The Two-Parameter Extreme-value Distribution S. Maghsoodloo

Suppose the TTF (time to failure), T, of a component has a Weibull distribution with the cdf given by

$$\mathbf{F}(t) = \mathbf{1} - \mathbf{e}^{-\left(\frac{t-\delta}{\theta-\delta}\right)^{\beta}}, \quad 0 \le t < \infty$$
(1)

where  $\delta$  is the minimum life,  $\theta$  is the characteristic life, and  $\beta$  is the Weibull slope.

In equation (1), make the transformation  $x = \beta \ln(\frac{t-\delta}{\theta-\delta})$ . Then the cdf of the rv X at point

x is given by

$$W(x) = P(X \le x) = \left[\beta \ln\left(\frac{T-\delta}{\theta-\delta}\right) \le x\right] = P\left[T \le \delta + (\theta-\delta)e^{x/\beta}\right] = 1 - e^{-\left(\frac{\delta+(\theta-\delta)e^{x/\beta}-\delta}{\theta-\delta}\right)^{\beta}}$$
$$= 1 - e^{-\left(e^{x/\beta}\right)^{\beta}} = 1 - e^{-\left(e^{x}\right)}, \quad -\infty < x < \infty$$
(2)

The cdf, W(x), is the standard distribution function for an extreme-value, where by extreme-value we mean the log base e of a component life-time. To obtain the pdf of the standard extreme-value distribution, differentiate Eq. (2) with respect to x.

pdf (x) = w(x) = 
$$e^{-(e^x)}e^x$$
,  $-\infty < x < \infty$  (3)

In order to convert the standard extreme-value density in (3) to the two-parameter case, let  $x = (y - \xi)/\theta$  and G(y) represent the cdf of Y at y. Then, by definition,

$$G(\mathbf{y}) = \mathbf{P}(\mathbf{Y} \le \mathbf{y}) = \mathbf{P}[(\mathbf{\theta} \ \mathbf{X} + \mathbf{\xi}) \le \mathbf{y}] = \mathbf{P}[\mathbf{X} \le (\mathbf{y} - \mathbf{\xi})/\mathbf{\theta}] = \mathbf{W}[(\mathbf{y} - \mathbf{\xi})/\mathbf{\theta}] =$$
$$= \mathbf{1} - \mathbf{e}^{-\mathbf{e}(\mathbf{y} - \mathbf{\xi})/\mathbf{\theta}} = \mathbf{1} - \mathbf{e}^{-\mathbf{Exp}[(\mathbf{y} - \mathbf{\xi})/\mathbf{\theta}]}, \quad -\infty < \mathbf{y} < \infty$$
(4)

Equation (4) represents the cdf of the two-parameter extreme-value distribution. To obtain its density function g(y), differentiate equation (4) wrt y.

$$g(\mathbf{y}) = \frac{1}{\theta} e^{-\mathbf{e}^{(\mathbf{y}-\boldsymbol{\xi})/\theta}} e^{(\mathbf{y}-\boldsymbol{\xi})/\theta} , \quad -\infty < \mathbf{y} < \infty$$
 (5)

The parameter  $\xi$  is a measure of location and because at  $y = \xi$  the value of the cdf in (4)

becomes  $G(\xi) = 1 - e^{-e^0} = 1 - e^{-1} = 0.6321205588$ , then  $-\infty < \xi < \infty$  is the 63.21206 percentile

of the extreme-value pdf in (5). The parameter  $\theta > 0$  is related to the spread (or standard deviation) of the density function in (5).

To obtain the inverse (or percentile) function, solve y in terms of G from equation (4), which yields

$$y = \xi + \theta \times \ln[\ln(\frac{1}{1-G})] \rightarrow G^{-1}(y) = \xi + \theta \times \ln[\ln(\frac{1}{1-y})], 0 \le y \le 1, i.e., \text{ the percentile}$$

function is given by

$$\mathbf{y}_{\mathbf{p}} = \boldsymbol{\xi} + \boldsymbol{\theta} \times \ln\left[\ln\left(\frac{1}{1-\mathbf{p}}\right)\right] \tag{6}$$

For example, the median of the distribution is obtained by putting p = 0.50 in equation (6), i.e.,

 $y_{0.50} = \xi + \theta \times \ln(\ln\frac{1}{0.5}) = \xi + \theta \times \ln(\ln 2) = \xi - 0.366513\theta.$  The 25<sup>th</sup> and 75<sup>th</sup> percentiles are  $y_{0.25} = \xi + \theta \times \ln\ln(\frac{1}{1 - 0.25}) = \xi - 1.2459\theta \text{ and } y_{0.75} =$ 

 $\xi + \theta \times \ln \ln \left(\frac{1}{1 - 0.75}\right) = \xi + 0.3266343\theta \rightarrow IQR = 1.5725336\theta.$ 

The mean of Y is given by [after the transformation  $u = e^{(y - \xi)/\theta}$ ]

$$E(Y) = \int_{-\infty}^{\infty} y \frac{1}{\theta} e^{-\mathbf{e}^{(y-\xi)/\theta}} e^{(y-\xi)/\theta} dy = \int_{0}^{\infty} (\xi + \theta \ln u) e^{-u} du = \xi + \theta \int_{0}^{\infty} (\ln u) e^{-u} du.$$
(7)

This last integral apparently can be shown to become  $\int_{0}^{\infty} (\ln u) e^{-u} du = -\gamma$  so that  $E(Y) = \xi$  –

 $\theta \times \gamma$ , where  $\gamma = 0.5772157$  is the Euler's constant. This proof is very difficult. Further, it can be

shown that 
$$V(Y) = \frac{\pi^2 \theta^2}{6}$$
 so that  $\sigma_Y = \frac{\pi \theta}{\sqrt{6}} = 1.28255 \ \theta$  and the  $CV(y) = \frac{1.28255\theta}{\xi - 0.5772157\theta}$ 

Setting the 1<sup>st</sup> derivative of g(y) in (5) equal to zero yields the modal point of the extreme-value distribution to be MO =  $\xi$ . Since the distribution is negatively skewed ( $\alpha_3 < 0$ ), then MO =  $\xi > y_{0.50}$  =  $\xi - 0.366513\theta > E(Y) = \xi - 0.5772157 \theta$ . Finally, the hazard function (or failure rate function)

is given by 
$$\mathbf{h}(\mathbf{y}) = \frac{\mathbf{f}(\mathbf{y})}{\mathbf{R}(\mathbf{y})} = \frac{\frac{1}{\theta} e^{-\mathbf{e}^{(\mathbf{y}-\boldsymbol{\xi})/\theta}} e^{(\mathbf{y}-\boldsymbol{\xi})/\theta}}{e^{-\mathbf{e}^{(\mathbf{y}-\boldsymbol{\xi})/\theta}}} = \frac{1}{\theta} e^{(\mathbf{y}-\boldsymbol{\xi})/\theta}$$