixtures such as hayfields, and natural or semi-natural
grasslands (including native nonagricultural species), affected factors other than
production of biomass may be important. Such endpoints include biodiversity or
species composition, and effects on those endpoints may be indirect, resulting, for
example, from competitive interactions among plants in mixed-species communities.
Most of the available data on non-crop herbaceous species are for grasslands, with
many of the recent studies conducted in Europe. See Section 9.4.7.2 for a review of
the recent literature on O₃ effects on competition and biodiversity in grasslands.

**Root growth**

Although O₃ does not penetrate soil, it could alter root development by decreasing
C assimilation via photosynthesis leading to less C allocation to the roots (Andersen,
2003). The response of root development to O₃ exposure depends on available
photosynthate within the plant and could vary over time. Many biotic and abiotic
factors, such as community dynamics and drought stress, have been found to alter
root development under elevated O₃. Generally, there is clear evidence that O₃
reduces C allocation to roots; however, results of a few recent individual studies have
shown negative (Jones et al., 2010), non-significant (Andersen et al., 2010; Phillips
et al., 2009) and positive effects (Pregitzer et al., 2008; Grebenc and Kraigher, 2007)
on root biomass and root: shoot ratio.
An earlier study at the Aspen FACE experiment found that elevated O\textsubscript{3} reduced coarse root and fine roots biomass in young stands of paper birch and trembling aspen (King et al., 2001). However, this reduction disappeared several years later. Ozone significantly increased fine-root production (<1.0 mm) in the aspen community (Pregitzer et al., 2008). This increase in fine root production was due to changes in community composition, such as better survival of the O\textsubscript{3}-tolerant aspen genotype, birch, and maple, rather than changes in C allocation at the individual tree level (Pregitzer et al., 2008; Zak et al., 2007). In an adult European beech/Norway spruce forest in Germany, drought was found to nullify the O\textsubscript{3}-driven stimulation of fine root growth. Ozone stimulated fine-root production of beech during the humid year, but had no significant impact on fine root production in the dry year (Matyssek et al., 2010; Nikolova et al., 2010).

Using a non-destructive method, Vollnes et al. (2010) studied the in vivo root development of subterranean clover (Trifolium subterraneum) before, during and after short-term O\textsubscript{3} exposure. It was found that O\textsubscript{3} reduced root tip formation, root elongation, the total root length, and the ratios between below- and above-ground growth within one week after exposure. Those effects persisted for up to three weeks; however, biomass and biomass ratios were not significantly altered at the harvest five weeks after exposure.

Several recent meta-analyses have generally indicated that O\textsubscript{3} reduced C allocated to roots. In one meta-analysis, Grantz et al. (2006) estimated the effect of O\textsubscript{3} on the root:shoot allometric coefficient (k), the ratio between the relative growth rate of the root and shoot. The results showed that O\textsubscript{3} reduced the root:shoot allometric coefficient by 5.6%, and the largest decline of the root:shoot allometric coefficient was observed in slow-growing plants. In another meta-analysis including 263 publications, Wittig et al. (2009) found that current O\textsubscript{3} exposure had no significant impacts on root biomass and root:shoot ratio when compared to pre-industrial O\textsubscript{3} exposure. However, if O\textsubscript{3} concentrations rose to 81-101 ppb (projected O\textsubscript{3} levels in 2100), both root biomass and root:shoot ratio were found to significantly decrease. Gymnosperms and angiosperms differed in their responses, with gymnosperms being less sensitive to elevated O\textsubscript{3}. In two other meta-analyses, Wang et al. (2010) found elevated O\textsubscript{3} reduced biomass allocation to roots by 8.3% at ambient CO\textsubscript{2} and 6.0% at elevated CO\textsubscript{2}, and Morgan et al. (2003) found O\textsubscript{3} reduced root dry weight of soybean.

### 9.4.3.2 Summary

The previous O\textsubscript{3} AQCDs concluded that there is strong and consistent evidence that ambient concentrations of O\textsubscript{3} decrease photosynthesis and growth in numerous plant species across the United States. Studies published since the last review continue to support that conclusion.

The meta-analyses by Wittig et al. (2009; 2007) demonstrate the coherence of O\textsubscript{3} effects on plant photosynthesis and growth across numerous studies and species.
using a variety of experimental techniques. Furthermore, recent meta-analyses have generally indicated that O₃ reduced C allocation to roots (Wittig et al., 2009; Grantz et al., 2006). Since the 2006 O₃ AQCD, several studies were published based on the Aspen FACE experiment using “free air,” O₃, and CO₂ exposures in a planted forest in Wisconsin. Overall, the studies at the Aspen FACE experiment were consistent with many of the open-top chamber (OTC) studies that were the foundation of previous O₃ NAAQS reviews. These results strengthen the understanding of O₃ effects on forests and demonstrate the relevance of the knowledge gained from trees grown in open-top chamber studies.

Evidence is sufficient to conclude that there is a causal relationship between ambient O₃ exposure and reduced growth of native woody and herbaceous vegetation.

### 9.4.3.3 Reproduction

Studies during recent decades have demonstrated O₃ effects on various stages of plant reproduction. The impacts of O₃ on reproductive development, as reviewed by Black et al. (2000), can occur by influencing (1) age at which flowering occurs, particularly in long-lived trees that often have long juvenile periods of early growth without flower and seed production; (2) flower bud initiation and development; (3) pollen germination and pollen tube growth; (4) seed, fruit, or cone yields; and (5) seed quality (Table 9-1) (U.S. EPA, 2006b). Several recent studies since the 2006 O₃ AQCD further demonstrate the effects of O₃ on reproductive processes in herbaceous and woody plant species. Although there have been documented effects of O₃ on reproductive processes, a knowledge gap still exists pertaining to the exact mechanism of these responses.

Ramo et al. (2007) exposed several meadow species to elevated O₃ (40-50 ppb) and CO₂ (+100 ppm), both individually and combined, over three growing seasons in ground-planted mesocosms, using OTCs. Elevated O₃ delayed flowering of *Campanula rotundifolia* and *Vicia cracca*. Ozone also reduced the overall number of produced flowers and decreased fresh weight of individual *Fragaria vesca* berries.

Black et al. (2007) exposed *Brassica campestris* to 70 ppb for two days during late vegetative growth or ten days during most of the vegetative phase. The two-day exposure had no effect on growth or reproductive characteristics, while the 10 day exposure reduced vegetative growth and reproductive site number on the terminal raceme, emphasizing the importance of exposure duration and timing. Mature seed number and weight per pod were unaffected due to reduced seed abortion, suggesting that, although O₃ affected reproductive processes, indeterminate species such as *B. campestris* possess enough compensatory flexibility to avoid reduced seed production (Black et al., 2007).

In the determinate species, *Plantago major*, Black et al. (2010) found that O₃ may have direct effects on reproductive development in populations of differing
sensitivity. Only the first flowering spike was exposed to 120 ppb O₃ for 7 hours per
day on 9 successive days (corresponding to flower development) while the leaves
and second spike were exposed to charcoal-filtered air. Exposure of the first spike to
O₃ affected seed number per capsule on both spikes even though spike two was not
exposed. The combined seed weight of spikes one and two was increased by 19% in
the two resistant populations, suggesting an overcompensation for injury, whereas, a
decrease of 21% was observed in the most sensitive population (Black et al., 2010).
The question remains as to whether these effects are true direct O₃-induced effects or
compensatory responses.

Studies by Darbah et al. (2008; 2007) of paper birch (Betula papyrifera) trees at the
Aspen FACE site in Rhinelander, WI investigated the effects of elevated O₃ and/or
CO₂ on reproductive fitness. Elevated O₃ increased flowering, but decreased seed
weight and germination success rate of seeds from the exposed trees. These results
suggest that O₃ can dramatically affect flowering, seed production, and seed quality
of paper birch, ultimately affecting its reproductive fitness (Darbah et al., 2008;
Darbah et al., 2007).

Table 9-1  Ozone effects on plant reproductive processes.

<table>
<thead>
<tr>
<th>Species</th>
<th>Condition Measures</th>
<th>References</th>
</tr>
</thead>
<tbody>
<tr>
<td>Apocynum androsaemifolium</td>
<td>Flowering time</td>
<td>Bergweiler et al. (1999)</td>
</tr>
<tr>
<td>(spreading dogbane)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Buddleia davidii</td>
<td>Flowering time</td>
<td>Findley et al. (1997)</td>
</tr>
<tr>
<td>(butterfly bush)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rubus cuneifolius</td>
<td>Pollen germination</td>
<td>Chappelka et al. (2002)</td>
</tr>
<tr>
<td>(sand blackberry)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Plantago major (plantain)</td>
<td>Pollen tube elongation</td>
<td>Stewart et al. (1998)</td>
</tr>
<tr>
<td>Fragaria × ananassa</td>
<td>Fruit yield</td>
<td>Drogoudi and Ashmore (2001; 2000)</td>
</tr>
<tr>
<td>(cultivated strawberry)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Plantago major (plantain)</td>
<td>Seed yield</td>
<td>Lyons and Barnes (1998); Pearson et al. (1996); Reiling and Davison (1992); Whittfield et al. (1997)</td>
</tr>
<tr>
<td>Understory herbs</td>
<td>Seed yield</td>
<td>Harward and Treshow (1975)</td>
</tr>
</tbody>
</table>

Source: Derived from Table AX9-22 of the 2006 O₃ AQCD.

9.4.3.4  Ecosystem Productivity and Carbon Sequestration

During the previous NAAQS review, there were limited studies that investigated the
effect of O₃ exposure on ecosystem productivity and C sequestration. Recent studies
from long-term FACE experiments provide more evidence of the association of O₃
exposure and changes in productivity at the ecosystem level of organization.
In addition to experimental studies, model studies also assessed the impact of O₃
exposure on productivity and C sequestration from stand to global scales.
In this section productivity of ecosystems is expressed in different ways depending on the model or the measurements of a study. The most common metric of productivity is Gross Primary Productivity. Gross Primary Productivity (GPP) is total carbon that enters the ecosystem through photosynthesis by plants. Plants return a larger portion of this carbon back to the atmosphere through respiration from roots and aboveground portions of plants ($R_{plant}$). Net primary production (NPP) is the difference between total carbon gain (GPP) and carbon loss through $R_{plant}$. Net ecosystem production (NEP) is the difference between NPP and carbon loss through heterotrophic respiration ($R_{het}$) (mostly decomposition of dead organic matter) (Lambers et al., 1998). Similarly net ecosystem exchange (NEE) is the net flux of carbon between the land and the atmosphere, typically measured using eddy covariance techniques. Positive values of NEE usually refer to carbon released to the atmosphere (i.e., a source), and negative values refer to carbon uptake (i.e., a sink). Other studies have calculated net carbon exchange (NCE). NCE is defined as NPP minus $R_{net}$, $E_c$ (the carbon emission during the conversion of natural ecosystems to agriculture) and $E_p$ (the sum of carbon emission from the decomposition of agricultural products). For natural vegetation, $E_c$ and $E_p$ are equal to 0, so NCE is equal NEP (Felzer et al., 2005). In general, modeling studies take into account the effect of O$_3$ on C fixation of a system and there is generally not an effect on $R_{plant}$, $R_{net}$, $E_c$ or $E_p$. Therefore, decreases in GPP, NPP, NEP, NEE and NCE indicate a general decrease in productivity of an ecosystem.

Two types of models are most often used to study the ecological consequences of O$_3$ exposure: (1) single plant growth models such as TREGRO (Tree Growth Model) and PnET–II (Photosynthetic EvapoTranspiration-II model) (Hogsett et al., 2008; Martin et al., 2001; Ollinger et al., 1997b), and (2) process-based ecosystem models such as PnET-CN, Dynamic Land Ecosystem Model (DLEM), Terrestrial Ecosystem Model (TEM), or Met Office Surface Exchange Scheme - Top-down Representation of Interactive Foliage and Flora Including Dynamics (MOSES-TRIFFID) (Felzer et al., 2009; Ren et al., 2007b; Sitch et al., 2007; Ollinger et al., 2002) (Table 9-2). In these models, carbon uptake is simulated through photosynthesis (TREGRO, PnET –II, PnET- CN, DLEM and MOSES-TRIFFID) or gross primary production (TEM). Photosynthesis rate at leaf level is modeled by a function of stomatal conductance and other parameters in TREGRO, PnET –II, PnET- CN, DLEM and MOSES-TRIFFID. Photosynthesis at canopy level is calculated by summing either photosynthesis of different leaf types (TREGRO, DLEM, and MOSES-TRIFFID) or photosynthesis of different canopy layers (PnET –II, PnET- CN). The detrimental effect of O$_3$ on plant growth is often simulated by multiplying photosynthesis rate by a coefficient that is dependent on stomatal conductance and cumulative O$_3$ uptake (Table 9-2). Different plant functional groups (PFGs, such as deciduous trees, coniferous trees or crops) show different responses to O$_3$ exposure. PnET-II, PnET-CN, TEM, DLEM and MOSES-TRIFFID estimate this difference by modifying net photosynthesis with coefficients that represent the O$_3$ induced fractional reduction of photosynthesis for each functional group. The coefficients used in PnET-II, PnET-CN, TEM, DLEM are derived from the functions of O$_3$ exposure (AOT40) versus photosynthesis reduction from Reich et al. (1987) and Tjoelker et al. (1995). The coefficients used in MOSES-TRIFFID are derived from the O$_3$ dose-
photosynthesis response function from Pleijel et al. (2004a) and Karlsson et al. (2004), where O₃ dose is estimated by a metric named CUOt (cumulative stomatal uptake of O₃). The O₃ threshold of CUOt is 1.6 nmol/m²/sec for woody PFT and 5 nmol/m²/sec for grass PFT, and is different from AOT40, which has an O₃ threshold level of 40 ppb for all PFTs. Experimental and model studies on ecosystem productivity and C sequestration at the forest stand scale as well as regional and global scales are reviewed in the following section.
Table 9-2  Comparison of models used to simulate the ecological consequences of O\textsubscript{3} exposure.

<table>
<thead>
<tr>
<th>Model</th>
<th>Model feature</th>
<th>Carbon uptake</th>
<th>Ozone effect</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>TREGRO</td>
<td>Hourly or daily step, single plant model simulating vegetation growth process</td>
<td>Leaf: leaf photosynthesis is a function of stomatal conductance, mesophyll conductance and the gradient of CO\textsubscript{2} from atmosphere to the mesophyll cells. Canopy: Leaf is divided into different ages. The canopy photosynthesis rate is the sum of the photosynthesis of all foliage groups.</td>
<td>The effect of O\textsubscript{3} on photosynthesis is simulated by reducing mesophyll conductance, and increasing respiration. The degree of O\textsubscript{3} damage is determined by ambient O\textsubscript{3} exposure, and the threshold O\textsubscript{3} concentration below which O\textsubscript{3} does not affect mesophyll conductance and respiration.</td>
<td>Hogsett et al. (2008); Weinstein et al. (2005); Tingey et al. (2004)</td>
</tr>
</tbody>
</table>
| PnET-II and PnET-CN | PnET-II: Monthly time-step, single plant model  
PnET-CN: Monthly time-step, ecosystem model   | Leaf: Maximum photosynthesis rate is determined by a function of foliar N concentration, and stomatal conductance is determined by a function of the actual rate of the photosynthesis. Canopy: canopy is divided into multiple, even-mass layers and photosynthesis is simulated by a multilayered canopy submodel. | The effect of O\textsubscript{3} on photosynthesis is simulated by an equation of stomatal conductance and O\textsubscript{3} dose (AOT40). The model assumes that photosynthesis and stomatal conductance remain coupled under O\textsubscript{3} exposure, with a reduction in photosynthesis for a given month causing a proportion reduction in stomatal conductance. | Ollinger et al. (2002; 1997b); Pan et al. (2009) |
<p>| TEM         | Monthly time-step, ecosystem model                                            | Ecosystem: TEM is run at a 0.5°0.5 degree resolution. Each grid cell is classified by vegetation type and soil texture, and vegetation and detritus are assumed to distribute homogeneously within grid cells. Carbon flows into ecosystem via gross primary production, which is a function of maximum rate of assimilation, photosynthetically active radiation, the leaf area relative to the maximum annual leaf area, mean monthly air temperate, and nitrogen availability. | The direct O\textsubscript{3} reduction on GPP is simulated by multiplying GPP by f(O\textsubscript{3})t, where f(O\textsubscript{3})t is determined by evapotranspiration, mean stomatal conductance, ambient AOT40, and O\textsubscript{3} response coefficient empirically derived from previous publications. | Felzer et al. (2005; 2004) |
| DLEM        | Daily time-step ecosystem model                                              | Leaf: photosynthesis is a function of 6 parameters: photosynthetic photon flux density, stomatal conductance, daytime temperature, the atmospheric CO\textsubscript{2} concentration, the leaf N content and the length of daytime. Canopy: Photosynthetic rates for sunlit leaf and shaded leaf scale up to the canopy level by multiplying the estimated leaf area index. Ecosystem: GPP is the sum of gross C fixation of different plant function groups. | The detrimental effect of O\textsubscript{3} is simulated by multiplying the rate of photosynthesis by O\textsubscript{3}eff, where O\textsubscript{3}eff is a function of stomatal conductance, ambient AOT40, and O\textsubscript{3} sensitive coefficient. Ozone’s indirect effect on stomatal conductance is also simulated, with a reduction in photosynthesis for a given month causing a reduction in stomatal conductance, and therefore canopy conductance. | Ren et al. (2007b; 2007a); Zhang et al. (2007a) |</p>
<table>
<thead>
<tr>
<th>Model</th>
<th>Model feature</th>
<th>Carbon uptake</th>
<th>Ozone effect</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>MOSES-TRIFFID</td>
<td>30 minute time-step, dynamic global vegetation model</td>
<td>Leaf: photosynthesis is a function of environmental and leaf parameters and stomatal conductance; Stomatal conductance is a function of the concentration of CO₂ and H₂O in air at the leaf surface and the current rate of photosynthesis of the leaf</td>
<td>The effect of O₃ is simulated by multiplying the rate of photosynthesis by F, where F depends upon stomatal conductance, O₃ exposure, a critical threshold for O₃ damage, and O₃ sensitive coefficient (functional type dependent)</td>
<td>Sitch et al. (2007)</td>
</tr>
</tbody>
</table>

**Local scale**

Both experimental and modeling studies have provided new information on effects of O₃ exposure at the stand or site level, i.e., at the local scale. The above- and below-ground biomass and net primary production (NPP) were measured at the Aspen FACE site after 7 years of O₃ exposure. Elevated O₃ caused 23, 13 and 14% reductions in total biomass relative to the control in the aspen, aspen–birch and aspen–maple communities, respectively (King et al., 2005). At the Kranzberg Forest FACE experiment in Germany, O₃ reduced annual volume growth by 9.5 m³/ha in a mixed mature stand of Norway spruce and European beech (Pretzsch et al., 2010).

At the grassland FACE experiment at Alp Flix, Switzerland, O₃ reduced the seasonal mean rates of ecosystem respiration and GPP by 8%, but had no significant impacts on aboveground dry matter productivity or growing season net ecosystem production (NEP) (Volk et al., 2011). Ozone also altered C accumulation and turnover in soil, as discussed in Section 9.4.6.

Changes in forest stand productivity under elevated O₃ were assessed by several model studies. TREGRO (Table 9-2) has been widely used to simulate the effects of O₃ on the growth of several species in different regions in the United States. Hogsett et al. (2008) used TREGRO to evaluate the effectiveness of various forms and levels of air quality standards for protecting tree growth in the San Bernardino Mountains of California. They found that O₃ exposures at the Crestline site resulted in a mean 20.9% biomass reduction from 1980 to 1985 and 10.3% biomass reduction from 1995 to 2000, compared to the “background” O₃ concentrations (O₃ concentration in Crook County, Oregon). The level of vegetation protection projected was different depending on the air quality scenarios under consideration. Specifically, when air quality was simulated to just meet the California 8-h average maximum of 70 ppb and the maximum 3 months 12-h SUM06 of 25 ppm-h, annual growth reductions were limited to 1% or less, while air quality that just met a previous NAAQS (the
2nd-highest 1-h max [125 ppb]) resulted in 6-7% annual reduction in growth, resulting in the least protection relative to background O₃ (Hogsett et al., 2008).

ZELIG is a forest succession gap model, and has been used to evaluate the dynamics of natural stand succession. Combining TREGRO with ZELIG, Weinstein et al. (2005) simulated the effects of different O₃ levels (0.5, 1.5, 1.75, and 2 times [×] ambient) on the growth and competitive interactions of white fir and ponderosa pine at three sites in California: Lassen National Park, Yosemite National Park, and Crestline. Their results suggested that O₃ had little impact on white fir, but greatly reduced the growth of ponderosa pine. If current O₃ concentrations continue over the next century, ambient O₃ exposure (SUM06 of 110 ppm-h) at Crestline was predicted to decrease individual tree C budget by 10% and decrease ponderosa pine abundance by 16%. Effects at Lassen National Park and Yosemite National Park sites were found to be smaller because of lower O₃ exposure levels (Weinstein et al., 2005).

To evaluate the influence of interspecies competition on O₃ effects, the linked TREGRO and ZELIG modeling system was used to predict the effects of O₃ over 100 years on the basal area of species in a Liriodendron tulipifera-dominated forest in the Great Smoky Mountains National Park (Weinstein et al., 2001). Ambient O₃ was predicted to decrease individual tree C budget by 28% and reduce the basal area of L. tulipifera by 10%, whereas a 1.5×-ambient exposure was predicted to cause a 42% decrease in the individual tree C budget and a 30% reduction in basal area. Individual tree C balance for Acer rubrum decreased 14% and 23% under ambient and 1.5×-ambient exposure, respectively. Prunus serotina was predicted to have less than a 2% decrease in tree C balance in all scenarios, but its basal area was greatly altered by the O₃ effects on the other tree species. Basal area of A. rubrum and P. serotina was predicted to increase for some years, but then decrease by up to 30%, depending on the scenario. The authors cautioned that the simulation results were heavily dependent on the assumption that only three of ten species studied could directly respond O₃ exposure and the rest of the species only indirectly responded through competitive interactions. Very different predictions of stand dynamics may have been simulated if more species could be parameterized to directly respond to O₃ exposure.

Some results from models that include competitive interactions between tree species, such as the linked TREGRO and ZELIG modeling system, may differ from empirical modeling based on short-term single-species O₃ exposure experiments. Single species experiments were often performed on tree seedlings for one to three years in open top chambers (OTCs) and indoor chambers (see Section 9.2). For example, OTC-based experiments that were used to create O₃ concentration-response relationships (discussed in Section 9.6.2) were the basis of estimated tree seedling biomass loss reported in studies by Hogsett et al. (1997) and in the 2007 EPA Staff Paper (2007b). These illustrative biomass loss analyses covered one or two years of O₃ exposure based on historical monitoring that was interpolated across regions of the United States. In contrast, competition models, such as the linked TREGRO and ZELIG modeling system, use empirical data to parameterize the simulation of growth.
from a seedling into mature trees in competition with other trees (Weinstein et al., 2005; Weinstein et al., 2001). These simulations may be run for 100 years or more and have modeled annual exposures O₃ across those years. Complicated competitive interactions emerge across many decades of the simulation. For example, long-term competition simulations can take into account competition for space, light, water, tree longevity, disturbance, shade tolerance as well as the differential effects of O₃ on each species. As a result, a particular species may appear to grow poorly under O₃ exposure in short-term seedling studies, but may grow relatively well under long-term model scenarios with competition added to the analysis. It is important to note that both of these approaches provide useful information about the long and short term affects of O₃ exposure on trees forest stands. However, it is very difficult to validate the results of the long-term simulation of the effects of O₃ on forest composition.

The effects of O₃ on stand productivity and dynamics were also studied by other tree growth or stand models, such as ECOPHYS, INTRASTAND and LINKAGES. ECOPHYS is a functional-structural tree growth model. The model used the linear relationship between the maximum capacity of carboxylation and O₃ dose to predict the relative effect of O₃ on leaf photosynthesis (Martin et al., 2001). Simulations with ECOPHYS found that O₃ decreased stem dry matter production, stem diameter and leaf dry matter production, induced earlier leaf abscission, and inhibited root growth (Martin et al., 2001). INTRASTAND is an hourly time step model for forest stand carbon and water budgets. LINKAGES is a monthly time step model simulating forest growth and community dynamics. Linking INTRASTAND with LINKAGES, Hanson et al. (2005) found that a simulated increase in O₃ concentration in 2100 (a mean 20-ppb increase over the current O₃ concentration) yields a 35% loss of carbon (C) in the net ecosystem exchange (NEE) with respect to the current conditions (174 g C/m²·year).

Regional and global scales

Since the publication of the 2006 O₃ AQCD, there is additional evidence suggesting that O₃ exposure alters ecosystem productivity and biogeochemical cycling at the regional scale, i.e., at scales ranging from watershed to subcontinental divisions, and at continental and global scales. Most of those studies were conducted by using process-based ecosystem models (Table 9-2) and are briefly reviewed in the following sections.

Ollinger et al. (1997a) simulated the effect of O₃ on hardwood forest productivity of 64 hardwood sites in the northeastern U.S. with PnET-II (Table 9-2). Their simulations indicated that O₃ caused a 3-16% reduction in NPP from 1987 to 1992 (Table 9-3). The interactive effects of O₃, N deposition, elevated CO₂ and land use history on C dynamics were estimated by PnET-CN (Table 9-2) (Ollinger et al., 2002). The results indicated that O₃ offset the increase in net C exchange caused by elevated CO₂ and N deposition by 13% (25.0 g C/m²·year) under agriculture site history, and 23% (33.6 g C/m²·year) under timber harvest site history. PnET-CN was
also used to assess changes in C sequestration of U.S. Mid-Atlantic temperate forest. Pan et al. (2009) designed a factorial modeling experiment to separate the effects of changes in atmospheric composition, historical climatic variability and land-disturbances on the C cycle. They found that O₃ acted as a negative factor, partially offsetting the growth stimulation caused by elevated CO₂ and N deposition in U.S. Mid-Atlantic temperate forest. Ozone decreased NPP of most forest types by 7-8%. Among all the forest types, spruce-fir forest was most resistant to O₃ damage, and NPP decreased by only 1% (Pan et al., 2009).

Felzer et al. (2004) developed TEM 4.3 (Table 9-2) to simulate the effects of O₃ on plant growth and estimated effects of O₃ on NPP and C sequestration of deciduous trees, conifers and crops in the conterminous United States. The results indicated that O₃ reduced NPP and C sequestration in the U.S. (Table 9-3) with the largest decreases (over 13% in some locations) in NPP occurring in the Midwest agricultural lands during the mid-summer. TEM was also used to evaluate the magnitude of O₃ damage at the global scale (Felzer et al., 2005). Simulations for the period 1860 to 1995 show that the largest reductions in NPP and net C exchange occurred in the mid western U.S., eastern Europe, and eastern China (Felzer et al., 2005). DLEM (Table 9-2) was developed to simulate the detrimental effect of O₃ on ecosystems, and has been used to examine the O₃ damage on NPP and C sequestration in Great Smoky Mountains National Park (Zhang et al., 2007a), grassland ecosystems and terrestrial ecosystems in China (Ren et al., 2007b; Ren et al., 2007a). Results of those simulations are listed in Table 9-3.

Instead of using AOT40 as their O₃ exposure metric as PnET, TEM and DLEM did, Sitch et al. (2007) incorporated a different O₃ metric named CUOt (cumulative stomatal uptake of O₃), derived from Pleijel et al. (2004a), into the MOSES-TRIFFID coupled model (Table 9-2). In the CUOt metric, the fractional reduction of plant production is dependent on O₃ uptake by stomata over a critical threshold for damage with this threshold level varying by plant functional type. Consistent with previous studies, their model simulation indicated that O₃ reduced global gross primary production (GPP), C-exchange rate and C sequestration (Table 9-3). The largest reductions in GPP and land-C storage were projected over North America, Europe, China and India. In the model, reduced ecosystem C uptake due to O₃ damage results in additional CO₂ accumulation in the atmosphere and an indirect radiative forcing of climate change. Their simulations indicated that the indirect radiative forcing caused by O₃ (0.62-1.09 W/m²) could have even greater impact on global warming than the direct radiative forcing of O₃ (0.89 W/m²) (Sitch et al., 2007).

Results from the various model studies presented in Table 9-3 are difficult to compare because of the various spatial and temporal scales used. However, all the studies showed that O₃ exposure decreased ecosystem productivity and C sequestration. These results are consistent and coherent with experimental results obtained from studies at the leaf, plant and ecosystem scales (Sitch et al., 2007; Felzer et al., 2005). Many of the models use the same underlying function to simulate the effect of O₃ exposure to C uptake. For example the functions of O₃ exposure
(AOT40) versus photosynthesis reduction for PnET-II, PnET-CN, TEM, DLEM were all from Reich et al. (1987) and Tjoelker et al. (1995). Therefore, it is not surprising that the results are similar. While these models can be improved and more evaluation with experimental data can be done, these models represent the state of the science for estimating the effect of O₃ exposure on productivity and C sequestration.

### 9.4.3.5 Summary

During the previous NAAQS reviews, there were very few studies that investigated the effect of O₃ exposure on ecosystem productivity and C sequestration. Recent studies from long-term FACE experiments, such as Aspen FACE, SoyFACE and the Kranzberg Forest (Germany), provide evidence of the association of O₃ exposure and reduced productivity at the ecosystem level of organization. Studies at the leaf and plant scales show that O₃ decreased photosynthesis and plant growth, which provides coherence and biological plausibility for the decrease in ecosystem productivity. Results across different ecosystem models, such as TREGRO, PnET, TEM and DLEM, are consistent with the FACE experimental evidence, which show that O₃ reduced productivity of various ecosystems. Productivity is measured by various metrics such as GPP, NPP, NEP, NCE, NEE and individual tree biomass gain. All these metrics indicate a decrease in CO₂ fixation by the systems that were studied.

Although O₃ generally causes negative effects on plant growth, the magnitude of the response varies among plant communities. For example, O₃ had little impact on white fir, but greatly reduced growth of ponderosa pine in southern California (Weinstein et al., 2005). Ozone decreased net primary production (NPP) of most forest types in the Mid-Atlantic region, but had small impacts on spruce-fir forest (Pan et al., 2009).

In addition to plant growth, other indicators that are typically estimated by model studies include net ecosystem CO₂ exchange (NEE), C sequestration, and crop yield. Model simulations consistently found that O₃ exposure caused negative impacts on these indicators, but the severity of these impacts was influenced by multiple interactions of biological and environmental factors. The suppression of ecosystem C sinks results in more CO₂ accumulation in the atmosphere. Globally, the indirect radiative forcing caused by O₃ exposure through lowering the ecosystem C sink could have an even greater impact on global warming than the direct radiative forcing of O₃ (Sitch et al., 2007). Ozone could also affect regional C budgets through interacting with multiple factors, such as N deposition, elevated CO₂ and land use history. Model simulations suggested that O₃ partially offset the growth stimulation caused by elevated CO₂ and N deposition in both Northeast- and Mid-Atlantic-region forest ecosystems of the U.S. (Pan et al., 2009; Ollinger et al., 2002).

The evidence is sufficient to infer that there is a causal relationship between O₃ exposure and reduced productivity, and a likely causal relationship between O₃ exposure and reduced carbon sequestration in terrestrial ecosystems.
### Table 9-3  Modeled effects of O₃ on primary production, C exchange, and C sequestration.

<table>
<thead>
<tr>
<th>Scale</th>
<th>Model</th>
<th>Index</th>
<th>O₃ Impacts</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>GPP</strong></td>
<td>MOSES-TRIFFID</td>
<td>CUOt</td>
<td>Decreased by 14-23% over the period 1901-2100</td>
<td>Sitch et al. (2007)</td>
</tr>
<tr>
<td><strong>NPP</strong></td>
<td>TEM</td>
<td>AOT40</td>
<td>Decreased by 0.8% without agricultural management and a decrease of 2.9% with optimal agricultural management</td>
<td>Felzer et al. (2005)</td>
</tr>
<tr>
<td>U.S.</td>
<td>TEM</td>
<td>AOT40</td>
<td>Reduced by 2.3% without optimal N fertilization and 7.2% with optimal N fertilization from 1983-1993</td>
<td>Felzer et al. (2005)</td>
</tr>
<tr>
<td>U.S.</td>
<td>TEM</td>
<td>AOT40</td>
<td>Reduced by 2.6-6.8% during the late 1980s to early 1990s.</td>
<td>Felzer et al. (2004)</td>
</tr>
<tr>
<td>Northeastern U.S.</td>
<td>PnET</td>
<td>AOT40</td>
<td>A reduction of 3-16% from 1987-1992</td>
<td>Ollinger et al. (1997a)</td>
</tr>
<tr>
<td>U.S. Mid-Atlantic</td>
<td>PnET</td>
<td>AOT40</td>
<td>Decreased NPP of most forest types by 7-8%</td>
<td>Pan et al. (2009)</td>
</tr>
<tr>
<td>China</td>
<td>DLEM</td>
<td>AOT40</td>
<td>Reduced NPP of grassland in China by 8.5 Tg³C from 1960s to 1990s</td>
<td>Ren et al. (2007a)</td>
</tr>
<tr>
<td><strong>C exchange</strong></td>
<td>TEM</td>
<td>AOT40</td>
<td>Reduced net C exchange (1950–1995) by 0.1 Pg C/yr without agricultural management and 0.3 Pg C/yr with optimal agricultural management</td>
<td>Felzer et al. (2005)</td>
</tr>
<tr>
<td>Global</td>
<td>MOSES-TRIFFID</td>
<td>CUOt</td>
<td>Decreased global mean land–atmosphere C fluxes by 1.3 Pg C/yr and 1.7 Pg C/yr for the ‘high’ and ‘low’ plant O₃ sensitivity models, respectively</td>
<td>Sitch et al. (2007)</td>
</tr>
<tr>
<td><strong>C sequestration</strong></td>
<td>MOSES-TRIFFID</td>
<td>CUOt</td>
<td>Reduced land-C storage accumulation by between 143 Pg C/yr and 263 Pg C/yr from 1900–2100</td>
<td>Sitch et al. (2007)</td>
</tr>
<tr>
<td>U.S.</td>
<td>TEM</td>
<td>AOT40</td>
<td>Reduced C sequestration by 18–38 Tg C/yr from 1950 to 1995</td>
<td>Felzer et al. (2004)</td>
</tr>
<tr>
<td>GSM National Park</td>
<td>DLEM</td>
<td>AOT40</td>
<td>Decreased the ecosystem C storage of deciduous forests by 2.5% and pine forest by 1.4% from 1971 to 2001</td>
<td>Zhang et al. (2007a)</td>
</tr>
<tr>
<td>China</td>
<td>DLEM</td>
<td>AOT40</td>
<td>Reduced total C storage by 0.06% in 1960s and 1.6% in 1990s in China’s terrestrial ecosystems</td>
<td>Ren et al. (2007b)</td>
</tr>
<tr>
<td>China</td>
<td>DLEM</td>
<td>AOT40</td>
<td>O₃ exposure reduced the net C sink of China’s terrestrial ecosystem by 7% from 1961 to 2005</td>
<td>Tian et al. (2011)</td>
</tr>
<tr>
<td>China</td>
<td>DLEM</td>
<td>AOT40</td>
<td>Ozone induced net carbon exchange reduction ranged from 0.4-43.1%, depending on different forest type</td>
<td>Ren et al. (2011)</td>
</tr>
</tbody>
</table>

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*CUOt is defined as the cumulative stomatal uptake of O₃, using a constant O₃-uptake rate threshold of t nmol/m²/sec.*

³Pg equals 1 × 10¹⁵ grams.
9.4.4 Crop Yield and Quality in Agricultural Systems

The detrimental effect of O$_3$ on crop production has been recognized since the 1960s and a large body of research has stemmed from that recognition. Previous O$_3$ AQCDs have extensively reviewed this body of literature. Table 9-4 summarizes recent experimental studies of O$_3$ effects on agricultural crops, exclusive of growth and yield. Growth and yield results are summarized in Table 9-17.

The actual concentration and duration threshold for O$_3$ damage varies from species to species and sometimes even among genotypes of the same species (Guidi et al., 2009; Sawada and Kohno, 2009; Biswas et al., 2008; Ariyaphanphitak et al., 2005; Dalstein and Vas, 2005; Keutgen et al., 2005). A number of comprehensive reviews and meta-analyses have recently been published discussing both the current understanding of the quantitative effects of O$_3$ concentration on a variety of crop species and the potential focus areas for biotechnological improvement to a future growing environment that will include higher O$_3$ concentrations (Bender and Weigel, 2011; Booker et al., 2009; Van Dingenen et al., 2009; Ainsworth, 2008; Feng et al., 2008b; Hayes et al., 2007; Mills et al., 2007a; Grantz et al., 2006; Morgan et al., 2003). Since the 2006 O$_3$ AQCD (U.S. EPA, 2006b), exposure-response indices for a variety of crops have been suggested (Mills et al., 2007a) and many reports have investigated the effects of O$_3$ concentration on seed or fruit quality to extend the knowledge base beyond yield quantity. This section will outline the key findings from these papers as well as highlight some of the recent research addressing the endpoints such as yields and crop quality.

This section will also highlight recent literature that focuses on O$_3$ damage to crops as influenced by other environmental factors. Genetic variability is not the only factor that determines crop response to O$_3$ damage. Ozone concentration throughout a growing-season is not homogeneous and other environmental conditions such as elevated CO$_2$ concentrations, drought, cold or nutrient availability may alleviate or exacerbate the oxidative stress response to a given O$_3$ concentration.

9.4.4.1 Yield

It is well known that yield is negatively impacted in many crop species in response to high O$_3$ concentration. However, the concentrations at which damage is observed vary from species to species. Numerous analyses of experiments conducted in OTCs and with naturally occurring gradients demonstrate that the effects of O$_3$ exposure also vary depending on the growth stage of the plant; plants grown for seed or grain are often most sensitive to exposure during the seed or grain-filling period (Soja et al., 2000; Pleijel et al., 1998; Younglove et al., 1994; Lee et al., 1988a). AX9.5.4.1 of the 2006 O$_3$ AQCD summarized many previous studies on crop yield.
Field studies and meta-analyses

The effect of O₃ exposure on U.S. crops remains an important area of research and several studies have been published on this topic since the 2006 O₃ AQCD (U.S. EPA, 2006b) (Table 9-4 and Table 9-17). For example, one study with cotton in a crop-weed interaction study (Grantz and Shrestha, 2006) utilizing OTCs suggests that 12-hour average O₃ concentrations of 79.9 ppb decreased cotton biomass by 25% and 12-hour average O₃ concentration of 122.7 ppb decreased cotton biomass by 75% compared to charcoal filtered control (12-h avg: 12.8 ppb). Further, this study suggests that the weed, yellow nutsedge, was less sensitive to increasing O₃ concentration, which would increase weed competition (Grantz and Shrestha, 2006).

In a study of peanuts in North Carolina, near ambient and elevated exposures of O₃ reduced photosynthesis and yield compared to very low O₃ conditions (Booker et al., 2007; Burkey et al., 2007). In another study, Grantz and Vu (2009) reported that sugarcane biomass growth significantly declined under O₃ exposure.

The average yield loss reported across a number of meta-analytic studies have been published recently for soybean (Morgan et al., 2003), wheat (Feng et al., 2008b), rice (Ainsworth, 2008), semi-natural vegetation (Hayes et al., 2007), potato, bean and barley (Feng and Kobayashi, 2009). Meta-analysis allows for the objective development of a quantitative consensus of the effects of a treatment across a wide body of literature. Further, this technique allows for a compilation of data across a range of O₃ fumigation techniques, durations and concentrations in order to assemble the existing literature in a meaningful manner.

