

## Final Solution

Saturday, March 1, 2025 8:20 AM

①  $X_1, X_2, \dots, X_n$  r.s.  $\text{Exp}(\theta)$

$$f(x|\theta) = \frac{1}{\theta} e^{-x/\theta} \quad \begin{array}{l} \theta > 0 \\ x > 0 \end{array}$$

MLE

$$L(\theta) = f(x_1, \dots, x_n | \theta) = \prod_{i=1}^n f(x_i | \theta)$$

$$= \prod_{i=1}^n \frac{1}{\theta} e^{-x_i/\theta}$$

$$= \theta^{-n} e^{-\sum x_i/\theta}$$

$$l(\theta) = \log L(\theta)$$

$$= -n \log \theta - \sum x_i / \theta = -n \log \theta - \sum x_i \cdot \theta^{-1}$$

$$\frac{d}{d\theta} l(\theta) = -\frac{n}{\theta} + \frac{\sum x_i}{\theta^2}$$

$$\frac{d}{d\theta} l(\theta) = 0 \quad \text{find maximum}$$

$$-\frac{n}{\theta} + \frac{\sum x_i}{\theta^2} = 0$$

$$\underline{\sum x_i} = \underline{n}$$

$$\theta^2 = \theta$$

$$\bar{x} = \theta \quad \therefore \hat{\theta} = \bar{x}$$

Fisher's Information:

$$I(\theta) = -n E \left[ \frac{d^2}{d\theta^2} \log f(x|\theta) \right]$$

start with the density

$$f(x|\theta) = \frac{1}{\theta} e^{-x/\theta} = \theta^{-1} e^{-x/\theta}$$

$$\log f(x|\theta) = -\log \theta - \frac{x}{\theta} = -\log \theta - x\theta^{-1}$$

$$\frac{d}{d\theta} \log f(x|\theta) = -\frac{1}{\theta} + \frac{x}{\theta^2} = -\theta^{-1} + x\theta^{-2}$$

$$\frac{d^2}{d\theta^2} \log f(x|\theta) = \theta^{-2} - 2x\theta^{-3}$$

$$E \left[ \frac{d^2}{d\theta^2} \log f(x|\theta) \right] = E \left[ \theta^{-2} - 2x\theta^{-3} \right]$$

