

# Appendix C

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## ***Fitting Quantiles by Combining Nonlinear and Linear Regression***

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The approach outlined below was motivated by a July 3, 1997, memorandum from Timothy Barry, Senior Analyst, Office of Policy and Re-Invention, to Jackie Moya, Environmental Engineer, Office of Research and Development.

Let  $F(x)$  be the cumulative distribution function (CDF) of a nonnegative continuous random variable  $X$ , that is,  $F(x) = P[X \leq x]$  = the probability of a value  $\leq x$ . Since  $X$  is continuous,  $F$  is continuous and strictly increasing, and its inverse  $F^{-1}$  exists, so that  $F[F^{-1}(p)] = p$  and  $F^{-1}[F(x)] = x$ . Let  $Y = aX^r$  be a power transform of  $X$  with both  $a$  and  $r$  strictly positive ( $a > 0, r > 0$ ), and let  $G(y) = P[Y \leq y]$  be the CDF of  $Y$ . Recall that  $y_p$  is the  $p$ th quantile of  $Y$  iff  $G(y_p) = p$ . Here iff denotes logical equivalence (“if and only if”).

Using basic algebra, set theory, and probability, it can be shown that

$$\log(y_p) = r \log[F^{-1}(p)] + \log(a). \quad (\text{C.1})$$

Hence, if  $F$  and its inverse  $F^{-1}$  are known, and there are empirical quantiles  $y_p$  for several different values of  $p$ , then the power transform parameters  $a$  and  $r$  by linear regression of  $\log(y_p)$  on  $\log[F^{-1}(p)]$  can be estimated. This is easily extended to cover distributions that are nonnegative and continuous except for a point mass  $M$  at zero. To see this, let  $H(y) = 0$  for  $y < 0$ ,  $H(y) = M + (1-M)G(y)$  for  $y \geq 0$ , and note that  $H(y_p) = p$  iff  $G(y_p) = (p-M)/(1-M)$ . Hence for  $p > M$ , the  $p$ th quantile  $y_p$  for  $H$  is obtained by solving  $G(y_p) = p_1$ , where  $p_1 = (p-M)/(1-M)$ . This leads to

$$\log(y_p) = r \log[F^{-1}(p_1)] + \log(a). \quad (\text{C.2})$$

These arguments suggest the following combined nonlinear/linear regression approach to fitting the five-parameter generalized F distribution with a point mass  $M$  at zero.

Let  $p_{\min}$  be the smallest  $p$  for which a positive empirical quantile  $y_p$  exceeds zero. Then  $M$  should not exceed  $p_{\min}$ .

1. Perform an outer search on  $M$ , or simply use a grid of  $M$  values, such as  
 $M = 0, 0.1 p_{\min}, 0.2 p_{\min}, \dots, 0.9 p_{\min}$ .
2. For a given value of  $M$ , perform a two-dimensional search on the degrees-of-freedom parameters  $df_1, df_2$  of the generalized F distribution.
3. Given  $M, df_1,$  and  $df_2,$  estimate  $a$  and  $r$  by solving the linear regression problem defined by Equation C.2.