Morgan et al. (2003) reported an average seed yield loss for soybean of 24% compared to charcoal filtered air across all O₃ concentrations used in the 53 compiled studies. The decrease in seed yield appeared to be the product of nearly equal decreases (7-12%) in seed weight, seed number and pod number. As would be expected, the lowest O₃ concentration (30-59 ppb) resulted in the smallest yield losses, approximately 8%, while the highest O₃ concentration (80-120 ppb) resulted in the largest yield losses, approximately 35% (Morgan et al., 2003). Further, the oil/protein ratio within the soybean seed was altered due to growth at elevated O₃ concentrations, with a decrease in oil content. The studies included in this meta-analysis all used enclosed fumigation systems or growth chambers which may have altered the coupling of the atmosphere to the lower plant canopy (McLeod and Long, 1999), although the results of Morgan et al. (2006), Betzelberger (2010), and the comparisons presented in Section 9.6.3 strongly suggest that decreases in yield between ambient and elevated exposures are not affected by exposure method.

Utilizing the Soybean Free Air gas Concentration Enrichment Facility (SoyFACE; www.soyface.illinois.edu). Morgan et al. (2006) reported a 20% seed yield loss due to a 23% increase in average daytime O₃ concentration (56-69 ppb) within a single soybean cultivar across two growing seasons in Illinois, which lies within the range predicted by the meta-analysis. A further breakdown of the effects of current O₃ concentrations (AOT40 of 4.7 ppm-h) on bean seed quality (Phaseolus vulgaris) has identified that growth at current O₃ concentrations compared to charcoal-filtered air raised total lipids, total crude protein and dietary fiber content (Iriti et al., 2009).
An increase in total phenolics was also observed, however the individual phenolic compounds responded differently, with significant decreases in anthocyanin content. The seeds from ambient O₃ exposed plants also displayed increased total antioxidant capacity compared to charcoal-filtered air controls (Iriti et al., 2009). Betzelberger et al. (2010) has recently utilized the SoyFACE facility to compare the impact of elevated O₃ concentrations across 10 soybean cultivars to investigate intraspecific variability of the O₃ response to find physiological or biochemical markers for eventual O₃ tolerance breeding efforts (Betzelberger et al., 2010). They report an average 17% decrease in yield across all 10 cultivars across two growing seasons due to a doubling of ambient O₃ concentrations, with the individual cultivar responses ranging from -7% to -36%. The exposure-response functions derived for these 10 current cultivars were similar to the response functions derived from the NCLAN studies conducted in the 1980s (Heagle, 1989), suggesting there has not been any selection for increased tolerance to O₃ in more recent cultivars. More complete comparisons between yield predictions based on data from cultivars used in NCLAN studies, and yield data for modern cultivars from SoyFACE are reported in Section 9.6.3 of this document. They confirm that the response of soybean yield to O₃ exposure has not changed in current cultivars.

A meta-analysis has also been performed on studies investigating the effects of O₃ concentrations on wheat (Feng et al., 2008b). Across 23 studies included, elevated O₃ concentrations (ranging from a 7-h daily average of 31-200 ppb) decreased grain yield by 29%. Winter wheat and spring wheat did not differ in their responses; however the response in both varieties to increasing O₃ concentrations resulted in successively larger decreases in yield, from a 20% decrease in 42 ppb to 60% in 153 ppb O₃. These yield losses were mainly caused by a combination of decreases in individual grain weight (-18%), ear number per plant (-16%), and grain number per ear (-11%). Further, the grain starch concentration decreased by 8% and the grain protein yield decreased by 18% due to growth at elevated O₃ concentrations as well. However, increases in grain calcium and potassium levels were reported (Feng et al., 2008b).

A recent meta-analysis found that growth at elevated O₃ concentrations negatively impacts nearly every aspect of rice performance as well (Ainsworth, 2008). While rice is not a major crop in the U.S., it provides a staple food for over half of the global population (IRRI, 2002) and the effects of rising O₃ concentrations on rice yields merit consideration. On average, rice yields decreased 14% in 62 ppb O₃ compared to charcoal-filtered air. This yield loss was largely driven by a 20% decrease in grain number (Ainsworth, 2008).

Feng and Kobayashi (2009) have recently compiled yield data for six major crop species, potato, barley, wheat, rice, bean and soybean and grouped the O₃ treatments used in those studies into three categories: baseline O₃ concentrations (<26 ppb), current ambient 7- or 12-h daily O₃ concentrations (31-50 ppb), and future ambient 7- or 12-h daily O₃ concentrations (51-75 ppb). Using these categories, they have effectively characterized the effects of current O₃ concentrations and the effects of future O₃ concentrations compared to baseline O₃ concentrations. At current O₃ concentrations, the...
concentrations, which ranged from 41-49 ppb in the studies included, soybean (-7.7%), bean (-19.0%), barley (-8.9%), wheat (-9.7%), rice (-17.5%) and potato (-5.3%) all had yield losses compared to the baseline O3 concentrations (<26 ppb). At future O3 concentrations, averaging 63 ppb, soybean (-21.6%), bean (-41.4%), barley (-14%), wheat (-28%), rice (-17.5%) and potato (-11.9%) all had significantly larger yield losses compared to the losses at current O3 concentrations (<26 ppb) (Feng and Kobayashi, 2009).

A review of OTC studies has determined the AOT40 critical level that causes a 5% yield reduction across a variety of agricultural and horticultural species (Mills et al., 2007a). The authors classify the species studied into three groups: sensitive, moderate and tolerant. The sensitive crops, including watermelon, beans, cotton, wheat, turnip, onion, soybean, lettuce, and tomato, respond with a 5% reduction in yield under a 3-month AOT40 of 6 ppm-h. Watermelon was the most sensitive with a critical level of 1.6 ppm-h. The moderately sensitive crops, including sugar beet, oilseed rape, potato, tobacco, rice, maize, grape and broccoli, responded with a 5% reduction in yield between 8.6 and 20 ppm-h. The crops classified as tolerant, including strawberry, plum and barley, responded with a 5% yield reduction between 62-83.3 ppm-h (Mills et al., 2007a).

Feng and Kobayashi (2009) compared their exposure-response results to those published by Mills et al. (2007a) and found the ranges of yield loss to be similar for soybean, rice and bean. However, Feng and Kobayashi (2009) reported smaller yield losses for potato and wheat and larger yield losses for barley compared to the dose-response functions published by Mills et al. (2007a), which they attributed to their more lenient criteria for literature inclusion.

While the studies investigating the impact of various O3 concentrations on yield are important and aid in determining the vulnerability of various crops to a variety of O3 concentrations, there is still uncertainty as to how these crops respond under field conditions with interacting environmental factors such as temperature, soil moisture, CO2 concentration, and soil fertility (Booker et al., 2009). Further, there appears to be a distinct developmental and genotype dependent influence on plant sensitivity to O3 that has yet to be fully investigated across O3 concentrations in a field setting. The potentially mitigating effect of breeding selection for O3 resistance has received very little attention in the published scientific literature. Anecdotal reports suggest that such selection may have occurred in recent decades for some crops in areas of the country with high ambient exposures. However, the only published literature available is on soybean and these studies indicate that sensitivity has not changed in cultivars of soybean between the 1980s and the 2000s (Betzelberger et al., 2010). This conclusion for soybeans is confirmed by comparisons presented in Section 9.6.3 of this document.

**Yield loss at regional and global scales**

Because O3 is heterogeneous in both time and space and O3 monitoring stations are predominantly near urban areas, the impacts of O3 on current crop yields at large
geographical scales are difficult to estimate. Fishman et al. (2010) have used satellite observations to estimate O₃ concentrations in the contiguous tri-state area of Iowa, Illinois and Indiana and have combined that information with other measured environmental variables to model the historical impact of O₃ concentrations on soybean yield across the 2002-2006 growing seasons. When soybean yield across Iowa, Indiana and Illinois was modeled as a function of seasonal temperature, soil moisture and O₃ concentrations, O₃ had the largest contribution to the variability in yield for the southern-most latitudes included in the dataset. Fishman et al. (2010) determined that O₃ concentrations significantly reduced soybean yield by 0.38 to 1.63% for every additional ppb of exposure across the 5 years. This value is consistent with previous chamber studies (Heagle, 1989) and results from SoyFACE (Morgan et al., 2006). Satellite estimates of tropospheric O₃ concentrations exist globally (Fishman et al., 2008), therefore utilizing this historical modeling approach is feasible across a wider geographical area, longer time-span and perhaps for more crop species.

The detrimental effects of O₃ on crop production at regional or global scales were also assessed by several model studies. Two large scale field studies were conducted in the U.S. (NCLAN) and in Europe (European Open Top Chamber Programme, EOTCP) to assess the impact of O₃ on crop production. Ozone exposure-response regression models derived from the two programs have been widely used to estimate crop yield loss (Avery et al., 2011a, b; Van Dingenen et al., 2009; Tong and Mauzerall, 2008; Wang and Mauzerall, 2004). Those studies found that O₃ generally reduced crop yield and that different crops showed different sensitivity to O₃ pollution (Table 9-5). Ozone was calculated to induce a possible 45-82 million metric tons loss for wheat globally. Production losses for rice, maize and soybean were on the order of 17-23 million metric tons globally (Van Dingenen et al., 2009). The largest yield losses occur in high-production areas exposed to high O₃ concentrations, such the Midwest and the Mississippi Valley regions in the U.S., Europe, China and India (Van Dingenen et al., 2009; Tong et al., 2007).

9.4.4.2 Crop Quality

In general, it appears that increasing O₃ concentrations above current ambient concentrations can cause species-dependent biomass losses, decreases in root biomass and nutritive quality, accelerated senescence and shifts in biodiversity. A study conducted with highbush blackberry has demonstrated decreased nutritive quality with increasing O₃ concentration despite no change in biomass between charcoal-filtered control, ambient O₃ and 2 × ambient O₃ exposures (Ditchkoff et al., 2009). A study conducted with sedge using control (30 ppb), low (55 ppb), medium (80 ppb) and high (105 ppb) O₃ treatments has demonstrated decreased root biomass and accelerated senescence in the medium and high O₃ treatments (Jones et al., 2010). Alfalfa showed no biomass changes across two years of double ambient O₃ concentrations (AOT40 of 13.9 ppm-h) using FACE fumigation (Maggio et al., 2009). However a modeling study has demonstrated that 84% of the variability in the
relative feed value in high-yielding alfalfa was due to the variability in mean $O_3$ concentration from 1998-2002 (Lin et al., 2007). Further, in a managed grassland FACE system, the reduction in total biomass harvest over five years decreased twice as fast in the elevated treatment (AOT40 of 13-59 ppm-h) compared to ambient (AOT40 of 1-20.7 ppm-h). Compared with the ambient control, loss in annual dry matter yield was 23% after 5 year. Further, functional groups were differentially affected, with legumes showing the strongest negative response (Volk et al., 2006). However, a later study by Stampfli and Fuhrer (2010) at the same site suggested that Volk et al. (2006) likely overestimated the effects of $O_3$ on yield reduction because the overlapping effects of species dynamics caused by heterogeneous initial conditions and a change in management were not considered by these authors. An OTC study conducted with *Trifolium subterraneum* exposed to filtered (<15 ppb), ambient, and 40 ppb above ambient $O_3$ demonstrated decreases in biomass in the highest $O_3$ treatment as well as 10-20% decreased nutritive quality which was mainly attributed to accelerated senescence (Sanz et al., 2005). A study conducted with Eastern gamagrass and big bluestem in OTCs suggested that big bluestem was not sensitive to $O_3$, but gamagrass displayed decreased nutritive quality in the 2 × ambient $O_3$ treatment, due to higher lignin content and decreased N (Lewis et al., 2006).

### 9.4.4.3 Summary

The detrimental effect of $O_3$ on crop production has been recognized since the 1960s and a large body of research has subsequently stemmed from those initial findings. Previous $O_3$ AQCDs have extensively reviewed this body of literature (U.S. EPA, 2006b). Current $O_3$ concentrations across the U.S. are high enough to cause yield loss for a variety of agricultural crops including, but not limited to, soybean, wheat, potato, watermelon, beans, turnip, onion, lettuce, and tomato. Continued increases in $O_3$ concentration may further decrease yield in these sensitive crops. Despite the well-documented yield losses due to increasing $O_3$ concentration, there is still a knowledge gap pertaining to the exact mechanisms of $O_3$-induced yield loss. Research has linked increasing $O_3$ concentration to decreased photosynthetic rates and accelerated senescence, which are related to yield.

Recent research has highlighted the effects of $O_3$ on crop quality. Increasing $O_3$ concentration decreases nutritive quality of grasses, decreases macro- and micro-nutrient concentrations in fruits and vegetable crops, and decreases cotton fiber quality. It is important to note that these effects, as well as those mentioned above can occur without the expression of visible injury on the leaves. These areas of research require further investigation to determine mechanisms and exposure-response relationships.

During the previous NAAQS reviews, there were very few studies that estimated $O_3$ impacts on crop yields at large geographical scales. Recent modeling studies found that $O_3$ generally reduced crop yield, but the impacts varied across regions and crop
species. For example, the largest O₃-induced crop yield losses occurred in high-production areas exposed to high O₃ concentrations, such as the Midwest and the Mississippi Valley regions of the U.S. (Van Dingenen et al., 2009). Among crop species, the estimated yield loss for wheat and soybean were higher than for rice and maize (Van Dingenen et al., 2009). Using satellite air-column observations with direct air-sampling O₃ data, Fishman et al. (2010) modeled the yield-loss due to O₃ over the continuous tri-state area of Illinois, Iowa, and Wisconsin. They determined that O₃ concentrations significantly reduced soybean yield, which further reinforces previous results from FACE-type experiments and OTC experiments. Evidence is sufficient to conclude that **there is a causal relationship between O₃ exposure and reduced yield and quality of agricultural crops.**
Table 9-4  Summary of recent studies of O₃ effects on crops (exclusive of growth and yield).

<table>
<thead>
<tr>
<th>Species</th>
<th>Facility</th>
<th>Location</th>
<th>Exposure Duration</th>
<th>Ozone Exposurea (Additional treatment)</th>
<th>Variable(s) measured</th>
<th>Percent (% change from CFb (% change from ambient))</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Alfalfa (Medicago sativa cv. Beaver) Growth chambers</td>
<td>1, 2 or 4 days</td>
<td>3 or 5 hours/day 85 ppb (Exposure duration)</td>
<td>Relative feed value</td>
<td>n.s. *high variability among treatment groups (N/A)</td>
<td>Muntifering et al. (2006b)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bean (Phaseolus vulgaris l. cv Borlottto) OTC, ground-planted Cumo, Italy</td>
<td>4 months</td>
<td>Seasonal AOT40: CF = 0.5 ppm-h; Ambient = 4.6 ppm-h (N/A)</td>
<td>Seed lipid, Protein content Fiber content</td>
<td>+28.5 (N/A) +7.88 (N/A) +14.54 (N/A)</td>
<td>Iriti et al. (2009)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Big Blue Stem (Andropogon gerardii) OTC Alabama, U.S.</td>
<td>4 months</td>
<td>12-h avg: CF = 14 ppb; Ambient = 29 ppb; Elevated = 71 ppb (N/A)</td>
<td>Relative feed value</td>
<td>n.s. (n.s.)</td>
<td>Lewis et al. (2006)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Brassica napus Growth chambers Belgium</td>
<td>4 days</td>
<td>CF &amp; 176 ppb for 4 hours/day (N/A)</td>
<td>Glucosinolates</td>
<td>−41 (N/A)</td>
<td>Gielen et al. (2006)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Brassica napus cv. Westar Growth chambers Finland</td>
<td>17-26 days</td>
<td>8-h avg: CF &amp; 100 ppb (Bt/non-Bt; herbivory)</td>
<td>VOC emissions</td>
<td>−30.7 (N/A); −34 (N/A)</td>
<td>Himanen et al. (2009b)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Eastern Gamagrass (Tripsacum dactyloides) OTC Alabama, U.S.</td>
<td>4 months</td>
<td>12-h avg: CF = 14 ppb; Ambient = 29 ppb; Elevated = 71 ppb (N/A)</td>
<td>Relative feed value</td>
<td>−17 (−12)</td>
<td>Lewis et al. (2006)</td>
<td></td>
<td></td>
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<tr>
<td>Lettuce (Lactuca sativa) OTC Carcaixent Experimental Station, Spain</td>
<td>30 days</td>
<td>12-h mean: CF = 10.2 ppb; NF = 30.1 ppb; NF+O₃ = 62.7 ppb (4 cultivars)</td>
<td>Lipid peroxidation; Root length</td>
<td>+77 (+38) −22 (−14)</td>
<td>Calatayud et al. (2002)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Peanut (Arachis hypogaea) OTC Raleigh, NC; U.S.</td>
<td>3 yr</td>
<td>12-h avg: CF = 22 ppb; Ambient = 46 ppb; Elevated = 75 ppb (CO₂: 375 ppm; 548 ppm; 730 ppm)</td>
<td>Harvest biomass</td>
<td>−40 (−10)</td>
<td>Booker et al. (2007)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Species</td>
<td>Exposure Duration</td>
<td>Ozone Exposure* (Additional treatment)</td>
<td>Variable(s) measured</td>
<td>Percent (%) change from CFb (% change from ambient)</td>
<td>Reference</td>
<td></td>
<td></td>
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<tr>
<td><em>Poa pratensis</em></td>
<td>3 yr; 4-5 weeks in the spring</td>
<td>8-h avg: CF+25 = 21.7 ppb; NF+50 = 73.1 ppb (Competition)</td>
<td>Relative feed value</td>
<td>N/A (n.s.; −8)</td>
<td>Bender et al. (2006)</td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Potato (Solanum tuberosum cv. Bintje)</em></td>
<td>2 yr</td>
<td>CF = 10 ppb; Ambient = 25 ppb; Ambient(+) = (36 ppb); Ambient(++) = (47 ppb) (N/A)</td>
<td>[K], [Ca], [Mg], [P], [N] per dry weight of tubers *dose-response regression, report significant positive or negative slope with increasing [O3]</td>
<td>N[P][Ca] n.s.; [K] &amp; [Mg] sig + (N/A)</td>
<td>Piikki et al. (2007)</td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Potato (Solanum tuberosum cv. Indira)</em></td>
<td>8 weeks</td>
<td>CF = 10 ppb; Ambient = 50 ppb; 2× Ambient = 100 ppb (CO2: 400 ppm &amp; 700 ppm)</td>
<td>Pathogen infestation using percent necrosis</td>
<td>+52 (n.s.)</td>
<td>Plessl et al. (2007)</td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Soybean (Glycine max cv. 93B15)</em></td>
<td>3 yr May-Oct</td>
<td>AOT40: CF = 0 ppm-h; Ambient = 3.4 ppm-h; Elevated = 9.0 ppm-h (Well-watered &amp; water-stressed)</td>
<td>Daily evapotranspiration</td>
<td>−28 (−14)</td>
<td>Bou Jaoudé et al. (2008a)</td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Soybean (Glycine max cv. 93B15)</em></td>
<td>4 months</td>
<td>8-h avg: Ambient = 38.5 ppb; Elevated = 52 ppb (Herbivory)</td>
<td>Herbivory defense-related genes</td>
<td>N/A (N/A)</td>
<td>Casteel et al. (2008)</td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Soybean (Glycine max cv. Essex)</em></td>
<td>2 yr</td>
<td>12-h avg: CF = 21 ppb; 1.5× Ambient = 74 ppb (CO2: 370 ppm &amp; 714 ppm)</td>
<td>Post-harvest residue</td>
<td>N/A (−15.46)</td>
<td>Booker et al. (2005)</td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Soybean (Glycine max cv. Essex)</em></td>
<td>3 months</td>
<td>12-h avg: CF = 18 ppb; Elevated = 72 ppb (CO2: 367 &amp; 718)</td>
<td>Water-use efficiency</td>
<td>n.s. (N/A)</td>
<td>Booker et al. (2004b)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Species</td>
<td>Facility</td>
<td>Location</td>
<td>Exposure Duration</td>
<td>Ozone Exposure* (Additional treatment)</td>
<td>Variable(s) measured</td>
<td>Percent (%) change from CFb (% change from ambient)</td>
<td>Reference</td>
</tr>
<tr>
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</tr>
<tr>
<td>Soybean (Glycine max) 10 cultivars</td>
<td>SoyFACE</td>
<td>Urbana, IL; U.S.</td>
<td>2 yr</td>
<td>8-h avg (ppb): Ambient = 46.3 &amp; 37.9; Elevated = 82.5 &amp; 61.3 (Cultivar comparisons)</td>
<td>Total antioxidant capacity</td>
<td>N/A (+19)</td>
<td>Betzelberger et al. (2010)</td>
</tr>
<tr>
<td>Spring Wheat (Triticum aestivum cv. Minaret; Satu; Drabant; Dragon) OTCs Belgium, Finland, &amp; Sweden</td>
<td></td>
<td></td>
<td>7 yr</td>
<td>Seasonal AOT40s ranged from: 0 to16 ppm-h (N/A)</td>
<td>Seed protein content; 1,000-seed weight regressed across all experiments</td>
<td>N/A (Significant negative correlation)</td>
<td>Piikki et al. (2008b)</td>
</tr>
<tr>
<td>Strawberry (Fragaria x ananassa Duch. Cv. Korona &amp; Elsanta) Growth chambers Bonn, Germany</td>
<td></td>
<td></td>
<td>2 months</td>
<td>8-h avg: CF = 0 ppb; Elevated = 78 ppb (N/A)</td>
<td>Total leaf area</td>
<td>−16 (N/A)</td>
<td>Keutgen et al. (2005)</td>
</tr>
<tr>
<td>Sweet Potato Growth Chambers Bonn, Germany</td>
<td></td>
<td></td>
<td>4 weeks</td>
<td>8-h avg: CF = 0 ppb; Ambient &lt;40 ppb; Elevated = 255 ppb (N/A)</td>
<td>Tuber weight</td>
<td>−14 (−11.5)</td>
<td>Keutgen et al. (2008)</td>
</tr>
<tr>
<td>Tomato (Lycopersicon esculentum) OTC Valencia, Spain</td>
<td></td>
<td></td>
<td>133 days</td>
<td>8- mean: CF = 16.3 ppb; NF = 30.1 ppb; NF(+) = 83.2 ppb (Various cultivars; early &amp; late harvest)</td>
<td>Brix degree</td>
<td>−7.2 (−3.6)</td>
<td>Dalstein et al. (2005)</td>
</tr>
<tr>
<td>Trifolium repens &amp; Trifolium pratense Aspen FACE Rhinelander, WI; U.S.</td>
<td></td>
<td></td>
<td>3 months</td>
<td>3-mo daylight avg: Ambient = 34.8 ppb; 1.2× Ambient = 42.23 ppb (CO2; 560 ppm)</td>
<td>Lignin; Dry-matter digestibility</td>
<td>N/A (+19.3) N/A (−4.2)</td>
<td>Muntifering et al. (2006a)</td>
</tr>
</tbody>
</table>

*Ozone exposure in ppb unless otherwise noted.

bCF = Carbon-filtered air; NF = Non-filtered air.
Table 9-5  Modeled effects of O₃ on crop yield loss at regional and global scales.

<table>
<thead>
<tr>
<th>Scale</th>
<th>Index</th>
<th>O₃ Impacts</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Global</td>
<td>M7a;</td>
<td>Reduced by 7.3% to 12.3% for wheat, 5.4% to 15.6% for soybean, 2.8% to</td>
<td>Van Dingenen et al. (2009)</td>
</tr>
<tr>
<td></td>
<td>M12b;</td>
<td>3.7% for rice, and 2.4% to 4.1% for maize in year 2000.</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AOT40</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Global</td>
<td>M12b;</td>
<td>O₃-induced global yield reductions ranged from 8.5-14% for soybean, 3.9-</td>
<td>Avnery et al. (2011a)</td>
</tr>
<tr>
<td></td>
<td>AOT40</td>
<td>15% for wheat, and 2.2-5.5% for maize in year 2000. Global crop</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>production losses totaled 79-121 million metric tons, worth $11-18 billion</td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>annually (in U.S. Dollars; 2000).</td>
<td></td>
</tr>
<tr>
<td>U.S.</td>
<td>M7;</td>
<td>Reduced by 4.1% to 4.4% for wheat, 7.1% to 17.7% for soybean, 2.6% to</td>
<td>Van Dingenen et al. (2009)</td>
</tr>
<tr>
<td></td>
<td>M12;</td>
<td>3.2% for rice, and 2.2% to 3.6% for maize in year 2000.</td>
<td></td>
</tr>
<tr>
<td></td>
<td>AOT40</td>
<td></td>
<td></td>
</tr>
<tr>
<td>U.S.</td>
<td>SUM06</td>
<td>Caused a loss of 53.8 million to 438 million bushels in soybean production,</td>
<td>Tong et al. (2007)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>which account for 1.7–14.2% of total U.S. soybean production in 2005</td>
<td></td>
</tr>
<tr>
<td>East</td>
<td>M7;</td>
<td>Reduced the yield of wheat, rice and corn by 1–9% and soybean by 23–</td>
<td>Wang et al. (2004)</td>
</tr>
<tr>
<td>Asia</td>
<td>M12</td>
<td>27% in China, Japan and South Korea in 1990.</td>
<td></td>
</tr>
</tbody>
</table>

ₐM7 is defined as 7-hour mean O₃ concentration (ppb).
₇M12 is defined as 12-hour mean O₃ concentration (ppb).

9.4.5  Water Cycling

Ozone can affect water use in plants and ecosystems through several mechanisms including damage to stomatal functioning and loss of leaf area. Figure 9-7 provides a simple illustration of potential effects of O₃ exposure on water cycling. Section 9.3.2 reviewed possible mechanisms for effects of O₃ exposure on stomatal functioning. This section on water cycling discusses how this alteration of stomatal functioning may affect water use in leaves, whole plants, a planted forest and watersheds.
In the literature, there is not a clear consensus on the nature of leaf-level stomatal conductance response to O₃ exposure. At the leaf level, O₃ exposure is known to result in stomatal patchiness (Paoletti and Grulke, 2005; Omasa et al., 1987; Ellenson and Amundson, 1982), i.e., the heterogeneous aperture widths of stomata on the leaf surface, and, as a result, the collective response of groups of stomata on leaves and canopies determines larger-scale responses to O₃. When measured at steady-state high light conditions, leaf-level stomatal conductance is often found to be reduced when exposed to O₃. For example, a meta-analysis of 55 studies found that O₃ reduced stomatal conductance by 11% (Wittig et al., 2007). However, these steady-state measurements were generally taken at saturating light conditions and steady-state vapor pressure deficit (VPD). Saturating light and steady-state VPD conditions are not common in the field since many parts of the plant canopy are shaded throughout the day. When studied under varying environmental conditions, many
studies have reported incomplete stomatal closure with elevated O₃ exposure during the day (Mills et al., 2009; Gruelke et al., 2007b; Matyssek et al., 1995; Wieser and Havranek, 1995) or at night (Gruelke et al., 2004). This may be due to sluggish stomatal response. Sluggish stomatal response, defined as a delay in stomatal response to changing environmental factors relative to controls (Paoletti and Gruelke, 2010) has also been documented by several researchers (Gruelke et al., 2007c; Matyssek et al., 1995; Pearson and Mansfield, 1993; Wallin and Skärby, 1992; Lee et al., 1990; Skarby et al., 1987; Keller and Häser, 1984; Reich and Lassoie, 1984). Sluggish stomatal response associated with O₃ exposure suggests an uncoupling of the normally tight relationship between carbon assimilation and stomatal conductance as measured under steady-state conditions (Gregg et al., 2006; Paoletti and Gruelke, 2005). Several tree and ecosystem models, such as TREGRO, PnET and DLEM, rely on this tight relationship to simulate water and carbon dynamics. The O₃-induced impairment of stomatal control may be more pronounced for plants growing under water stress (Wilkinson and Davies, 2010; Gruelke et al., 2007a; Paoletti and Gruelke, 2005; Bonn et al., 2004; Kellomaki and Wang, 1997; Tjoelker et al., 1995; Reich and Lassoie, 1984). Since leaf-level stomatal regulation is usually assessed in a steady state rather than as a dynamic response to changing environmental conditions, steady state measurements cannot detect sluggish stomatal response. Because of sluggish stomatal responses, water loss from plants could be greater or reduced under dynamic environmental conditions over days and months. In situations where stomata fail to close under low light or water stressed conditions, water loss may be greater over time. In other situations, it is possible that sluggish stomata may fail to completely open in response to environmental stimuli and result in decreased water loss.

In addition to the impacts on stomatal performance, O₃-induced physiological changes, such as reduced leaf area index and accelerated leaf senescence could alter water use efficiency. It is well established from chamber and field studies that O₃ exposure is correlated with lower foliar retention (Karnosky et al., 2003; Topa et al., 2001; Pell et al., 1999; Gruelke and Lee, 1997; Karnosky et al., 1996; Miller et al., 1972; Miller et al., 1963). However, Lee et al. (2009a) did not find changes in needle area of ponderosa pine and reported that greater canopy conductance followed by water stress under elevated O₃ may have been caused by stomatal dysfunction. At the Aspen FACE experiment, stand-level water use, as indicated by sap flux per unit ground area, was not significantly affected by elevated O₃ despite a 22% decrease in leaf area index and 20% decrease in basal area (Uddling et al., 2008). The same study reported a substantial increase in maximum sap flow per unit leaf area under elevated O₃, indicating higher canopy conductance compared to controls. A subsequent study at Aspen FACE (Uddling et al., 2009) reported that leaf-level conductance was not reduced by elevated O₃ as observed in most short-term experiments on tree seedlings (Wittig et al., 2007). The mean values of leaf-level conductance were always higher in elevated O₃ compared to controls, although this increase was not always statistically significant (Uddling et al., 2009). The authors also reported a less sensitive stomatal closure response to increasing vapor pressure deficit in pure aspen stands exposed to elevated O₃. This indicated that there was some evidence of impaired stomatal control. These studies at Aspen FACE also suggested that long-
term cumulative effects of elevated O₃ on tree and stand structure may be more important than the primary stomatal responses for understanding the effect of O₃ on stand-level water use (Uddling et al., 2009; 2008). Elevated O₃ could also affect evapotranspiration by altering tree crown interception of precipitation. Ozone was shown to change branch architectural parameters, and the effects were species-dependent at the Aspen FACE experiment (Rhea et al., 2010). The authors found that there was a significant correlation between canopy architecture parameters and stemflow (the flow of intercepted water down the stem of a tree) for birch but not aspen.

It is difficult to scale up physiology measurements from leaves to ecosystems. Thus, the current understanding of how stomatal response at the leaf scale is integrated at the scale of whole forest canopies, and therefore how it influences tree and forest stand water use is limited. Field studies by (McLaughlin et al., 2007a; 2007b) provided valuable insight into the possible consequences of stomatal sluggishness for ecosystem water cycling. McLaughlin et al. (2007a; 2007b) indicated that O₃ increased water use in a mixed deciduous forest in eastern Tennessee. McLaughlin et al. (2007a; 2007b) found that O₃, with daily maximum levels ranging from 69 to 83 ppb, reduced stem growth by 30-50% in the high-O₃ year 2002. The decrease in growth rate was caused in part by amplification of diurnal cycles of water loss and recovery. Peak hourly O₃ exposure increased the rate of water loss through transpiration as indicated by the increased stem sap flow. The authors suggested that a potential mechanism for the increased sap flow could be altered stomatal regulation from O₃ exposure, but this was inferred through sap flow measurements and was not directly measured. Alternatively, stomatal conductance may have increased under higher O₃ conditions (Paoletti and Grulke, 2010). The increased canopy water loss resulted in higher water uptake by the trees as reflected in the reduced soil moisture in the rooting zone. The change in tree water use led to further impacts on the hydrological cycle at the landscape level. Increased water use under high O₃ exposure was reported to reduce late-season modeled streamflow in three forested watersheds in eastern Tennessee (McLaughlin et al., 2007b).

Felzer et al. (2009) used TEM-Hydro to assess the interactions of O₃, climate, elevated CO₂ and N limitation on the hydrological cycle in the eastern United States. They found that elevated CO₂ decreased evapotranspiration by 2-4% and increased runoff by 3-7%, as compared to the effects of climate alone. When O₃ damage and N limitation were included, evapotranspiration was reduced by an additional 4-7% and runoff was increased by an additional 6-11% (Felzer et al., 2009). Based upon simulation with INTRAST and LINKAGES, Hanson et al. (2005) found that increasing O₃ concentration by 20 ppb above the current ambient level yields a modest 3% reduction in water use. Those ecological models were generally built on the assumption that O₃ induces stomatal closure and have not incorporated possible stomatal sluggishness due to O₃ exposure. Because of this assumption, results of those models normally found that O₃ reduced water use.
9.4.5.1 Summary

Although the evidence was from a limited number of field and modeling studies, findings showed an association between O₃ exposure and alteration of water use and cycling in vegetation, and at the watershed level. There is not a clear consensus on the nature of leaf-level stomatal conductance response to O₃ exposure. When measured under steady-state high light conditions, leaf-level stomatal conductance is often found to be reduced when plants are exposed to O₃. However, measurements of stomatal conductance under dynamic light and VPD conditions indicate sluggish responses under elevated O₃ exposure, which could potentially lead to increased water loss from vegetation in some situations. Field studies conducted by McLaughlin et al. (2007a; 2007b) suggested that peak hourly O₃ exposure increased the rate of water loss from several tree species, and led to a reduction in the late-season modeled stream flow in three forested watersheds in eastern Tennessee. Sluggish stomatal responses during O₃ exposure was suggested as a possible mechanism for increased water loss during peak O₃ exposure. Currently, the O₃-induced reduction in stomatal aperture is the biological assumption for most process-based models. Because of this assumption, results of those models normally found that O₃ reduced water loss. For example, Felzer et al. (2009) found that O₃ damage and N limitation together reduced evapotranspiration and increased runoff.

Although the direction of the response differed among studies, the evidence is sufficient to conclude that there is likely to be a causal relationship between O₃ exposure and the alteration of ecosystem water cycling.

9.4.6 Below-Ground Processes

Above-ground and below-ground processes are tightly interconnected. Because roots and soil organisms are not exposed directly to O₃, below-ground processes are affected by O₃ through alterations in the quality and quantity of C supply from photosynthates and litterfall (Andersen, 2003). Ozone can decrease leaf C uptake by reducing photosynthesis (Section 9.3). Ozone can also increase metabolic costs by stimulating the production of chemical compounds for defense and repair processes, and by increasing the synthesis of antioxidants to neutralize free radicals (see Section 9.3), both of which increase the allocation of carbon for above-ground processes. Therefore, O₃ could significantly reduce the amount of C available for allocation to below-ground by decreasing C uptake while increasing C consumption of above-ground processes (Andersen, 2003).

Since the 2006 O₃ AQCD, there is additional evidence for O₃ effects on below-ground processes. Ozone has been found to alter root growth, soil food web structure, decomposer activities, C turnover, water cycling and nutrient flow (Figure 9-8). Ozone effects on root development and root biomass production and soil food web structure are reviewed in Section 9.4.3.1 and Section 9.4.9.2, respectively. The focus
in this section is on the response of litter input, decomposer activities, soil respiration, soil C formation and nutrient cycling.

![Conceptual diagram showing where O₃ alters C, water and nutrient flow in a tree-soil system, including transfer between biotic and abiotic components below ground that influence soil physical and chemical properties.](image)

Note: Arrows denote C flux pathways that are affected by O₃. Dashed lines indicate where the impact of O₃ is suspected but unknown.

Source: Modified from Andersen et al. (2003).

**Figure 9-8** Conceptual diagram showing where O₃ alters C, water and nutrient flow in a tree-soil system, including transfer between biotic and abiotic components below ground that influence soil physical and chemical properties.

### 9.4.6.1 Litter Carbon Chemistry, Litter Nutrient and Their Ecosystem Budgets

Consistent with previous findings, recent studies show that, although the responses are often species-dependent, O₃ tends to alter litter chemistry (U.S. EPA, 2006b). Alterations in chemical parameters, such as changes in C chemistry and nutrient concentrations, were observed in both leaf and root litter (Table 9-6).
At the Aspen FACE site, several studies investigated litter chemistry changes (Parsons et al., 2008; Johnson and Pregitzer, 2007; Chapman et al., 2005; Liu et al., 2005). In both aspen and birch leaf litter, elevated O$_3$ increased the concentrations of soluble sugars, soluble phenolics and condensed tannins (Parsons et al., 2008; Liu et al., 2005). Compared to other treatments, aspen litter under elevated O$_3$ had the highest fiber concentration, with the lowest concentration associated with the birch litter under the same conditions (Parsons et al., 2008). Chapman et al. (2005) measured chemical changes in fine root litter and found that elevated O$_3$ decreased lignin concentration. O$_3$-induced chemistry changes were also reported from other experimental sites. Results from an OTC study in Finland suggested that elevated O$_3$ increased the concentration of acid-soluble lignin, but had no significant impact on other chemicals such as total sugars, hemicelluloses, cellulose or total lignin in the litter of silver birch (Kasurinen et al., 2006). Results from the free air canopy O$_3$ exposure experiment at Kranzberg Forest showed that O$_3$ increased starch concentrations but had no impact on cellulose and lignin in beech and spruce leaf litter (Aneja et al., 2007). The effect of O$_3$ on three antioxidants (ascorbate, glutathione and α-tocopherol) in fine roots of beech was also assessed at Kranzberg Forest. The results indicated that O$_3$ had no significant effect on α-tocopherol and ascorbate concentrations, but decreased glutathione concentrations in fine roots (Haberer et al., 2008). In addition to changing C chemistry, O$_3$ also altered nutrient concentrations in green leaves and litter (Table 9-6).

The combined effects of O$_3$ on biomass productivity and chemistry changes may alter C chemicals and nutrient contents at the canopy or stand level. For example, although O$_3$ had different impacts on their concentrations, annual fluxes of C chemicals (soluble sugar, soluble phenolics, condensed tannins, lipid and hemicelluloses), macro nutrients (N, P, K and S) and micro nutrients (Mg, B, Cu and Zn) to soil were all reduced due to lower litter biomass productivity at Aspen FACE (Liu et al., 2007a; Liu et al., 2005). In a 2-year growth chamber experiment in Germany, N content of a spruce canopy in a mixed culture and Ca content of a beech canopy in a monoculture was increased due to elevated O$_3$, although leaf production was not significantly altered by O$_3$ (Rodenkirchen et al., 2009).
Table 9-6  The effect of elevated $O_3$ on leaf/litter nutrient concentrations.

<table>
<thead>
<tr>
<th>Study Site</th>
<th>Species</th>
<th>O$_3$ Concentration</th>
<th>Response</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Suonenjoki Research Station, Finland</td>
<td>Silver birch</td>
<td>Ambient: 10-60 ppb Elevated: 2× ambient</td>
<td>Decreased the concentration of P, Mn, Zn and B in leaf litter</td>
<td>Kasurinen et al. (2006)</td>
</tr>
<tr>
<td>Aspen FACE</td>
<td>Aspen and birch</td>
<td>Ambient: 50-60 ppb Elevated: 1.5× ambient</td>
<td>Decreased the concentrations of P, S, Ca and Zn, but had no impact on the concentrations of N, K, Mg, Mn, B and Cu in leaf litter.</td>
<td>Liu et al. (2007a)</td>
</tr>
<tr>
<td>Aspen FACE</td>
<td>Birch</td>
<td>Ambient: 50-60 ppb Elevated: 1.5× ambient</td>
<td>Increase N concentration in birch litter</td>
<td>Parsons et al. (2008)</td>
</tr>
<tr>
<td>Kranzberg Forest, Germany</td>
<td>Beech and spruce</td>
<td>Ambient: 9-41 ppb Elevated: 2× ambient</td>
<td>Increased N concentration in beach leaf, but not in spruce needle</td>
<td>Kozovits et al. (2006)</td>
</tr>
<tr>
<td>Kranzberg Forest, Germany</td>
<td>Beech and spruce</td>
<td>Ambient: 9-41 ppb Elevated: 2× ambient</td>
<td>(1) Had no significant effects on spruce needle chemistry; (2) increased Ca concentration in beech leaves in monoculture, but had no impacts on other nutrients</td>
<td>Rodenkirchen et al. (2009)</td>
</tr>
<tr>
<td>Salerno, Italy</td>
<td>Holm oak</td>
<td>Non-filtered OTC: 29 ppb Filtered OTC: 17ppb</td>
<td>$O_2$ had no significant impacts on litter C, N, lignin and cellulose concentrations</td>
<td>Baldantoni et al. (2011)</td>
</tr>
<tr>
<td>Kuopio University Research Garden, Finland</td>
<td>Red Clover</td>
<td>Ambient: 25.7 ppb Elevated: 1.5× ambient</td>
<td>Increased the total phenolic content of leaves and had minor effects on the concentrations of individual phenolic compounds</td>
<td>Saviranta et al. (2010)</td>
</tr>
</tbody>
</table>

9.4.6.2  Decomposer Metabolism and Litter Decomposition

The above- and below-ground physiological changes caused by O$_3$ exposure cascade through the ecosystem and affect soil food webs. In the 2006 O$_3$ AQCD, there were very few studies on the effect of O$_3$ on the structure and function of soil food webs, except two studies conducted by Larson et al. (2002) and Phillips et al. (2002). Since the last O$_3$ AQCD, new studies have provided more information on how O$_3$ affects the metabolism of soil microbes and soil fauna.