$$= \theta^{-2} - 2 E[x] \theta^{-3}$$

$$= \theta^{-2} - 2\theta \theta^{-3}$$

$$= \theta^{-2} - 2\theta^{-2}$$

$$= \frac{1}{\theta^2} - \frac{2}{\theta^2}$$

$$= -\frac{1}{\theta^2}$$

$$I_X(\theta) = -E\left[-\frac{1}{\theta^2}\right] = \frac{1}{\theta^2}$$

single X

AV of the MLE

$$AV(\hat{\theta}) = \frac{1}{n I_X(\theta)} = \frac{\theta^2}{n}$$

95% CI for  $\theta$

$$\hat{\theta} \pm z_{\alpha/2} \sqrt{AV(\hat{\theta})}$$

$$\hat{\theta} \pm z_{\alpha/2} \sqrt{\frac{\hat{\theta}^2}{n}}$$

$$\hat{\theta} \pm z_{\alpha/2} \frac{\hat{\theta}}{\sqrt{n}}$$

(2)  $X_1, X_2, \dots, X_n$  iid Pareto distribution

✓ with shape parameter  $\alpha$

$$f(x|\alpha) = \alpha 2^\alpha x^{-(\alpha+1)} \quad \alpha > 0$$

$$x > 2$$

a) joint density of  $X_1, \dots, X_n$

$$L(\alpha) = f(x_1, \dots, x_n | \alpha) = \prod_{i=1}^n f(x_i | \alpha)$$

$$= \prod_{i=1}^n \alpha 2^\alpha x_i^{-(\alpha+1)}$$

$$= \alpha^n 2^{n\alpha} \left( \prod_{i=1}^n x_i \right)^{-(\alpha+1)}$$



b) MLE

$$l(\alpha) = \log L(\alpha)$$

$$= n \log(\alpha) + n\alpha \log(2)$$

$$- (\alpha+1) \log \left( \prod_{i=1}^n x_i \right)$$

$$= n \log(\alpha) + n\alpha \log(2)$$

$$\begin{aligned}
 & -(\alpha+1) \sum_{i=1}^n \log(x_i) \\
 = & n \log(\alpha) + n\alpha \log(2) \\
 & - \alpha \sum \log(x_i) - \sum \log(x_i)
 \end{aligned}$$

$$\frac{d}{d\alpha} l(\alpha) = \frac{n}{\alpha} + n \log(2) - \sum \log(x_i)$$

$$\frac{d}{d\alpha} l(\alpha) = 0 \quad \text{find maximum}$$

$$\frac{n}{\alpha} + n \log(2) - \sum \log(x_i) = 0$$

$$\frac{1}{n} \cdot \frac{n}{\alpha} = \left( \sum \log(x_i) - n \log(2) \right) \frac{1}{n}$$

$$\frac{1}{\alpha} = \frac{1}{n} \left( \sum \log(x_i) - n \log(2) \right)$$

$$\hat{\alpha} = n \left( \sum \log(x_i) - n \log(2) \right)^{-1} \quad \checkmark$$

c) Fisher's Information.

$f(x|\alpha)$  density

$$f(x|\alpha) = \alpha 2^\alpha x^{-(\alpha+1)}$$

$$\begin{aligned} \log f(x|\alpha) &= \log(\alpha) + \alpha \log(2) \\ &\quad - (\alpha+1) \log(x) \\ &= \log(\alpha) + \alpha \log(2) \\ &\quad - \alpha \log(x) - \log(x) \end{aligned}$$

$$\frac{d}{d\alpha} \log f(x|\alpha) = \frac{1}{\alpha} + \log(2)$$

$$\frac{d}{d\alpha} \log f(x|\alpha) = -\log(x)$$

$$\frac{d^2}{d\alpha^2} \log f(x|\alpha) = -\frac{1}{\alpha^2}$$

$$\begin{aligned} I(\alpha) &= -E\left[\frac{d^2}{d\alpha^2} \log f(x|\alpha)\right] \\ &= -E\left[-\frac{1}{\alpha^2}\right] = \frac{1}{\alpha^2} \end{aligned}$$

AV

$$AV\left(\frac{1}{\alpha}\right) = \frac{1}{n I(\alpha)} = \frac{\alpha^2}{n}$$

d) approximate 95% CI for  $\alpha$

$$\hat{\alpha} \pm z_{\alpha/2} \sqrt{\widehat{AV}(\hat{\alpha})}$$

$$\hat{\alpha} \pm z_{\alpha/2} \sqrt{\frac{\hat{\alpha}^2}{n}}$$

$$\hat{\alpha} \pm z_{\alpha/2} \frac{\hat{\alpha}}{\sqrt{n}}$$

(3)

$X_1, \dots, X_n$  i.i.d. Pareto distribution

$$f(x|\alpha) = \alpha 2^\alpha x_i^{-(\alpha+1)} \quad \begin{array}{l} \alpha > 0 \\ x > 2 \end{array}$$

a)  $H_0: \alpha = \alpha_0$   
 $H_1: \alpha = \alpha_1 \quad \alpha_0 < \alpha_1$

N.P. Lemma

$$\Lambda = \frac{L(\alpha_0)}{L(\alpha_1)} = \frac{\alpha_0^n 2^{n\alpha_0} (\prod x_i)^{-(\alpha_0+1)}}{\alpha_1^n 2^{n\alpha_1} (\prod x_i)^{-(\alpha_1+1)}}$$

$$= \left(\frac{\alpha_0}{\alpha_1}\right)^n 2^{n(\alpha_0 - \alpha_1)} (\prod x_i)^{-(\alpha_0+1) + (\alpha_1+1)}$$

$$= \left(\frac{\alpha_0}{\alpha_1}\right)^n 2^{n(\alpha_0 - \alpha_1)} (\prod x_i)^{\alpha_1 - \alpha_0}$$

