

EEB 596z, Problem Set 5

Due 28 March 2002

1 : The **Weibul distribution** (after the Swedish physicist Waloddi Weibul, who proposed this distribution in 1939 for the breaking strenght of materials), has density function

$$p(x) = \lambda x^{\lambda-1} \exp(-x^\lambda) \quad \text{for } x, \lambda > 0$$

[As an aside, note that the Weibul arises by assuming  $y = x^\lambda$  follows an exponential distribution,  $p(y) = \theta \exp(-\theta y)$ ].

Suppose we sample  $n$  values  $(x_1, \dots, x_n)$  independently from a Weibul with parameter  $\lambda$ .

- (a) What is the resulting likelihood function  $\ell(\lambda | x_1, \dots, x_n)$ , for  $\lambda$ ?

$$\ell(\lambda | x_1, \dots, x_n) = \lambda^n \left( \prod_{i=1}^n x_i^{\lambda-1} \right) \exp \left( - \sum_{i=1}^n x_i^\lambda \right)$$

- (b) What is the resulting log-likelihood function?

$$\ln(\ell(\lambda | x_1, \dots, x_n)) = n \ln \lambda + (\lambda - 1) \sum_{i=1}^n \ln x_i - \sum_{i=1}^n x_i^\lambda$$

- (c) What is the score function for  $\lambda$ ?

$$S(\lambda) = \frac{\partial \ln(\ell(\lambda | x_1, \dots, x_n))}{\partial \lambda} = \frac{\partial n \ln \lambda + (\lambda - 1) \sum_{i=1}^n \ln x_i - \sum_{i=1}^n x_i^\lambda}{\partial \lambda}$$

Recalling that  $d x^\lambda / d \lambda = x^\lambda \ln(x)$ , which follows since

$$x^\lambda = e^{\ln(x^\lambda)} = e^{\lambda \ln(x)}, \text{ so that } \frac{d}{d \lambda} [e^{\lambda \ln(x)}] = \ln(x) e^{\lambda \ln(x)} = \ln(x) x^\lambda$$

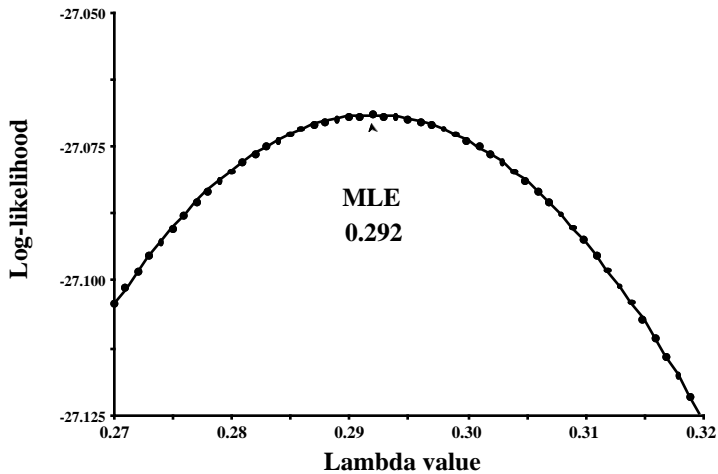
gives

$$= \frac{n}{\lambda} + \sum_{i=1}^n \ln(x_i) - \sum_{i=1}^n x_i^\lambda \ln(x_i)$$

- (d) What is the information (i.e., the derivate of the score function, which is the second derivative of the log-likelihood function)

$$\frac{\partial S(\lambda)}{\partial \lambda} = - \left( \frac{n}{\lambda^2} + \sum_{i=1}^n x_i^\lambda \ln(x_i)^2 \right)$$

- (e) Suppose we observe 5  $x$  values: 12, 15.5, 17.5, 21.3, and 13.5. Plot the log-likelihood as a function of  $\lambda$  (you can use Excel, or whatever program you wish).



- (f) What is the MLE in this case? (You can use the previous graph)

MLE for  $\lambda$  is  $\hat{\lambda} = 0.292$

- (g) Using your results from parts (d) and (f), what is the variance of your MLE estimator?

Recall from the notes (Equation A4.6), that

$$\sigma^2(\hat{\lambda}) = - \left( \frac{\partial^2 \ln(\ell(\lambda | x_1, \dots, x_n))}{\partial \lambda^2} \right)^{-1}$$

when the information (the second derivative) is evaluated at the MLE. Hence,

$$\begin{aligned} \sigma^2(\hat{\lambda}) &= - \left( -\frac{n}{\lambda^2} - \sum_{i=1}^n x_i^\lambda \ln(x_i)^2 \right)^{-1} \\ &= - \left( -\frac{5}{0.292^2} + \sum_{i=1}^n x_i^{0.292} \ln(x_i)^2 \right)^{-1} \\ &= -(-144.357)^{-1} = 0.00692 \end{aligned}$$

Hence, note than an approximate 95% confidence interval for  $\lambda$  is

$$\hat{\lambda} + \pm 1.96 \sqrt{0.00692} = 0.292 \pm 0.163$$

- (h) What is the probability value for the likelihood-ratio test for this data that  $\lambda = 1$ ? For  $\lambda = 4$ ?

The likelihood ratio (LR) test for  $\lambda = 1$  is

$$\text{LR} = 2 \ln \left( \frac{0.292^5 \left( \prod_{i=1}^5 x_i^{0.292-1} \right) \exp \left( -\sum_{i=1}^5 x_i^{0.292} \right)}{1^5 \left( \prod_{i=1}^5 x_i^{1-1} \right) \exp \left( -\sum_{i=1}^5 x_i^1 \right)} \right) = 105.461$$

LT is distributed as a  $\chi^2$  distribution with one degree of freedom.  $\Pr(\chi_1^2 > 105.461) \simeq 0$ .  
The likelihood ratio (LR) test for  $\lambda = 0.4$  gives LR = 1.66.  $\Pr(\chi_1^2 > 1.66) = 0.198$ .