Chung et al. (2006) found that the activity of the cellulose-degrading enzyme 1,4-β-glucosidase was reduced by 25% under elevated O$_3$ at Aspen FACE. The decrease in cellulose-degrading enzymatic activity was associated with the lower cellulose availability under elevated O$_3$ (Chung et al., 2006). However, a later study at the same site, which was conducted in the 10th year of the experiment, found that O$_3$ had no impact on cellulolytic activity in soil (Edwards and Zak, 2011). In a lysimeter study of beech trees (Fagus sylvatica) in Germany, soil enzyme activity was found to be suppressed by O$_3$ exposure (Esperschutz et al., 2009; Pritsch et al., 2009). Except for xylosidase, enzyme activities involved in plant cell wall
degradation (cellbiohydrolase, beta-glucosidase and glucuronidase) were decreased in rhizosphere soil samples under elevated O₃ (2 × ambient level) (Pritsch et al., 2009). Similarly, Chen et al. (2009) found O₃ exposure, with a 3-month AOT40 of 21-44 ppm-h, decreased the microbial metabolic capability in the rhizosphere and bulk soil of wheat, although the observed reduction in bulk soil was not significant.

Ozone-induced change in soil organisms’ activities could affect litter decomposition rates. Results of recent studies indicated that O₃ slightly reduced or had no impacts on litter decomposition (Liu et al., 2009b; Parsons et al., 2008; Kasurinen et al., 2006) (Baldantoni et al., 2011). The responses varied among species, sites and exposure length. Parsons et al. (2008) collected litter from aspen and birch seedlings at Aspen FACE site, and conducted a 23-month field litter incubation starting in 1999. They found that elevated O₃ had different impacts on the decomposition of aspen and birch litter. Elevated O₃ was found to reduce aspen litter decomposition. However, O₃ accelerated birch litter decomposition under ambient CO₂, but reduced it under elevated CO₂ (Parsons et al., 2008). Liu et al. (2009b) conducted another litter decomposition study at Aspen FACE from 2003 to 2006, when stand leaf area index (LAI) reached its maximum. During the 935-day field incubation, elevated O₃ was shown to reduce litter mass loss in the first year, but not in the second year. They suggested that higher initial tannin and phenolic concentrations under elevated O₃ reduced microbial activity in the first year (Liu et al., 2009b). In an OTC experiment, Kasurinen et al. (2006) collected silver birch leaf litter from three consecutive growing seasons and conducted three separate litter-bag incubation experiments. Litter decomposition was not affected by O₃ exposure in the first two incubations, but a slower decomposition rate was found in the third incubation. Their principle component analysis indicated that the litter chemistry changes caused by O₃ (decreased Mn, P, B and increased C:N) might be partially responsible for the decreased mass loss of their third incubation. In another OTC experiment, Baldantoni et al. (2011) found that O₃ significantly reduced leaf litter decomposition of Quercus ilex L, although litter C, N, lignin and cellulose concentrations were not altered by O₃ exposure.

### 9.4.6.3 Soil Respiration and Carbon Formation

Ozone could reduce the availability of photosynthates for export to roots, and thus, indirectly increase root mortality and turnover rates. Ozone has also been shown to reduce above-ground litter productivity and alter litter chemistry, which would affect the quality and quantity of the C supply to soil organisms (Section 9.4.6.1).

The complex interactions among those changes make it difficult to predict the response of soil C cycling under elevated O₃. The 2006 O₃ AQCD concluded that O₃ had no consistent impact on soil respiration (U.S. EPA, 2006b). Ozone could increase or decrease soil respiration, depending on the approach and timing of the measurements. Ozone may also alter soil C formation. However, very few experiments directly measured changes in soil organic matter content under O₃ fumigation (U.S. EPA, 2006b). Recent studies on soil respiration and soil C content
also found mixed responses. Most importantly, recent results from long-term fumigation experiments, such as the Aspen FACE experiment, suggest that ecosystem response to O₃ exposure can change over time. Observations made during the late exposure years can be inconsistent with those during the early years, highlighting the need for caution when assessing O₃ effects based on short-term studies (Table 9-7).

**Soil Respiration**

Ozone has shown inconsistent impacts on soil respiration. A sun-lit controlled-environment chamber study found that O₃ had no significant effects on soil respiration, fine root biomass or any of the soil organisms in a reconstructed ponderosa pine/soil-litter system (Tingey et al., 2006). In an adult European beech/Norway spruce forest at Kranzberg Forest, the free air O₃ fumigation (AOT40 of 10.2-117 ppm-h) increased soil respiration under both beech and spruce during a humid year (Nikolova et al., 2010). The increased soil respiration under beech has been accompanied by the increase in fine root biomass and ectomycorrhizal fungi diversity and turnover (Grebenec and Kraigher, 2007). The stimulating effect on soil respiration disappeared under spruce in a dry year, which was associated with a decrease in fine root production in spruce under drought. This finding suggested that drought was a more dominant stress than O₃ for spruce (Nikolova et al., 2010). Andersen et al. (2010) labeled the canopies of European beech and Norway spruce with CO₂ depleted in 13C at the same site. They did not observe any significant changes in soil respiration for either species.

The nearly 10 year long studies at Aspen FACE indicated that the response of soil respiration to O₃ interacted with CO₂ exposure and varied temporally (Table 9-7) (Pregitzer et al., 2008; Pregitzer et al., 2006; King et al., 2001). Ozone treatment alone generally had the lowest mean soil respiration rates, although those differences between control and elevated O₃ were usually not significant. However, soil respiration rates were different with O₃ alone and when acting in combination with elevated CO₂. In the first five years (1998-2002), soil respiration under +CO₂+O₃ treatment was similar to that under control and lower than that under +CO₂ treatment (Pregitzer et al., 2006; King et al., 2001). Since 2003, +CO₂+O₃ treatment started to show the greatest impact on soil respiration. Compared to elevated CO₂, soil respiration rate under +CO₂+O₃ treatment was 15-25% higher from 2003-2004, and 5-10% higher from 2005-2007 (Pregitzer et al., 2008; Pregitzer et al., 2006). Soil respiration was highly correlated with the biomass of roots with diameters of <2 mm and <1 mm, across plant community and atmospheric treatments. The authors suggested that the increase in soil respiration rate may be due to +CO₂+O₃ increased fine root (<1.0 mm) biomass production (Pregitzer et al., 2008).
Table 9-7  The temporal variation of ecosystem responses to O₃ exposure at Aspen FACE site

<table>
<thead>
<tr>
<th>Endpoint</th>
<th>Period of Measurement</th>
<th>Response</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Litter decomposition</td>
<td>1999-2001</td>
<td>O₃ reduced aspen litter decomposition. However, O₃ accelerated birch litter decomposition under ambient CO₂, but reduced it under elevated CO₂</td>
<td>Parsons et al. (2008)</td>
</tr>
<tr>
<td></td>
<td>2003-2006</td>
<td>O₃ reduced litter mass loss in the first year, but not in the second year.</td>
<td>Liu et al. (2009b)</td>
</tr>
<tr>
<td>Fine root production</td>
<td>1999</td>
<td>O₃ had no significant impact on fine root biomass</td>
<td>King et al. (2001)</td>
</tr>
<tr>
<td></td>
<td>2002, 2005</td>
<td>O₃ increased fine root biomass</td>
<td>Pregitzer et al. (2008)</td>
</tr>
<tr>
<td>Soil respiration</td>
<td>1998-1999</td>
<td>Soil respiration under +CO₂+O₃ treatment was lower than that under +CO₂ treatment</td>
<td>King et al. (2001)</td>
</tr>
<tr>
<td></td>
<td>2003-2007</td>
<td>Soil respiration under +CO₂+O₃ treatment was 5-25% higher than under elevated CO₂ treatment</td>
<td>Pregitzer et al. (2008; 2006)</td>
</tr>
<tr>
<td>Soil C formation</td>
<td>1998-2001</td>
<td>O₃ reduced the formation rates of total soil C by 51% and acid-insoluble soil C by 48%</td>
<td>Loya et al. (2003)</td>
</tr>
<tr>
<td></td>
<td>2004-2008</td>
<td>No significant effect of O₃ on the new C formed under elevated CO₂</td>
<td>Talhelm et al. (2009)</td>
</tr>
</tbody>
</table>

Changes in leaf chemistry and productivity due to O₃ exposure have been shown to affect herbivore growth and abundance (see Section 9.4.9.1). Canopy insects could affect soil carbon and nutrient cycling through frass deposition, or altering chemistry and quantity of litter input to the forest floor. A study at the Aspen FACE found that although elevated O₃ affected the chemistry of frass and greenfall, these changes had small impact on microbial respiration and no effect on nitrogen leaching (Hillstrom et al., 2010a). However, respiratory carbon loss and nitrate immobilization were nearly double in microcosms receiving herbivore inputs than those receiving no herbivore inputs (Hillstrom et al., 2010a).

**Soil Carbon Formation**

Ozone-induced reductions in plant growth can result in reduced C input to soil and therefore soil C content (Andersen, 2003). The simulations of most ecosystem models support this prediction (Ren et al., 2007b; Zhang et al., 2007a; Felzer et al., 2004). However, very few studies have directly measured soil C dynamics under elevated O₃. After the first four years of fumigation (from 1998 to 2001) at the Aspen FACE site, Loya et al. (2003) found that forest stands exposed to both elevated O₃ and CO₂ accumulated 51% less total soil C, and 48% less acid-insoluble soil C compared to stands exposed only to elevated CO₂. Soil organic carbon (SOC) was continuously monitored at the Aspen FACE site, and the later data showed that the initial reduction in new C formation (soil C derived from plant litter since the start of the experiment) by O₃ under elevated CO₂ is only a temporary effect (Table 9-7) (Talhelm et al., 2009). The amount of new soil C in the elevated CO₂ and
the combined elevated CO₂ and O₃ treatments has converged since 2002. There was no significant effect of O₃ on the new C formed under elevated CO₂ over the last four years of the study (2004-2008). Talhelm et al. (2009) suggested the observed reduction in the early years of the experiment might be driven by a suppression of C allocated to fine root biomass. During the early exposure years, O₃ had no significant impact on fine root production (King et al., 2001). However, the effect of O₃ on fine root biomass was observed later in the experiment. Ozone increased fine root production and the highest fine root biomass was observed under the combined elevated CO₂ and O₃ treatment in the late exposure years (Table 9-7) (Pregitzer et al., 2006). This increase in fine root production was due to changes in community composition, such as better survival of an O₃-tolerant aspen genotype, birch and maple, rather than changes in C allocation at the individual tree level (Pregitzer et al., 2008; Zak et al., 2007).

9.4.6.4 Nutrient Cycling

Ozone can affect nutrient cycling by changing nutrient release from litter, nutrient uptake by plants, and soil microbial activity. Nitrogen is the limiting nutrient for most temperate ecosystems, and several studies examined N dynamics under elevated O₃. Nutrient mineralization from decomposing organic matter is important for sustaining ecosystem production. Holmes et al. (2006) found that elevated O₃ decreased gross N mineralization at the Aspen FACE site, indicating that O₃ may reduce N availability. Other N cycling processes, such as NH₄⁺ immobilization, gross nitrification, microbial biomass N and soil organic N, were not affected by elevated O₃ (Holmes et al., 2006). Similarly, Kanerva et al. (2006) found total N, NO₃⁻, microbial biomass N, potential nitrification and denitrification in their meadow mesocosms were not affected by elevated O₃ (40-50 ppb). Ozone was found to decrease soil mineral N content at SoyFACE, which was likely caused by a reduction in plant material input and increased denitrification (Pujol Pereira et al., 2011). Ozone also showed small impact on other micro and macro nutrients. Liu et al. (2007a) assessed N, P, K, S, Ca, Mg, Mn, B, Zn and Cu release dynamics at Aspen FACE, and they found that O₃ had no effects on most nutrients, except to decrease N and Ca release from litter. These studies reviewed above suggest that soil N cycling processes are not affected or slightly reduced by O₃ exposure. However, in a lysimeter study with young beech trees, Stoelken et al. (2010) found that elevated O₃ stimulated N release from litter which was largely attributed to an enhanced mobilization of inert nitrogen fraction.

Using the SImple NItrogen Cycle model (SINIC), Hong et al. (2006) evaluated the impacts of O₃ exposure on soil N dynamics and streamflow nitrate flux. The detrimental effect of O₃ on plant growth was found to reduce plant uptake of N and therefore increase nitrate leaching. Their model simulation indicated that ambient O₃ exposure increased the mean annual stream flow nitrate export by 12% (0.042 g N/m²·year) at the Hubbard Brook Experimental Watershed from 1964-1994 (Hong et al., 2006).
9.4.6.5 Dissolved Organic Carbon and Biogenic Trace Gases Emission

The O₃-induced changes in plant growth, C and N fluxes to soil and microbial metabolism can alter other biogeochemical cycling processes, such as soil dissolved organic carbon (DOC) turnover and trace gases emission.

Jones et al. (2009) collected fen cores from two peatlands in North Wales, UK and exposed them to one of four levels of O₃ (AOT40 of 0, 3.69, 5.87 and 13.80 ppm-h for 41 days). They found the concentration of porewater DOC in fen cores was significantly decreased by increased O₃ exposure. A reduction of the low molecular weight fraction of DOC was concurrent with the observed decrease in DOC concentration. Their results suggested that O₃ damage to overlying vegetation may decrease utilizable C flux to soil. Microbes, therefore, have to use labile C in the soil to maintain their metabolism, which, the authors hypothesized, leads to a decreased DOC concentration with a shift of the DOC composition to more aromatic, higher molecular weight organic compounds.

Several studies since the 2006 O₃ AQCD have examined the impacts of O₃ on nitrous oxide (N₂O) and methane (CH₄) emission. Kanerva et al. (2007) measured the fluxes of N₂O and CH₄ in meadow mesocosms, which were exposed to elevated CO₂ and O₃ in OTCs in south-western Finland. They found that the daily N₂O fluxes were decreased in the NF+O₃ (non-filtered air + elevated O₃, 40-50 ppb) after three seasons of exposure. Elevated O₃ alone or combined with CO₂ did not have any significant effect on the daily fluxes of CH₄ (Kanerva et al., 2007). In another study conducted in central Finland, the 4 year open air O₃ fumigation (AOT40 of 20.8-35.5 ppm-h for growing season) slightly increased potential CH₄ oxidation by 15% in the peatland microcosms, but did not affect the rate of potential CH₄ production or net CH₄ emissions, which is the net result of the potential CH₄ production and oxidation (Morsky et al., 2008). However, several studies found that O₃ could significantly reduce CH₄ emission. Toet et al. (2011) exposed peatland mesocosms to O₃ in OTCs for two years, and found that CH₄ emissions were significantly reduced by about 25% during midsummer periods of both years. In an OTC study of rice paddy, Zheng et al. (2011) found that the daily mean CH₄ emissions were significantly lower under elevated O₃ treatments than those in charcoal-filtered air and nonfiltered air treatments. They found that the seasonal mean CH₄ emissions were negatively related with AOT40, but positively related to the relative rice yield, aboveground biomass and underground biomass.

9.4.6.6 Summary

Since the 2006 O₃ AQCD, more evidence has shown that although the responses are often site specific, O₃ altered the quality and quantity of litter input to soil, microbial community composition, and C and nutrient cycling. Biogeochemical cycling of below-ground processes is fueled by C input from plants. Studies at the leaf and plant
level have provided biologically plausible mechanisms, such as reduced photosynthetic rates, increased metabolic cost, and reduced root C allocation for the association of O₃ exposure and the alteration of below-ground processes.

Results from Aspen FACE and other experimental studies consistently found that O₃ reduced litter production and altered C chemistry, such as soluble sugars, soluble phenolics, condensed tannins, lignin, and macro/micro nutrient concentration in litter (Parsons et al., 2008; Kasurinen et al., 2006; Liu et al., 2005). Under elevated O₃, the changes in substrate quality and quantity could alter microbial metabolism and therefore soil C and nutrient cycling. Several studies indicated that O₃ suppressed soil enzyme activities (Pritsch et al., 2009; Chung et al., 2006). However, the impact of O₃ on litter decomposition was inconsistent and varied among species, sites and exposure length. Similarly, O₃ had inconsistent impacts on dynamics of micro and macro nutrients.

Studies from the Aspen FACE experiment suggested that the response of below-ground C cycle to O₃ exposure, such as litter decomposition, soil respiration and soil C content, changed over time. For example, in the early part of the experiment (1998-2003), O₃ had no impact on soil respiration but reduced the formation rates of total soil C under elevated CO₂. However, after 10-11 years of exposure, O₃ was found to increase soil respiration but have no significant impact on soil C formation under elevated CO₂.

The evidence is sufficient to infer that there is a causal relationship between O₃ exposure and the alteration of below-ground biogeochemical cycles.

**9.4.7 Community Composition**

The effects of O₃ on species competition (AX9.3.3.4) and community composition (AX9.6.4) were summarized in the 2006 O₃ AQCD. Plant species differ in their sensitivity to O₃. Further, different genotypes of a given species also vary in their sensitivity. This differential sensitivity could change the competitive interactions that lead to loss in O₃ sensitive species or genotypes. In addition, O₃ exposure has been found to alter reproductive processes in plants (see Section 9.4.3.3). Changes in reproductive success could lead to changes in species composition. However, since ecosystem-level responses result from the interaction of organisms with one another and with their physical environment, it takes longer for a change to develop to a level of prominence at which it can be identified and measured. A shift in community composition in forest and grassland ecosystems noted in the 2006 O₃ AQCD has continued to be observed from experimental and gradient studies. Additionally, research since the last review has shown that O₃ can alter community composition and diversity of soil microbial communities.
9.4.7.1 Forest

In the San Bernardino Mountains in southern California, O₃ pollution caused a significant decline in ponderosa pine (*Pinus ponderosa*) and Jeffrey pine (*Pinus jeffreyi*) (U.S. EPA, 2006b). Pine trees in the young mature age class group exhibited higher mortality rates compared with mature trees at a site with severe O₃ visible foliar injury. The vulnerability of young mature pines was most likely caused by the fact that trees in this age class were emerging into the canopy, where higher O₃ concentrations were encountered (McBride and Laven, 1999). Because of the loss of O₃-sensitive pines, mixed forests of ponderosa pine, Jeffery Pine and white fir (*Abies concolor*) shifted to predominantly white fir (Miller, 1973). Ozone may have indirectly caused the decline in understory diversity in coniferous forests in the San Bernardino Mountains through an increase in pine litterfall. This increase in litterfall from O₃ exposure results in an understory layer that may prohibit the establishment of native herbs, but not the exotic annual *Galium aparine* (Allen et al., 2007).

Ozone damage to conifer forests has also been observed in several other regions. In the Valley of Mexico, a widespread mortality of sacred fir (*Abies religiosa*) was observed in the heavily polluted area of the Desierto de los Leones National Park in the early 1980s (de Lourdes de Bauer and Hernandez-Tejeda, 2007; Fenn et al., 2002). Ozone damage was widely believed to be an important causal factor in the dramatic decline of sacred fir. In alpine regions of southern France and the Carpathians Mountains, O₃ was also considered as the major cause of the observed decline in cembran pine (*Pinus cembra*) (Wieser et al., 2006). However, many environmental factors such as light, temperature, nutrient and soil moisture, and climate extremes such as unusual dry and wet periods could interact with O₃ and alter the response of forest to O₃ exposure. For those pollution gradient studies, several confounding factors, such as drought, insect outbreak and forest management, may also contribute to or even be the dominant factors causing the mortality of trees (de Lourdes de Bauer and Hernandez-Tejeda, 2007; Wieser et al., 2006).

Recent evidence from long-term free O₃ fumigation experiments provided additional support for the potential impacts of O₃ on species competition and community composition changes in forest ecosystems. At the Aspen FACE site, community composition at both the genetic and species levels was altered after seven years of fumigation with O₃ (Kubiske et al., 2007). In the pure aspen community, O₃ fumigation reduced growth and increased mortality of sensitive clone 259, while the O₃ tolerant clone 8L emerged as the dominant clone. Growth of clone 8L was even greater under elevated O₃ compared to controls, probably due to O₃ alleviated competitive pressure on clone 8L by reducing growth of other clones. In the mixed aspen-birch and aspen-maple communities, O₃ reduced the competitive capacity of aspen compared to birch and maple (Kubiske et al., 2007). In a phytotron study, O₃ fumigation reduced growth of beech but not spruce in mixed culture, suggesting a higher susceptibility of beech to O₃ under interspecific competition (Kozovits et al., 2005).
9.4.7.2 Grassland and Agricultural Land

The response of managed pasture, often cultivated as a mixture of grasses and clover, to O₃ pollution has been studied for many years. The tendency for O₃-exposure to shift the biomass of grass-legume mixtures in favor of grass species, reported in the previous O₃ AQCD has been generally confirmed by recent studies. In a mesocosm study, *Trifolium repens* and *Lolium perenne* mixtures were exposed to an episodic rural O₃ regime within solardomes for 12 weeks. *T. repens* showed significant changes in biomass but not *L. perenne*, and the proportion of *T. repens* decreased in O₃-exposed mixtures compared to the control (Hayes et al., 2009). The changes in community composition of grass-legume-forb mixtures were also observed at the Le Mouret FACE experiment, Switzerland. During the 5-year O₃ fumigation (AOT40 of 13.3-59.5 ppm-h), the dominance of legumes in fumigated plots declined more quickly than those in the control plots (Volk et al., 2006). However, Stampfli and Fuhrer (2010) reanalyzed the species and soil data and suggested that Volk et al. (2006) overestimated the O₃ effect. Stampfli and Fuhrer (2010) found that the difference in the species dynamics between control and O₃ treatment was more caused by heterogeneous initial conditions than O₃ exposure. Several studies also suggested that mature/species-rich ecosystems were more resilient to O₃ exposure. At another FACE experiment, located at Alp Flix, Switzerland, O₃ fumigation (AOT40 of 15.2-64.9 ppm-h) showed no significant impact on community composition of this species-rich pasture (Bassin et al., 2007b). Although most studies demonstrated an increase in grass:forb ratio with O₃ exposure (Hayes et al., 2009; U.S. EPA, 2006b), a study on a simulated upland grassland community showed that O₃ reduced the grass:forb ratio (Hayes et al., 2010) which may be due to the grass species in this community. The grass species studied by Hayes et al. (2010), *Anthoxanthum odoratum*, was more sensitive to O₃ than other grass species such as *L. perenne* (Hayes et al., 2009). Pfleeger et al. (2010) collected seed bank soil from an agricultural field and examined how the plant community responded over several generations to elevated O₃ exposures. Sixty plant species from 22 families emerged in the chambers over their four year study. Overall, they found that O₃ appeared to have small impacts on seed germination and only a minor effect on species richness of pioneer plant communities.

Several review papers have discussed the physiological and ecological characteristics of O₃-sensitive herbaceous plants. Hayes et al. (2007) assessed species traits associated with O₃ sensitivity by the changes in biomass caused by O₃ exposure. Plants of the therophyte (e.g., annual) life form were particularly sensitive to O₃. Species with higher mature leaf N concentration tended to be more sensitive than those with lower leaf N concentration. Plants growing under high oxidative stress environments, such as high light or high saline, were more sensitive to O₃. Using the same dataset from Hayes et al. (2007), Mills et al. (2007b) identified the O₃ sensitive communities. They found that the largest number of these O₃ sensitive communities were associated with grassland ecosystems. Among grassland ecosystems, alpine grassland, sub-alpine grassland, woodland fringe, and dry grassland were identified as the most sensitive communities.
Several methods have been used to study microbial composition changes associated with elevated O₃. Phospholipid fatty acid (PLFA) analysis is widely used to determine whether O₃ elicits an overall effect on microbial community composition. However, since PLFA markers cover a broad range of different fungi, resolution of this method may be not fine enough to detect small changes in the composition of fungal communities. Methods, such as microscopic analyses and polymerase chain reaction–denaturing gradient gel electrophoresis (PCR–DGGE), have better resolution to specifically analyze the fungal community composition. The resolution differences among those methods needs to be considered when assessing the O₃ impact on microbial community composition.

Kanerva et al. (2008) found that elevated O₃ (40-50 ppb) decreased total, bacterial, actinobacterial and fungal PLFA biomass values as well as fungal:bacterial PLFA biomass ratio in their meadow mesocosms in south-western Finland. The relative proportions of individual PLFAs between the control and elevated O₃ treatments were significantly different, suggesting that O₃ modified the structure of the microbial community. Morsky et al. (2008) exposed boreal peatland microcosms to elevated O₃, with growing season AOT40 of 20.8-35.3 ppm-h, in an open-air O₃ exposure field in Central Finland. They also found that microbial composition was altered after three growing seasons with O₃ fumigation, as measured by PLFA. Ozone tended to increase the presence of Gram-positive bacteria and the biomass of fungi in the peatland microcosms. Ozone also resulted in higher microbial biomass, which co-occurred with the increases in concentrations of organic acids and leaf density of sedges (Morsky et al., 2008). In a lysimeter experiment in Germany, O₃ was found to alter the PLFA profiles in the upper 0-20 cm rhizosphere soil of European beech. Elevated O₃ reduced bacterial abundance but had no detectable effect on fungal abundance (Pritsch et al., 2009). Using microscopic analyses, Kasurinen et al. (2005) found that elevated O₃, with 5 or 6 months of AOT40 of 20.6-30.9 ppm-h, decreased the proportions of black and liver-brown mycorrhizas and increased that of light brown/orange mycorrhizas. In an herbaceous plant study, SSCP (single-strand conformation polymorphism) profiles indicated that O₃ stress (about 75 ppb) had a very small effect on the structural diversity of the bacterial community in rhizospheres (Dohrmann and Tebbe, 2005). At the Aspen FACE site, O₃ had no significant effect on fungal relative abundance, as indicated by PLFA profile. However, elevated O₃ altered fungal community composition, according to the identification of 39 fungal taxonomic units from soil using polymerase chain reaction–denaturing gradient gel electrophoresis (PCR-DGGE) (Chung et al., 2006). In another study at Aspen FACE, phylogenetic analysis suggested that O₃ exposure altered the agaricomycete community. The ectomycorrhizal communities developing under elevated O₃ had higher proportions of Cortinarius and Inocybe species, and lower proportions of Laccaria and Tomentella (Edwards and Zak, 2011). Ozone was found to change microbial community composition in an agricultural system. Chen et al. (2010b) found elevated O₃ (100-150 ppb) had significant effects on soil microbial composition expressed as PLFA percentage in a rice paddy in China.
9.4.7.4 Summary

In the 2006 O₃ AQCD, the impact of O₃ exposure on species competition and community composition was assessed. Ozone was found to cause a significant decline in ponderosa and Jeffrey pine in the San Bernardino Mountains in southern California. Ozone exposure also tended to shift the grass-legume mixtures in favor of grass species (U.S. EPA, 2006b). Since the 2006 O₃ AQCD, more evidence has shown that O₃ exposure changed the competitive interactions and could lead to loss of O₃ sensitive species or genotypes. Studies at plant level found that the severity of O₃ damage on growth, reproduction, and foliar injury varied among species, which provided the biological plausibility for the alteration of community composition. Additionally, research since the last review has shown that O₃ can alter community composition and diversity of soil microbial communities.

The decline of conifer forests under O₃ exposure was continually observed in several regions. Ozone damage was believed to be an important causal factor in the dramatic decline of sacred fir in the valley of Mexico (de Lourdes de Bauer and Hernandez-Tejeda, 2007), as well as cembran pine in southern France and the Carpathian Mountains (Wieser et al., 2006). Results from the Aspen FACE site indicated that O₃ could alter community composition of broadleaf forests as well. At the Aspen FACE site, O₃ reduced growth and increased mortality of a sensitive aspen clone, while the O₃ tolerant clone emerged as the dominant clone in the pure aspen community. In the mixed aspen-birch and aspen-maple communities, O₃ reduced the competitive capacity of aspen compared to birch and maple (Kubiske et al., 2007).

The tendency for O₃-exposure to shift the biomass of grass-legume mixtures in favor of grass species, was reported in the 2006 O₃ AQCD and has been generally confirmed by recent studies. However, in a high elevation mature/species-rich grass-legume pasture, O₃ fumigation showed no significant impact on community composition (Bassin et al., 2007b).

Ozone exposure not only altered community composition of plant species, but also microorganisms. The shift in community composition of bacteria and fungi has been observed in both natural and agricultural ecosystems, although no general patterns could be identified (Kanerva et al., 2008; Morsky et al., 2008; Kasurinen et al., 2005).

The evidence is sufficient to conclude that there is likely to be a causal relationship between O₃ exposure and the alteration of community composition of some ecosystems.

9.4.8 Factors that Modify Functional and Growth Response

Many biotic and abiotic factors, including insects, pathogens, root microbes and fungi, temperature, water and nutrient availability, and other air pollutants, as well as elevated CO₂, influence or alter plant response to O₃. These modifying factors were
comprehensively reviewed in AX9.3 of the 2006 O₃ AQCD and thus, this section serves mainly as a brief summary of the previous findings. A limited number of new studies published since the 2006 O₃ AQCD add to the understanding of the role of these interactions in modifying O₃-induced plant responses. Many of these modifying factors and interactions are integrated into discussions elsewhere in this chapter and the reader is directed to those sections.

9.4.8.1 Genetics

It is well known that species vary greatly in their responsiveness to O₃. Even within a given species, individual genotypes or populations can also vary significantly with respect to O₃ sensitivity (U.S. EPA, 2006b). Therefore, caution should be taken when considering a species’ degree of sensitivity to O₃. Plant response to O₃ is determined by genes that are directly related to oxidant stress and to an unknown number of genes that are not specifically related to oxidants, but instead control leaf and cell wall thickness, stomatal conductance, and the internal architecture of the air spaces. It is rarely the case that single genes are responsible for O₃ tolerance. Studies using molecular biological tools and transgenic plants have positively verified the role of various genes and gene products in O₃ tolerance and are continuing to increase the understanding of O₃ toxicity and differences in O₃ sensitivity. See Section 9.3.3.2 of this document for a discussion of recent studies related to gene expression changes in response to O₃.

9.4.8.2 Environmental Biological Factors

As stated in the 2006 O₃ AQCD, the biological factors within the plant’s environment that may influence its response to O₃ encompass insects and other animal pests, diseases, weeds, and other competing plant species. Ozone may influence the severity of a disease or infestation by a pest or weed, either by direct effects on the causal species, or indirectly by affecting the host, or both. In addition, the interaction between O₃, a plant, and a pest, pathogen, or weed may influence the response of the target host species to O₃ (U.S. EPA, 2006b). Several recent studies on the effects of O₃ on insects via their interactions with plants are discussed in Section 9.4.9.1. In addition, O₃ has also been shown to alter soil fauna communities (Section 9.4.9.2).

In contrast to detrimental biological interactions, there are mutually beneficial relationships or symbioses involving higher plants and bacteria or fungi. These include (1) the nitrogen-fixing species Rhizobium and Frankia that nodulate the roots of legumes and alder and (2) the mycorrhizae that infect the roots of many crop and tree species, all of which may be affected by exposure of the host plants to O₃. Some discussion of mycorrhizae can be found in Section 9.4.6.
In addition to the interactions involving animal pests, O\textsubscript{3} also has indirect effects on higher herbivorous animals, e.g., livestock, due to O\textsubscript{3}-induced changes in feed quality. Recent studies on the effects of O\textsubscript{3} on nutritive quality of plants are discussed in Section 9.4.4.2.

Intra- and interspecific competition are also important factors in determining vegetation response to O\textsubscript{3}. Plant competition involves the ability of individual plants to acquire the environmental resources needed for growth and development: light, water, nutrients, and space. Intraspecific competition involves individuals of the same species, typically in monoculture crop situations, while interspecific competition refers to the interference exerted by individuals of different species on each other when they are in a mixed culture. This topic was previously reviewed in AX9.3.3.4 of the 2006 O\textsubscript{3} AQCD. Recent studies on competition and its implications for community composition are discussed in Section 9.4.7.

### 9.4.8.3 Physical Factors

Physical or abiotic factors play a large role in modifying plant response to O\textsubscript{3}, and have been extensively discussed in previous O\textsubscript{3} AQCDs. This section summarizes those findings as well as recent studies published since the last review.

Although some studies have indicated that O\textsubscript{3} impact significantly increases with increased ambient temperature (Ball et al., 2000; Mills et al., 2000), other studies have indicated that temperature has little effect (Balls et al., 1996; Fredericksen et al., 1996). A recent study by Riikonen et al. (2009) at the Ruohoniemi open air exposure field in Kuopio, Finland found that the effects of temperature and O\textsubscript{3} on total leaf area and photosynthesis of *Betula pendula* were counteractive. Elevated O\textsubscript{3} reduced the saplings’ ability to utilize the warmer growth environment by increasing the stomatal limitation for photosynthesis and by reducing the redox state of ascorbate in the apoplast in the combination treatment as compared to temperature alone (Riikonen et al., 2009).

Temperature affects the rates of all physiological processes based on enzyme catalysis and diffusion; each process and overall growth (the integral of all processes) has a distinct optimal temperature range. It is important to note that a plant’s response to changes in temperature will depend on whether it is growing near its optimum temperature for growth or near its maximum temperature (Rowland-Bamford, 2000). However, temperature is very likely an important variable affecting plant O\textsubscript{3} response in the presence of the elevated CO\textsubscript{2} levels contributing to global climate change. In contrast, some evidence suggests that O\textsubscript{3} exposure sensitizes plants to low temperature stress (Colls and Unsworth, 1992) and, also, that O\textsubscript{3} decreases below-ground carbohydrate reserves, which may lead to responses in perennial species ranging from rapid demise to impaired growth in subsequent seasons (i.e., carry-over effects) (Andersen et al., 1997).
Light, a component of the plant’s physical environment, is an essential “resource” of energy content that drives photosynthesis and C assimilation. It has been suggested that increased light intensity may increase the O\textsubscript{3} sensitivity of light-tolerant species while decreasing that of shade-tolerant species, but this appears to be an oversimplification with many exceptions. Several studies suggest that the interaction between O\textsubscript{3} sensitivity and light environment is complicated by the developmental stage as well as the light environment of individual leaves in the canopy (Kitao et al., 2009; Topa et al., 2001; Chappelka and Samuelson, 1998).

Although the relative humidity of the ambient air has generally been found to increase the effects of O\textsubscript{3} by increasing stomatal conductance (thereby increasing O\textsubscript{3} flux into the leaves), abundant evidence also indicates that the ready availability of soil moisture results in greater O\textsubscript{3} sensitivity (Mills, 2002). The partial “protection” against the effects of O\textsubscript{3} afforded by drought has been observed in field experiments (Low et al., 2006) and modeled in computer simulations (Broadmeadow and Jackson, 2000). Conversely, drought may exacerbate the effects of O\textsubscript{3} on plants (Pollastrini et al., 2010; Gruulke et al., 2003b). There is also some evidence that O\textsubscript{3} can predispose plants to drought stress (Maier-Maercker, 1998). Hence, the nature of the response is largely species-specific and will depend to some extent upon the sequence in which the stressors occur.

9.4.8.4 Interactions with Other Pollutants

Ozone-nitrogen interactions

Elevated O\textsubscript{3} exposure and N deposition often co-occur. However, the interactions of O\textsubscript{3} exposure and N deposition on vegetation are complex and less well understood compared to their independent effects. Consistent with the conclusion of the 2006 O\textsubscript{3} AQCD, the limited number of studies published since the last review indicated that the interactive effects of N and O\textsubscript{3} varied among species and ecosystems (Table 9-8). Nitrogen deposition could stimulate relative growth rate (RGR), and lead to increased stomatal conductance. Therefore, plants might become more susceptible to O\textsubscript{3} exposure. Alternatively, N deposition may increase the availability of photosynthates for use in detoxification and plants could become more tolerant to O\textsubscript{3} (Bassin et al., 2007a). Elevated O\textsubscript{3} exposure and N deposition could also act in concert to increase plant susceptibility to disease (von Tiedemann, 1996). To better understand these interactions in ecosystems across the U.S., more information is needed considering combined O\textsubscript{3} exposure and N deposition related effects.

Only a few recent studies have investigated the interactive effects of O\textsubscript{3} and N in the United States. Gruulke et al. (2005) measured stomatal conductance of California black oak (Quercus kelloggii) at a long-term N-enrichment site located in the San Bernardino Mountains, which is accompanied by high O\textsubscript{3} exposure (80 ppb, 24-h avg. over a six month growing season). The authors found that N amendment led to poor stomatal control in full sun in midsummer of the average precipitation.
years, but enhanced stomatal control in shade leaves of California black oak. In an
OTC study, Handley and Grulke (2008) found that O₃ lowered photosynthetic ability
and water-use efficiency, and increased leaf chlorosis and necrosis of California
black oak. Nitrogen fertilization tended to reduce plant sensitivity to O₃ exposure;
however, the interaction was not statistically significant. In another study, Grulke et
al. (2008) reported that various lines of phenomenological and experimental evidence
indicate that N deposition and O₃ pollution contribute to the susceptibility of forests
to wildfire in the San Bernadino Mountains by increasing stress due to drought,
weakening trees, and predisposing them to bark beetle infestation (U.S. EPA, 2008b).

Studies conducted outside the U.S. are also summarized in Table 9-8. Generally, the
responses were species specific. The O₃-induced reduction in photosynthetic rate and
biomass loss were greater in the relatively high N treatment for watermelon (Citrullus
lanants) (Calatayud et al., 2006) and Japanese beech (Fagus crenata) seedlings
(Yamaguchi et al., 2007). However, there was no significant interactive effect of O₃
and N on biomass production for Quercus serrata seedlings (Watanabe et al., 2007),
young Norway spruce (Picea abies) trees (Thomas et al., 2005), and young European
beech (Fagus sylvatica) trees Thomas et al. (2006).
Table 9-8  Response of plants to the interactive effects of elevated O₃ exposure and nitrogen enrichment.

