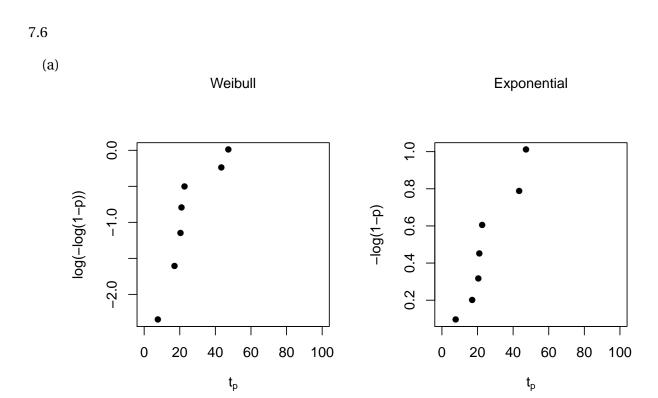
## Reliability Analysis Assignment 5 (personal)

## Kuan-I Chung

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(b) The exponential probability plot fits better than the Weibull one.

(c) Let  $X_{(1)},...,X_{(r)}$  be the exact-timed data and  $X_{(r+1)},...,X_{(n)}$  be the right censored data. Assume both set of data have exponential distributions with  $\theta$ , then

$$\begin{split} L(\theta) &= \left(\prod_{i=1}^{r} \frac{1}{\theta} e^{-\frac{x_{(i)}}{\theta}}\right) \left(\prod_{i=r+1}^{n} \frac{1}{\theta} e^{-\frac{x_{(i)}}{\theta}}\right) = \left(\frac{1}{\theta}\right)^{r} e^{-\frac{\sum_{i=1}^{n} x_{(i)}}{\theta}} = \left(\frac{1}{\theta}\right)^{r} e^{-\frac{\sum_{i=1}^{n} x_{i}}{\theta}} \\ &\Rightarrow log L(\theta) = log \left(\frac{1}{\theta}\right)^{r} - \frac{\sum_{i=1}^{n} x_{i}}{\theta} \\ \text{Let } \frac{\partial log L(\theta)}{\partial \theta} &= -\frac{r}{\theta} + \frac{\sum_{i=1}^{n} x_{i}}{\theta^{2}} = 0 \Rightarrow \widehat{\theta} = \frac{\sum_{i=1}^{n} x_{i}}{r} \\ &\because \frac{\partial^{2} log L(\theta)}{\partial \theta^{2}} \Big|_{\theta = \widehat{\theta}} &= \frac{\sum_{i=1}^{n} x_{i} - 2\sum_{i=1}^{n} x_{i}}{\theta^{3}} < 0 \\ &\therefore \widehat{\theta}_{mle} &= \frac{\sum_{i=1}^{n} x_{i}}{r} \approx 82.8 \\ &\Rightarrow Var \left(\widehat{\theta}_{mle}\right) &= Var \left(\frac{\sum_{i=1}^{n} x_{i}}{r}\right) &= Var \left(\frac{2 \cdot \frac{\sum_{i=1}^{n} x_{i}}{\theta} \cdot \frac{\theta}{2}}{r}\right) \\ &\because 2 \cdot \frac{\sum_{i=1}^{n} x_{i}}{\theta} \sim \chi_{2r}^{2} \\ &\Rightarrow Var \left(\widehat{\theta}_{mle}\right) &= \frac{\theta^{2}}{4r^{2}} \cdot 4r = \frac{\theta^{2}}{r} \\ &\Rightarrow se_{\widehat{\theta}_{mle}} &= \frac{\widehat{\theta}}{\sqrt{r}} &= \frac{\sum_{i=1}^{n} x_{i}}{r \sqrt{r}} \approx 31.3 \end{split}$$