<table>
<thead>
<tr>
<th>Site</th>
<th>Species</th>
<th>Ozone exposure</th>
<th>N addition</th>
<th>Responses</th>
<th>References</th>
</tr>
</thead>
<tbody>
<tr>
<td>San Bernardino Mountains, U.S.</td>
<td>California black oak (Quercus kelloggii)</td>
<td>80 ppb</td>
<td>0, and 50 kg N/ha·yr</td>
<td>N-amended trees had lower late summer C gain and greater foliar chlorosis in the drought year, and poor stomatal control and lower leaf water use efficiency and in midsummer of the average precipitation year.</td>
<td>Grulke et al. (2005)</td>
</tr>
<tr>
<td>San Bernardino Mountains, U.S.</td>
<td>California black oak (Quercus kelloggii)</td>
<td>0, 75, and 150 ppb</td>
<td>0, and 50 kg N/ha·yr</td>
<td>N fertilization tended to reduce plant sensitivity to O₃ exposure; however the interaction was not statistically significant.</td>
<td>Handley and Gruke (2008)</td>
</tr>
<tr>
<td>Switzerland</td>
<td>Spruce trees (Picea abies)</td>
<td>Filtered (19.4-28.1 ppb); Ambient (37.6-47.4 ppb)</td>
<td>0, 20, 40 and 80 kg N/ha·yr</td>
<td>Higher N levels alleviated the negative impact of O₃ on root starch concentrations</td>
<td>Thomas et al. (2005)</td>
</tr>
<tr>
<td>Switzerland</td>
<td>Beech trees (Fagus sylvatica)</td>
<td>Filtered (19.4-28.1 ppb); Ambient (37.6-47.4 ppb)</td>
<td>0, 20, 40 and 80 kg N/ha·yr</td>
<td>N addition amplified the negative effects of O₃ on leaf area and shoot elongation.</td>
<td>Thomas et al. (2006)</td>
</tr>
<tr>
<td>Switzerland Alpine pasture</td>
<td>Ambient (AOT40 of 11.1-12.6 ppm-h); 1.2 ambient (AOT40 of 15.2-29.5 ppm-h); 1.6 ambient (28.4-64.9 ppm-h)</td>
<td>0, 5, 10, 25, 50 kg N/ha·yr</td>
<td>The positive effects of N addition on canopy greening were counteracted by accelerated leaf senescence in the highest O₃ treatment.</td>
<td>Bassin et al. (2007b)</td>
<td></td>
</tr>
<tr>
<td>Switzerland Alpine pasture</td>
<td>Ambient (AOT40 of 11.1-12.6 ppm-h); 1.2 ambient (AOT40 of 15.2-29.5 ppm-h); 1.6 ambient (28.4-64.9 ppm-h)</td>
<td>0, 5, 10, 25, 50 kg N/ha·yr</td>
<td>Only a small number of species showed significant O₃ and N interactive effects on leaf chlorophyll concentration, leaf weight and change in ¹⁸O, and the patterns were not consistent.</td>
<td>Bassin et al. (2009)</td>
<td></td>
</tr>
<tr>
<td>Switzerland Alpine pasture</td>
<td>Ambient (AOT40 of 11.1-12.6 ppm-h); 1.2 ambient (AOT40 of 15.2-29.5 ppm-h); 1.6 ambient (28.4-64.9 ppm-h)</td>
<td>0, 5, 10’ 25, 50 kg N/ha·yr</td>
<td>Highest N addition resulted in carbon loss, but there was no interaction between O₃ and N treatments.</td>
<td>Volk et al. (2011)</td>
<td></td>
</tr>
<tr>
<td>Site</td>
<td>Species</td>
<td>Ozone exposure</td>
<td>N addition</td>
<td>Responses</td>
<td>References</td>
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<tr>
<td>Spain</td>
<td>Watermelon <em>(Citrullus lanatus)</em></td>
<td>O₃ free (AOT₄₀ of 0 ppm-h), Ambient (AOT₄₀ of 5.1-6.3 ppm-h), Elevated O₃ (AOT₄₀ of 32.5-35.6 ppm-h)</td>
<td>140, 280, and 436 kg N/ha·yr</td>
<td>High N concentration enhanced the detrimental effects of O₃ on Chlorophyll a fluorescence parameters, lipid peroxidation, and the total yield.</td>
<td>Calatayud et al. (2006)</td>
</tr>
<tr>
<td>Spain</td>
<td>Clover <em>(Trifolium striatum)</em></td>
<td>Filtered (24-h avg. of 8-22 ppb), Ambient (29-34 ppb), Elevated O₃ (35-56 ppb)</td>
<td>10, 30, and 60 kg N/ha·yr</td>
<td>O₃ reduced total aerial biomass. N fertilization counterbalanced O₃-induced effects only when plants were exposed to moderate O₃ levels (ambient) but not under elevated O₃ concentrations.</td>
<td>Sanz et al. (2007)</td>
</tr>
<tr>
<td>Japan</td>
<td>Japanese beech seedlings <em>(Fagus crenata)</em></td>
<td>Filtered (24-h avg. of 10.3-13.2 ppb), Ambient (42.0-43.3 ppb), 1.5 Ambient (62.6-63.9 ppb), 2.0 ambient (82.7-84.7 ppb)</td>
<td>0, 20 and 50 kg N/ha·yr</td>
<td>The O₃-induced reduction in net photosynthesis and whole-plant dry mass were greater in the relatively high N treatment than that in the low N treatment.</td>
<td>Yamaguchi et al. (2007)</td>
</tr>
<tr>
<td>Japan</td>
<td>Japanese tree seedlings <em>(Quercus serrata)</em></td>
<td>Filtered (24-h avg. of 10.3-13.2 ppb), Ambient (42.0-43.3 ppb), 1.5 ambient (62.6-63.9 ppb), 2.0 ambient (82.7-84.7 ppb)</td>
<td>0, 20 and 50 kg N/ha·yr</td>
<td>No significant interactive effects of O₃ and N load on the growth and net photosynthetic rate were detected.</td>
<td>Watanabe et al. (2007)</td>
</tr>
</tbody>
</table>

**Ozone-carbon dioxide interactions**

Several decades of research has shown that exposure to elevated CO₂ increases photosynthetic rates (Bernacchi et al., 2006; Bernacchi et al., 2005; Tissue et al., 1999; Tissue et al., 1997; Will and Ceulemans, 1997), decreases stomatal conductance (Ainsworth and Rogers, 2007; Paoletti et al., 2007; Bernacchi et al., 2006; Leakey et al., 2006; Medlyn et al., 2001) and generally increases the growth of plants (McCarthy et al., 2009; Norby et al., 2005). This is in contrast to the decrease in photosynthesis and growth in many plants that are exposed to elevated O₃. The interactive effects on vegetation have been the subject of research in the past two decades due to the implications on productivity and water use of ecosystems. This area of research was discussed in detail in AX9.3.8.1 of the 2006 O₃ AQCD and the conclusions made then are still relevant (U.S. EPA, 2006b).
The bulk of the available evidence shows that, under the various experimental conditions used (which almost exclusively employed abrupt or “step” increases in CO₂ concentration, as discussed below), increased CO₂ levels (ambient + 200 to 400 ppm) may protect plants from the negative effects of O₃ on growth. This protection may be afforded in part by CO₂ acting together with O₃ in inducing stomatal closure, thereby reducing O₃ uptake, and in part by CO₂ reducing the negative effects of O₃ on Rubisco and its activity in CO₂-fixation. Although both CO₂-induced and O₃-induced decreases in stomatal conductance have been observed primarily in short-term studies, recent data show a long-term and sustained reduction in stomatal conductance under elevated CO₂ for a number of species (Ainsworth and Long, 2005; Ellsworth et al., 2004; Gunderson et al., 2002). Instances of increased stomatal conductance have also been observed in response to O₃ exposure, suggesting partial stomatal dysfunction after extended periods of exposure (Paoletti and Gruiske, 2010; Gruiske et al., 2007a; Maier-Maercker, 1998).

Important caveats must be raised with regard to the findings presented in published research. The first caveat concerns the distinctly different natures of the exposures to O₃ and CO₂ experienced by plants in the field. Changes in the ambient concentrations of these gases have very different dynamics. In the context of climate change, CO₂ levels increase relatively slowly (globally 2 ppm/year) and may change little over several seasons of growth. On the other hand, O₃ presents a fluctuating stressor with considerable hour-to-hour, day-to-day and regional variability (Polle and Pell, 1999). Almost all of the evidence presented comes from experimentation involving plants subjected to an abrupt step increase to a higher, steady CO₂ concentration. In contrast, the O₃ exposure concentrations usually varied from day to day. Luo and Reynolds (1999), Hui et al. (2002), and Luo (2001) noted the difficulties in predicting the likely effects of a gradual CO₂ increase from experiments involving a step increase or those using a range of CO₂ concentrations. It is also important to note that the levels of elevated CO₂ in many of the studies will not be experienced in the field for 30 or 40 years, but elevated levels of O₃ can occur presently in several areas of the United States. Therefore, the CO₂ × O₃ interaction studies may be less relevant for current ambient conditions.

Another caveat concerns the interactions of O₃ and CO₂ with other climatic variables, such as temperature and precipitation. In light of the key role played by temperature in regulating physiological processes and modifying plant response to increased CO₂ levels (Morison and Lawlor, 1999; Long, 1991) and the knowledge that relatively modest increases in temperature may lead to dramatic consequences in terms of plant development (Lawlor, 1998), it is important to consider that studying CO₂ and O₃ interactions alone may not create a complete understanding of effects on plants under future climate change.
Insects and Other Wildlife

9.4.9.1 Insects

Insects may respond indirectly to changes in plants (i.e., increased reactive oxygen species, altered phytochemistry, altered nutrient content) that occur under elevated O₃ conditions, or O₃ can have a direct effect on insect performance (Menendez et al., 2009). Effects of O₃ on insects occur at the species level (i.e., growth, survival, reproduction, development, feeding behavior) and at the population and community-level (i.e., population growth rate, community composition). In general, effects of O₃ on insects are highly context- and species-specific (Lindroth, 2010; Bidart-Bouzat and Imeh-Nathaniel, 2008). Furthermore, plant responses to O₃ exposure and herbivore attack have been demonstrated to share signaling pathways, complicating characterization of these stressors (Lindroth, 2010; Menendez et al., 2010, 2009). Although both species-level and population and community-level responses to elevated O₃ are observed in field and laboratory studies discussed below, there is no consensus on how insects respond to feeding on O₃–exposed plants.

Species-level responses

In considering insect growth, survival and reproduction in elevated O₃ conditions, several studies have indicated an effect while others have found no correlation. The performance of five herbivore species (three moths and two weevils) was assessed in an OTC experiment at 2 × ambient concentration (Peltonen et al., 2010). Growth of larvae of the Autumnal moth, Epirrita autumna, was significantly decreased in the O₃ treatment while no effects were observed in the other species. In an aphid oviposition preference study using birch buds grown in a three year OTC experiment, O₃ had neither a stimulatory or deterring effect on egg-laying (Peltonen et al., 2006). Furthermore, changes in birch bud phenolic compounds associated with the doubled ambient concentrations of O₃ did not correlate with changes in aphid oviposition (Peltonen et al., 2006). Reproduction in Popillia japonica, that were fed soybeans and grown under elevated O₃ appeared to be unaffected (O'Neill et al., 2008). In a meta-analysis of effects of elevated O₃ on 22 species of trees and 10 species of insects, the rates of survival, reproduction and food consumption were typically unaffected while development times were reduced and pupal masses were increased (Valkama et al., 2007).

At the Aspen FACE site insect performance under elevated (50-60 ppb) O₃ conditions (approximately 1.5 × background ambient levels of 30-40 ppb O₃) have been considered for several species. Cumulative fecundity of aphids (Cepegillettea betulaefoliae), that were reared on O₃-exposed paper birch (Betula papyrifera) trees, was lower than aphids from control plots (Awmack et al., 2004). No effects on growth, development, adult weight, embryo number and birth weight of newborn nymphs were observed. In a study conducted using three aspen genotypes,
performance of the aspen beetle (*Chrysomela crochi*) decreased across all parameters measured (development time, adult mass and survivorship) under elevated O₃ (Vigue and Lindroth, 2010). There was an increase in the development time of male and female aspen beetle larvae although the percentages varied across genotypes. Decreased beetle adult mass and survivorship was observed across all genotypes under elevated O₃ conditions. Another study from the Aspen FACE site did not find any significant effects of elevated O₃ on performance (longevity, fecundity, abundance) of the invasive weevil (*Polydrusus sericeus*) (Hillstrom et al., 2010b).

Since the 2006 O₃ AQCD, several studies have considered the effect of elevated O₃ on feeding behavior of insects. In a feeding preference study, the common leaf weevil (*Phyllobius pyri*) consumed significantly more leaf discs from one aspen clone when compared to a second clone under ambient air conditions (Freiwald et al., 2008). In a moderately elevated O₃ environment (1.5 × ambient), this preference for a certain aspen clone was less evident, however, leaves from O₃-exposed trees were significantly preferred to leaves grown under ambient conditions. Soybeans grown under enriched O₃ had significantly less loss of leaf tissue to herbivory in August compared to earlier in the growing season (July) when herbivory was not affected (Hamilton et al., 2005). Other plant-herbivore interactions have shown no effects of elevated O₃ on feeding. Feeding behavior of Japanese beetles (*P. japonica*) appeared to be unchanged when beetles were fed soybean leaves grown under elevated O₃ conditions (O'Neill et al., 2008). At the Aspen FACE site, feeding by the invasive weevil (*Polydrusus sericeus*), as measured by leaf area consumption, was not significantly different between foliage that was grown under elevated O₃ versus ambient conditions (Hillstrom et al., 2010b).

**Population-level and community-level responses**

Recent data on insects provide evidence of population-level and community-level responses to O₃. Elevated levels of O₃ can affect plant phytochemistry and nutrient content which in turn can alter population density and structure of the associated herbivorous insect communities and impact ecosystem processes (Cornelissen, 2011; Lindroth, 2010). In 72-hour exposures to elevated O₃, mean relative growth rate of the aphid *Diuraphis noxia* increased with O₃ concentration suggesting that more rapid population growth may occur when atmospheric O₃ is elevated (Summers et al., 1994). In a long-term study of elevated O₃ on herbivore performance at the Aspen FACE site, individual performance and population-level effects of the aphid *C. betulaefolii* were assessed. Elevated O₃ levels had a strong positive effect on the population growth rates of the aphids; although effects were not detected by measuring growth, development, adult weight, embryo number or birth weight of newborn nymphs (Awmack et al., 2004). Conversely, a lower rate of population growth was observed in aphids previously exposed to O₃ in an OTC (Menendez et al., 2010). No direct effects of O₃ were observed; however, nymphs born from adults exposed to and feeding on O₃ exposed plants were less capable of infesting new plants when compared to nymphs in the control plots (Menendez et al., 2010). Elevated O₃ reduced total arthropod abundance by 17% at Aspen FACE, largely as a
result of the negative effects on parasitoids, although phloem-feeding insects may benefit (Hillstrom and Lindroth, 2008). Herbivore communities affected by O₃ and N were sampled along an air pollution gradient in the Los Angeles basin (Jones and Paine, 2006). Abundance, diversity, and richness of herbivores were not affected. However, a shift in community structure, from phloem-feeding to chewing dominated communities, was observed along the gradient. No consistent effect of elevated O₃ on herbivory or insect population size was detected at SoyFACE (O'Neill et al., 2010; Dermody et al., 2008).

Evidence of modification of insect populations and communities in response to elevated O₃ includes genotypic and phenotypic changes. In a study conducted at the Aspen FACE site, elevated O₃ altered the genotype frequencies of the pea aphid (*Acyrthosiphon pisum*) grown on red clover (*Trifolium pratense*) over multiple generations (Mondor et al., 2005). Aphid color was used to distinguish between the two genotypes. Ozone increased the genotypic frequencies of pink-morph:green-morph aphids from 2:1 to 9:1, and depressed wing-induction responses more strongly in the pink than the green genotype (Mondor et al., 2005). Growth and development of individual green and pink aphids reared as a single genotype or mixed genotypes were unaffected by elevated O₃ (Mondor et al., 2010). However, growth of pea aphid populations is not readily predictable using individual growth rates.

### 9.4.9.2 Wildlife

#### Herpetofauna

Since the 2006 O₃ AQCD, direct effects of O₃ exposure including physiological changes and alterations of ecologically important behaviors such as feeding and thermoregulation have been observed in wildlife. These studies have been conducted in limited laboratory exposures, and the levels of O₃ treatment (e.g., 0.2-0.8 ppm) were often unrealistically higher than the ambient levels. Amphibians may be especially vulnerable to airborne oxidants due to the significant gas exchange that occurs across the skin (Andrews et al., 2008; Dohm et al., 2008). Exposure to 0.2 ppm to 0.8 ppm O₃ for 4 hours resulted in a decrease of oxygen consumption and depressed lung ventilation in the California tree frog *Pseudacris cadaverina* (Mautz and Dohm, 2004). Following a single 4-h inhalation exposure to 0.8 ppm O₃, reduced pulmonary macrophage phagocytosis was observed at 1 and 24 hours postexposure in the marine toad (*Bufo marinus*) indicating an effect on immune system function (Dohm et al., 2005). There was no difference in macrophage function at 48 hours postexposure in exposed and control individuals.

Behavioral effects of O₃ observed in amphibians include responses to minimize the surface area of the body exposed to the air and a decrease in feeding rates (Dohm et al., 2008; Mautz and Dohm, 2004). The adoption of a low-profile “water conservation posture” during O₃ exposure was observed in experiments with the
California tree frog (Mautz and Dohm, 2004). Marine toads, *Bufo marinus*, exposed to 0.06 µL/L (ppm) O₃ for 4 hours ate significantly fewer mealworms at 1 hour and 48 hours postexposure than control toads (Dohm et al., 2008). In the same study, escape/exploratory behavior as measured by total distance moved was not negatively affected in the O₃-exposed individuals as compared to the controls (Dohm et al., 2008).

Water balance and thermal preference in herpetofauna are altered with elevated O₃. Marine toads exposed to 0.8 ppm O₃ for 4 hours exhibited behavioral hypothermia when temperature selection in the toads was assessed at 1, 24 and 48 hours postexposure (Dohm et al., 2001). Ozone-exposed individuals lost almost 5g more body mass on average than controls due to evaporative water loss. At 24 hours after exposure, the individuals that had lost significant body mass selected lower body temperatures (Dohm et al., 2001). Behavioral hypothermia was also observed in reptiles following 4-h exposures to 0.6 ppm O₃. Exposure of the Western Fence Lizard (*Sceloporus occidentalis*) at 25°C induced behavioral hypothermia that recovered to control temperatures by 24 hours (Mautz and Dohm, 2004). The behavioral hypothermic response persisted in lizards exposed to O₃ at 35°C at 24 hours postexposure resulting in a mean body temperature of 3.3°C over controls.

**Soil fauna communities**

Ozone has also been shown to alter soil fauna communities (Meehan et al., 2010; Kasurinen et al., 2007; Loranger et al., 2004). Abundance of Acari (mites and ticks) decreased by 47% under elevated O₃ at Aspen FACE site, probably due to the higher secondary metabolites and lower N concentrations in litter and foliage under elevated O₃ (Loranger et al., 2004). In another study from the Aspen FACE site, leaf litter collected from aspen grown under elevated O₃ conditions was higher in fiber and lignin concentrations than litter from trees grown under ambient conditions. These chemical characteristics of the leaves were associated with increased springtail population growth following 10 weeks in a laboratory microcosm (Meehan et al., 2010). Consumption rates of earthworms fed on leaf litter for 6 weeks from trees grown under elevated O₃ conditions and ambient air did not vary significantly between treatments (Meehan et al., 2010). In another study on juvenile earthworms *Lumbricus terrestris*, individual growth was reduced when worms were fed high-O₃ birch litter from trees exposed for three years to elevated O₃ in an OTC system (Kasurinen et al., 2007). In the same study no significant growth or mortality effects were observed in isopods.

### 9.4.9.3 Indirect Effects on Wildlife

In addition to the direct effects of O₃ exposure on physiological and behavioral endpoints observed in the laboratory, there are indirect effects to wildlife. These effects include changes in biomass and nutritive quality of O₃-exposed plants (reviewed in Section 9.4.4) that are consumed by wildlife. Reduced digestibility of
O₃-exposed plants may alter dietary intake and foraging strategies in herbivores. In a study using native highbush blackberry (*Rubus argutus*) relative feed value of the plants decreased in bushes exposed to double ambient concentrations of O₃ (Ditchkoff et al., 2009). Indirect effects of elevated O₃ on wildlife include changes in chemical signaling important in ecological interactions reviewed below.

**Chemical signaling in ecological interactions**

Ozone has been shown to degrade or alter biogenic VOC signals important to ecological interactions including; (1) attraction of pollinators and seed dispersers; (2) defense against herbivory; and (3) predator-prey interactions (Pinto et al., 2010; McFrederick et al., 2009; Yuan et al., 2009; Pinto et al., 2007a; Pinto et al., 2007b). Each signal released by emitters has an atmospheric lifetime and a unique chemical signature comprised of different ratios of individual hydrocarbons that are susceptible to atmospheric oxidants such as O₃ (Yuan et al., 2009; Wright et al., 2005). Under elevated O₃ conditions, these olfactory cues may travel shorter distances before losing their specificity (McFrederick et al., 2009; McFrederick et al., 2008). Additional non-phytogenic VOC-mediated interrelationships with the potential to be modified by O₃ include territorial marking, pheromones for attraction of mates and various social interactions including scent trails, nestmate recognition and signals involved in aggregation behaviors (McFrederick et al., 2009). For example, the alcohols, ketones and aldehydes comprising sex pheromones in moths could be especially vulnerable to degradation by O₃, since some males travel >100 meters to find mates (Carde and Haynes, 2004). In general, effects of O₃ on scent-mediated ecological interactions are highly context- and species-specific (Lindroth, 2010; Bidart-Bouzat and Imeh-Nathaniel, 2008).

**Pollination and seed dispersal**

Phytogenic VOC’s attract pollinators and seed dispersers to flowers and fruits (Dudareva et al., 2006; Theis and Raguso, 2005). These floral scent trails in plant-insect interactions may be destroyed or transformed by O₃ (McFrederick et al., 2008). Using a Lagrangian model, the rate of destruction of phytogenic VOC’s was estimated in air parcels at increasing distance from a source in response to increased regional levels of O₃, hydroxyl and nitrate radicals (McFrederick et al., 2008). Based on the model, the ability of pollinators to locate highly reactive VOCs from emitting flowers may have decreased from kilometers during pre-industrial times to <200 meters at current ambient conditions (McFrederick et al., 2008). Scents that travel shorter distances (0-10 meters) are less susceptible to air pollutants, while highly reactive scents that travel longer distances (10 to 100s of meters), are at a higher risk for degradation (McFrederick et al., 2009). For example, male euglossine bees can detect bait stations from a distance of at least one kilometer (Dobson, 1994).
Defense against herbivory

Ozone can alter the chemical signature of VOCs emitted by plants and these VOCs are subsequently detected by herbivores (Blande et al., 2010; Iriti and Faoro, 2009; Pinto et al., 2007a; Vuorinen et al., 2004; Jackson et al., 1999; Cannon, 1990). These modifications can make the plant either more attractive or repellant to phytophagous insects (Pinto et al., 2010). For example, under elevated O₃, the host plant preference by forest tent caterpillars increased for birch compared to aspen (Agrell et al., 2005). Ozone-induced emissions from red spruce needles were found to repel spruce budworm larvae (Cannon, 1990). Transcriptional profiles of field grown soybean (Glycine max) grown in elevated O₃ conditions were altered due to herbivory by Japanese beetles. The herbivory resulted in a higher number of transcripts in the leaves of O₃-exposed plants and upregulation of antioxidant metabolism associated with plant defense (Casteel et al., 2008).

Ozone may modify signals involved in plant-to-plant interactions and plant defense against pathogens (Blande et al., 2010; Pinto et al., 2010; McFrederick et al., 2009; Yuan et al., 2009). In a recent study with lima beans, 80 ppb O₃ degraded several herbivore-induced VOCs, reducing the distance over which plant-to-plant signaling occurred (Blande et al., 2010).

Predator-prey interactions

Elevated O₃ conditions are associated with disruption of pheromone-mediated interactions at higher trophic levels (e.g., predators and parasitoids of herbivores). In a study from the Aspen FACE site, predator escape behaviors of the aphid (Chatophorus stevensis) were enhanced on O₃-fumigated aspen trees although the mechanism of this response remains unknown (Mondor et al., 2004). The predatory mite Phytoseiulus persimilis can distinguish between the VOC signature of ozonated lima bean plants and ozonated lima bean plants simultaneously damaged by T. urticae (Vuorinen et al., 2004) however, other tritrophic interactions have shown no effect (Pinto et al., 2007b).

There are few studies that consider host location behaviors of parasites under elevated O₃. In closed chambers fumigated with O₃, the searching efficiency and proportion of the host larval fruit flies parasitized by Asobara tabida declined when compared to filtered air controls (Gate et al., 1995). The host location behavior and rate of parasitism of the wasp (Coesia plutellae) on Plutella xylostella-infested potted cabbage plants was tested under ambient and doubled O₃ conditions in an open-air fumigation system (Pinto et al., 2008). The number of wasps found in the field and the percentages of parasitized larvae were not significantly different from controls under elevated O₃.

Elevated O₃ has the potential to perturb specialized food-web communication in transgenic crops. In insect-resistant oilseed rape Brassica napus grown under 100 ppb O₃ in a growth chamber, reduced feeding damage by Plutella xylostella led to deceased attraction of the endoparasitoid (Costesia vestalis), however this
tritrophic interaction was influenced by the degree of herbivore feeding (Himanen et al., 2009a; Himanen et al., 2009b). Under chronic O₃-exposure, the insect resistance trait BT cry1Ac in transgenic B. napus was higher than the control (Himanen et al., 2009c). There was a negative relative growth rate of the Bt target herbivore, P. xylostella, in all O₃ treatments.

### 9.4.9.4 Summary

Recent information on O₃ effects on insects and other wildlife is limited to a few species and there is no consensus on how these organisms respond to elevated O₃. Studies published since the last review show impacts of elevated O₃ on both species-level responses (reproduction, growth, feeding behavior) and community and ecosystem-level responses (population growth, abundance, shift in community structure) in some insects and soil fauna. Changes in ecologically important behaviors such as feeding and thermoregulation have recently been observed with O₃ exposure in amphibians and reptiles, however, these responses occur at concentrations of O₃ much higher than ambient levels.

Recent information available since the last review considers the effects of O₃ on chemical signaling in insect and wildlife interactions. Specifically, studies on O₃ effects on pollination and seed dispersal, defenses against herbivory and predator-prey interactions all consider the ability of O₃ to alter the chemical signature of VOCs emitted during these pheromone-mediated events. The effects of O₃ on chemical signaling between plants, herbivores and pollinators as well as interactions between multiple trophic levels is an emerging area of study that may result in further elucidation of O₃ effects at the species, community and ecosystem-level.

### 9.5 Effects-based Air Quality Exposure Indices and Dose Modeling

#### 9.5.1 Introduction

Exposure indices are metrics that quantify exposure as it relates to measured plant response (e.g., reduced growth). They are summary measures of monitored ambient O₃ concentrations over time, intended to provide a consistent metric for reviewing and comparing exposure-response effects obtained from various studies. Such indices may also provide a basis for developing a biologically-relevant air quality standard for protecting vegetation and ecosystems. Effects on plant growth and/or yield have been a major focus of the characterization of O₃ impacts on plants for purposes of the air quality standard setting process (U.S. EPA, 2007b, 1996e, 1986). The relationship of O₃ and plant responses can be characterized quantitatively as “dose-response” or “exposure-response.” The distinction is in how the pollutant concentration is
expressed: “dose” is the pollutant concentration absorbed by the leaf over some time period, and is very difficult to measure directly, whereas “exposure” is the ambient air concentration measured near the plant over some time period, and summarized for that period using an index. Exposure indices have been most useful in considering the form of the secondary O₃ NAAQS, in large part because they only require ambient air quality data rather than more complex indirect calculations of dose to the plant. The attributes of exposure indices that are most relevant to plant response are the weighting of O₃ concentrations and the daily and seasonal time-periods. Several different types of exposure indices are discussed in Section 9.5.2.

From a theoretical perspective, a measure of plant O₃ uptake or dose from ambient air (either rate of uptake or cumulative seasonal uptake) might be a better predictor of plant response to O₃ than an exposure index and may be useful in improving risk assessment. An uptake estimate would have to integrate all those environmental factors that influence stomatal conductance, including but not limited to temperature, humidity, and soil water status (Section 9.5.4). Therefore, uptake values are generally obtained with simulation models that require knowledge of species- and site-specific values for the variables mentioned. However, a limitation of modeling dose is that environmental variables are poorly characterized. In addition, it has also been recognized that O₃ detoxification processes and the temporal dynamics of detoxification must be taken into account in dose modeling (Heath et al., 2009) (Section 9.5.4). Because of this, research has focused historically on predictors of O₃ damage to plants based only on exposure as a summary measure of monitored ambient pollutant concentration over some integral of time, rather than dose (U.S. EPA, 1996c; Costa et al., 1992; Lee et al., 1988b; U.S. EPA, 1986; Lefohn and Benedict, 1982; O’Gara, 1922).

### 9.5.2 Description of Exposure Indices Available in the Literature

Mathematical approaches for summarizing ambient air quality information in biologically meaningful forms for O₃ vegetation effects assessment purposes have been explored for more than 80 years (U.S. EPA, 1996b; O’Gara, 1922). In the context of national standards that protect for “known or anticipated” effects on many plant species in a variety of habitats, exposure indices provide a numerical summary of very large numbers of ambient observations of concentration over extended periods. Like any summary statistic, exposure indices retain information on some, but not all, characteristics of the original observations. Several indices have been developed to attempt to incorporate some of the biological, environmental, and exposure factors that influence the magnitude of the biological response and contribute to observed variability (Hogsett et al., 1988). In the 1996 O₃ AQCD (U.S. EPA, 1996a), the exposure indices were arranged into five categories; (1) One event, (2) Mean, (3) Cumulative, (4) Concentration weighted, and (5) Multicomponent, and were discussed in detail (Lee et al., 1989). Figure 9-9 illustrates how several of the indices weight concentration and accumulate exposure. For example, the SUM06 index (panel a) is a threshold-based approach wherein concentrations below
0.06 ppm are given a weight of zero and concentrations at or above 0.06 ppm are given a weight of 1.0 that is summed, usually over 3 to 6 months. The Sigmoid approach (panel b), which is similar to the W126 index (Lefohn et al., 1988; Lefohn and Runeckles, 1987), is a non-threshold approach wherein all concentrations are given a weight that increases from zero to 1.0 with increasing concentration and summed.

Figure 9-9  Diagrammatic representation of several exposure indices illustrating how they weight concentration and accumulate exposure.
This section will primarily discuss SUM06, W126 and AOTx exposure metrics. Below are the definitions of the three cumulative index forms:

- **SUM06**: Sum of all hourly O₃ concentrations greater than or equal to 0.06 ppm observed during a specified daily and seasonal time window (Figure 9-9, Panel A).

- **AOTx**: Sum of the differences between hourly O₃ concentrations greater than a specified threshold during a specified daily and seasonal time window. For example, AOT40 is sum of the differences between hourly concentrations above 0.04 ppm.

- **W126**: Sigmoidally weighted sum of all hourly O₃ concentrations observed during a specified daily and seasonal time window (Lefohn et al., 1988; Lefohn and Runeckles, 1987), similar to Figure 9-9, Panel B. The sigmoidal weighting of hourly O₃ concentration is given in the equation below, where \( C \) is the hourly O₃ concentration in ppm:

\[
W_c = \frac{1}{1 + 4403e^{-126C}}
\]

Equation 9-1

These indices have a variety of relevant time windows that may be applied and are discussed in Section 9.5.3.

Various factors with known or suspected bearing on the exposure-response relationship, including concentration, time of day, respite time, frequency of peak occurrence, plant phenology, predisposition, etc., have been weighted with various functions in a large set of indices. The resulting indices were evaluated by ranking them according to the goodness-of-fit of a regression model of growth or yield response (Lee et al., 1989). The statistical evaluations for each of these indices were completed using growth or yield response data from many earlier exposure studies (e.g., NCLAN). This retrospective approach was necessary because there were no studies specifically designed to test the goodness-of-fit of the various indices. The goodness-of-fit of a set of linear and nonlinear models for exposure-response was ranked as various proposed indices were used in turn to quantify exposure. This approach provided evidence for the best indices. The results of retrospective analyses are described below.

Most of the early retrospective studies reporting regression approaches used data from the NCLAN program or data from Corvallis, Oregon or California (Costa et al., 1992; Lee et al., 1988b; Lefohn et al., 1988; Musselman et al., 1988; Lee et al., 1987; U.S. EPA, 1986). These studies were previously reviewed by the EPA (U.S. EPA, 1996c; Costa et al., 1992) and were in general agreement that the best fit to the data resulted from using cumulative concentration-weighted exposure indices (e.g., W126, SUM06). Lee et al. (1987) suggested that exposure indices that included all the 24-h data performed better than those that used only 7 hours of data; this was
consistent with the conclusions of Heagle et al. (1987) that plants receiving exposures for an additional 5 hours/day showed 10% greater yield loss than those exposed for 7 hours/day. In an analysis using the National Crop Loss Assessment Network (NCLAN) data, Lee et al. (1988b) found several indices which only cumulated and weighted higher concentrations (e.g., W126, SUM06, SUM08, and AOT40) performed very well. Amongst this group no index had consistently better fits than the other indices across all studies and species (Heagle et al., 1994b; Lefohn et al., 1988; Musselman et al., 1988). Lee et al. (1988b) found that adding phenology weighting to the index somewhat improved the performance of the indices. The “best” exposure index was a phenologically weighted cumulative index, with sigmoid weighting on concentration and a gamma weighting function as a surrogate for plant growth stage. This index provided the best statistical fit when used in the models under consideration, but it required data on species and site conditions, making specification of weighting functions difficult for general use.

Other factors, including predisposition time (Hogsett et al., 1988; McCool et al., 1988) and crop development stage (Tingey et al., 2002; Heagle et al., 1991) contributed to variation in the biological response and suggested the need for weighting O\(_3\) concentrations to account for predisposition time and phenology. However, the roles of predisposition and phenology in plant response vary considerably with species and environmental conditions; therefore, specification of a weighting function for general use in characterizing plant exposure has not been possible.

European scientists took a similar approach in developing indices describing growth and yield loss in crops and tree seedlings, using OTCs with modified ambient exposures, but many fewer species and study locations were employed in the European studies. There is evidence from some European studies that a lower (Pleijel et al., 1997) or higher (Finnan et al., 1997; Finnan et al., 1996) cutoff value in indices with a threshold may provide a better statistical fit to the experimental data. Finnan et al. (1997) used seven exposure studies of spring wheat to confirm that cumulative exposure indices emphasizing higher O\(_3\) concentrations were best related to plant response and that cumulative exposure indices using weighting functions, including cutoff concentrations, allometric and sigmoidal, provided a better fit and that the ranking of these indices differed depending on the exposure-response model used. Weighting those concentrations associated with sunshine hours in an attempt to incorporate an element of plant uptake did not improve the index performance (Finnan et al., 1997). A more recent study using data from several European studies of Norway spruce, analyzed the relationship between relative biomass accumulation and several cumulative, weighted indices, including the AOT40 (area over a threshold of 40ppb) and the SUM06 (Skarby et al., 2004). All the indices performed relatively well in regressing biomass and exposure index, with the AOT20 and AOT30 doing slightly better than others (\(r^2 = 0.46-0.47\)). In another comparative study of four independent data sets of potato yield and different cumulative uptake indices with different cutoff values, a similarly narrow range of \(r^2\) was observed (\(r^2 = 0.3-0.4\)) (Pleijel et al., 2004b).
In Europe, the cutoff concentration-weighted index AOT40 was selected in developing exposure-response relationships based on OTC studies of a limited number of crops and trees (Grunhage and Jager, 2003). The United Nations Economic Commission for Europe (UNECE, 1988) adopted the critical levels approach for assessment of O₃ risk to vegetation across Europe. As used by the UNECE, the critical levels are not like the air quality regulatory standards used in the U.S., but rather function as planning targets for reductions in pollutant emissions to protect ecological resources. Critical levels for O₃ are intended to prevent long-term deleterious effects on the most sensitive plant species under the most sensitive environmental conditions, but not intended to quantify O₃ effects. A critical level was defined as “the concentration of pollutant in the atmosphere above which direct adverse effects on receptors, such as plants, ecosystems, or materials may occur according to present knowledge” (UNECE, 1988). The nature of the “adverse effects” was not specified in the original definition, which provided for different levels for different types of harmful effect (e.g., visible injury or loss of crop yield). There are also different critical levels for crops, forests, and semi-natural vegetation. The caveat, “according to present knowledge” is important because critical levels are not rigid; they are revised periodically as new scientific information becomes available. For example, the original critical level for O₃ specified concentrations for three averaging times, but further research and debate led to the current critical level being stated as the cumulative exposure (concentration × hours) over a cutoff concentration of 40 ppb (AOT40) (Fuhrer et al., 1997).

More recently in Europe, a decision was made to work toward a flux-based approach (see Section 9.5.4) for the critical levels (“Level II”), with the goal of modeling O₃ flux-effect relationships for three vegetation types: crops, forests, and semi-natural vegetation (Grunhage and Jager, 2003). Progress has been made in modeling flux (U.S. EPA, 2006b) and the Mapping Manual is being revised (Ashmore et al., 2004a, b; Grennfelt, 2004; Karlsson et al., 2003). The revisions may include a flux-based approach for three crops: wheat, potatoes, and cotton. However, because of a lack of flux-response data, a cumulative, cutoff concentration-based (AOTx) exposure index will remain in use for the near future for most crops and for forests and semi-natural herbaceous vegetation (Ashmore et al., 2004b).

In both the U.S. and Europe, the adequacy of these numerical summaries of exposure in relating biomass and yield changes have, for the most part, all been evaluated using data from studies not necessarily designed to compare one index to another (Skarby et al., 2004; Lee et al., 1989; Lefohn et al., 1988). Very few studies in the U.S. have addressed this issue since the 2006 O₃ AQCD. McLaughlin et al. (2007a) reported that the cumulative exposure index of AOT60 related well to reductions in growth rates at forest sites in the southern Appalachian Mountains. However, the authors did not report an analysis to compare multiple indices. Overall, given the available data from previous O₃ AQCDs and the few recent studies, the cumulative, concentration-weighted indices perform better than the peak or mean indices. It is still not possible, however, to distinguish the differences in performance among the cumulative, concentration-weighted indices.
The main conclusions from the 1996 and 2006 O₃ AQCDs regarding an index based on ambient exposure are still valid. No information has come forth since the 2006 O₃ AQCD to alter those conclusions. These key conclusions can be restated as follows:

- ozone effects in plants are cumulative;
- higher O₃ concentrations appear to be more important than lower concentrations in eliciting a response;
- plant sensitivity to O₃ varies with time of day and plant development stage;
- quantifying exposure with indices that accumulate the O₃ hourly concentrations and preferentially weight the higher concentrations improves the explanatory power of exposure/response models for growth and yield, over using indices based on mean and peak exposure values.

Following the 2006 criteria review process (U.S. EPA, 2006b), the EPA proposed an alternative form of the secondary NAAQS for O₃ using a cumulative, concentration-weighted exposure index to protect vegetation from damage (72 FR 37818). The EPA considered two specific concentration-weighted indices: the cutoff concentration weighted SUM06 and the sigmoid-weighted W126 exposure index (U.S. EPA, 2007b). These two indices performed equally well in predicting the exposure-response relationships observed in the crop and tree seedlings studies (Lee et al., 1989). At a workshop convened to consider the science supporting these indices (Heck and Cowling, 1997) there was a consensus that these cumulative concentration-weighted indices being considered were equally capable of predicting plant response. It should be noted that there are some important differences between the SUM06 and W126. When considering the response of vegetation to O₃ exposures represented by the threshold (e.g., SUM06) and non-threshold (e.g., W126) indices, the W126 metric does not have a cut-off in the weighting scheme as does SUM06 and thus it includes consideration of potentially damaging exposures below 60 ppb. The W126 metric also adds increasing weight to hourly concentrations from about 40 ppb to about 100 ppb (Lefohn et al., 1988; Lefohn and Runeckles, 1987). This is unlike cut-off metrics such as the SUM06 where all concentrations above 60 ppb are treated equally. This is an important feature of the W126 since as hourly concentrations become higher, they become increasingly likely to overwhelm plant defenses and are known to be more detrimental to vegetation (see Section 9.5.3.1).

### 9.5.3 Important Components of Exposure Indices

In the previous O₃ AQCDs it was established that higher hourly concentrations have greater effects on vegetation than lower concentrations (U.S. EPA, 2006b, 1996c). Further, it was determined that the diurnal and seasonal duration of exposure is important for plant response. Weighting of hourly concentrations and the diurnal and seasonal time window of exposure are the most important variables in a cumulative exposure index and will be discussed below. However, these variables should be looked at in the context of plant phenology, diurnal conductance rates, plant canopy structure, and detoxification mechanisms of vegetation as well as the climate and...
meteorology, all of which are determinants of plant response. These more specific factors will be discussed in the uptake and dose modeling Section 9.5.4.

9.5.3.1 Role of Concentration

The significant role of peak O3 concentrations was established based on several experimental studies (U.S. EPA, 1996c). Several studies (Oksanen and Holopainen, 2001; Yun and Laurence, 1999; Nussbaum et al., 1995) have added support for the important role that peak concentrations, as well as the pattern of occurrence, plays in plant response to O3. Oksanen and Holopainen (2001) found that the peak concentrations and the shape of the O3 exposure (i.e., duration of the event) were important determinants of foliar injury in European white birch saplings, but growth reductions were found to be more related to total cumulative exposure. Based on air quality data from 10 U.S. cities, three 4-week exposure treatments having the same SUM06 value were constructed by Yun and Laurence (1999). The authors used different exposure regimes to explore effects of treatments with variable versus uniform peak occurrence during the exposure period. The authors reported that the variable peak exposures were important in causing injury, and that the different exposure treatments, although having the same SUM06, resulted in very different patterns of foliar injury. Nussbaum et al. (1995) also found peak concentrations and the pattern of occurrence to be critical in determining the measured response. The authors recommended that to describe the effect on total forage yield, peak concentrations >0.11 ppm must be emphasized by using an AOT with higher threshold concentrations.

A greater role for peak concentrations in effects on plant growth might be inferred based on air quality analyses for the southern California area (Tingey et al., 2004; Lee et al., 2003a). In the late 1960s and 1970s, extremely high O3 concentrations had impacted the San Bernardino National Forest. However, over the past 20+ years, significant reductions in O3 exposure have occurred (Bytnerowicz et al., 2008; Lee et al., 2003a; Lefohn and Shadwick, 2000; Davidson, 1993). An illustration of this improvement in air quality is shown by the 37-year history of O3 air quality at the Crestline site in the San Bernardino Mountains (Figure 9-10) (Lee et al., 2003a). Ozone exposure increased from 1963 to 1979 concurrent with increased population and vehicular miles, followed by a decline to the present mirroring decreases in precursor emissions. The pattern in exposure was evident in various exposure indices including the cumulative concentration weighted (SUM06), as well as maximum peak event (1-h peak), and the number of days having hourly averaged O3 concentrations greater than or equal to 95 ppb. The number of days having hourly averaged O3 concentrations greater than or equal to 95 ppb declined significantly from 163 days in 1978 to 103 days in 1997. The changes in ambient O3 air quality for the Crestline site were reflected in the changes in frequency and magnitude of the peak hourly concentration and the duration of exposure (Figure 9-10). Considering the role of exposure patterns in determining response, the seasonal and diurnal
patterns in hourly $O_3$ concentration did not vary appreciably from year to year over the 37-year period (Lee et al., 2003a).

The potential importance of exposure to peak concentrations comes both from results of measures of tree conditions on established plots and from results of model simulations. Across a broad area of the San Bernardino National Forest, the Forest Pest Management (FPM) method of injury assessment indicated an improvement in crown condition from 1974 to 1988; and the area of improvement in injury assessment is coincident with an improvement in $O_3$ air quality (Miller and Rechel, 1999). A more recent analysis of forest changes in the San Bernardino National Forest, using an expanded network of monitoring sites, has verified significant changes in growth, mortality rates, basal area, and species composition throughout the area since 1974 (Arbaugh et al., 2003). A model simulation of ponderosa pine growth over the 40-year period in the San Bernardino National Forest showed a significant impact of $O_3$ exposure on tree growth and indicates improved growth with reduced $O_3$ concentrations. This area has also experienced elevated N deposition and based on a number of environmental indicators, it appears that this area is experiencing N saturation (Fenn et al., 1996). To account for this potential interaction, the model simulations were conducted under conditions of unlimited soil N. The actual interactions are not known. The improvement in growth over the years was attributed to improved air quality, but no distinction was made regarding the relative role of “mid-range” and higher hourly concentrations, only that improved growth tracked decreasing SUM06, maximum peak concentration, and number of days of hourly $O_3 > 95$ ppb (Tingey et al., 2004). A summary of air quality data from 1980 to 2000 for the San Bernardino National Forest area of the number of “mid-range” hourly concentrations indicated no dramatic changes over this 20-year period, ranging from about 1,500 to 2,000 hours per year (Figure 9-11). There was a slow increase in the number of “mid-range” concentrations from 1980 to 1986, which corresponds to the period after implementation of the air quality standard. Another sharper increase was observed in the late 1990s. This pattern of occurrence of mid-range hourly concentrations suggests a lesser role for these concentration ranges compared to the higher values in either of the ground-level tree injury observations of the model simulation of growth over the 40-year period.
Figure 9-10  Trends in May to September: 12-hour SUM06, Peak 1-hour O₃ concentration and number of daily exceedances of 95 ppb for the Crestline site in 1963 to 1999; in relation to trends in mean daily maximum temperature for Crestline and daily reactive organic gases (ROG) and oxides of nitrogen (NOₓ) for San Bernardino County.
9.5.3.2  Diurnal and Seasonal Exposure

Diurnal Exposure

The diurnal patterns of maximal leaf/needle conductance and occurrence of higher ambient concentrations can help determine which hours during the day over a season should be included in an exposure index. Stomatal conductance is species and phenology dependent and is linked to both diurnal and seasonal meteorological activity as well as to soil/site conditions (e.g., VPD, soil moisture). Daily patterns of leaf/needle conductance are often highest in midmorning, whereas higher ambient O₃ concentrations generally occur in early to late afternoon when stomata are often partially closed and conductances are lower. Total O₃ flux depends on atmospheric and boundary layer resistances, both of which exhibit variability throughout the day. Experimental studies with tree species demonstrated the decoupling of ambient O₃ exposure, peak occurrence, and gas exchange, particularly in areas of drought (Panek, 2004). Several studies have suggested that ponderosa pine trees in the southern and northern Sierra Nevada Mountains may not be as susceptible to high O₃ concentrations as to lower concentrations, due to reduced needle conductance and O₃ uptake during the period when the highest concentrations occur (Panek et al., 2002; Panek and Goldstein, 2001; Bauer et al., 2000; Arbaugh et al., 1998). Panek et al.
(2002) compared direct O₃ flux measurements into a canopy of ponderosa pine and demonstrated a lack of correlation of daily patterns of conductance and O₃ occurrence, especially in the late season drought period; the authors concluded that a consideration of climate or season was essential, especially considering the role of soil moisture and conductance/uptake. In contrast, Grulke et al. (2002) reported high conductance when O₃ concentrations were high in the same species, but under different growing site conditions. The longer-term biological responses reported by Miller and Rechel (1999) for ponderosa pine in the same region, and the general reduction in recent years in ambient O₃ concentrations, suggest that stomatal conductance alone may not be a sufficient indicator of potential vegetation injury or damage. Another consideration for the effect of O₃ uptake is the diurnal pattern of detoxification capacity of the plant. The detoxification capacity may not follow the same pattern as stomatal conductance (Heath et al., 2009).

The use of a 12-h (8:00 a.m. to 8:00 p.m.) daylight period for a W126 cumulating exposure was based primarily on evidence that the conditions for uptake of O₃ into the plant occur mainly during the daytime hours. In general, plants have the highest stomatal conductance during the daytime and in many areas atmospheric turbulent mixing is greatest during the day as well (Uddling et al., 2010; U.S. EPA, 2006b). However, notable exceptions to maximum daytime conductance are cacti and other plants with crassulacean acid metabolism (CAM photosynthesis) which only open their stomata at night. This section will focus on plants with C₃ and C₄ photosynthesis, which generally have maximum stomatal conductance during the daytime.

Recent reviews of the literature reported that a large number of species had varying degrees of nocturnal stomatal conductance (Caird et al., 2007; Dawson et al., 2007; Musselman and Minnick, 2000). The reason for night-time water loss through stomata is not well understood and is an area of active research (e.g., Christman et al., 2009; Howard et al., 2009). Night-time stomatal opening may be enhanced by O₃ damage that could result in loss of stomatal control, and less complete closure of stomata, than under low O₃ conditions (Caird et al., 2007; Grulke et al., 2007b). In general, the rate of stomatal conductance at night is much lower than during the day (Caird et al., 2007). Atmospheric turbulence at night is also often low, which results in stable boundary layers and unfavorable conditions for O₃ uptake into vegetation (Finkelstein et al., 2000). Nevertheless, nocturnal turbulence does intermittently occur and may result in non-negligible O₃ flux into the plants. In addition, plants might be more susceptible to O₃ exposure at night than during the daytime, because of potentially lower plant defenses (Heath et al., 2009; Loreto and Fares, 2007; Musselman et al., 2006; Musselman and Minnick, 2000). For significant nocturnal stomatal flux and O₃ effects to occur, specific conditions must exist. A susceptible plant with nocturnal stomatal conductance and low defenses must be growing in an area with relatively high night-time O₃ concentrations and appreciable nocturnal atmospheric turbulence. It is unclear how many areas there are in the U.S. where these conditions occur. It may be possible that these conditions exist in mountainous areas of southern California, front-range of Colorado (Turnipseed et al., 2009) and the Great Smoky Mountains of North Carolina and Tennessee. Tobiessen
(1982) found that shade intolerant tree species showed opening of stomata in the dark and did not find this in shade tolerant species. This may indicate shade intolerant trees may be more likely to be susceptible to O₃ exposure at night. More information is needed in locations with high night-time O₃ to assess the local O₃ patterns, micrometeorology and responses of potentially vulnerable plant species.

Several field studies have attempted to quantify night-time O₃ uptake with a variety of methods. However, many of these studies have not linked the night-time flux to measured effects on plants. Grulke et al. (2004) showed that the stomatal conductance at night for ponderosa pine in the San Bernardino National Forest (CA) ranged from one tenth to one fourth that of maximum daytime stomatal conductance. In June, at a high-elevation site, it was calculated that 11% of the total daily O₃ uptake of pole-sized trees occurred at night. In late summer, however, O₃ uptake at night was negligible. However, this study did not consider the turbulent conditions at night. Finkelstein et al. (2000) investigated O₃ deposition velocity to forest canopies at three different sites. The authors found the total flux (stomatal and non-stomatal) to the canopy to be very low during night-time hours as compared to day-time hours. However, the authors did note that higher nocturnal deposition velocities at conifer sites may be due to some degree of stomatal opening at night (Finkelstein et al., 2000). Work by Mereu et al. (2009) in Italy on Mediterranean species indicated that nocturnal uptake was from 10 to 18% of total daily uptake during a weak drought and up to 24% as the drought became more pronounced. The proportion of night-time uptake was greater during the drought due to decreases in daytime stomatal conductance (Mereu et al., 2009). In a study conducted in California, (Fares et al., 2011) reported that calculated mean percentages of nocturnal uptake were 5%, 12.5%, 6.9% of total O₃ uptake for lemon, mandarin, and orange, respectively. In another recent study at the Aspen FACE site in Wisconsin, calculated leaf-level stomatal O₃ flux was near zero from the night-time hours of 8:00 p.m. to 5:00 a.m. (Uddling et al., 2010). This was likely due to low horizontal wind speed (>1 meter/sec) and low O₃ concentrations (<25 ppb) during those same night-time hours (Figure 9-12).
Note: Subscripts "max" and "min" refer to stomatal fluxes calculated neglecting and accounting for potential non-stomatal O₃ flux, respectively.

Source: Reprinted with permission of Elsevier Ltd. (Uddling et al., 2010).

Figure 9-12  Diurnal (a) conductance through boundary layer and stomata (gsb), (b) ozone concentration, and leaf-level stomatal O₃ flux (Fst0l) in control plots from mid-June through August, in (c) 2004 and (d) 2005 in the Aspen FACE experiment.

A few studies have tested the biological effects of night-time O₃ exposure on vegetation in controlled chambers. Biomass of ponderosa pine seedlings was significantly reduced when seedlings were exposed to either daytime or nighttime episodic profiles (Lee and Hogsett, 1999). However, the biomass reductions were much greater with daytime peak concentrations than with nighttime peak concentrations. Similarly, birch cuttings grown in field chambers that were exposed to O₃ at night only, daytime only, and 24 hours showed similar reductions in biomass in night only and day only treatments. Birch seedling showed greater reductions in growth in 24-h exposures than those exposed to O₃ at night or day only (Matyssek et al., 1995). Field mustard (Brassica rapa) plants exposed to O₃ during the day or night showed little significant difference in the amounts of injury or reduced growth response to O₃ treatment, although the stomatal conductance was 70-80% lower at
night (Winner et al., 1989). These studies show that effects can be seen with nighttime exposures to O₃ but when atmospheric conditions are stable at night, it is uncertain how these exposures may affect plants and trees with complex canopies in the field.

**Seasonal exposure**

Vegetation across the U.S. has widely varying periods of physiological activity during the year due to variability in climate and phenology. In order for a particular plant to be vulnerable to O₃ pollution, it must have foliage and be physiologically active. Annual crops are typically grown for periods of two to three months. In contrast, perennial species may be photosynthetically active longer (up to 12 months each year for some species) depending on the species and where it is grown. In general, the period of maximum physiological activity and thus, potential O₃ uptake for vegetation coincides with some or all of the intra-annual period defined as the O₃ season, which varies on a state-by-state basis (Figure 3-24). This is because the high temperature and high light conditions that typically promote the formation of tropospheric O₃ also promote physiological activity in vegetation. There are very limited exceptions to this pattern where O₃ can form in the winter in areas in the western U.S. with intense natural gas exploration (Pinto, 2009), but this is typically when plants are dormant and there is little chance of O₃ uptake. Given the significant variability in growth patterns and lengths of growing season among the wide range of vegetation species that may experience adverse effects associated with O₃ exposure, no single time window of exposure can work perfectly for all types of vegetation.

Various intra-annual averaging and accumulation time periods have been considered for the protection of vegetation. The 2007 proposal for the secondary O₃ standard (75 FR 37818) proposed to use the maximum consecutive 3-month period within the O₃ season. The U.S. Forest Service and federal land managers have used a 24-h W126 accumulated for 6 months from April through September (U.S. Forest Service, 2000). However, some monitors in the U.S. are operational for as little as four months and would not have enough data for a 6-month seasonal window. The exposure period in the vast majority of O₃ exposure studies conducted in the U.S. has been much shorter than 6 months. Most of the crop studies done through NCLAN had exposures less than three months with an average of 77 days. Open-top chamber studies of tree seedlings, compiled by the EPA, had an average exposure of just over three months or 99 days. In more recent FACE experiments, SoyFACE exposed soybeans for an average of approximately 120 days per year and the Aspen FACE experiment exposed trees to an average of approximately 145 days per year of elevated O₃, which included the entire growing season at those particular sites. Despite the possibility that plants may be exposed to ambient O₃ longer than 3 months in some locations, there is generally a lack of exposure experiments conducted for longer than 3 months.
In an analysis of the 3- and 6-month maximum W126 values calculated for over 1,200 AQS (Air Quality System) and CASTNET (Clean Air Status and Trend Network) EPA monitoring sites for the years 2008-2009, it was found that these 2 accumulation periods resulted in highly correlated metrics (Figure 9-13). The two accumulation periods were centered on the yearly maximum for each monitoring site, and it is possible that this correlation would be weaker if the two periods were not temporally aligned. In the U.S., W126 cumulated over 3 months, and W126
cumulated over 6 months are proxies of one another, as long as the period in which daily W126 is accumulated corresponds to the seasonal maximum. Therefore, it is expected that either statistic will predict vegetation response equally well. In other words, the strength of the correlation between maximum 3-month W126 and maximum 6-month W126 is such that there is no material difference in their predictive value for vegetation response.

9.5.4 Ozone Uptake/Dose Modeling for Vegetation

Another approach for improving risk assessment of vegetation response to ambient O3 is based on estimating the O3 concentration from the atmosphere that enters the leaf (i.e., flux or deposition). Interest has been increasing in recent years, particularly in Europe, in using mathematically tractable flux models for O3 assessments at the regional, national, and European scale (Matyssek et al., 2008; Paoletti and Manning, 2007; ICP M&M, 2004; Emberson et al., 2000b; Emberson et al., 2000a). Some researchers have claimed that using flux models can be used to better predict vegetation responses to O3 than exposure-based approaches (Matyssek et al., 2008). However, other research has suggested that flux models do not predict vegetation responses to O3 better than exposure-based models, such as AOT40 (Gonzalez-Fernandez et al., 2010). While some efforts have been made in the U.S. to calculate O3 flux into leaves and canopies (Fares et al., 2010a; Turnipseed et al., 2009; Uddling et al., 2009; Bergweiler et al., 2008; Hogg et al., 2007; Grulke et al., 2004; Grantz et al., 1997; Grantz et al., 1995), little information has been published relating these fluxes to effects on vegetation. The lack of flux data in the U.S. and the lack of understanding of detoxification processes have made this technique less viable for vulnerability and risk assessments in the U.S.

Flux calculations are data intensive and must be carefully implemented. Reducing uncertainties in flux estimates for areas with diverse surface or terrain conditions to within ± 50% requires “very careful application of dry deposition models, some model development, and support by experimental observations” (Wesely and Hicks, 2000). As an example, the annual average deposition velocity of O3 among three nearby sites in similar vegetation was found to vary by ± 10%, presumably due to terrain (Brook et al., 1997). Moreover, the authors stated that the actual variation was even greater, because stomatal uptake was unrealistically assumed to be the same among all sites, and flux is strongly influenced by stomatal conductance (Brook et al., 1997; Massman and Grantz, 1995; Fuentes et al., 1992; Reich, 1987; Leuning et al., 1979). This uptake-based approach to quantify the vegetation impact of O3 requires inclusion of those factors that control the diurnal and seasonal O3 flux to vegetation (e.g., climate patterns, species and/or vegetation-type factors and site-specific factors). The models have to distinguish between stomatal and non-stomatal components of O3 deposition to adequately estimate actual concentration reaching the target tissue of a plant to elicit a response (Uddling et al., 2009). Determining this O3 uptake via canopy and stomatal conductance relies on models to predict flux and ultimately the “effective” flux (Grunhage et al., 2004; Massman, 2004; Massman et
“Effective flux” has been defined as the balance between O₃ flux and detoxification processes (Heath et al., 2009; Musselman and Massman, 1999; Grunhage and Haenel, 1997; Dammgen et al., 1993). The time-integrated “effective flux” is termed “effective dose.” The uptake mechanisms and the resistances in this process, including stomatal conductance and biochemical defense mechanisms, are discussed below. The flux-based index is the goal for the “Level II” critical level for assessment of O₃ risk to vegetation and ecosystems across Europe (Ashmore et al., 2004a).

An important consideration in both O₃ exposure and uptake is how the O₃ concentration at the top of low vegetation such as, crops and tree seedlings may be lower than the height at which the measurement is taken. Ambient monitor inlets in the U.S. are typically at heights of 3 to 5 meters. During daytime hours, the vertical O₃ gradient can be relatively small because turbulent mixing maintains the downward flux of O₃. For example, Horvath et al. (1995) calculated a 7% decrease in O₃ going from a height of 4 meters down to 0.5 meters above the surface during unstable (or turbulent) conditions in a study over low vegetation in Hungary [see Section AX3.3.2. of the 2006 O₃ AQCD (U.S. EPA, 2006b)]. There have been several studies indicating decreased O₃ concentrations under tree canopies (Kolb et al., 1997; Samuelson and Kelly, 1997; Joss and Graber, 1996; Fredericksen et al., 1995; Lorenzini and Nali, 1995; Enders, 1992; Fontan et al., 1992; Neufeld et al., 1992). In contrast, for forests, measured data may underestimate O₃ concentration at the top of the canopy. The difference between measurement height and canopy height is a function of several factors, the intensity of turbulent mixing in the surface layer and other meteorological factors, canopy height and total deposition to the canopy. Some researchers have used deposition models to estimate O₃ concentration at canopy-top height based on concentrations at measurement height (Emberson et al., 2000a). However, deposition models usually require meteorological data inputs that are not always available or well characterized across large geographical scales.

Soil moisture is a critical factor in controlling O₃ uptake through its effect on plant water status and stomatal conductance. In an attempt to relate uptake, soil moisture, and ambient air quality to identify areas of potential risk, available O₃ monitoring data for 1983 to 1990 were used along with literature-based seedling exposure-response data from regions within the southern Appalachian Mountains that might have experienced O₃ exposures sufficient to inhibit growth (Lefohn et al., 1997). In a small number of areas within the region, O₃ exposures and soil moisture availability were sufficient to possibly cause growth reductions in some O₃ sensitive species (e.g., black cherry). The conclusions were limited, however, because of the uncertainty in interpolating O₃ exposures in many of the areas and because the hydrologic index used might not reflect actual water stress.

The non-stomatal component of plant defenses are the most difficult to quantify, but some studies are available (Heath et al., 2009; Barnes et al., 2002; Plochl et al., 2000; Chen et al., 1998; Massman and Grantz, 1995). Massman et al. (2000) developed a conceptual model of a dose-based index to determine how plant injury response to O₃ relates to the traditional exposure-based parameters. The index used time-
varying-weighted fluxes to account for the fact that flux was not necessarily correlated with plant injury or damage. The model applied only to plant foliar injury and suggested that application of flux-based models for determining plant damage (yield or biomass) would require a better understanding and quantification of the relationship between injury and damage.

9.5.5 Summary

Exposure indices are metrics that quantify exposure as it relates to measured plant damage (i.e., reduced growth). They are summary measures of monitored ambient O₃ concentrations over time intended to provide a consistent metric for reviewing and comparing exposure-response effects obtained from various studies. No recent information is available since 2006 that alters the basic conclusions put forth in the 2006 and 1996 O₃ AQCDs. These AQCDs focused on the research used to develop various exposure indices to help quantify effects on growth and yield in crops, perennials, and trees (primarily seedlings). The performance of indices was compared through regression analyses of earlier studies designed to support the estimation of predictive O₃ exposure-response models for growth and/or yield of crops and tree (seedling) species.

Another approach for improving risk assessment of vegetation response to ambient O₃ is based on determining the O₃ concentration from the atmosphere that enters the leaf (i.e., flux or deposition). Interest has been increasing in recent years, particularly in Europe, in using mathematically tractable flux models for O₃ assessments at the regional, national, and European scale (Matyssek et al., 2008; Paoletti and Manning, 2007; ICP M&M, 2004; Emberson et al., 2000b; Emberson et al., 2000a). While some efforts have been made in the U.S. to calculate O₃ flux into leaves and canopies (Turnipseed et al., 2009; Uddling et al., 2009; Bergweiler et al., 2008; Hogg et al., 2007; Grulke et al., 2004; Grantz et al., 1997; Grantz et al., 1995), little information has been published relating these fluxes to effects on vegetation. There is also concern that not all O₃ stomatal uptake results in a yield reduction, which depends to some degree on the amount of internal detoxification occurring with each particular species. Those species having high amounts of detoxification potential may, in fact, show little relationship between O₃ stomatal uptake and plant response (Musselman and Massman, 1999). The lack of data in the U.S. and the lack of understanding of detoxification processes have made this technique less viable for vulnerability and risk assessments in the U.S.
The main conclusions from the 1996 and 2006 O₃ AQCDs regarding indices based on ambient exposure are still valid. These key conclusions can be restated as follows:

- Ozone effects in plants are cumulative;
- higher O₃ concentrations appear to be more important than lower concentrations in eliciting a response;
- plant sensitivity to O₃ varies with time of day and plant development stage;
- quantifying exposure with indices that accumulate the O₃ hourly concentrations and preferentially weight the higher concentrations improves the explanatory power of exposure/response models for growth and yield, over using indices based on mean and peak exposure values.

Various weighting functions have been used, including threshold-weighted (e.g., SUM06) and continuous sigmoid-weighted (e.g., W126) functions. Based on statistical goodness-of-fit tests, these cumulative, concentration-weighted indices could not be differentiated from one another using data from previous exposure studies. Additional statistical forms for O₃ exposure indices have been discussed in Lee et al. (1988b). The majority of studies published since the 2006 O₃ AQCD do not change earlier conclusions, including the importance of peak concentrations, and the duration and occurrence of O₃ exposures in altering plant growth and yield.

Given the current state of knowledge and the best available data, exposure indices that cumulate and differentially weight the higher hourly average concentrations and also include the “mid-level” values continue to offer the most defensible approach for use in developing response functions and comparing studies, as well as for defining future indices for vegetation protection.

### 9.6 Ozone Exposure-Plant Response Relationships

#### 9.6.1 Introduction

The adequate characterization of the effects of O₃ on plants for the purpose of setting air quality standards is contingent not only on the choice of the index used (i.e., SUM06, W126) to summarize O₃ concentrations (Section 9.5), but also on quantifying the response of the plant variables of interest at specific values of the selected index. The many factors that determine the response of plants to O₃ exposure have been discussed in previous sections. They include species, genotype and other genetic characteristics (Section 9.3), biochemical and physiological status (Section 9.3), previous and current exposure to other stressors (Section 9.4.8), and characteristics of the exposure itself (Section 9.5). Establishing a secondary air quality standard entails the capability to generalize those observations, in order to obtain predictions that are reliable enough under a broad variety of conditions, taking into account these factors. This section reviews results that have related specific quantitative observations of O₃ exposure with quantitative observations of plant
responses, and the predictions of responses that have been derived from those observations through empirical models.

For four decades, exposure to $O_3$ at ambient concentrations found in many areas of the U.S. has been known to cause detrimental effects in plants (U.S. EPA, 2006b, 1996b, 1984, 1978a). Results published after the 2006 $O_3$ AQCD continue to support this finding, and the following sections deal with the quantitative characterizations of the relationship, and what new insights may have appeared since 2006. Detrimental effects on plants include visible injury, decreases in the rate of photosynthesis, reduced growth, and reduced yield of marketable plant parts. Most published exposure-response data have reported $O_3$ effects on the yield of crops and the growth of tree seedlings, and those two variables have been the focus of the characterization of ecological impacts of $O_3$ for the purpose of setting secondary air quality standards. In order to support quantitative modeling of exposure-response relationships, data should preferably include more than three levels of exposure, and some control of potential confounding or interacting factors should be present in order to model the relationship with sufficient accuracy. Letting potential confounders, such as other stressors, vary freely when generating $O_3$ exposure-response data might improve the ‘realism’ of the data, but it also greatly increases the amount of data necessary to extract a clear quantitative description of the relationship. Conversely however, experimental settings should not be so exhaustively restrictive as to make generalization outside of them problematic. During the last four decades, many of the studies of the effects of $O_3$ on growth and yield of plants have not included enough levels of $O_3$ to parameterize more than the simplest linear model. The majority of these studies have only contrasted two levels, ambient and elevated, or sometimes three by adding carbon filtration in OTC studies, with little or no consideration of quantitatively relating specific values of exposure to specific values of growth or yield. This is not to say that studies that did not include more than two or three levels of $O_3$ exposure, or studies that were conducted in uncontrolled environments, do not provide exposure-response information that is highly relevant to reviewing air quality standards. In fact, they can be essential in verifying the agreement between predictions obtained through the empirical models derived from experiments such as NCLAN, and observations. The consensus of model predictions and observations from a variety of studies conducted in other locations, at other times, and using different exposure methods, greatly increases confidence in the reliability of both. Furthermore, if they are considered in the aggregate, studies with few levels of exposure or high unaccounted variability can provide additional independent estimates of decrements in plant growth and yield, at least within a few broad categories of exposure.

Extensive exposure-response information on a wide variety of plant species has been produced by two long-term projects that were designed with the explicit aim of obtaining quantitative characterizations of the response of such an assortment of crop plants and tree seedlings to $O_3$ under North American conditions: the NCLAN project for crops, and the EPA National Health and Environmental Effects Research Laboratory, Western Ecology Division tree seedling project (NHEERL/WED). The NCLAN project was initiated by the EPA in 1980 primarily to improve
estimates of yield loss under field conditions and to estimate the magnitude of crop losses caused by \( \text{O}_3 \) throughout the U.S. (Heck et al., 1991; Heck et al., 1982). The cultural conditions used in the NCLAN studies approximated typical agronomic practices, and the primary objectives were: (1) to define relationships between yields of major agricultural crops and \( \text{O}_3 \) exposure as required to provide data necessary for economic assessments and development of \( \text{O}_3 \) NAAQS; (2) to assess the national economic consequences resulting from \( \text{O}_3 \) exposure of major agricultural crops; and (3) to advance understanding of cause-and-effect relationships that determine crop responses to pollutant exposures.

NCLAN experiments yielded 54 exposure-response curves for 12 crop species, some of which were represented by multiple cultivars at several of 6 locations throughout the United States. The NHEERL/WED project was initiated by EPA in 1988 with similar objectives for tree species, and yielded 49 exposure-responses curves for multiple genotypes of 11 tree species grown for up to three years in Oregon, Michigan, and the Great Smoky Mountains National Park. Both projects used OTCs to expose plants to three to five levels of \( \text{O}_3 \). Eight of the 54 crop datasets were from plants grown under a combination of \( \text{O}_3 \) exposure and experimental drought conditions. Figure 9-14 through Figure 9-17 summarizes some of the NCLAN and NHEERL/WED results.

It should be noted that data from FACE experiments might also be used for modeling exposure-response. They only use two levels of \( \text{O}_3 \) (ambient concentration at the site and a multiple of it), but given that the value of both levels of exposure changes every year, and that they are typically run for many consecutive years, aggregating data over time produces twice as many levels of \( \text{O}_3 \) as there are years. As described in Section 9.2.4, FACE experiments seek to impose fewer constraints on the growth environment than OTCs. As a consequence, FACE studies have to contend with larger variability, especially year-to-year variability, but the difference in experimental conditions between the two methodologies makes comparisons between their results especially useful.

Growth and yield of at least one crop (soybean) has been investigated in yearly experiments since 2001 at a FACE facility in Illinois (UIUC, 2010; Morgan et al., 2006). However, almost all analyses of SoyFACE published so far have been based on subsets of one or two years, and have only contrasted ambient versus elevated \( \text{O}_3 \) as categorical variables. They have not modeled the response of growth and yield to \( \text{O}_3 \) exposure continuously over the range of exposure values that have occurred over time. The only exception is a study by Betzelberger et al. (2010), who used a linear regression model on data pooled over 2 years. Likewise, trees of three species (trembling aspen, paper birch, and sugar maple) were grown between 1998 and 2009 in a FACE experiment located in Rhinelander, Wisconsin (Pregitzer et al., 2008; Dickson et al., 2000). The Aspen FACE experiment has provided extensive data on responses of trees beyond the seedling stage under long-term exposure, and also on ecosystem-level responses (Section 9.4), but the only attempt to use those data in a continuous model of the response of tree growth to \( \text{O}_3 \) exposure (Percy et al., 2007) suffered from severe methodological problems, some of which are discussed in
Section 9.6.3. Finally, one experiment was able to exploit a naturally occurring gradient of O₃ concentrations to fit a linear regression model to the growth of cottonwood (Gregg et al., 2006, 2003). Factors such as genotype, soil type and soil moisture were under experimental control, and the authors were able to partition out the effects of potential confounders such as temperature, atmospheric N deposition, and ambient CO₂.

A serious difficulty in assessing results of exposure-response research is the multiplicity of O₃ metrics that have been used in reporting. As described in Section 9.5, metrics that entail either weighting or thresholding of hourly values cannot be algebraically converted into one another, or into unweighted metrics such as hourly average. When computing O₃ exposure using weighted or thresholded metrics, each metric has to be computed separately from the original hourly data. Comparisons of exposure-response models can only be made between studies that used the same metric, and the value of exposure at which a given plant response is expected using one metric of exposure cannot be exactly converted to another metric. Determining the exposure value at which an effect would be observed in a different metric can only be accomplished by first computing the experimental exposures in this metric from the hourly data, then estimating (fitting) model coefficients again. This problem is irremediable, although useful comparisons might be made using categorical exposures such as ‘current ambient exposure’ or ‘2050 projected exposure’, which can serve as a common reference for quantitative values expressed in various metrics. Studies that contained growth or yield exposure-response data at few levels of exposure, and/or using metrics other than W126 are summarized in Table 9-17 and Table 9-18.

9.6.2 Estimates of Crop Yield Loss and Tree Seedling Biomass Loss in the 1996 and 2006 Ozone AQCDs


$$Y = \alpha e^{-(\frac{W126}{\eta})^\beta}$$

Equation 9-2

is the most appropriate model for the response of absolute yield and growth to O₃ exposure, because of the interpretability of its parameters, its flexibility (given the small number of parameters), and its tractability for estimation. In addition, removing the intercept \( \alpha \) results in a model of relative yield (yield relative to [yield at exposure=0]) without any further reparameterization. Formulating the model in terms
of relative yield or relative yield loss (yield loss=[1 − relative yield]) is essential in comparing exposure-response across species, genotypes, or experiments for which absolute values of the response may vary greatly. In the 1996 and 2006 O\textsubscript{3} AQCDs, the two-parameter model of relative yield was used in deriving common models for multiple species, multiple genotypes within species, and multiple locations.

Given the disparate species, genotypes, and locations that were included in the NCLAN and NHEERL/WED projects, and in the absence of plausible distributional assumptions with respect to those variables, a three step process using robust methods was used to obtain parameter estimates that could be generalized. The models that were derived for each species or group of species were referred to as median composite functions. In the first step, the three parameters of the Weibull model were computed for absolute yield or biomass data from each NCLAN and NHEERL/WED experiment (54 crop datasets and 49 tree seedling datasets), using nonlinear regression. When data were only available for three levels of exposure because of experimental problems, the shape parameter \( \beta \) was constrained to 1, reducing the model to an exponential decay model. In the second step, \( \alpha \) was dropped, and predicted values of relative yield or biomass were then computed for 12-hour W126 exposures between 0 and 60 ppm-h. At each of these W126 exposure values, the 25th, 50th, and 75th percentiles of the response were identified among the predicted curves of relative response. For example, for the 34 NCLAN studies of 12 crop species grown under non-droughted conditions for a complete cropping cycle (Figure 9-14), the 3 quartiles of the response were identified at every integer value of W126 between 0 and 60. The third step fitted a two-parameter Weibull model to those percentiles, yielding the median composite function for the relative yield or biomass response to O\textsubscript{3} exposure for each grouping of interest (e.g., all crops, all trees, all datasets for one species), as well as composite functions for the other quartiles. In the 1996 and 2006 O\textsubscript{3} AQCDs this modeling of crop yield loss and tree seedling biomass loss was conducted using the SUM06 metric for exposure. This section updates those results by using the 12-hour W126 as proposed in 2007 (72 FR 37818) and 2010 (75 FR 2938, page 3, 003). Figure 9-14 through Figure 9-17 present quantiles of predicted relative yield or biomass loss at seven values of the 12-h W126 for some representative groupings of NCLAN and NHEERL/WED results. Table 9-9 through Table 9-11 give the 90-day 12-h W126 O\textsubscript{3} exposure values at which 10 and 20\% yield or biomass losses are predicted in 50 and 75\% of crop or tree species using the composite functions.
Note: Quantiles of the predicted relative yield loss at 7 values of 12-hour W126 for 34 Weibull curves estimated using nonlinear regression on data from 34 studies of 12 crop species grown under well-watered conditions for the full duration of 1 cropping cycle.


Figure 9-14  Quantiles of predicted relative yield loss for 34 NCLAN crop experiments.
Notes: Quantiles of the predicted relative yield loss at 7 values of 12-h W126 for Weibull curves estimated using nonlinear regression for 4 species grown under well-watered conditions for the full duration of 1 cropping cycle. The number of studies available for each species is indicated on each plot.


**Figure 9-15**  Quantiles of predicted relative yield loss for 4 crop species in NCLAN experiments.
Note: Quantiles of the predicted relative above-ground biomass loss at 7 values of 12-h W126 for 49 Weibull curves estimated using nonlinear regression on data from 49 studies of 11 tree species grown under well-watered conditions for 1 or 2 years. Curves were standardized to 90-day W126.


Figure 9-16  Quantiles of predicted relative biomass loss for 49 studies of 11 tree species in NHEERL/WED experiments.
Note: Quantiles of the predicted relative above-ground biomass loss at 7 exposure values of 12-h W126 for Weibull curves estimated using nonlinear regression on data for 4 tree species grown under well-watered conditions for 1 or 2 year. Curves were standardized to 90-day W126. The number of studies available for each species is indicated on each plot.


**Figure 9-17** Quantiles of predicted relative biomass loss for 4 tree species in NHEERL/WED experiments.
Table 9-9  
Ozone exposures at which 10 and 20% yield loss is predicted for 50 and 75% of crop species.

<table>
<thead>
<tr>
<th>Predicted Yield Loss for Crop Speciesa</th>
<th>90-day 12-h W126 for 10% yield loss (ppm-h)</th>
<th>90-day 12-h W126 for 20% yield loss (ppm-h)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model for the 50th Percentile of 34 curves</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Relative yield=exp(-(W126/104.82)**1.424)</td>
<td>22</td>
<td>37</td>
</tr>
<tr>
<td>Model for the 75th Percentile of 34 curves</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Relative yield=exp(-(W126/78.12)**1.415)</td>
<td>16</td>
<td>27</td>
</tr>
</tbody>
</table>

aBased on composite functions for the 50th and 75th percentiles of 34 Weibull curves for relative yield loss data from 34 non-droughted NCLAN studies of 12 crop species; curves were standardized to 90-day W126.
Source of parameters for the 34 curves: Lee and Hogsett (1996).

Table 9-10  
Ozone exposures at which 10 and 20% yield loss is predicted for 50 and 75% of crop species (Droughted versus Watered conditions).

<table>
<thead>
<tr>
<th>Predicted Yield Loss for Crop Speciesa</th>
<th>90 day 12-h W126 for 10% yield loss (ppm-h)</th>
<th>90 day 12-h W126 for 20% yield loss (ppm-h)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model for the 50th Percentile of 2×8 curves</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Watered Relative yield=exp(-(W126/132.86)**1.170)</td>
<td>19</td>
<td>37</td>
</tr>
<tr>
<td>Droughted Relative yield=exp(-(W126/179.84)**1.713)</td>
<td>48</td>
<td>75</td>
</tr>
<tr>
<td>Model for the 75th Percentile of 2×8 curves</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Watered Relative yield=exp(-(W126/90.43)**1.310)</td>
<td>16</td>
<td>29</td>
</tr>
<tr>
<td>Droughted Relative yield=exp(-(W126/105.16)**1.833)</td>
<td>31</td>
<td>46</td>
</tr>
</tbody>
</table>

aUnder drought conditions and adequate moisture based on composite functions for the 50th and 75th percentiles of 16 Weibull curves for relative yield loss data from 8 NCLAN studies that paired droughted and watered conditions for the same genotype; curves were standardized to 90-day W126.
Table 9-11  Ozone exposures at which 10 and 20% biomass loss is predicted for 50 and 75% of tree species.

<table>
<thead>
<tr>
<th>Predicted Biomass Loss for Tree Species&lt;sup&gt;a&lt;/sup&gt;</th>
<th>90 day 12 h W126 for 10% yield loss (ppm-h)</th>
<th>90 day 12 h W126 for 20% yield loss (ppm-h)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model for the 50th Percentile of 49 curves</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Relative yield=(\exp(-\left(W_{126}/131.57\right)^{1.242}))</td>
<td>21</td>
<td>39</td>
</tr>
<tr>
<td>Model for the 75th Percentile of 49 curves</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Relative yield=(\exp(-\left(W_{126}/65.49\right)^{1.500}))</td>
<td>15</td>
<td>24</td>
</tr>
</tbody>
</table>

<sup>a</sup>Based on composite functions for the 50th and 75th percentiles of 49 Weibull curves for relative above-ground biomass loss data from 49 studies of 11 tree species grown under well-watered conditions for 1 or 2 year; curves were standardized to 90-day W126. Source of parameters for the 49 curves: Lee and Hogsett (1996).