(d)

$$\frac{\widehat{\theta} - \theta}{\widehat{se}_{\widehat{\theta}}} \sim N(0, 1) \Rightarrow P\left(\left|\frac{\widehat{\theta} - \theta}{\widehat{se}_{\widehat{\theta}}}\right| < z_{1 - \frac{\alpha}{2}}\right) = 1 - \alpha$$

$$\Rightarrow \widehat{\theta} - z_{1 - \frac{\alpha}{2}} \widehat{se}_{\widehat{\theta}} < \theta < \widehat{\theta} + z_{1 - \frac{\alpha}{2}} \widehat{se}_{\widehat{\theta}}$$

$$\therefore \text{ The 95\% C.I. of } \theta \text{ is } [21.452, 144.148]$$

(e)

$$F(t_p) = 1 - e^{\frac{t_p}{\theta}} \Rightarrow t_p = -\theta \log(1 - p) :: \widehat{t}_{0.1_{mle}} \approx 8.7$$

$$t_{0.1} = -\theta \log(1 - 0.1) \approx 0.105\theta$$

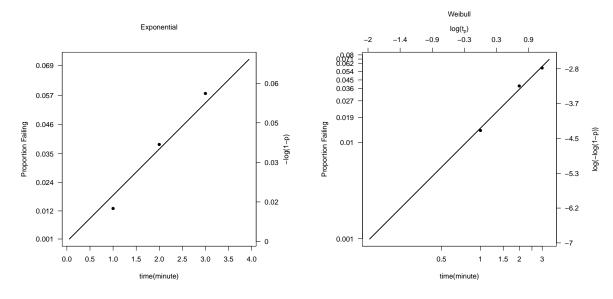
$$0.105\left(\widehat{\theta} - z_{1-\frac{\alpha}{2}}\,\widehat{se}_{\widehat{\theta}}\right) < 0.105\theta < 0.105\left(\widehat{\theta} + z_{1-\frac{\alpha}{2}}\,\widehat{se}_{\widehat{\theta}}\right)$$

$$0.105 \cdot 21.452 < t_{0.1} < 0.105 \cdot 144.148$$

Therefore, [2.252, 15.136] is the 95% C.I. of  $t_{0.1}$ .

8.2

(a) and (b)



(c) Both of the log-likelihood of the two distributions are -54.8, thus there are no difference between them.

8.3

By the equation (8.3), we perform a likelihood-ratio test.

$$H_0: (\beta, \eta) = (\beta_0, \eta_0) \; ; \; H_1: (\beta, \eta) \neq (\beta_0, \eta_0)$$

If

$$-2log\left[\frac{L(\beta_0,\eta_0)}{L(\widehat{\beta}_0,\widehat{\eta}_0)}\right] > \chi^2_{1-\alpha,2}$$

than,  $H_0$  would be rejected. Under the condition of the null hypothesis, the likelihood ratio would be

$$-2log\left\{\frac{\prod_{i=1}^{3}\left[\left(1-e^{-\left(\frac{t_{i}}{\widehat{\eta}^{i}}\right)}\right)-\left(1-e^{-\left(\frac{t_{i-1}}{\widehat{\eta}^{i}}\right)}\right)\right]^{l_{i}}\prod_{j=1}^{3}\left[e^{-\left(\frac{t_{j}}{\widehat{\eta}^{i}}\right)}\right]^{l_{j}}}{\prod_{i=1}^{3}\left[\left(1-e^{-\left(\frac{t_{i}}{\widehat{\eta}}\right)^{\widehat{\beta}}}\right)-\left(1-e^{-\left(\frac{t_{i-1}}{\widehat{\eta}}\right)^{\widehat{\beta}}}\right)\right]^{l_{i}}\prod_{j=1}^{3}\left[e^{-\left(\frac{t_{j}}{\widehat{\eta}}\right)^{\widehat{\beta}}}\right]^{l_{j}}}\right\}=0.7232654$$

, where  $\widehat{\beta}$  and  $\widehat{\eta}$  are the mle of parameters of Weibull distribution and  $\widehat{\eta}$  is the mle of parameters of exponential distribution. Let  $\alpha=0.05$ ,  $\chi^2_{1-\alpha,2}=\chi^2_{0.95,2}=5.9915$ . Hence,  $H_0$  is not rejected.