9.6.3 Validation of 1996 and 2006 Ozone AQCD Models and Methodology Using the 90-day 12-h W126 and Current FACE Data

Since the completion of the NCLAN and NHEERL/WED projects, almost no studies have been published that could provide a basis for estimates of exposure-response that can be compared to those of the 1996 and 2006 O<sub>3</sub> AQCDs. Most experiments, regardless of exposure methodology, include only two levels of exposure. In addition, very few studies have included measurements of exposure using the W126 metric, or the hourly O<sub>3</sub> concentration data that would allow computing exposure using the W126. Two FACE projects, however, were conducted over multiple years, and by adding to the number of exposure levels over time, can support independent model estimation and prediction using the same model and the same robust process as summarized in Section 9.6.2. Hourly O<sub>3</sub> data were available from both FACE projects.

The SoyFACE project is situated near Champaign, IL, and comprises 32 octagonal rings (20m-diameter), 4 of which in a given year are exposed to ambient conditions, and 4 of which are exposed to elevated O<sub>3</sub> as a fixed proportion of the instantaneous ambient concentration (Betzelberger et al., 2010; UIUC, 2010; Morgan et al., 2006; Morgan et al., 2004). Since 2002, yield data have been collected for up to 8 genotypes of soybean grown in subplots within each ring. The Aspen FACE project is situated in Rhinelander, WI, and comprises 12 rings (30m-diameter), 3 of which are exposed to ambient conditions, and 3 of which are exposed to O<sub>3</sub> as a fixed proportion of the instantaneous ambient concentration (Pregitzer et al., 2008; Karnosky et al., 2005; Dickson et al., 2000). In the summer of 1997, half the area of each ring was planted with small (five to seven leaf sized) clonally propagated plants of five genotypes of trembling aspen, which were left to grow in those environments until 2009. Biomass data are currently available for the years 1997-2005 (King et al., 2005). Ozone exposure in these two FACE projects can be viewed as a categorical
variable with two levels: ambient, and elevated. However, this overlooks the facts that not only do both ambient and elevated exposure vary from year to year, but the proportionality between them also changes yearly. This change has two sources: first, the dispensing of O₃ into the elevated exposure rings varies from the set point for the ambient/elevated proportionality to some extent, and for SoyFACE, the set point changed between years. Second, when using threshold or concentration-weighted cumulative metrics (such as AOT40, SUM06 or W126), the proportionality does not propagate regularly from the hourly data to the yearly value. For example, hourly average elevated exposures that are a constant 1.5 times greater than ambient do not result in AOT40, SUM06 or W126 values that are some constant multiple of the ambient values of those indices. Depending on the fraction of hourly values that are above the threshold or heavily weighted, the same average yearly exposure will result in different exposure values when using thresholded or weighted metrics. In some years, elevated exposures in FACE experiments experience many more values above the threshold, or more heavily weighted than the ambient exposures; thus in those years, the distance between ambient and elevated exposure values increases relative to other years. As a consequence, the number of exposure levels in multi-year experiments is twice the number of years. In the case of SoyFACE for the period between 2002 and 2008, ambient exposure in the highest year was approximately equal to elevated exposure in the lowest year, with 14 levels of O₃ exposure evenly distributed from lowest to highest. The particular conditions of the Aspen FACE experiment resulted in 12 exposure levels between 1998 and 2003, but they were not as evenly distributed between minimum and maximum over the 6-year period.

There are necessary differences in the modeling of exposure-response in annual plants such as soybean, and in perennial plants such as aspen trees, when exposure takes place over multiple years. In annual plants, responses recorded at the end of the life cycle, i.e., yearly, are analyzed in relationship to that year’s exposure. Yield of soybeans is affected by exposure during the year the crop was growing, and a new crop is planted every year. Thus an exposure-response relationship can be modeled from yearly responses matched to yearly exposures, with those exposure-response data points having been generated in separate years. For perennial organisms, which are not harvested yearly and continue to grow from year to year, such pairing of exposure and response cannot be done without accounting for time. Not only does the size of the organism at the beginning of each year of exposure increase, but size is also dependent on the exposure from previous years. Therefore the relationship of response and exposure must be analyzed either one year at a time, or by standardizing the response as a yearly increment relative to size at the beginning of each year. Furthermore, the relevant measurement of exposure is cumulative, or cumulative yearly average exposure, starting in the year exposure was initiated, up to the end of the year of interest. When analyzing the growth of trees over several years, it would be evidently incorrect to pair the exposure level in every discrete year with absolute size of the trees that year, and posit a direct relationship between them, without taking increasing age into consideration. In the Aspen FACE experiment, for example, one could not establish an exposure-response relationship by matching 12 yearly exposures and 12 yearly tree sizes, while disregarding age as if size did not
also depend on it. This is the basis of the 2007 study of Aspen FACE data by Percy et al. (2007), which compares the size of trees of various ages as if they were all the same age, and was therefore not informative.

9.6.3.1 Comparison of NCLAN-Based Prediction and SoyFACE Data

For this ISA, EPA conducted a comparison between yield of soybean as predicted by the composite function three-step process (Section 9.6.2) using NCLAN data, and observations of yield in SoyFACE. The median composite function for relative yield was derived for the 11 NCLAN soybean Weibull functions for non-droughted studies, and comparisons between the predictions of the median composite and SoyFACE observations were conducted as follows.

For the years 2007 and 2008, SoyFACE yield data were available for 7 and 6 genotypes, respectively. The EPA used those data to compare the relative change in yield observed in SoyFACE in a given year between ambient O₃ and elevated O₃, versus the relative change in yield predicted by the NCLAN-based median composite function between those same two values of O₃ exposure. The two parameter median composite function for relative yield of soybean based on NCLAN data was used to predict yield response at the two values of exposure observed in SoyFACE in each year, and the change between yield under ambient and elevated was compared to the change observed in SoyFACE for the relevant year (Table 9-12). This approach results in a direct comparison of predicted versus observed change in yield. Because the value of relative response between any two values of O₃ exposure is independent of the intercept α, this comparison does not require prediction of the absolute values of the responses.

Since comparisons of absolute values might be of interest, the predictive functions were also scaled to the observed data: SoyFACE data were used to compute an intercept α while the shape and scale parameters (β and η) were held at their value in the NCLAN predictive model. This method gives a comparison of prediction and observation that takes all the observed information into account to provide the best possible estimate of the intercept, and thus the best possible scaling (Table 9-13 and Figure 9-18). For the comparison of NCLAN and SoyFACE, this validation was possible for 2007 and 2008, where data for 7 and 6 soybean genotypes, respectively, were available. The median composite function for relative yield was derived for the 11 NCLAN soybean Weibull functions for nondroughted studies, and the values of median yield under ambient exposure at SoyFACE in 2007 and 2008 were used to obtain an estimate of the intercept α for the NCLAN median function in each of the two years. Table 9-12 presents the results of ambient/elevated relative yield comparisons between the NCLAN-derived predictions and SoyFACE observations. Table 9-13 and Figure 9-18 present the results of comparisons between NCLAN-derived predictions and SoyFACE observations of yield, with the predictive function scaled to provide absolute yield values.
Table 9-12  Comparison between change in yield observed in the SoyFACE experiment between elevated and ambient O₃, and change predicted at the same values of O₃ by the median composite function for NCLAN.

<table>
<thead>
<tr>
<th>Year</th>
<th>Ambient (g/m²)</th>
<th>Elevated (g/m²)</th>
<th>Predicted by NCLANa</th>
<th>Observed at SoyFACE (g/m²)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2007</td>
<td>4.39</td>
<td>46.23</td>
<td>75</td>
<td>76</td>
</tr>
<tr>
<td>2008</td>
<td>3.23</td>
<td>28.79</td>
<td>85</td>
<td>88</td>
</tr>
</tbody>
</table>

aTwo-parameter relative yield model.

Table 9-13  Comparison between yield observed in the SoyFACE experiment and yield predicted at the same values of O₃ by the median composite function for NCLAN.

<table>
<thead>
<tr>
<th>Year</th>
<th>Ambient (g/m²)</th>
<th>Elevated (g/m²)</th>
<th>Yield predicted by NCLANa (g/m²)</th>
<th>Yield observed at SoyFACE (g/m²)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2007</td>
<td>4.39</td>
<td>46.23</td>
<td>309.2</td>
<td>305.2</td>
</tr>
<tr>
<td>2008</td>
<td>3.23</td>
<td>28.79</td>
<td>350.3</td>
<td>344.8</td>
</tr>
</tbody>
</table>

aThree-parameter absolute yield model with intercept scaled to SoyFACE data.
Finally, a composite function for the 25th, 50th, and 75th percentiles was developed from SoyFACE annual yield data, and compared to the NCLAN-based function. The process described in Section 9.6.2 was applied to SoyFACE data for individual genotypes, aggregated over the years during which each was grown; one genotype from 2003 to 2007, and six genotypes in 2007 and 2008. First, the three parameter Weibull model described in Section 9.6.2 was estimated using nonlinear regression on exposure-yield data for each genotype separately, over the years for which data were available, totaling seven curves. The 25th, 50th, and 75th percentiles of the predicted values for the two parameter relative yield curves were then identified at every integer of W126 between 0 and 60, and a two-parameter Weibull model estimated by regression for the three quartiles. The comparison between these composite functions for the quartiles of relative yield loss in SoyFACE and the corresponding composite functions for NCLAN is presented in Figure 9-19.
Figure 9-19  Comparison of composite functions for the quartiles of 7 curves for 7 genotypes of soybean grown in the SoyFACE experiment, and for the quartiles of 11 curves for 5 genotypes of soybean grown in the NCLAN project.

As seen in Table 9-13 and Table 9-14 and in Table 9-18, the agreement between predictions based on NCLAN data and SoyFACE observations was notably close in single-year comparisons. Together with the very high agreement between median composite models for NCLAN and SoyFACE (Figure 9-19), it provides very strong mutual confirmation of those two projects’ results with respect to the response of yield of soybeans to O₃ exposure. It is readily apparent from these results that the methodology described in Section 9.6.2 for obtaining predictions of yield or yield loss from NCLAN data is strongly validated by SoyFACE results. As described in Section 9.2, the exposure technologies used in the two projects were in sharp contrast, specifically with respect to the balance each achieved between control of potential interacting factors or confounders, and fidelity to natural conditions. The comparisons that EPA conducted therefore demonstrate that the methodology used in developing the composite functions is resistant to the influence of nuisance variables and that predictions are reliable. They may also suggest that the aspects in which the two exposure technologies differ have less influence on exposure-response than initially supposed. These results are also in agreement with comparative studies reviewed in Section 9.2.6.
EPA also conducted two comparisons between prediction of above-ground biomass loss based on NHEERL/WED results and observations from Aspen FACE. The median composite function was developed from NHEERL/WED data for 11 studies that used wild-type seedlings of aspen as well as four clonally propagated genotypes. All plants were grown in OTCs for one growing season before being destructively harvested. Aspen FACE data were from clonally propagated trees of five genotypes grown from 1998 to 2003, with above-ground biomass calculated using allometric equations derived from data for trees harvested destructively in 2000 and 2002 (King et al., 2005).

The two parameter median composite function for relative biomass was used to predict biomass response under the observed elevated exposure, relative to its value under observed ambient exposure, for each separate year of Aspen FACE. EPA first compared Aspen FACE observations of the change in biomass between ambient and elevated exposure with the corresponding prediction at the same values of exposure. Comparisons between observed and predicted absolute biomass values were then conducted for each year by scaling the predictive function to yearly Aspen FACE data as described for soybean data in Section 9.6.3.1. In all cases, yearly 90 day 12-hour W126 values for Aspen FACE were computed as the cumulative average from the year of planting up to the year of interest. A comparison of composite functions between NHEERL/WED and Aspen FACE, similar to the one performed for NCLAN and SoyFACE, was not possible: as discussed in the introduction to Section 9.6, the pairing of 12 exposure values from separate years and 12 values of biomass cannot be the basis for a model of exposure-response, because the trees continued growing for the six-year period of exposure. Because the same trees were used for the entire duration, and continued to grow, data could not be aggregated over years. Table 9-14 presents the results of ambient/elevated relative biomass comparisons between the NHEERL/WED-derived predictions and Aspen FACE observations. Table 9-15 and Figure 9-20 present the results of the comparison between NHEERL/WED-derived predictions and Aspen FACE observations for absolute biomass, using Aspen FACE data to scale the NHEERL/WED-derived composite function.
Table 9-14  Comparison between change in above-ground biomass elevated and ambient O$_3$ in Aspen FACE experiment in 6 year, and change predicted at the same values of O$_3$ by the median composite function for NHEERL/WED.

<table>
<thead>
<tr>
<th>Year</th>
<th>Ambient</th>
<th>Elevated</th>
<th>Predicted by NHEERL/WED*</th>
<th>Observed at Aspen FACE</th>
</tr>
</thead>
<tbody>
<tr>
<td>1998</td>
<td>3.19</td>
<td>30.08</td>
<td>74</td>
<td>75</td>
</tr>
<tr>
<td>1999</td>
<td>2.61</td>
<td>33.85</td>
<td>70</td>
<td>70</td>
</tr>
<tr>
<td>2000</td>
<td>2.43</td>
<td>30.16</td>
<td>74</td>
<td>71</td>
</tr>
<tr>
<td>2001</td>
<td>2.55</td>
<td>31.00</td>
<td>73</td>
<td>71</td>
</tr>
<tr>
<td>2002</td>
<td>2.51</td>
<td>30.27</td>
<td>74</td>
<td>69</td>
</tr>
<tr>
<td>2003</td>
<td>2.86</td>
<td>29.12</td>
<td>75</td>
<td>71</td>
</tr>
</tbody>
</table>

*Two-parameter relative biomass model

Table 9-15  Comparison between above-ground biomass observed in Aspen FACE experiment in 6 year and biomass predicted by the median composite function based on NHEERL/WED.

<table>
<thead>
<tr>
<th>Year</th>
<th>Ambient</th>
<th>Elevated</th>
<th>Biomass Predicted by NHEERL/WED* (g/m$^2$)</th>
<th>Biomass Observed at Aspen FACE (g/m$^2$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1998</td>
<td>3.19</td>
<td>30.08</td>
<td>276.0</td>
<td>203.2</td>
</tr>
<tr>
<td>1999</td>
<td>2.61</td>
<td>33.85</td>
<td>958.7</td>
<td>668.3</td>
</tr>
<tr>
<td>2000</td>
<td>2.43</td>
<td>30.16</td>
<td>1,382.4</td>
<td>1,022.8</td>
</tr>
<tr>
<td>2001</td>
<td>2.55</td>
<td>31.00</td>
<td>1,607.0</td>
<td>1,173.7</td>
</tr>
<tr>
<td>2002</td>
<td>2.51</td>
<td>30.27</td>
<td>2,079.0</td>
<td>1,532.1</td>
</tr>
<tr>
<td>2003</td>
<td>2.86</td>
<td>29.12</td>
<td>2,640.1</td>
<td>1,981.2</td>
</tr>
</tbody>
</table>

*Three-parameter absolute biomass model with intercept scaled to Aspen FACE data.
Figure 9-20  **Comparison between above-ground biomass observed in Aspen FACE experiment in 6 year and biomass predicted by the median composite function based on NHEERL/WED.**

As in the comparisons between NCLAN and SoyFACE, the agreement between predictions based on NHEERL/WED data and Aspen FACE observations was very close. The results of the two projects strongly reinforce each other with respect to the response of aspen biomass to O₃ exposure. The methodology used for obtaining the median composite function is shown to be capable of deriving a predictive model despite potential confounders, and despite the added measurement error that is expected from calculating biomass using allometric equations. In addition, the function based on one year of growth was shown to be applicable to subsequent years.

The results of experiments that used different exposure methodologies, different genotypes, locations, and durations converged to the same values of response to O₃ exposure for each of two very dissimilar plant species, and predictions based on the earlier experiments were validated by the data from current ones. However, in these comparisons, the process used in establishing predictive functions involved aggregating data over variables such as time, locations, and genotypes, and the use of a robust statistic (quartiles) for that aggregation. The validating data, from SoyFACE and Aspen FACE, were in turn aggregated over the same variables. The accuracy of
predictions is not expected to be conserved for individual values of those variables over which aggregation occurred. For example, the predicted values for soybean, based on data for five genotypes, are not expected to be valid for each genotype separately. As shown in the validation, however, aggregation that occurred over different values of the same variable did not affect accuracy: composite functions based on one set of genotypes were predictive for another set, as long as medians were used for both sets. A study of cottonwood (Populus deltoides) conducted using a naturally occurring gradient of O₃ exposure (Gregg et al., 2006, 2003) may provide an illustration of the response of an individual species whose response is far from the median response for an aggregation of species.

9.6.3.3 Exposure-Response in a Gradient Study

Gregg et al. (2003) grew saplings of one clonally propagated genotype of cottonwood (Populus deltoides) in seven locations within New York City and in the surrounding region between July and September in 1992, 1993 and 1994, and harvested them 72 days after planting. Owing to regional gradients of atmospheric O₃ concentration, the experiment yielded eight levels of exposure (Figure 9-21), and the authors were able to rule out environmental variables other than O₃ to account for the large differences in biomass observed after one season of growth. The deficit in growth increased substantially faster with increasing O₃ exposure than has been observed in aspen, another species of the same genus (Populus tremuloides, Section 9.6.3.2). Using a three parameter Weibull model (Figure 9-21), the biomass of cottonwood at a W126 exposure of 15 ppm-h, relative to biomass at 5 ppm-h, is estimated to be 0.18 (18% of growth at 5 ppm-h). The relative biomass of trembling aspen within the same 5-15 ppm-h range of exposure is estimated to be 0.92, using the median composite model for aspen whose very close agreement with Aspen FACE data was shown in Section 9.6.3.2. Using a median composite function for all deciduous trees in the NHEERL/WED project (6 species in 21 studies) also gives predictions that are very distant from the cottonwood response observed in this experiment. For all deciduous tree species in NHEERL/WED, biomass at a W126 exposure of 15 ppm-h, relative to biomass at 5 ppm-h, was estimated to be 0.87.
As shown in Section 9.6.2, the median models available for trembling aspen and soybean have verifiable predictive ability for those particular species. This suggests that the corresponding NCLAN- and NHEERL/WED-based models for multiple crop and tree species can provide reliable estimates of losses for similar assortments of species. However, their predictive ability would likely be poor for individual species not tested.

The cottonwood data of Gregg et al. (2003) show an extremely severe response to O₃. They are consistent with the expectation that among species and genotypes, some are likely to be substantially more sensitive than a median measure, such as the estimate produced by NHEERL/WED (Figure 9-16), but the sensitivity of this particular species has not been studied elsewhere.

An alternative hypothesis for the difference between the response of cottonwood in this experiment and deciduous tree species in NHEERL/WED, or the difference between the response of cottonwood and aspen in NHEERL/WED and Aspen FACE, could be the presence of confounding factors in the environments where the experiment was conducted. However, variability in temperature, moisture, soil fertility, and atmospheric deposition of N were all ruled out by Gregg et al. (2003) as contributing to the observed response to O₃. In addition, this hypothesis would imply

Figure 9-21  Above-ground biomass for one genotype of cottonwood grown in seven locations for one season in 3 years.
that the unrecognized confounder(s) were either absent from both OTC and FACE studies, or had the same value in both. This is not impossible, but the hypothesis that cottonwood is very sensitive to O₃ exposure is more parsimonious, and sufficient.

### 9.6.3.4 Meta-analyses of Growth and Yield Studies

Since the 2006 O₃ AQCD, five studies have used meta-analytic methods to integrate results from experimental studies of crops or tree species relevant to the United States. It is possible to obtain exposure-response data for growth and yield from those meta-analyses, but because all of them provided summary measurements of O₃ exposure as hourly averages of various lengths of exposures, comparisons with exposure-response results where exposure is expressed as W₁₂₆ are problematic. Table 9-16 summarizes the characteristics of the five meta-analyses. They all included studies conducted in the U.S. and other locations worldwide, and all of them expressed responses as comparative change between levels of exposure to O₃, with carbon filtered air (CF) among those levels. Using hourly average concentration to summarize exposure, CF rarely equates with absence of O₃, although it almost always near zero when exposure is summarized as W₁₂₆, SUM06, or AOT40.

<table>
<thead>
<tr>
<th>Study</th>
<th>Number of articles included</th>
<th>Years of publication surveyed</th>
<th>Crop, species or genera</th>
<th>Response</th>
<th>Number of O₃ levels</th>
<th>Duration of exposure</th>
</tr>
</thead>
<tbody>
<tr>
<td>Feng et al. (2008b)</td>
<td>53</td>
<td>1980-2007</td>
<td>Wheat</td>
<td>Yield</td>
<td>5</td>
<td>&gt;10 days</td>
</tr>
<tr>
<td>Feng and Kobayashi (2009)</td>
<td>All crops together: 81</td>
<td>1980-2007</td>
<td>Potato, barley, wheat, rice, bean, soybean</td>
<td>Yield</td>
<td>3</td>
<td>&gt;10 days</td>
</tr>
<tr>
<td>Grantz et al. (2006)</td>
<td>16</td>
<td>1992-2004</td>
<td>34 Herbaceous dicots 21 Herbaceous monocots 5 Tree species</td>
<td>Relative Growth Rate</td>
<td>2</td>
<td>2-24 weeks</td>
</tr>
<tr>
<td>Wittig et al. (2009)</td>
<td>All responses:263 Articles that included biomass:unreported</td>
<td>1970-2006</td>
<td>4 Gymnosperm tree genera 11 Angiosperm tree genera</td>
<td>Total biomass</td>
<td>4</td>
<td>&gt;7 days</td>
</tr>
</tbody>
</table>

Table 9-16 Meta-analyses of growth or yield studies published since 2005.
The only effect of O₃ exposure on yield of rice reported in Ainsworth (2008) was a decrease of 14% with exposure increasing from CF to 62 ppb average concentration. Feng et al. (2008b) were able to separate exposure of wheat into four classes with average concentrations of 42, 69, 97, and 153 ppb, in data where O₃ was the only treatment. Mean responses relative to CF were yield decreases of 17, 25, 49, and 61% respectively. Feng et al. (2008b) observed that wheat yield losses were smaller under conditions of drought, and that Spring wheat and Winter wheat appeared similarly affected. However, mean exposure in studies of Winter wheat was substantially higher than in studies of Spring wheat (86 versus 64 ppb), which suggests that the yield of Spring wheat was in fact more severely affected, since yield was approximately the same, even though Spring wheat was exposed to lower concentrations. Exposures of the six crops considered in Feng and Kobayahi (2009) were classified into two ranges, each compared to CF air. In the lower range of exposure (41-49 ppb), potato studies had the highest average exposure (45 ppb) and wheat and rice the lowest (41 ppb). In the higher range (51-75 ppb), wheat studies had the highest average exposure (65 ppb), and potato, barley and rice the lowest (63 ppb). In other words, across the studies included, all crops were exposed to very similar levels of O₃. At approximately 42 ppb, the yield of potato, barley, wheat, rice, bean, and soybean declined by 5.3, 8.9, 9.7, 17.5, 19, and 7.7% respectively, relative to CF air. At approximately 64 ppb O₃, declines were 11.9, 12.5, 21.1, 37.5, 41.4, and 21.6%. Grantz et al. (2006) reported Relative Growth Rate (RGR) rather than growth, and did not report O₃ exposure values in a way that would allow calculation of mean exposure for each of the three categories of plants for which RGR changes are reported. All studies used only two levels of exposure, with CF air as the lower one, and most used elevated exposure in the range of 40 to 70 ppb. Decline in RGR was 8.2% for the 34 herbaceous dicots, 4.5% for the 21 herbaceous monocots, and 17.9% for the 5 tree species. Finally, Wittig et al. (2009) divided the studies analyzed into three classes of comparisons: CF versus ambient, CF versus elevated, and ambient versus elevated, but reported comparisons between three average levels of exposure besides CF: 40 ppb, 64 ppb, and 97 ppb. Corresponding decreases in total biomass relative to CF were 7, 17, and 17%.

These meta-analyses provide very strong confirmation of EPA’s conclusions from previous O₃ AQCDs: compared to lower levels of ambient O₃, current levels in many locations are having a substantial detrimental effect on the growth and yield of a wide variety of crops and natural vegetation. They also confirm strongly that decreases in growth and yield continue at exposure levels higher than current ambient levels. However, direct comparisons with the predictions of exposure-response models that use concentration-weighted cumulative metrics are difficult.
9.6.3.5 Additional Exposure-Response Data

The studies summarized in Table 9-17 and Table 9-18 contain growth or yield exposure-response data at too few levels of exposure for exposure-response models, and/or used metrics other than W126. These tables update Tables AX9-16 through AX9-19 of the 2006 O₃ AQCD.

9.6.4 Summary

None of the information on effects of O₃ on vegetation published since the 2006 O₃ AQCD has modified the assessment of quantitative exposure-response relationships that was presented in that document. This assessment updates the 2006 exposure-response models by computing them using the W126 metric, cumulated over 90 days. Almost all of the experimental research on the effects of O₃ on growth or yield of plants published since 2006 used only two levels of exposure. In addition, hourly O₃ concentration data that would allow calculations of exposure using the W126 metric are generally unavailable. However, two long-term experiments, one with a crop species (soybean), one with a tree species (aspen), have produced data that can be used to validate the exposure-response models presented in the 2006 O₃ AQCD, and methodology used to derive them.

Quantitative characterization of exposure-response in the 2006 O₃ AQCD was based on experimental data generated for that purpose by the National Crop Loss Assessment Network (NCLAN) and EPA National Health and Environmental Effects Research Laboratory, Western Ecology Division (NHEERL-WED) projects, using OTCs to expose crops and trees seedling to O₃. In recent years, yield and growth results for two of the species that had provided extensive exposure-response information in those projects have become available from studies that used FACE technology, which is intended to provide conditions much closer to natural environments (Pregitzer et al., 2008; Morgan et al., 2006; Morgan et al., 2004; Dickson et al., 2000). The robust methods that were used previously with exposure measured as SUM06 were applied to the NCLAN and NHEERL-WED data with exposure measured as W126, in order to derive single-species median models for soybean and aspen from studies involving different genotypes, years, and locations. The resulting models were used to predict the change in yield of soybean and biomass of aspen between the two levels of exposure reported in recent FACE experiments. Results from these new experiments were exceptionally close to predictions from the models. The accuracy of model predictions for two widely different plant species provides support for the validity of the corresponding multiple-species models for crops and trees in the NCLAN and NHEERL-WED projects. However, variability among species in those projects indicates that the range of sensitivity is likely quite wide. This was confirmed by a recent experiment with cottonwood in a naturally occurring gradient of exposure (Gregg et al., 2006), which established the occurrence of species with responses substantially more severe.
under currently existing conditions than are predicted by the median model for multiple species.

Results from several meta-analyses have provided approximate values for responses of yield of soybean, wheat, rice and other crops under broad categories of exposure, relative to charcoal-filtered air (Ainsworth, 2008; Feng et al., 2008b; Morgan et al., 2003). Likewise, Feng and Kobayashi (2009) have summarized yield data for six crop species under various broad comparative exposure categories, while Wittig et al. (2009) reviewed 263 studies that reported effects on tree biomass. However, these analyses have proved difficult to compare with exposure-response models, especially given that exposure was not expressed in the same W126 metric.

Table 9-17 Summary of studies of effects of O3 exposure on growth and yield of agricultural crops.

<table>
<thead>
<tr>
<th>Species</th>
<th>Facility Location</th>
<th>Exposure Duration</th>
<th>O3 Exposure (Additional Treatment)</th>
<th>Response Measured</th>
<th>Percent Change from CF (Percent Change from Ambient)</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>Alfalfa (Medicago sativa)</td>
<td>OTC; 0.27m³ pots Federico, Italy</td>
<td>2 yr, 2005, 2006</td>
<td>AOT40: CF 0 ppm-h 13.9 ppm-h (2005), 10.1 ppm-h (2006) (NaCl: 0.29, 0.65, 0.83, 1.06 deciSiemens/meter)</td>
<td>Total shoot yield</td>
<td>n.s. (N/A)</td>
<td>Maggio et al. (2009)</td>
</tr>
<tr>
<td>Bean (Phaseolus vulgaris l. cv Borlotto)</td>
<td>OTC; ground-planted Curno, Italy</td>
<td>3 months, 2006</td>
<td>Seasonal AOT40: CF (0.5 ppm-h); ambient (4.6 ppm-h) (N/A)</td>
<td># Seeds per plant; 100-seed weight</td>
<td>-33 (N/A) n.s. (N/A)</td>
<td>Gerosa et al. (2009)</td>
</tr>
<tr>
<td>Big Blue Stem (Andropogon gerardii)</td>
<td>OTC Alabama</td>
<td>4 months, 2003</td>
<td>12-h avg: CF (14 ppb), Ambient (29 ppb), Elevated (71 ppb) (N/A)</td>
<td>Final harvest biomass; RVF</td>
<td>n.s. (n.s.) -7 (-7)</td>
<td>Lewis et al. (2006)</td>
</tr>
<tr>
<td>Brassica napus cv. Westar</td>
<td>Growth chambers Finland</td>
<td>17-26 days</td>
<td>8-h avg: CF (0 ppb), 100 ppb (Bt/non-Bt; herbivory)</td>
<td>Shoot biomass</td>
<td>-30.70 (N/A)</td>
<td>Himanen et al. (2009b)</td>
</tr>
<tr>
<td>Corn (Zea mays cv. Chambord)</td>
<td>OTC France</td>
<td>33 days</td>
<td>AOT40 ppm-h: 1.1; 1.3; 4.9; 7.2; 9.3; 12.8 (N/A)</td>
<td>Total above-ground biomass</td>
<td>N/A (Highest treatment caused - 26% change)</td>
<td>Leitao et al. (2007a)</td>
</tr>
<tr>
<td>Cotton cv. Pima</td>
<td>OTC; 9-L pots San Joaquin Valley, CA</td>
<td>8 weeks</td>
<td>12-h avg: 12.8 ± 0.6; 79.9 ± 6.3; 122.7 ± 9.7 (N/A)</td>
<td>Above-ground biomass</td>
<td>-76 (n.s.)</td>
<td>Grantz and Shrestha (2006)</td>
</tr>
<tr>
<td>Species</td>
<td>Facility</td>
<td>Exposure Duration</td>
<td>O₃ Exposure (Additional Treatment)</td>
<td>Response Measured</td>
<td>Percent Change from CF (Percent Change from Ambient)</td>
<td>Reference</td>
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<tr>
<td>Eastern Gamagrass (<em>Tripsacum dactyloides</em>) OTC Alabama</td>
<td>4 months, 2003</td>
<td>12-h avg: CF (14ppb), Ambient (29 ppb), Elevated (71 ppb) (N/A)</td>
<td>Final harvest biomass; RVF</td>
<td>+68 (+42); -17 (-12)</td>
<td>Lewis et al. (2006)</td>
<td></td>
</tr>
<tr>
<td>Grapevine (<em>Vitis vinifera</em>) OTC Austria</td>
<td>3 yr, May-Oct</td>
<td>AOT40 ppm-h: CF (0), Ambient (7-20), Elevated. 1 (20-30), Elevated. 2 (38-48)</td>
<td>Total fruit yield/ Sugar yield</td>
<td>-20 to -80 in different yr (-20 to -90 in different yr)</td>
<td>Soja et al. (2004)</td>
<td></td>
</tr>
<tr>
<td>Mustard (<em>Brassica campestris</em>) Chambers; 7.5-cm pots</td>
<td>10 days</td>
<td>CF &amp; 67.8 ppb for 7 h (N/A)</td>
<td>Seeds/plant</td>
<td>n.s. (N/A)</td>
<td>Black et al. (2007)</td>
<td></td>
</tr>
<tr>
<td>Oilseed Rape (<em>Brassica napus</em>) OTC Yangtze Delta, China</td>
<td>39 days</td>
<td>Daily avg: 100 ppb, one with diurnal variation and one with constant concentration (N/A)</td>
<td>Biomass and pods per plant</td>
<td>Diurnal variability reduced both biomass and pod number more than constant fumigation (N/A)</td>
<td>Wang et al. (2008)</td>
<td></td>
</tr>
<tr>
<td>Peanut (<em>Arachis hypogaea</em>) OTC Raleigh, NC</td>
<td>3 yr</td>
<td>12-h avg: CF (22 ppb), Ambient (46 ppb), Elevated (75ppb) (CO₂: 375 ppm; 548 ppm; 730 ppm)</td>
<td>Yield (seed weight, g/m)</td>
<td>-33 (-8)</td>
<td>Burkey et al. (2007)</td>
<td></td>
</tr>
<tr>
<td>Poa pratensis OTC Braunschweig, Germany</td>
<td>2000-2002: 4-5 weeks in the Spring</td>
<td>8-h avg: CF+25 (21.7), NF+50 (73.1) (Competition)</td>
<td>Total biomass (g DW/pot)</td>
<td>N/A (n.s.)</td>
<td>Bender et al. (2006)</td>
<td></td>
</tr>
<tr>
<td>Potato (<em>Solanum tuberosum</em>) OTC; CHIP 6 northern European locations</td>
<td>1988,1999. Emergence to harvest</td>
<td>AOT40:CF (0); Ambient (0.27-5.19); NF (0.002-2.93) NF+ (3.10-24.78) (N/A)</td>
<td>Tuber yield averaged across 5 field-sites; Tuber starch content regressed against [O₃] report sig. ± slope with increasing [O₃]</td>
<td>N/A (-27% +27%, most comparisons n.s.) Linear regression slope = -0.0098</td>
<td>Vandermeiren et al. (2005)</td>
<td></td>
</tr>
<tr>
<td>Rice (<em>Oryza sativa</em>) OTC Raleigh, NC</td>
<td>1997-1998, June-September</td>
<td>12-h mean ppb: CF (27.5), Elevated (74.8) (CO₂)</td>
<td>Total biomass; Seed yield</td>
<td>-25(N/A) -13 to 20 (N/A)</td>
<td>Reid and Fiscus (2008)</td>
<td></td>
</tr>
<tr>
<td>Species</td>
<td>Facility</td>
<td>Location</td>
<td>Exposure Duration</td>
<td>O₃ Exposure (Additional Treatment)</td>
<td>Response Measured</td>
<td>Percent Change from CF (Percent Change from Ambient)</td>
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</tr>
<tr>
<td>Rice ((Oryza sativa)) 20 Asian cultivars OTC Gunma Prefecture, Japan</td>
<td>2008 growing season</td>
<td>Daily avg (ppb): CF (2), 0.8×ambient (23); 1×ambient (28); 1.5×ambient (42); 2×ambient (57) (Cultivar comparisons)</td>
<td>Yield</td>
<td>From n.s. to -30 across all cultivars</td>
<td>Sawada and Kohno (2009)</td>
<td></td>
</tr>
<tr>
<td>Seminatural grass FACE Le Mouret, Switzerland</td>
<td>5 yr</td>
<td>Seasonal AOT40: Ambient (0.1-7.2 ppm-h); Elevated. (1.8-24.1 ppm-h) (N/A)</td>
<td>Relative annual yield</td>
<td>N/A (2×faster decrease in yield/yr)</td>
<td>Volk et al. (2006)</td>
<td></td>
</tr>
<tr>
<td>Soybean OTC; CRA Bari, Italy</td>
<td>2003-2005 growing seasons</td>
<td>Seasonal AOT40 ppm-h: CF (0), Ambient (3.4), High (9.0) (Drought)</td>
<td>Yield</td>
<td>-46 (-9)</td>
<td>Bou Jaoudé et al. (2006b)</td>
<td></td>
</tr>
<tr>
<td>Soybean ((Glycine max \text{ cv. Essex})) Chambers; 21 L Raleigh, NC</td>
<td>2×3 months</td>
<td>12-h avg: CF (28), Elevated (79), Elevated flux (112) (CO₂: 365 &amp; 700)</td>
<td>Seed mass per plant</td>
<td>-30 (N/A)</td>
<td>Booker and Fiscus (2005)</td>
<td></td>
</tr>
<tr>
<td>Soybean ((Glycine max \text{ cv. Essex})) OTCs; 21-L pots Raleigh, NC</td>
<td>2×3 months</td>
<td>12-h avg: CF (18); Elevated (72) (CO₂: 367 &amp; 718)</td>
<td>Seed mass per plant</td>
<td>-34 (N/A)</td>
<td>Booker et al. (2004b)</td>
<td></td>
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<tr>
<td>Soybean ((Glycine max \text{ cv. Tracaja})) Chambers; pots Brazil</td>
<td>20 days</td>
<td>12-h avg: CF &amp; 30 ppb (N/A)</td>
<td>Biomass</td>
<td>-18 (N/A)</td>
<td>Bulbovas et al. (2007)</td>
<td></td>
</tr>
<tr>
<td>Soybean ((Glycine max)) 10 cultivars SoyFACE Urbana, IL</td>
<td>2007 &amp; 2008</td>
<td>8-h avg: Ambient (46.3 &amp; 37.9), Elevated (82.5 &amp; 61.3) (Cultivar comparisons)</td>
<td>Yield</td>
<td>N/A (-17.20)</td>
<td>Betzelberger et al. (2010)</td>
<td></td>
</tr>
<tr>
<td>Spring Wheat ((Triticum aestivum \text{ cv. Minaret; Satu; Drabant; Dragon})) OTCs Belgium, Finland, &amp; Sweden</td>
<td>1990-2006</td>
<td>Seasonal AOT40s ranged from 0 to 16 ppm-h (N/A)</td>
<td>Seed protein content; 1,000-seed weight regressed across all experiments</td>
<td>N/A (significant negative correlation)</td>
<td>Piikki et al. (2008b)</td>
<td></td>
</tr>
<tr>
<td>Strawberry ((Fragaria x ananassa Duch. Cv Korona &amp; Elsanta)) Growth chambers Bonn, Germany</td>
<td>2 months</td>
<td>8-h avg: CF (0 ppb) &amp; Elevated (78 ppb) (N/A)</td>
<td>Fruit yield (weight/plant)</td>
<td>-16 (N/A)</td>
<td>Keutgen et al. (2005)</td>
<td></td>
</tr>
<tr>
<td>Species</td>
<td>Facility</td>
<td>Location</td>
<td>Exposure Duration</td>
<td>O3 Exposure (Additional Treatment)</td>
<td>Response Measured</td>
<td>Percent Change from CF (Percent Change from Ambient)</td>
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</tr>
<tr>
<td>Sugarbeet</td>
<td>OTC</td>
<td>Belgium</td>
<td>2003, 2004; 5 months</td>
<td>8-h avg: Ambient (36 ppb); Elevated (62 ppb) (N/A)</td>
<td>Sugar yield</td>
<td>N/A (-9)</td>
</tr>
<tr>
<td>Sugarcane</td>
<td>CSTR</td>
<td>San Joaquin Valley, CA</td>
<td>2007; 11-13 weeks.</td>
<td>12-h avg: CF (4 ppb); Ambient (58); Elevated (147) (N/A)</td>
<td>Total biomass (g/plant)</td>
<td>-40 (-30)</td>
</tr>
<tr>
<td>Sweet Potato</td>
<td>Growth chambers</td>
<td>Bonn, Germany</td>
<td>4 weeks</td>
<td>8-h avg: CF (0 ppb), Ambient (&lt;40 ppb); Elevated (255 ppb) (N/A)</td>
<td>Tuber weight</td>
<td>-14 (-11.5)</td>
</tr>
<tr>
<td>Tomato</td>
<td>OTC</td>
<td>Valencia, Spain</td>
<td>133 days in 1998</td>
<td>8-h mean ppb: CF 16.3, NF 30.1, NF+ 83.2 (Various cultivars; early &amp; late harvest)</td>
<td>Yield</td>
<td>n.s (n.s.)</td>
</tr>
<tr>
<td>Trifolium</td>
<td>OTC; 2.5-L pots</td>
<td>Madrid, Spain</td>
<td>29 days</td>
<td>12-h avg: CF (&lt;7.9 ± 6.3); Ambient (34.4 ± 10.8); Elevated (56.4 ± 22.3) (N: 5, 15 &amp; 30 kg/ha)</td>
<td>Above-ground biomass</td>
<td>-45 (-35)</td>
</tr>
<tr>
<td>Watermelon</td>
<td>OTC</td>
<td>Valencia, Spain</td>
<td>2000, 2001; 90 days</td>
<td>AOT40: CF (0 ppm-h); Ambient (5.7 ppm-h); Elevated (34.1 ppm-h) (N:0, 14.0 &amp; 29.6 g/pot)</td>
<td>Total fruit yield (kg)</td>
<td>n.s. (54)</td>
</tr>
<tr>
<td>Yellow Nutsedge</td>
<td>OTC; 9-L pots</td>
<td>San Joaquin Valley, CA</td>
<td>8 weeks</td>
<td>12-h avg: 12.8 ± 0.6; 79.9 ± 3.3; 122.7 ± 9.7 (N/A)</td>
<td>Above-ground biomass</td>
<td>n.s. (n.s.)</td>
</tr>
</tbody>
</table>

In studies where variables other than O₃ were included in the experimental design, response to O₃ is only provided for the control level of those variables.
### Table 9-18 Summary of studies of effects of O₃ exposure on growth of natural vegetation.

<table>
<thead>
<tr>
<th>Species</th>
<th>Facility Location</th>
<th><strong>O₃ Exposure (Additional Treatment)</strong></th>
<th><strong>Response Measured</strong></th>
<th><strong>Response</strong></th>
<th><strong>Reference</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>Yellow nutsedge (<em>Cyperus esculentus</em>)</td>
<td>CSTR San Joaquin Valley, CA</td>
<td>12-h mean ppb: CF (4); CF+ (60); CF2+ (115)</td>
<td>Above-ground biomass; tubers (g/plant)</td>
<td>ns; CF(4.1) CF+(3.9) CF2+(2.7)</td>
<td>Grantz et al. (2010a)</td>
</tr>
<tr>
<td>35 herbaceous species</td>
<td>OTC Corvallis, OR</td>
<td>4-yr avg; yearly W126 ppm-h: CF (0), CF+ (21), CF 2+ (49.5)</td>
<td>Total community above-ground biomass (35 species) after 4 years</td>
<td>CF (459 g/m²), CF+ (457 g/m²), CF2+ (398 g/m²)</td>
<td>Pfleeger et al. (2010)</td>
</tr>
<tr>
<td>Highbush blackberry (<em>Rubus argutus</em>)</td>
<td>OTC Auburn, AL</td>
<td>12-h mean ppb: CF (21.7), Ambient (32.3), Elevated (73.3)</td>
<td>Vegetative regrowth after pruning</td>
<td>CF (75.1 g/plant), Ambient (76.4 g/plant), Elevated (73.1 g/plant)</td>
<td>Ditchkoff et al. (2009)</td>
</tr>
<tr>
<td>Horseweed (<em>Conyza canadensis</em>)</td>
<td>CSTR San Joaquin Valley, CA</td>
<td>W126 ppm-h: CF(0), CF+ (11), CF 2+ (30) (Glyphosate resistance)</td>
<td>Total biomass (g/plant)</td>
<td>Glyphosate sensitive: CF (0.354) CF+ (0.197) CF2+ (0.106) Glyphosate resistant: CF(0.510) CF+ (0.313) CF2+ (0.143)</td>
<td>Grantz et al. (2008)</td>
</tr>
<tr>
<td>Red Oak (<em>Quercus rubrum</em>)</td>
<td>Forest sites Look Rock &amp; Twin Creeks Forests, TN</td>
<td>AOT60: 2001 (11.5), 2002 (24.0), 2003 (11.7) (Observational study with multiple environmental variables)</td>
<td>Annual circumference increment (change relative to 2001 in year 2002;2003)</td>
<td>-42.8%; +1%</td>
<td>McLaughlin et al. (2007a)</td>
</tr>
<tr>
<td>Pine species</td>
<td>Forest sites Look Rock Forest, TN</td>
<td>AOT60: 2001 (11.5), 2002 (24.0), 2003 (11.7) (Observational study with multiple environmental variables)</td>
<td>Annual circumference increment (change relative to 2001 in year 2002;2003)</td>
<td>-62.5%; -2.9%</td>
<td>McLaughlin et al. (2007a)</td>
</tr>
<tr>
<td>Species</td>
<td>Facility Location</td>
<td>Exposure Duration</td>
<td>O₃ Exposure (Additional Treatment)</td>
<td>Response Measured</td>
<td>Response</td>
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<tr>
<td>Black Cherry (<em>Prunus rigida</em>)</td>
<td>Twin Creeks Forest, TN</td>
<td>2002-2003, April-October</td>
<td>AOT60: 2002 (24.0), 2003 (11.7)</td>
<td>Annual circumference increment (change relative to 2003 in year 2002)</td>
<td>-75%</td>
</tr>
<tr>
<td>Hemlock (<em>Tsuga canadensis</em>)</td>
<td>Twin Creeks Forest, TN</td>
<td>2002-2003, April-October</td>
<td>AOT60: 2002 (24.0), 2003 (11.7)</td>
<td>Annual circumference increment (change relative to 2003 in year 2002)</td>
<td>-21.9%</td>
</tr>
<tr>
<td>Red Maple (<em>Acer rubrum</em>)</td>
<td>Twin Creeks Forest, TN</td>
<td>2002-2003, April-October</td>
<td>AOT60: 2002 (24.0), 2003 (11.7)</td>
<td>Annual circumference increment (change relative to 2003 in year 2002)</td>
<td>-59.6%</td>
</tr>
<tr>
<td>Species</td>
<td>Facility Location</td>
<td>Exposure Duration</td>
<td>O$_3$ Exposure (Additional Treatment)</td>
<td>Response Measured</td>
<td>Response</td>
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</tr>
<tr>
<td>Sugar Maple (<em>Acer saccharum</em>)</td>
<td>Forest sites Twin Creeks Forest, TN</td>
<td>2002-2003, April-October</td>
<td>AOT60: 2002 (24.0), 2003 (11.7) (Observational study with multiple environmental variables)</td>
<td>Annual circumference increment (change relative to 2003 in year 2002)</td>
<td>-63.8%</td>
</tr>
<tr>
<td>Trembling aspen (<em>Populus tremuloides</em>), 5 genotypes</td>
<td>Aspen FACE Rhinelander, WI</td>
<td>1998-2004, May-October</td>
<td>Cumulative avg 90-day 12-h W126. Ambient 3.1 ppm-h Elevated: 27.2 ppm-h (Competition with birch, maple)</td>
<td>main stem volume after 7 years</td>
<td>Ambient: 6.22 dm$^3$; Elevated: 4.73 dm$^3$</td>
</tr>
<tr>
<td>Hybrid Poplar (<em>Populus trichocarpa</em> x <em>Populus deltoides</em>)</td>
<td>OTC Seattle, WA</td>
<td>2003, 3 months</td>
<td>Daily mean (μg/g): CF(&lt;9), Elevated (85-128)</td>
<td>Total biomass</td>
<td>CF to elevated: -12.9%</td>
</tr>
</tbody>
</table>

In studies where variables other than O$_3$ were included in the experimental design, response to O$_3$ is only provided for the control level of those variables.

### 9.7 Summary and Conclusions

Based on the evidence presented in Chapter 9 and summarized here, O$_3$ is causally related or likely to be causally related to effects observed on vegetation and ecosystems. The evidence for these effects spans the entire continuum of biological organization, from the cellular and subcellular level to the whole plant, and up to ecosystem-level processes, and includes evidence for effects at lower levels of organization, leading to effects at higher levels. Given the current state of knowledge, exposure indices that cumulate and differentially weight the higher hourly average concentrations and also include the mid-level values are the most appropriate for use in developing response functions and comparing studies. The framework for causal determinations (see Preamble) has been applied to the body of scientific evidence to examine effects attributed to O$_3$ exposure collectively and the determinations are presented in Table 9-19.
### Table 9-19  Summary of O₃ causal determinations for vegetation and ecosystem effects.

<table>
<thead>
<tr>
<th>Vegetation and Ecosystem Effects</th>
<th>Conclusions from 2006 O₃ AQCD</th>
<th>Conclusions from this ISA</th>
</tr>
</thead>
<tbody>
<tr>
<td>Visible Foliar Injury Effects on Vegetation</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause impaired aesthetic quality of many native plants and trees by increasing foliar injury.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Vegetation Growth</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause decreased growth and biomass accumulation in annual, perennial and woody plants, including agronomic crops, annuals, shrubs, grasses, and trees.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Productivity in Terrestrial Ecosystems</td>
<td>There is evidence that O₃ is an important stressor of ecosystems and that the effects of O₃ on individual plants and processes are scaled up through the ecosystem, affecting net primary productivity.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Reduced Carbon (C) Sequestration in Terrestrial Ecosystems</td>
<td>Limited studies in the 2006 O₃ AQCD.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Reduced Yield and Quality of Agricultural Crops</td>
<td>Data published since the 1996 O₃ AQCD strengthen previous conclusions that there is strong evidence that current ambient O₃ concentrations cause decreased yield and/or nutritive quality in a large number of agronomic and forage crops.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Ecosystem Water Cycling</td>
<td>Ecosystem water quantity may be affected by O₃ exposure at the landscape level.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Below-ground Biogeochemical Cycles</td>
<td>Ozone-sensitive species have well known responses to O₃ exposure, including altered C allocation to below-ground tissues; and also altered rates of leaf and root production, turnover, and decomposition. These shifts can affect overall C and N loss from the ecosystem in terms of respired C, and leached aqueous dissolved organic and inorganic C and N.</td>
<td>Causal Relationship</td>
</tr>
<tr>
<td>Alteration of Terrestrial Community Composition</td>
<td>Ozone may be affecting above- and below-ground community composition through impacts on both growth and reproduction. Significant changes in plant community composition resulting directly from O₃ exposure have been demonstrated.</td>
<td>Likely to be a Causal Relationship</td>
</tr>
</tbody>
</table>
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10 THE ROLE OF TROPOSPHERIC OZONE IN CLIMATE CHANGE AND UV-B SHIELDING EFFECTS

10.1 Introduction

Atmospheric O₃ plays an important role in the Earth’s energy budget by interacting with incoming solar radiation and outgoing infrared radiation. Over mid-latitudes, approximately 90% of the total atmospheric O₃ column is located in the stratosphere (Kar et al., 2010; Crist et al., 1994). Therefore, tropospheric O₃ makes up a relatively small portion (~10%) of the total column of O₃ over mid-latitudes, but it does play an important role in the overall radiation budget. The next section (Section 10.2) briefly describes the physics of the earth’s radiation budget, providing background material for the subsequent two sections assessing how perturbations in tropospheric O₃ concentrations might affect (1) climate through its role as a greenhouse gas (Section 10.3), and (2) health, ecology and welfare through its role in shielding the earth’s surface from solar ultraviolet radiation (Section 10.4). The concluding section in this chapter (Section 10.5) includes a summary of effects assessed in this chapter along with their associated causal determinations.

10.2 Physics of the Earth’s Radiation Budget

Radiant energy from the sun enters the atmosphere in a range of wavelengths, but peaks strongly in the visible (400-750 nm) part of the spectrum. Longer wavelength infrared (750 nm-1 mm) and shorter wavelength ultraviolet (100-400 nm) radiation are also present in the solar electromagnetic spectrum. Since the energy possessed by a photon is inversely proportional to its wavelength, infrared (IR) radiation carries the least energy per photon, and ultraviolet (UV) radiation carries the most energy per photon. Ultraviolet radiation is further subdivided into classes (bands) based on wavelength: UV-A refers to wavelengths from 400-315 nm; UV-B from 315-280 nm; and UV-C from 280-100 nm. Within the UV spectrum, UV-A radiation is the least energetic band and UV-C is the most energetic band.

The wavelength of radiation also determines how the photons interact with the complex mixture of gases, clouds, and particles present in the atmosphere (see Figure 10-1). UV-A radiation can be scattered but is not absorbed to any meaningful degree by atmospheric gases including O₃. UV-B radiation is absorbed and scattered in part within the atmosphere. UV-C is almost entirely blocked by the Earth’s upper atmosphere, where it participates in photoionization and photodissociation processes including absorption by stratospheric O₃. Since UV-A radiation is less energetic and does not interact with O₃ in the troposphere or the stratosphere and UV-C radiation is almost entirely blocked by stratospheric O₃, UV-B radiation is the most important band to consider in relation to tropospheric O₃ shielding.
Tropospheric O$_3$ plays a “disproportionate” role in absorbing UV-B radiation compared with stratospheric O$_3$ on a molecule per molecule basis (Balis et al., 2002; Zerefos et al., 2002; Crist et al., 1994; Bruhl and Crutzen, 1989). This effect results from the higher atmospheric pressure present in the troposphere, resulting in higher concentrations of gas molecules present that can absorb or scatter radiation. For this reason, the troposphere is referred to as a “multiple scattering” regime for UV absorption, compared to the “single scattering” regime in the stratosphere. Thus, careful quantification of atmospheric absorbers and scatterers, along with a well-resolved description of the physics of these interactions, is necessary for predicting the effects of tropospheric O$_3$ on UV-B flux at the surface.

Solar flux at all wavelengths has a temporal dependence, while radiative scattering and absorption have strong wavelength, path length, and gas/particle concentration dependencies. These combine to create nonlinear effects on UV flux at the Earth’s surface. Chapter 10 of the 2006 O$_3$ AQCD (U.S. EPA, 2006b) describes in detail several key factors that influence the spatiotemporal distribution of ground-level UV radiation flux, including: (1) long-term solar activity including sunspot cycle; (2) solar rotation; (3) the position of the Earth in its orbit around the sun; (4) atmospheric absorption and scattering of UV radiation by gas molecules and aerosol particles; (5) absorption and scattering by stratospheric and tropospheric clouds; and (6) surface albedo. The efficiencies of absorption and scattering are highly dependent on the concentration of the scattering medium, particle size (for aerosols and clouds), and the altitude at which these processes are occurring. These properties are sensitive to meteorology, which introduces additional elements of spatial and temporal dependency in ground-level UV radiation flux.
Figure 10-1  Diagram of the factors that determine human exposure to ultraviolet radiation.

About 30% of incoming solar radiation is directly reflected back to space, mainly by clouds or surfaces with high albedo (reflectivity), such as snow, ice, and desert sand. Radiation that does penetrate to the Earth’s surface and is absorbed can be re-emitted in the longwave (infrared) portion of the spectrum; the rest goes into evaporating water or soil moisture or emerges as sensible heat. The troposphere is opaque to the outgoing longwave radiation. Polyatomic gases such as water vapor, CO₂, CH₄, and O₃ absorb and re-emit the radiation upwelling from the Earth’s surface, reducing the efficiency with which that energy returns to space. In effect, these gases act as a blanket warming the Earth’s surface. This phenomenon, known as the “Greenhouse Effect,” was first quantified in the 19th century (Arrhenius, 1896), and gives rise to the term “greenhouse gas.” The most important greenhouse gas is water vapor.
10.3 Effects of Tropospheric O₃ on Climate

10.3.1 Background

As a result of its interaction with incoming solar radiation and outgoing longwave radiation, tropospheric O₃ plays a major role in determining climate, and increases in its abundance may contribute to climate change (IPCC, 2007c). Models estimate that the global average concentration of O₃ in the troposphere has increased 30-70% since the pre-industrial era (Gauss et al., 2006), while observations indicate that in some regions tropospheric O₃ concentrations may have increased by factors as great as 4 or 5 (Marenco et al., 1994; Staehelin et al., 1994). These increases are tied to the rise in emissions of O₃ precursors from human activity, mainly fossil fuel consumption and agricultural processes.

The effect on climate of the tropospheric O₃ concentration change since pre-industrial times has been estimated to be about 25-40% of the anthropogenic CO₂ effect and about 75% of the anthropogenic CH₄ effect (IPCC, 2007c), ranking it third in importance behind these two major greenhouse gases. In the 21st century, as the Earth’s population continues to grow and energy technology spreads to developing countries, a further rise in the global concentration of tropospheric O₃ is likely, with associated consequences for human health and ecosystems relating to climate change.

To examine the science of a changing climate and to provide balanced and rigorous information to policy makers, the World Meteorological Organization (WMO) and the United Nations Environment Programme (UNEP) formed the Intergovernmental Panel on Climate Change (IPCC) in 1988. The IPCC supports the work of the Conference of Parties (COP) to the United Nations Framework Convention on Climate Change (UNFCCC). The IPCC periodically brings together climate scientists from member countries of WMO and the United Nations to review knowledge of the physical climate system, past and future climate change, and evidence of human-induced climate change. IPCC climate assessment reports are issued every five to seven years.

This section draws in part on the fourth IPCC Assessment Report (AR4) (IPCC, 2007c), as well as other peer-reviewed published research. Section 10.3.2 reviews evidence of climate change in the recent past and projections of future climate change. It also offers a brief comparison of tropospheric O₃ relative to other greenhouse gases. Section 10.3.3 describes factors that influence the magnitude of tropospheric O₃ effects on climate. Section 10.3.4 considers the competing effects of O₃ precursors on climate. Finally, Section 10.3.5 and Section 10.3.6 describe the effects of changing tropospheric O₃ concentrations on past and future climate. Downstream effects resulting from climate change, such as ecosystem responses, are outside the scope of this assessment, which focuses rather on the effects of changes in tropospheric O₃ concentrations on radiative forcing and climate.
10.3.2 Climate Change Evidence and the Influence of Tropospheric O₃

10.3.2.1 Climate Change in the Recent Past

From the end of the last ice age 12,000 years ago until the mid-1800s, observations from ice cores show that concentrations of the long-lived greenhouse gases CO₂, CH₄, and N₂O have been relatively stable. Unlike these greenhouse gases, O₃ is not preserved in ice, and no record of it before the late 1800s exists. Models, however, suggest that it, too, has remained relatively constant during this time period (Thompson et al., 1993; Thompson, 1992). The stable mix of these greenhouse gases in the atmosphere, together with water vapor, has kept the global mean temperature of the Earth close to 15°C. Without the presence of greenhouse gases in the atmosphere, the Earth’s global mean temperature would be about 30°C cooler, or -15°C.

Since the start of the Industrial Revolution, human activity has led to observable increases of greenhouse gases in the atmosphere, mainly through fossil fuel combustion. According to the IPCC AR4 (IPCC, 2007c), there is now “very high confidence” that the net effect of anthropogenic greenhouse gas emissions since 1750 has led to warming, and it is “very likely” that human activity contributed to the 0.76°C rise in global mean temperature observed over the last century. The increase of tropospheric O₃ abundance may have contributed 0.1-0.3°C warming to the global climate during this time period (Hansen et al., 2005; Mickley et al., 2004). Global cooling due to anthropogenic aerosols (IPCC, 2007c) has likely masked the full warming effect of the anthropogenic greenhouse gases on a global scale.

10.3.2.2 Projections of Future Climate Change

The IPCC AR4 projects a warming of ~0.2°C per decade for the remainder of the 21st century (IPCC, 2007c). Even at constant concentrations of greenhouse gases in the atmosphere, temperatures are expected to increase by about 0.1°C per decade, due to the slow response of oceans to the warming applied so far. It is likely that the Earth will experience longer and more frequent heat waves in the 21st century, together with more frequent droughts and/or heavy precipitation events in some regions, due to perturbations in the hydrological cycle that result from changing temperatures. Sea levels could increase by 0.3-0.8 meters by 2300 due to thermal expansion of the oceans. The extent of Arctic sea ice is expected to decline, and contraction of the Greenland ice sheet could further contribute to the sea level rise (IPCC, 2007c).

Projections of future climate change are all associated with some degree of uncertainty. A major uncertainty involves future trends in the anthropogenic emissions of greenhouse gases or their precursors. For the IPCC AR4 climate projections, a set of distinct “storylines” or emission pathways was developed (IPCC,
Each storyline took into account factors such as population growth, mix of energy technologies, and the sharing of technology between developed and developing nations, and each resulted in a different scenario for anthropogenic emissions. When these trends in emissions are applied to models, these scenarios yield a broad range of possible climate trajectories for the 21st century.

A second factor bringing large uncertainty to model projections of future climate is the representation of climate and, especially, climate feedbacks. A rise in surface temperatures would perturb a suite of other processes in the earth-atmosphere-ocean system, which may in turn either amplify the temperature increase (positive feedback) or diminish it (negative feedback). One important feedback involves the increase of water vapor content of the atmosphere that would accompany higher temperatures (Bony et al., 2006). Water vapor is a potent greenhouse gas; accounting for the water vapor feedback may increase the climate sensitivity to a doubling of CO₂ by nearly a factor of two (Held and Soden, 2000). The ice-albedo feedback is also strongly positive; a decline in snow cover and sea ice extent would diminish the Earth’s albedo, allowing more solar energy to be retained at the surface (Holland and Bitz, 2003; Rind et al., 1995). A final example of a climate feedback involves the effects of changing cloud cover in a warming atmosphere. Models disagree on the magnitude and even the sign of the cloud cover feedback on surface temperatures (Soden and Held, 2006).

10.3.2.3 Metrics of Potential Climate Change

Two metrics frequently used to estimate the potential climate effect of some perturbation such as a change in greenhouse gas concentration are: (1) radiative forcing; and (2) global warming potential (GWP). These metrics differ in a fundamental way as described below.

Radiative forcing is a change in the radiative balance at a particular level of the atmosphere or at the surface when a perturbation is introduced in the earth-atmosphere-ocean system. In the global mean, radiative forcing of greenhouse gases at the tropopause (top of the troposphere) is roughly proportional to the surface temperature response (Hansen et al., 2005; NRC, 2005). It thus provides a useful metric for policymakers for assessing the response of the earth’s surface temperature to a given change in the concentration of a greenhouse gas. Positive values of radiative forcing indicate warming in a test case relative to the control; negative values indicate cooling. The units of radiative forcing are energy flux per area, or W/m².

Radiative forcing requires just a few model years to calculate, and it shows consistency from model to model. However, radiative forcing does not take into account the climate feedbacks that could amplify or dampen the actual surface temperature response, depending on region. Quantifying the change in surface temperature requires a climate simulation in which all important feedbacks are accounted for. As some of these processes are not well understood, the surface
temperature response to a given radiative forcing can be highly uncertain and can vary greatly among models and even from region to region within the same model.

GWP indicates the integrated radiative forcing over a specified period (usually 100 years) from a unit mass pulse emission of a greenhouse gas or its precursor, and is reported as the magnitude of this radiative forcing relative to that of CO₂. GWP is most useful for comparing the potential climate effects of long-lived gases, such as N₂O or CH₄. Since tropospheric O₃ has a lifetime on the order of weeks to months, GWP is not seen as a valuable metric for quantifying the importance of O₃ on climate (Forster et al., 2007). Thus, this assessment focuses on radiative forcing as the metric of climate influence resulting from changes in tropospheric O₃ concentrations.

### 10.3.2.4 Tropospheric O₃ as a Greenhouse Gas

Tropospheric O₃ differs in important ways from other greenhouse gases. It is not emitted directly, but is produced through photochemical oxidation of CO, CH₄, and nonmethane volatile organic compounds (VOCs) in the presence of nitrogen oxide radicals (NOₓ = NO + NO₂; see Chapter 3, Section 3.2 for further details on the chemistry of O₃ formation). It is also supplied by vertical transport from the stratosphere. The lifetime of O₃ in the troposphere is typically a few weeks, resulting in an inhomogeneous distribution that varies seasonally; the distribution of the long-lived greenhouse gases like CO₂ and CH₄ are much more uniform. The longwave radiative forcing by O₃ is mainly due to absorption in the 9.6 μm window, where absorption by water vapor is weak. It is therefore less sensitive to local humidity than the radiative forcing by CO₂ or CH₄, for which there is much more overlap with the water absorption bands (Lenoble, 1993). And unlike other major greenhouse gases, O₃ absorbs in the shortwave as well as the longwave part of the spectrum.

Figure 10-2 shows the main steps involved in the influence of tropospheric O₃ on climate. Emissions of O₃ precursors including CO, VOCs, CH₄, and NOₓ lead to production of tropospheric O₃. A change in the abundance of tropospheric O₃ perturbs the radiative balance of the atmosphere, an effect quantified by the radiative forcing metric. The earth-atmosphere-ocean system responds to the radiative forcing with a climate response, typically expressed as a change in surface temperature. Finally, the climate response causes downstream climate-related health and ecosystem effects, such as redistribution of diseases or ecosystem characteristics due to temperature changes. Feedbacks from both the climate response and downstream effects can, in turn, affect the abundance of tropospheric O₃ and O₃ precursors through multiple mechanisms. Direct feedbacks are discussed further in Section 10.3.3.4; the downstream climate effects and their long-term feedbacks are extremely complex and outside the scope of this assessment.
Note: This figure includes the relationship between precursor emissions, changes in tropospheric O$_3$ abundance, radiative forcing, climate response, and climate effects. Units shown are those typical for each quantity illustrated. Feedbacks from both the climate response and climate effects can, in turn, affect the abundance of tropospheric O$_3$ and O$_3$ precursors through multiple feedback mechanisms. Climate effects and their feedbacks are deemphasized in the figure since these downstream effects are extremely complex and outside the scope of this assessment.

**Figure 10-2  Schematic illustrating the effects of tropospheric O$_3$ on climate.**

The IPCC (2007c) reported a radiative forcing of 0.35 W/m$^2$ for the change in tropospheric O$_3$ abundance since the pre-industrial era, ranking it third in importance after the greenhouse gases CO$_2$ (1.66 W/m$^2$) and CH$_4$ (0.48 W/m$^2$). Figure 10-3 shows the global average radiative forcing estimates and uncertainty ranges in 2005 for anthropogenic CO$_2$, CH$_4$, O$_3$ and other important agents and mechanisms. The error bars encompassing the tropospheric O$_3$ radiative forcing estimate in the figure range from 0.25 to 0.65 W/m$^2$, making it relatively more uncertain than the long-lived greenhouse gases.
Figure 10-3  Global average radiative forcing (RF) estimates and uncertainty ranges in 2005 for anthropogenic CO₂, CH₄, O₃, and other important agents and mechanisms.
10.3.3 Factors that Influence the Effect of Tropospheric $O_3$ on Climate

This section describes the main factors that influence the magnitude of the climate response to changes in tropospheric $O_3$ abundance. They include: (1) trends in the concentration of tropospheric $O_3$; (2) the effect of surface albedo on $O_3$ radiative forcing; (3) the effect of vertical distribution on $O_3$ radiative forcing; (4) feedback factors that can alter the climate response to $O_3$ radiative forcing; and (5) the indirect effects of tropospheric $O_3$ on the carbon cycle. Trends in stratospheric $O_3$ abundance may also affect temperatures at the Earth’s surface, but aside from issues relating to stratospheric-tropospheric exchange discussed in Chapter 3, Section 3.4.1.1, stratospheric $O_3$ assessment is beyond the scope of this document.

10.3.3.1 Trends in the Concentration of Tropospheric $O_3$

To first order, the effect of tropospheric $O_3$ on global surface temperature is proportional to the change in tropospheric $O_3$ concentration. The earth’s surface temperatures are most sensitive to $O_3$ abundance perturbations in the mid to upper troposphere. This section therefore focuses mainly on observed $O_3$ concentration trends in the free troposphere or in regions far from $O_3$ sources, where a change in $O_3$ concentrations may indicate change throughout the troposphere. Data from ozonesondes, mountaintops, and remote surface sites are discussed, as well as satellite data.

Observed Trends in $O_3$ Concentrations since the Pre-Industrial Era

Measurements of $O_3$ concentrations at two European mountain sites dating from the late 1800s to early 1900s show values at about 10 ppb, about one-fifth the values observed today at similar sites (Pavelin et al., 1999; Marenco et al., 1994). The accuracy of these early measurements is questionable however, in part because they exhibit $O_3$ concentrations equivalent to or only a couple of parts per billion greater than those observed at nearby low-altitude sites during the same time period (Mickley et al., 2001; Volz and Kley, 1988). A larger vertical gradient in tropospheric $O_3$ concentration would be expected because of its stratospheric source and its longer lifetime aloft. In another study, Staehelin et al. (1994) revisited observations made in the Swiss mountains during the 1950s and found a doubling in $O_3$ concentrations from that era to 1989-1991.

Routine observations of $O_3$ in the troposphere began in the 1970s with the use of balloon-borne ozonesondes, but even this record is sparse. Trends from ozonesondes have been highly variable and dependent on region (Logan et al., 1999). Over most sites in the U.S., ozonesondes reveal little trend. Over Canada, observations show a decline in $O_3$ concentrations between 1980 and 1990, then a rebound in the following decade (Tarasick et al., 2005). Ozonesondes over Europe give a mixed picture. European ozonesondes showed increases in the 1970s and 1980s, with smaller
increases or even declines since then (Oltmans et al., 2006; Logan et al., 1999). Over Japan, O₃ concentrations in the lower troposphere increased about 0.2-0.4 ppb/year during the 1990s (Naja and Akimoto, 2004).

Ground-based measurements in remote regions provide a record of tropospheric O₃ concentrations, but like ozonesonde data, such measurements are sparse before the 1970s. Springtime O₃ observations from several mountain sites in the western U.S. show a positive trend of about 0.5-0.7 ppb/year since the 1980s (Cooper et al., 2010; Jaffe et al., 2003). Ship-borne O₃ measurements for the time period 1977 to 2002 indicate increases of 0.1-0.7 ppb/year over much of the Atlantic south of 40ºN, but no appreciable change north of 40ºN (Lelieveld et al., 2004). The lack of trend for the North Atlantic would seem at odds with O₃ observations at Mace Head (53ºN) on the west coast of Ireland, which show a significant positive trend of about 0.5 ppb/year from 1987 to 2003 (Simmonds et al., 2004). Over Japan, O₃ concentrations at a remote mountain site have increased 1 ppb/year from 1998 to 2003 (Tanimoto, 2009), a rate more than double that recorded by ozonesondes in the lower troposphere over Japan during the 1990s (Naja and Akimoto, 2004).

At Zugspitze, a mountain site in Germany, O₃ concentrations increased by 12% per decade during the 1970s and 1980s, consistent with European ozonesondes (Oltmans et al., 2006). Since then, O₃ concentrations continue to increase at Zugspitze, but more slowly. What little data exist for the Southern Hemisphere point to measurable increases in tropospheric O₃ concentrations in recent decades, as much as ~15% at Cape Grim in the 1989-2004 time period (Oltmans et al., 2006).

The satellite record is now approaching a length that can be useful for diagnosing trends in the total tropospheric O₃ column (details on the use of satellites to measure tropospheric O₃ concentrations are covered in Chapter 3, Section 3.5.5.5). In contrast to the surface data from ships, tropospheric O₃ columns from the Total Ozone Mapping Spectrometer (TOMS) show no trend over the tropical Atlantic for the period 1980-1990 (Thompson and Hudson, 1999). Over the Pacific Ocean, a longer, 25 year record of TOMS data again reveals no trend over the tropics, but shows increases in tropospheric column O₃ of about 2-3 Dobson Units (DU)¹ at mid-latitudes in both hemispheres (Ziemke et al., 2005).

Interpreting these recent trends in tropospheric O₃ concentrations is challenging. The first difficulty is reconciling apparently contradictory trends in the observations, e.g., over tropical oceans. A second difficulty is that the O₃ concentration trends depend on several factors, not all of which can be well characterized. These factors include (1) trends in emissions of O₃ precursors, (2) variation in the stratospheric source of O₃, (3) changes in solar radiation resulting from stratospheric O₃ depletion, and (4) trends in tropospheric temperatures (Fusco and Logan, 2003). Recent positive trends in the western U.S. and over Japan are consistent with the rapid increase in emissions of O₃ precursors from mainland Asia and transport of pollution across the

¹ The Dobson Unit is a typical unit of measure for the total O₃ in a vertical column above the Earth's surface. One DU is equivalent to the amount of O₃ that would exist in a 1 μm (10⁻⁶ meter) thick layer of pure O₃ at standard temperature (0ºC) and pressure (1 atm), and corresponds to a column of O₃ containing 2.69 × 10²⁰ molecules/m². A typical value for the amount of O₃ in a column of the Earth's atmosphere, although highly variable, is 300 DU and approximately 10% (30 DU) of that exists in the troposphere at mid latitudes.
Pacific Ocean (Cooper et al., 2010; Tanimoto, 2009). The satellite trends over the northern mid-latitudes are consistent with this picture as well (Ziemke et al., 2005). Increases in tropospheric O$_3$ concentrations in the Southern Hemisphere are also likely due to increased anthropogenic NO$_X$ emissions, especially from biomass burning (Fishman et al., 1991). Recent declines in summertime O$_3$ concentrations over Europe can be partly explained by decreases in O$_3$ precursor emissions there (Jonson et al., 2005), while springtime increases at some European sites are likely linked to changes in stratospheric dynamics (Ordonez et al., 2007). Over Canada, Fusco and Logan (2003) found that O$_3$ depletion in the lowermost stratosphere may have reduced the stratospheric flux of O$_3$ into the troposphere by as much as 30% from the early 1970s to the mid 1990s, consistent with the trends in ozonesondes there.

**Calculation of O$_3$ Concentration Trends for the Recent Past**

Simulations of trends in tropospheric O$_3$ concentrations provide a means for testing current knowledge of O$_3$ processes and predicting with greater confidence trends in future O$_3$ concentrations. Time-dependent emission inventories of O$_3$ precursors have also been developed for 1850-2000 (Lamarque et al., 2010) and for 1890-1990 (Van Aardenne et al., 2001). These inventories allow for the calculation of changing O$_3$ concentration over time.

One recent multi-model study calculated an increase in the O$_3$ concentration since pre-industrial times of 8-14 DU, or about 30-70% (Gauss et al., 2006). The large spread in modeled estimates reveals the limitations in knowledge of processes in the pristine atmosphere. Models typically overestimate the late nineteenth and early twentieth century observations available in surface air and at mountain sites by 50-100% (Lamarque et al., 2005; Shindell et al., 2003; Mickley et al., 2001; Kiehl et al., 1999). Reconciling the differences between models and measurements will require more accurate simulation of the natural sources of O$_3$ (Mickley et al., 2001) and/or implementation of novel sinks such as bromine radicals, which may reduce background O$_3$ concentrations in the pristine atmosphere by as much as 30% (Yang et al., 2005c).

For the more recent past (since 1970), application of time-dependent emissions reveals an equatorward shift in the distribution of tropospheric O$_3$ in the Northern Hemisphere due to the industrialization of societies at low-latitudes (Lamarque et al., 2005; Berntsen et al., 2000). By constraining a model with historical (1950s-2000) observations, Shindell and Faluvegi (2002) calculated a large increase of 8.2 DU in tropospheric O$_3$ abundance over polluted continental regions since 1950. This trend is not captured in standard chemistry models, but is consistent with the change in tropospheric O$_3$ concentrations since pre-industrial times implied by the observations from the late 1800s (Pavelin et al., 1999; Marenco et al., 1994).
10.3.3.2 The Effect of Surface Albedo on O₃ Radiative Forcing

The Earth’s surface albedo plays a role in O₃ radiative forcing. Through most of the troposphere, absorption of incoming shortwave solar radiation by O₃ is small relative to its absorption of outgoing longwave terrestrial radiation. However, over surfaces characterized by high albedo (e.g., over snow, ice, or desert sand), incoming radiation is more likely to be reflected than over darker surfaces, and the probability that O₃ will absorb shortwave solar radiation is therefore larger. In other words, energy that would otherwise return to space may instead be retained in the atmosphere. Several studies have shown that transport of O₃ to the Arctic from mid-latitudes leads to radiative forcing estimates greater than 1.0 W/m² in the region, especially in summer (Shindell et al., 2006; Liao et al., 2004b; Mickley et al., 1999). Both the high surface albedo of the Arctic and the large solar zenith angles there (which increase the path length of incoming sunlight) lead to strong shortwave radiative forcing in the region. Because the Arctic is especially sensitive to radiative forcing through the ice-albedo feedback, the large contribution in the shortwave solar spectrum to the total radiative forcing in the region may be important.

10.3.3.3 The Effect of Vertical Distribution on O₃ Radiative Forcing

In the absence of feedbacks, O₃ increments near the tropopause produce the largest increases in surface temperature (Lacis et al., 1990; Wang et al., 1980). This is a result of the colder temperature of the tropopause relative to the rest of the troposphere and stratosphere. Since radiation emitted by the atmosphere is approximately proportional to the fourth power of its temperature\(^1\), the colder the added O₃ is relative to the earth’s surface, the weaker the radiation emitted and the greater the “trapping” of longwave radiation in the troposphere.

10.3.3.4 Feedback Factors that Alter the Climate Response to Changes in O₃ Radiative Forcing

Estimates of radiative forcing provide a first-order assessment of the effect of tropospheric O₃ on climate. In the atmosphere, climate feedbacks and transport of heat alter the sensitivity of Earth’s surface temperature to addition of tropospheric O₃. Assessment of the full climate response to increases in tropospheric O₃ concentrations requires use of a climate model to simulate these interactions.

Due to its short lifetime, O₃ is heterogeneously distributed through the troposphere. Sharp horizontal gradients exist in the radiative forcing of O₃, with the greatest radiative forcing since pre-industrial times occurring over the northern mid-latitudes (more on this in Section 10.3.5 and Section 10.3.6). If climate feedbacks are

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\(^1\) As described by the Stefan-Boltzmann law, an ideal blackbody—which the atmosphere approximates—absorbs at all wavelengths and re-radiates proportional to the fourth power of its temperature.
particularly powerful, they may obscure or even erase the correlation between regional radiative forcing and climate response (Harvey, 2004; Boer and Yu, 2003). The transport of heat through the atmosphere, though not technically a feedback, may also weaken the correlation between radiative forcing and climate response. Several model studies have reported that the horizontal pattern of surface temperature response from 2000-2100 trends in predicted short-lived species (including O\textsubscript{3}) closely matches the pattern from the trends in the long-lived greenhouse gases over the same time period (Levy et al., 2008; Shindell et al., 2008; Shindell et al., 2007). This correspondence occurs even though the patterns of radiative forcing for the short-lived and long-lived species differ substantially. In a separate paper, Shindell et al. (2007) found that Arctic temperatures are especially sensitive to the mid-latitude radiative forcing from tropospheric O\textsubscript{3}.

Other studies have found that the signature of warming due to tropospheric O\textsubscript{3} does show some consistency with the O\textsubscript{3} radiative forcing. For example, Mickley et al. (2004) examined the change in O\textsubscript{3} concentrations since pre-industrial times and found greater warming in the Northern Hemisphere than in the Southern Hemisphere (+0.4\degree C versus +0.2\degree C), as well as higher surface temperatures downwind of Europe and Asia and over the North American interior in summer. For an array of short-lived species including O\textsubscript{3}, Shindell and Faluvegi (2009) found that radiative forcing applied over northern mid-latitudes yield more localized responses due to local cloud, water vapor, and albedo feedbacks than radiative forcing applied over the tropics.

Climate feedbacks can also alter the sensitivity of surface temperature to the vertical distribution of tropospheric O\textsubscript{3}. The previous section (Section 10.3.3.3) described the greater effect of O\textsubscript{3} added to the upper troposphere (near the tropopause) on radiative forcing, relative to additions in the mid- to lower troposphere. However, warming induced by increased O\textsubscript{3} concentrations in the upper troposphere could stabilize the atmosphere to some extent, limiting the transport of heat to the Earth’s surface and mitigating the effect of the added O\textsubscript{3} on surface temperature (Joshi et al., 2003; Christiansen, 1999). Hansen et al. (1997) determined that allowing cloud feedbacks in a climate model meant that O\textsubscript{3} enhancements in the mid-troposphere had the greatest effect on surface temperature.