(a) Based on  $Z_{\widehat{F}(2)} \sim N(0, 1)$ , the 95% C.I. is

$$\left[\widehat{F}(2) \pm z_{1-\frac{0.05}{2}} \widehat{se}_{\widehat{F}(2)}\right].$$

$$\widehat{F}(2) = 1 - e^{-e^{\frac{log2-3.162}{0.743}}} = 0.035$$
, where  $\widehat{\mu}_{mle} = 3.162$  and  $\widehat{\sigma}_{mle} = 0.743$ 

By equation (8.10),

$$\widehat{se}_{\widehat{F}(2)} = \sqrt{\left(\frac{\partial F(2)}{\partial \mu}\right)^2 \widehat{Var}\left(\widehat{\mu}\right) + 2\left(\frac{\partial F(2)}{\partial \mu}\right) \left(\frac{\partial F(2)}{\partial \sigma}\right) \widehat{Cov}\left(\widehat{\mu},\widehat{\sigma}\right) + \left(\frac{\partial F(2)}{\partial \sigma}\right)^2 \widehat{Var}\left(\widehat{\sigma}\right)},$$

 $\widehat{se}_{\widehat{F}(2)} = 0.011$ . Thus, [0.014, 0.056] is the 95% C.I.

(b) Based on  $Z_{logit} \sim N(0,1)$ ,

$$\left[\frac{\widehat{F}(2)}{\widehat{F}(2) + \left(1 - \widehat{F}(2)\right)w}, \frac{\widehat{F}(2)}{\widehat{F}(2) + \frac{1 - \widehat{F}(2)}{w}}\right],$$

where

$$w = e^{z_{1 - \frac{0.05}{2}} \frac{\hat{se}_{\widehat{F}(2)}}{\widehat{F}(2)(1 - \widehat{F}(2))}} = 1.847.$$

Thus, [0.019, 0.063] is the 95% C.I. of F(2).

(c) The confidence interval based on  $Z_{\widehat{F}(2)} \sim N(0,1)$  is thinner than the other on under the same level.

9.3

(a) and (b)

$$\hat{\mu}^* = 10.3901$$
 and  $\hat{\sigma}^* = 0.3460$ 

(c)

$$\hat{\mu} = 10.2299 \text{ and } \hat{\sigma} = 0.3164$$

$$Z_{log(\hat{t}_{0.1}^*)} = \frac{log(\hat{t}_{0.1}^*) - log(\hat{t}_{0.1})}{\widehat{se}_{log(\hat{t}_{0.1}^*)}}$$

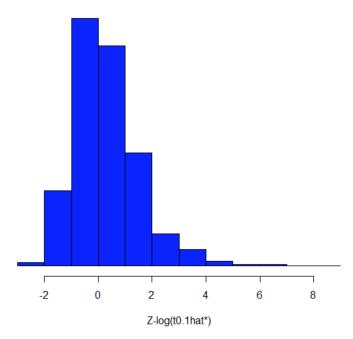
$$\hat{t}_{0.1} = e^{\hat{\mu} + \hat{\sigma} \Phi_{sev}^{-1}(0.1)}$$
 and  $\hat{t}_{0.1}^* = e^{\hat{\mu}^* + \hat{\sigma}^* \Phi_{sev}^{-1}(0.1)}$ 

 $log\left(\widehat{t}_{0.1}\right) = \widehat{\mu} + \widehat{\sigma}\Phi_{sev}^{-1}(0.1) = 9.518 \text{ and } log\left(\widehat{t}_{0.1}^*\right) = \widehat{\mu}^* + \widehat{\sigma}^*\Phi_{sev}^{-1}(0.1) = 9.611$ 

$$\widehat{se}_{log(\widehat{t}_{0.1}^*)} = \frac{\widehat{se}_{\widehat{t}_{0.1}^*}}{\widehat{t}_{0.1}^*}$$

, where  $\widehat{se}_{\widehat{t}^*_{0.1}}$  can be calculated by equation (8.10).

## (d) and (e) Shock Absorber Data (Both Failure Modes) Weibull Distribution Bootstrap-t log-transform



The 95% C.I. of  $t_{0.1}$  is [8472.97,17252.16] which differs from the C.I. in table 9.2 for the condition in this problem based on resampling from the original data.