Finally, climate feedbacks can amplify or diminish the climate response of one greenhouse gas relative to another. For example, Mickley et al. (2004) found a greater temperature response to CO\textsubscript{2} radiative forcing than to an O\textsubscript{3} radiative forcing of similar global mean magnitude, due in part to the relatively weak ice-albedo feedback for O\textsubscript{3} radiative forcing. Since CO\textsubscript{2} absorbs in the same bands as water vapor, CO\textsubscript{2} radiative forcing saturates in the middle troposphere and is also shifted toward the drier poles. A poleward shift in radiative forcing amplifies the ice-albedo feedback in the case of CO\textsubscript{2}, and the greater mid-troposphere radiative forcing allows for greater surface temperature response, relative to that for O\textsubscript{3}.
10.3.3.5 Indirect Effects of Tropospheric O$_3$ on the Carbon Cycle

A proposed indirect effect of tropospheric O$_3$ on climate involves the carbon cycle. By directly damaging plant life in ways discussed in Chapter 9, increases in tropospheric O$_3$ concentrations may depress the land-carbon sink of CO$_2$, leading to accumulation of CO$_2$ in the atmosphere and ultimately warming of the Earth’s surface. Sitch et al. (2007) calculated that this indirect warming effect of O$_3$ on climate has about the same magnitude as the O$_3$ direct effect. Their results suggest a doubled sensitivity of surface temperatures to O$_3$ radiative forcing, compared to current model estimates.

A large array of additional indirect effects involving biospheric responses to tropospheric O$_3$ concentrations are possible. For example, increasing temperature due to increases in tropospheric O$_3$ concentration may alter biodiversity, species migration, and consequent impacts on surface albedo. Such long-term feedbacks may play an important role in the eventual climate response to changes in tropospheric O$_3$ abundance, but a full evaluation of such long-term feedbacks on climate change is outside the scope of this assessment.

10.3.4 Competing Effects of O$_3$ Precursors on Climate

Changes in concentrations of O$_3$ precursors can affect the radiative balance of the atmosphere through multiple (and sometimes competing) mechanisms. For example, the O$_3$ precursor CH$_4$ is itself a powerful greenhouse gas. Ozone and its other precursors also exert a strong control on the oxidizing capacity of the troposphere, and so can affect the lifetime of gases such as CH$_4$ (Derwent et al., 2001). For example, an increase in CO or VOCs would lead to a decrease in hydroxyl (OH) concentrations. Since OH is a major sink for CH$_4$, a decline in OH would lengthen the CH$_4$ lifetime, enhance the CH$_4$ concentration, and amplify surface warming. A rise in NO$_X$ emissions, on the other hand, could lead to an increase in OH in certain locations, shortening the CH$_4$ lifetime and causing surface cooling (Fuglestvedt et al., 1999). Ozone can itself generate OH through (1) photolysis leading to excited oxygen atoms followed by reaction with water vapor and (2) reaction with HO$_2$.

Figure 10-4 shows the radiative forcing associated with a suite of anthropogenic emissions, including O$_3$ precursors (IPCC, 2007b). The emission-based radiative forcing for CH$_4$, which includes the CH$_4$ effect on O$_3$ production, is +0.9 W/m$^2$, or nearly double that of the CH$_4$ abundance-based radiative forcing shown in Figure 10-3. Figure 10-4 also shows a warming from anthropogenic CO and VOC emissions of +0.27 W/m$^2$ and a net cooling of -0.21 W/m$^2$ for NO$_X$ emissions. The net cooling for NO$_X$ occurs mainly due to the links between NO$_X$ and CH$_4$. Consistent with these results, Shindell and Faluvegi (2009) calculated positive (+0.25 W/m$^2$) radiative forcing from the increase in anthropogenic emissions of CO and VOCs since pre-industrial times, as well as for CH$_4$ (+1 W/m$^2$). In contrast,
Shindell and Faluvegi (2009) found negative (-0.29 W/m²) radiative forcing from anthropogenic emissions of NOₓ. Other studies have found a near cancellation of the positive O₃ radiative forcing and the negative CH₄ radiative forcing that arise from an incremental increase in anthropogenic NOₓ emissions (Naik et al., 2005; Fiore et al., 2002; Fuglestvedt et al., 1999). The net effect of aircraft NOₓ on climate is especially complex (Isaksen et al., 2001; Wild et al., 2001). Stevenson (2004) calculated that aircraft NOₓ leads to short-term net warming via O₃ production in the cool upper troposphere, but long-term net cooling because of CH₄ loss.

OH production from O₃ precursors can also affect regional sulfate air quality and climate by increasing gas-phase oxidation rates of SO₂. Using the A1B scenario in the IPCC AR4, Unger (2006) reported that by 2030, enhanced OH from the A1B O₃ precursors may increase surface sulfate aerosol concentrations by up to 20% over India and China, relative to the present-day, with a corresponding increase in radiative cooling over these regions. In this way, O₃ precursors may impose an indirect cooling via sulfate (Unger, 2006).

Taken together, these results point out the need for careful assessment of net radiative forcing involving multiple pollutants in developing climate change policy (Unger et al., 2008). Many studies point to CH₄ as a particularly attractive target for emissions control since CH₄ is itself an important precursor of O₃ (West et al., 2007; Fiore et al., 2002). Fiore et al. (2002) found that reducing anthropogenic CH₄ emissions by 50% would lead to a global negative (-0.37 W/m²) radiative forcing, mostly from CH₄. In later research, Fiore et al. (2008) reported that CH₄ reductions would most strongly affect tropospheric O₃ column amounts in regions of strong downwelling from the upper troposphere (e.g., around 30°N) and in regions of NOₓ-saturated conditions.

The magnitude of the radiative forcing from the change in tropospheric O₃ abundance since the pre-industrial era is uncertain. This uncertainty derives in part from the scarcity of early measurements and in part from limited knowledge regarding processes in the natural atmosphere. As noted previously, the IPCC AR4 reports a radiative forcing of 0.35 W/m² from the change in tropospheric O₃ abundance since 1750 (Forster et al., 2007), ranking it third in importance behind the greenhouse gases CO₂ and CH₄. The O₃ radiative forcing could, in fact, be as large as 0.7 W/m², if reconstructions of pre-industrial and mid-20th century O₃ concentrations based on the measurement record are valid (Shindell and Faluvegi, 2002; Mickley et al., 2001). In any event, Unger et al. (2010) showed that present-day O₃ radiative forcing can be attributed to emissions from many economic sectors, including on-road vehicles, household biofuel, power generation, and biomass burning. As much as one-third of the radiative forcing from the 1890 to 1990 change in tropospheric O₃ concentration could be due to increased biomass burning (Ito et al., 2007a).

These calculated radiative forcing estimates can be compared to those obtained from satellite data. Using data from TOMS, Worden et al. (2008) estimated a reduction in clear-sky outgoing longwave radiation of 0.48 W/m² by O₃ in the upper troposphere over oceans in 2006. This radiative forcing includes contributions from both
anthropogenic and natural sources of O₃. Assuming that the concentration of O₃ has roughly doubled since pre-industrial times (Gauss et al., 2006), the total O₃ radiative forcing estimated with TOMS is consistent with that obtained from models estimating just the anthropogenic contribution.

Note: Values represent radiative forcing in 2005 due to emissions and changes since 1750. (S) and (T) next to gas species represent stratospheric and tropospheric changes, respectively.

**Figure 10-4** Components of radiative forcing for emissions of principal gases, aerosols, aerosol precursors, and other changes.
10.3.5 Calculating Radiative Forcing and Climate Response to Past Trends in Tropospheric O\textsubscript{3} Concentrations

Calculation of the climate response to the O\textsubscript{3} radiative forcing is challenging due to complexity of feedbacks, as mentioned in Section 10.3.2.2 and Section 10.3.3.4. In their modeling study, Mickley et al. (2004) reported a global mean increase of 0.28°C since pre-industrial times, with values as large as 0.8°C in continental interiors. For the time period since 1870, Hansen et al. (2005) estimated a much smaller increase in global mean surface temperature (0.11°C), but they implemented 1880s anthropogenic emissions in their base simulation and also took into account trends in both stratospheric and tropospheric O\textsubscript{3} concentrations. The modeled decline of lower stratospheric O\textsubscript{3} concentrations, especially over polar regions, cooled surface temperatures in this study, counteracting the warming effect of increasing tropospheric O\textsubscript{3} concentrations.

Figure 10-5 shows the Hansen et al. (2005) results as reported in Shindell et al. (2006). In that figure, summertime O\textsubscript{3} has the largest radiative effect over the continental interiors of the Northern Hemisphere. Shindell et al. (2006) estimated that the change in tropospheric O\textsubscript{3} concentration over the 20th century could have contributed about 0.3°C to annual mean Arctic warming and as much as 0.4-0.5°C during winter and spring. Over eastern China, Chang et al. (2009) calculated a surface temperature increase of 0.4°C to the 1970-2000 change in tropospheric O\textsubscript{3} concentration. It is not clear, however, to what degree regional changes in O\textsubscript{3} concentration influenced this response, as opposed to more global changes.
10.3.6 Calculating Radiative Forcing and Climate Response to Future Trends in Tropospheric O₃ Concentrations

Future trends in tropospheric O₃ concentrations depend in large part on what pathways in energy technology the world's societies will follow in coming decades. The trends in O₃ concentration will also depend on the changes in a suite of climate-sensitive factors, such as the water vapor content of the atmosphere. This section describes the following issues: (1) projected trends in the anthropogenic emissions of O₃ precursors; (2) the effects of these emissions on the tropospheric O₃ concentrations; (3) the effects of changing climate on tropospheric O₃ concentrations; and (4) radiative forcing and climate response to 21st century trends in tropospheric O₃ concentrations.
10.3.6.1 Emissions of Anthropogenic O₃ Precursors Across the 21st Century

The IPCC SRES effort devised scenarios for short-lived O₃ precursors as well as the well-mixed greenhouse gases including NOₓ, CO, and VOCs (IPCC, 2000). Using the IMAGE socioeconomic model, Streets et al. (2004) provided speciation for NOₓ and VOCs and allocated the trends in emissions over 17 regions and 8 economic sectors for the 2000-2050 time period. The worst-case IPCC scenario, A2, features continued dependence on fossil fuels, rapid population growth, and little sharing of technology between developed and developing nations. By 2100 in this scenario, global NOₓ, CO and CH₄ emissions increase by a factor of 3.5, 2.6, and 2.9, respectively, relative to 2000 (IPCC, 2000). Most of these increases in emissions occur over developing countries. For example over Asia, NOₓ emissions in the A2 scenario increase by more than a factor of four by 2100. The more moderate A1B scenario has global NOₓ and CO emissions increasing by 25% and 90%, respectively by 2100, but global CH₄ emissions decreasing by 10%. In the B1 scenario, with its emphasis on clean and efficient technologies, global emissions of NOₓ, CO, and CH₄ all decrease by 2100, relative to the present day (-40%, -60%, and -30%, respectively).

Other emissions scenarios have been recently developed to describe trends in the short-term (up to 2030). The Current Legislation (CLE) scenario provides trends consistent with existing air quality regulations; the Maximum Feasible Reduction (MFR) scenario seeks to reduce emissions of O₃ precursors to the maximum extent possible. Emission source changes relative to the present day for CLE, MFR, and A2 are given in Stevenson et al. (2006).

For the Fifth Assessment Report (IPCC AR5), a new set of climate futures has been developed: the Representative Concentration Pathways (RCPs) (Moss et al., 2010). The RCPs will explore for the first time approaches to climate change mitigation. The RCPs are designed to achieve radiative forcing targets of 2.6, 4.5, 6.0 and 8.5 W/m² by 2100, and have been designated RCP 2.6, RCP 4.5, RCP 6.0, and RCP 8.5, respectively (RCP 2.6 is also known as RCP3-PD.) The trends in O₃ precursors for the RCP scenarios were determined by climate policies implicit in each scenario and by plausible assumptions regarding future air quality regulations. These scenarios were chosen to map the wide range of climate outcomes presented in the literature and represent only four of many possible scenarios that would lead to the specific radiative forcing targets; a wide range of socioeconomic conditions could be consistent with each radiative forcing pathway (Moss et al., 2010). Therefore, they should not be interpreted as forecasts of future conditions, but rather as plausible climate and socio-economic futures.

Plots and comparisons of the RCP trends are available on the RCP website (RCP, 2009). In all RCPs, global anthropogenic NOₓ emissions decline 30-50% during the 21st century, though RCP 8.5 shows a peak during the 2020s at a value ~15% greater than that of 2000. Global anthropogenic VOC and CO emissions are relatively flat during the 2000-2050 time range, and then decline by 30-50% by the end of the
century. For CH₄, global mean emission trends for the four RCP projections differ substantially across the 21st century, with RCP 8.5 showing a tripling of emissions by 2100, and RCP 2.6 showing the emissions cut by half in this time range. RCP 4.5 and 6.0 show a peak in CH₄ emissions in the middle of the century before dropping by the end of the century to just below 2000 emission levels. All these global trends, however, contain some regional variation. For example, Asian emissions of both NOₓ and VOCs show large increases in the near term (2030s to 2050s).

### 10.3.6.2 Impact of 21st Century Trends in Emissions on Tropospheric O₃ Concentrations

Due to its short lifetime, tropospheric O₃ concentrations will respond readily to changes in anthropogenic emissions of O₃ precursors. As shown in Table 10-1, a recent multi-model study found increases in the tropospheric O₃ concentration of 15% and 6% for the IPCC A2 and CLE scenarios respectively for the 2000-2030 time period, and a decrease for the MFR scenario of 5% (Stevenson et al., 2006). These results indicate that the growth in tropospheric O₃ concentrations between 2000 and 2030 could be reduced or even reversed, depending on emission controls. For the relatively moderate A1B emissions scenario over the 2000-2050 time period, Wu et al. (2008a) calculated a change in O₃ concentration of about 20%.

As noted above, the RCP scenarios show large variations in their future projections of global mean CH₄ emissions, but mainly declines in the emissions of the other O₃ precursors across the 21st century. In one of the first efforts to assess the effect of these emission trends on global O₃ abundances, Lamarque et al. (2011) found that the large CH₄ increase in the RCP 8.5 scenario would drive a 15% enhancement of the tropospheric O₃ abundance by 2100, relative to the present-day, leading to a global mean radiative forcing of +0.2 W/m². By contrast, the global O₃ abundance would decrease in the other three RCPs, with declines in radiative forcing ranging from -0.07 to -0.2 W/m².
Table 10-1 Changes in anthropogenic emissions, CH4 and tropospheric O3 concentrations between 2000 and 2030, and the associated tropospheric O3 radiative forcing for three scenarios.

<table>
<thead>
<tr>
<th>Scenario</th>
<th>IPCC A2a</th>
<th>Current Legislation (CLE)a</th>
<th>Maximum Feasible Reduction (MFR)b</th>
</tr>
</thead>
<tbody>
<tr>
<td>Percent change in NOX emissions</td>
<td>+96%</td>
<td>+18%</td>
<td>-53%</td>
</tr>
<tr>
<td>Percent change in CO emissions</td>
<td>+62%</td>
<td>-16%</td>
<td>-53%</td>
</tr>
<tr>
<td>Percent change in CH4 concentration</td>
<td>+23%</td>
<td>+19%</td>
<td>0%</td>
</tr>
<tr>
<td>Percent change in tropospheric O3 concentration</td>
<td>+15%</td>
<td>+6%</td>
<td>-5%</td>
</tr>
<tr>
<td>Radiative forcing due to O3 concentration changeb (W/m²)</td>
<td>0.3</td>
<td>0.18</td>
<td>-0.05</td>
</tr>
</tbody>
</table>

aValues are ensemble means.
bIncludes radiative forcing due to corresponding CH4 change.

Source: Adapted from Stevenson et al. (2006).

10.3.6.3 Impact of 21st Century Climate on Tropospheric O3 Concentrations

For the time period from the 1800s to the present-day, most of the increase in the concentration of tropospheric O3 can be traced to changing emissions. Model studies show that climate change so far has likely had little effect on the tropospheric O3 concentrations (e.g., Grenfell et al., 2001). In the future, however, climate change is expected to bring large changes in a suite of variables that could affect O3 production, loss, and transport. For example, increased water vapor in a warming atmosphere is expected to enhance OH concentrations, which in remote, NOX-poor regions will accelerate O3 loss rates (Johnson et al., 1999).

In the 2050s A1B climate, Wu et al. (2008b) calculated a 5 ppb decrease in surface O3 concentrations over oceans. A rise in temperatures will also likely promote emissions of isoprene, an important biogenic precursor of O3. Model studies have calculated 21st-century increases in isoprene emissions ranging from 25-50%, depending on climate scenario and time horizon (Wu et al., 2008a and references therein). These studies however did not take into account the effects of changing climate and CO2 concentration on vegetation extent, which could have large consequences for biogenic emissions (Heald et al., 2008; Sanderson et al., 2003). In any event, enhanced isoprene emissions will increase O3 concentrations in VOC-limited regions, but decrease O3 concentrations in NOX-limited regions (Wu et al., 2008a; Pyle et al., 2007; Sanderson et al., 2003). Convection frequencies and lightning flash rates will also likely change in a changing climate, with consequences for lightning NOX emissions and O3 concentrations in the upper troposphere (Sinha and Toumi, 1997; Price and Rind, 1994). While Wu et al. (2008a) calculated an
increase in lightning NO\textsubscript{X} by 2050 due to enhanced deep convection, Jacobson and Streets (2009) projected a decrease in lightning NO\textsubscript{X} due to a declining cloud ice in their future atmosphere. Finally, changes in transport processes will almost certainly accompany global climate change. For the 2050 A1B climate, Wu et al. (2008b) showed that flattening of the meridional temperature gradient in a warming world would lead to slower intercontinental transport of tropospheric O\textsubscript{3}. For the A2 climate in 2100, Zeng and Pyle (2003) projected an 80% increase in the flux of stratospheric O\textsubscript{3} into the troposphere, relative to the present-day.

Taken together, these climate-driven processes could have appreciable effects on the concentration and distribution of tropospheric O\textsubscript{3}. As shown in Wu et al. (2008b), model projections of the change in O\textsubscript{3} concentration due solely to future climate change range from -12% to +3%, depending on the model, scenario, and time horizon.

### 10.3.6.4 Radiative Forcing and Climate Response from 21st Century Trends in Tropospheric O\textsubscript{3} Concentrations

In the near term (2000-2030), Stevenson et al. (2006) estimated an O\textsubscript{3} radiative forcing of near zero for MFR, 0.18 W/m\textsuperscript{2} for CLE, and 0.3 W/m\textsuperscript{2} for the A2 scenario (Table 10-1). Menon et al. (2008), following the moderate A1B scenario, calculated a radiative forcing of 0.12 W/m\textsuperscript{2} from the 2000-2030 change in tropospheric O\textsubscript{3} concentrations, about the same as that derived by Stevenson et al. (2006) for the CLE scenario. Over the longer term (2000 to 2100) for the A1B scenario, Gauss et al. (2003) reported large positive radiative forcing (0.40 to 0.78 W/m\textsuperscript{2}) due to the change in tropospheric O\textsubscript{3} concentrations, as shown in Figure 10-6. Normalized radiative forcing for these model calculations fell within a relatively narrow range, 0.032 to 0.040 W/m\textsuperscript{2} DU, indicating that the largest uncertainty lies in the model-calculated changes in O\textsubscript{3} concentration. Applying the A2 scenario, Chen et al. (2007b) estimated a global mean radiative forcing of 0.65 W/m\textsuperscript{2} from tropospheric O\textsubscript{3} by 2100, consistent with the Gauss et al. (2003) results. These studies took into account only the effect of changing emissions on tropospheric O\textsubscript{3} concentrations. In their calculations of the 2000-2100 radiative forcing from O\textsubscript{3} in the A2 scenario, Liao et al. (2006) found that inclusion of climate effects on tropospheric O\textsubscript{3} reduced their radiative forcing estimate by 20%.
Figure 10-6  Global mean radiative forcing estimates calculated by a set of models for the 2000-2100 change in tropospheric O₃ concentrations.

Several studies have included tropospheric O₃ in their investigations of the response in the future atmosphere to a suite of short-lived species (e.g., Levy et al., 2008; Shindell et al., 2008; Shindell et al., 2007). Few studies, however, have calculated the climate response to changes in tropospheric O₃ concentrations alone in the future atmosphere. For the A2 atmosphere, Chen et al. (2007b) estimated a global mean surface temperature increase of +0.34°C by 2100 in response to the change in O₃ concentration. The largest temperature increases in this study, as much as 5°C, occurred over the populous regions of Asia and the Middle East and downwind of biomass burning regions in South Africa and South America.
10.4 UV-B Shielding Effects and Tropospheric O$_3$

10.4.1 Background

UV radiation emitted from the Sun contains sufficient energy when it reaches the Earth to break (photolyze) chemical bonds in molecules, thereby leading to damaging effects on living organisms and materials. Atmospheric O$_3$ plays a crucial role in reducing exposure to solar UV radiation at the Earth’s surface. Stratospheric O$_3$ is responsible for the majority of this shielding effect, as approximately 90% of total atmospheric O$_3$ is located there over mid-latitudes (Kar et al., 2010; Crist et al., 1994). Investigation of the supplemental shielding of UV-B radiation provided by tropospheric O$_3$ is necessary for quantifying UV-B exposure and the incidence of related human health effects, ecosystem effects, and materials damage. The role of tropospheric O$_3$ in shielding of UV-B radiation is discussed in this section.

10.4.2 Human Exposure and Susceptibility to Ultraviolet Radiation

The factors that potentially influence UV radiation exposure were discussed in detail in Chapter 10 of the 2006 O$_3$ AQCD (U.S. EPA, 2006b) and are summarized here. These factors included outdoor activity, occupation, age, sex, geography, and protective behavior. Outdoor activity and occupation both influenced the amount of time people spend outdoors during daylight hours, the predominant factor for exposure to solar UV radiation. Age and sex were found to be factors that influence human exposure to UV radiation, particularly by influencing other factors of exposure such as outdoor activity and risk behavior. Studies indicated that females generally spent less time outdoors and, consequently, had lower UV radiation exposure on average compared to males. Geography influences the degree of solar UV flux to the surface, and hence exposure to UV radiation. Higher solar flux at lower latitudes increased the annual UV radiation dose for people living in southern states relative to northern states. Altitude was also found to influence personal exposure to UV radiation. Protective behaviors such as using sunscreen, wearing protective clothing, and spending time in shaded areas were shown to reduce exposure to UV radiation. Given these and other factors that potentially influence UV radiation exposure, the 2006 O$_3$ AQCD (U.S. EPA, 2006b) listed the following subpopulations potentially at risk for higher exposures to UV radiation:

- Individuals who engage in high-risk behavior (e.g., sunbathing);
- Individuals who participate in outdoor sports and activities;
- Individuals who work outdoors with inadequate shade (e.g., farmers, construction workers, etc.);
- Individuals living in geographic areas with higher solar flux including lower latitudes (e.g., Honolulu, HI) and higher altitudes (e.g., Denver, CO).
The risks associated with all these factors are, of course, highly dependent on season and region (Sliney and Wengraitis, 2006).

### 10.4.3 Human Health Effects due to UV-B Radiation

Chapter 10 of the 2006 O$_3$ AQCD ([U.S. EPA, 2006b](#)) covered in detail the human health effects associated with solar UV-B radiation exposure. These effects include erythema, skin cancer, ocular damage, and immune system suppression. These adverse effects, along with protective effects of UV radiation through increased production of vitamin D are summarized in this section. For additional details, the reader is referred to Chapter 10 of the 2006 O$_3$ AQCD ([U.S. EPA, 2006b](#)) and references therein.

The most conspicuous and well-recognized acute response to UV radiation is erythema, or the reddening of the skin. Erythema is likely caused by direct damage to DNA by UV radiation. Many studies discussed in Chapter 10 of the 2006 O$_3$ AQCD ([U.S. EPA, 2006b](#)) found skin type to be a significant risk factor for erythema. Skin cancer is another prevalent health effect associated with UV radiation. Exposure to UV radiation is considered to be a major risk factor for all forms of skin cancer. Ocular damage from UV radiation exposure includes effects on the cornea, lens, iris, and associated epithelial and conjunctival tissues. The region of the eye affected by exposure to UV radiation depends on the wavelength of the incident UV radiation. Depending on wavelength, common health effects associated with UV radiation include photokeratitis (snow blindness; short wavelengths) and cataracts (opacity of the lens; long wavelengths).

Experimental studies reviewed in Chapter 10 of the 2006 O$_3$ AQCD ([U.S. EPA, 2006b](#)) have shown that exposure to UV radiation may suppress local and systemic immune responses to a variety of antigens. Results from controlled human exposure studies suggest that immune suppression induced by UV radiation may be a risk factor contributing to skin cancer induction. There is also evidence that UV radiation has indirect involvement in viral oncogenesis through the human papillomavirus, dermatomyositis, human immunodeficiency virus, and other forms of immunosuppression.

A potential health benefit of increased UV-B exposure relates to the production of vitamin D in humans. Most humans depend on sun exposure to satisfy their requirements for vitamin D. Vitamin D deficiency can cause metabolic bone disease among children and adults, and also may increase the risk of many common chronic diseases, including type I diabetes mellitus and rheumatoid arthritis. Substantial in vitro and toxicological evidence also support a role for vitamin D activity against the incidence or progression of various forms of cancer. In some studies, UV-B related production of vitamin D had potential beneficial immunomodulatory effects on multiple sclerosis, insulin-dependent diabetes mellitus, and rheumatoid arthritis. More details on UV-B protective studies are provided in Chapter 10 of the 2006 O$_3$ AQCD ([U.S. EPA, 2006b](#)).
In establishing guidelines on limits of exposure to UV radiation, the International Commission on Non-Ionizing Radiation Protection (ICNIRP) agreed that some low-level exposure to UV radiation has health benefits (ICNIRP, 2004). However, the adverse health effects of higher UV exposures necessitated the development of exposure limits for UV radiation. The ICNIRP recognized the challenge in establishing exposure limits that would achieve a realistic balance between beneficial and adverse health effects. As concluded by ICNIRP (2004), “[t]he present understanding of injury mechanisms and long-term effects of exposure to [UV radiation] is incomplete, and awaits further research.”

10.4.4 Ecosystem and Materials Damage Effects Due to UV-B Radiation

A 2009 progress report on the environmental effects of O3 depletion from the UNEP, Environmental Effects Assessment Panel (UNEP, 2009) lists many ecosystem and materials damage effects from UV-B radiation. An in-depth assessment of the global ecosystem and materials damage effects from UV-B radiation per se is out of the scope of this assessment. However, a brief summary of some mid-latitude effects is provided in this section to provide context for UV-B related issues pertaining to tropospheric O3. The reader is referred to the UNEP report (UNEP, 2009) and references therein for further details. All of these UV-B related ecosystem and materials effects can also be influenced by climate change through temperature and other meteorological alterations, making quantifiable predictions of UV-B shielding effects difficult.

**Terrestrial ecosystem effects** from increased UV-B radiation include reduced plant productivity and plant cover, changes in biodiversity, susceptibility to infection, and increases in natural UV protective responses. In general, however, these effects are small for moderate UV-B increases at mid-latitudes. A field study on wheat in southern Chile found no substantial changes in crop yield with moderate increases in UV-B radiation (Calderini et al., 2008). Similarly, field studies on silver birch (Betula pendula) in Finland found no measurable effects in photosynthetic function with increases in UV-B radiation (Aphalo et al., 2009). Subtle, but important, changes in habitat and biodiversity have also been linked to increases in UV-B radiation (Mazza et al., 2010; Obara et al., 2008; Wahl, 2008). Some plants have natural coping mechanisms for dealing with changes in UV-B radiation (Favory et al., 2009; Jenkins, 2009; Brown and Jenkins, 2008; Ioki et al., 2008), but these defenses may have costs in terms of reduced growth (Snell et al., 2009; Clarke and Robinson, 2008; Semerdjieva et al., 2003; Phoenix et al., 2000).

**Aquatic ecosystem effects** from increased UV-B radiation include sensitivity in growth, immune response, and behavioral patterns of aquatic organisms. One study looking at coccolithophores, an abundant phytoplankton group, found a 25% reduction in cellular growth with UV-B exposure (Gao et al., 2009a). Exposure to relevant levels of UV-B radiation has been shown to modify immune response, blood chemistry, and behavior in certain species of fish (Markkula et al., 2009; Holtby and
Adverse effects on growth and development from UV-B radiation have also been observed for amphibians, sea urchins, mollusks, corals, and zooplankton (Garcia et al., 2009b; Romansic et al., 2009; Croteau et al., 2008b; Croteau et al., 2008a; Marquis et al., 2008; Marquis and Miaud, 2008; Oromi et al., 2008). Increases in the flux of UV-B radiation may also result in an increase in the catalysis of trace metals including mercury, particularly in clear oligotrophic lakes with low levels of dissolved organic carbon to stop the penetration of UV-B radiation (Schindler et al., 1996). This could then alter the mobility of trace metals including the potential for increased mercury volatilization and transport within and among ecosystems.

**Biogeochemical cycles**, particularly the carbon cycle, can also be influenced by increased UV-B radiation. A study on high latitude wetlands found UV-induced increases in CO₂ uptake through soil respiration (Haapala et al., 2009) while studies on arid terrestrial ecosystems found evidence for UV-induced release of CO₂ through photodegradation of above-ground plant litter (Brandt et al., 2009; Henry et al., 2008; Caldwell et al., 2007; Zepp et al., 2007). Changes in solar UV radiation may also have effects on carbon cycling and CO₂ uptake in the oceans (Brewer and Peltzer, 2009; Meador et al., 2009; Fritz et al., 2008; Zepp et al., 2008; Hader et al., 2007) as well as release of dissolved organic matter from sediment and algae (Mayer et al., 2009; Riggsbee et al., 2008). Additional studies showing effects on these and additional biogeochemical cycles including the water cycle and halocarbon cycle can be found in the UNEP report (UNEP, 2009) and references therein.

**Materials damage** from increased UV-B radiation include UV-induced photodegradation of wood (Kataoka et al., 2007) and plastics (Pickett et al., 2008). These studies and others summarizing photo-resistant coatings and materials designed to reduce photodegradation of materials are summarized in the UNEP report (UNEP, 2009) and references therein.

The ecosystem, carbon cycle, and materials effects described in this section are for UV-B exposure in general. Only a small fraction of these effects would be offset by incremental decreases in UV-B exposure resulting from increases in tropospheric O₃ concentrations. Attribution of UV-B shielding effects to changes in tropospheric O₃ concentrations is a highly complex problem as discussed in the next section.

### 10.4.5 UV-B Shielding Effects Associated with Changes in Tropospheric O₃ Concentrations

There are multiple complexities in attempting to quantify the relationship between changes in tropospheric O₃ concentrations and UV-B exposure. The 2006 O₃ AQCD (U.S. EPA, 2006b) described a handful of studies addressing this relationship, but none reported quantifiable effects of tropospheric O₃ concentration fluctuations on UV-B exposure at the surface. Further quantifying the relationship between UV-B exposure and health or welfare effects is complicated by the uncertainties involved in the selection of an action spectrum and appropriate characterization of dose.
(e.g., peak or cumulative levels of exposure, timing of exposures, etc.) The lack of published studies that critically examined these issues together—that is the incremental health or welfare effects attributable specifically to UV-B changes resulting from changes in tropospheric O₃ concentrations—lead to the prior conclusion that the effect of changes in surface-level O₃ concentrations on UV-induced health outcomes could not be critically assessed within reasonable uncertainty (U.S. EPA, 2006b).

A recent study by Madronich et al. (2011) used CMAQ to estimate UV radiation response to changes in tropospheric O₃ concentrations under different control scenarios projected out to 2020. This study focused on southeastern U.S. and accounted for spatial and temporal variation in tropospheric O₃ concentration reductions, an important consideration since most controls are focused on reducing O₃ concentrations in populated urban areas. The contrasting control strategies considered in this study included a historical scenario designed to meet an 84 ppb 8-h daily max standard and a reduced scenario designed to bring areas predicted to exceed a similarly designed 70 ppb standard into attainment. A biologically effective irradiance was estimated by multiplying the modeled UV irradiance by a sensitivity function (action spectrum) for the induction of nonmelanoma skin cancer in mice corrected for human skin transmission, then integrating over UV wavelengths. The average relative change in skin cancer-weighted surface UV radiation between the two scenarios was 0.11 ± 0.03% over June, July and August. Weighting by population, this estimate increased to 0.19 ± 0.06%. Madronich et al. (2011) report that their estimated UV radiation increment is an order of magnitude less than that reported in an earlier study by Lutter and Wolz (1997) with the main reason for the discrepancy coming from the overly-simplified uniform 10 ppb reduction in O₃ concentrations assumed in the former study. Madronich et al. (2011) did not attempt to link their predicted increase in UV radiation to a predicted increase in skin cancer incidence, however, due to several remaining and substantial uncertainties.

Quantitatively estimating human health and welfare effects directly attributed to changes in UV-B penetration resulting from changes in ground-level O₃ concentrations will require both (a) a solid understanding of the multiple factors that define the extent of exposure to UV-B, and (b) well-defined and quantifiable links between UV-B exposure and human disease and welfare effects. Detailed information does not exist regarding the relevant type (e.g., peak or cumulative) and time period (e.g., developmental, lifetime, or current) of exposure, wavelength dependency of biological responses, and inter-individual variability in UV resistance.

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1 The reader is referred to the U.S. EPA 2003 Final Response to Court Remand (U.S. EPA, 2003) for detailed discussions of the data and scientific issues associated with the determination of public health benefits resulting from the attenuation of UV-B by surface-level O₃.
Although the UV-B related health effects attributed to marginal reductions in tropospheric or ground-level O$_3$ concentrations have not been directly assessed to date, they would be expected to be small based on current information indicating a negligibly small effect of potential future changes in tropospheric O$_3$ concentrations on ground-level UV-B radiation. In conclusion, the effect of changes in surface-level O$_3$ concentrations on UV-induced health and welfare outcomes cannot yet be critically assessed within reasonable uncertainty.

### 10.5 Summary and Causal Determinations

#### 10.5.1 Summary of the Effects of Tropospheric O$_3$ on Climate

Radiative forcing by a greenhouse gas or aerosol is a metric used to quantify the change in balance between radiation coming into and going out of the atmosphere caused by the presence of that substance. Tropospheric O$_3$ is a major greenhouse gas and radiative forcing agent; evidence from satellite data shows a sharp dip in the outgoing infrared radiation in the 9.6 µm O$_3$ absorption band. Models calculate that the global average concentration of tropospheric O$_3$ has doubled since the pre-industrial era, while observations indicate that in some regions O$_3$ may have increased by factors as great as 4 or 5. These increases are tied to the rise in emissions of O$_3$ precursors from human activity, mainly fossil fuel consumption and agricultural processes. Overall, the evidence supports a causal relationship between changes in tropospheric O$_3$ concentrations and radiative forcing.

While the developed world has successfully reduced emissions of O$_3$ precursors in recent decades, many developing countries have experienced large increases in precursor emissions and these trends are expected to continue, at least in the near term. Projections of radiative forcing due to changing O$_3$ concentrations over the 21st century show wide variation, due in large part to the uncertainty of future emissions of source gases. In the near-term (2000-2030), projections of O$_3$ radiative forcing range from near zero to $+0.3$ W/m$^2$, depending on the emissions scenario (Stevenson et al., 2006).

The impact of the tropospheric O$_3$ change since pre-industrial times on climate has been estimated to be about 25-40% of the anthropogenic CO$_2$ impact and about 75% of the anthropogenic CH$_4$ impact according to the IPCC, ranking it third in importance after CO$_2$ and CH$_4$. There are large uncertainties in the magnitude of the radiative forcing estimate attributed to tropospheric O$_3$, making the impact of tropospheric O$_3$ on climate more uncertain than the effect of the longer-lived greenhouse gases. Furthermore, radiative forcing does not take into account the climate feedbacks that could amplify or dampen the actual climate response (e.g., surface temperature change) that would result from a change in tropospheric O$_3$ concentrations. Quantifying the change in surface temperature requires a complex climate simulation in which all important feedbacks and interactions are accounted
for. As these processes are not well understood or easily modeled, the surface temperature response to a given radiative forcing is highly uncertain and can vary greatly among models and from region to region within the same model. Even with these uncertainties, global climate models indicate that tropospheric O₃ has contributed to observed changes in global mean and regional surface temperatures. As a result of such evidence presented in climate modeling studies, there is likely to be a causal relationship between changes in tropospheric O₃ concentrations and effects on climate as quantified through surface temperature response.

Reduction of tropospheric O₃ concentrations could therefore provide an important means to slow climate change in addition to the added benefit of improving surface air quality. However, the precursors of O₃ also have competing effects on the greenhouse gas CH₄, complicating emissions reduction strategies. A decrease in CO or VOC emissions would enhance OH concentrations, shortening the lifetime of CH₄, while a decrease in NOₓ emissions could depress OH concentrations in certain regions and lengthen the CH₄ lifetime. Abatement of CH₄ emissions would likely provide the most straightforward means to address climate change since CH₄ is itself an important O₃ precursor (West et al., 2007; West et al., 2006; Fiore et al., 2002). A reduction of CH₄ emissions would also improve air quality on its own right. A set of global abatement measures identified by West and Fiore (2005) could reduce CH₄ emissions by 10% at a cost savings, decrease background O₃ concentrations by about 1 ppb in the Northern Hemisphere summer, and lead to a global net cooling of 0.12 W/m². West et al. (2007) explored further the benefits of CH₄ abatement, finding that a 20% reduction in global CH₄ emissions would lead to greater cooling per unit reduction in surface O₃ concentration, compared to 20% reductions in VOCs or CO.

Important uncertainties remain regarding the effect of tropospheric O₃ on future climate change. To address these uncertainties, further research is needed to: (1) improve knowledge of the natural atmosphere; (2) interpret observed trends in O₃ concentrations in the free troposphere and remote regions; (3) improve understanding of the CH₄ budget, especially emissions from wetlands and agricultural sources; (4) understand the relationship between regional O₃ radiative forcing and regional climate change; and (5) determine the optimal mix of emissions reductions that would act to limit future climate change.

10.5.2 Summary of UV-B Related Effects on Human Health, Ecosystems, and Materials Relating to Changes in Tropospheric O₃ Concentrations

UV radiation emitted from the Sun contains sufficient energy when it reaches the Earth to break (photolyze) chemical bonds in molecules, thereby leading to damaging effects on living organisms and materials. Atmospheric O₃ plays a crucial role in reducing exposure to solar UV radiation at the Earth’s surface. Ozone in the stratosphere is responsible for the majority of this shielding effect, as approximately 90% of total atmospheric O₃ is located there over mid-latitudes. Ozone in the
troposphere provides supplemental shielding of radiation in the wavelength band from 280-315 nm, referred to as UV-B radiation. UV-B radiation has important effects on human health and ecosystems, and is associated with materials damage.

EPA has found no published studies that adequately examine the incremental health or welfare effects (adverse or beneficial) attributable specifically to changes in UV-B exposure resulting from perturbations in tropospheric O₃ concentrations. While the effects are expected to be small, they cannot yet be critically assessed within reasonable uncertainty. Overall, the evidence is inadequate to determine if a causal relationship exists between changes in tropospheric O₃ concentrations and effects on health and welfare related to UV-B shielding.

10.5.3 Summary of O₃ Causal Determinations

The evidence reviewed in this chapter describes the recent findings regarding the climate and UV-B shielding effects of changes in tropospheric O₃ concentrations. Table 10-2 provides an overview of the causal determinations for each of the categories evaluated including the effect of tropospheric O₃ on radiative forcing, climate change, and health and welfare effects related to UV-B shielding.

<table>
<thead>
<tr>
<th>Effects</th>
<th>Causal Determination</th>
</tr>
</thead>
<tbody>
<tr>
<td>Radiative Forcing</td>
<td>Causal relationship</td>
</tr>
<tr>
<td>Climate Change</td>
<td>Likely to be a causal relationship</td>
</tr>
<tr>
<td>Health and Welfare Effects Related to UV-B Shielding</td>
<td>Inadequate to determine if a causal relationship exists</td>
</tr>
</tbody>
</table>
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