Reject  $H_0$   $\Delta < k$

$$\left(\frac{\alpha_0}{\alpha_1}\right)^n 2^{n(\alpha_0 - \alpha_1)} (\prod x_i)^{\alpha_1 - \alpha_0} < k$$

since  $\frac{\alpha_0}{\alpha_1} < 1$   $\Leftrightarrow (\prod x_i)^{\alpha_1 - \alpha_0} < k'$

but  $\left(\frac{\alpha_0}{\alpha_1}\right)^n > 0$ ,

since  $\alpha_1 - \alpha_0 > 0$

and

$$2^{n(\alpha_0 - \alpha_1)} < 1$$

but  $2^{n(\alpha_0 - \alpha_1)} > 0$

$$\Leftrightarrow \prod x_i < k''$$



$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } \prod x_i < k \\ 0 & \text{otherwise} \end{cases}$$

b) Yes

For the test to be uniformly most powerful it must have MLR in the statistic.

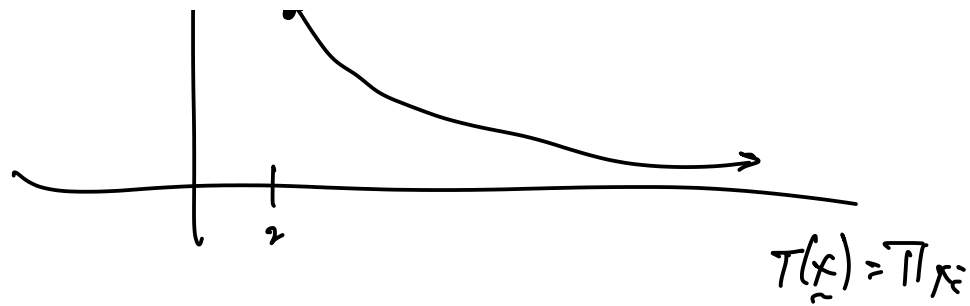
$$\alpha_1 < \alpha_2$$

$$\frac{L(\alpha_2)}{L(\alpha_1)} = \left(\frac{\alpha_2}{\alpha_1}\right)^n 2^{n(\alpha_2 - \alpha_1)} \left(\prod x_i\right)^{-(\alpha_2 - \alpha_1)}$$

$$T(\underline{x}) = \prod x_i$$

Since  $\alpha_2 - \alpha_1 > 0$

Then  $-(\alpha_2 - \alpha_1) < 0$



monotone decreasing

So answer Yes

$$\phi(\underline{x}) = \begin{cases} 1 & \text{if } \prod x_i \leq c \\ 0 & \text{otherwise} \end{cases} \quad \text{UMP}$$

for  $H_0: \alpha = \alpha_0$

$H_1: \alpha > \alpha_0$

(4)

$X_1, X_2, \dots, X_n$  Pareto distribution  
Shape  $\alpha$ , scale  $\theta$

$$f(x | \alpha, \theta) = \alpha \theta^\alpha (x + \theta)^{-(\alpha+1)} \quad \begin{matrix} \alpha > 0 \\ \theta > 0 \end{matrix}$$

$$x > 0$$

$$a) \quad \hat{\alpha} = 10.8783$$

$$\hat{\theta} = 2.2532$$

b) approximate 95% CI for  $\alpha$

$$\hat{\alpha} \pm 1.96 \text{ se}_{\hat{\alpha}}$$

4.45

$$10.8783 \pm 1.96 (2.27)$$

$$(6.43, 15.33)$$

approximate 95% CI for  $\theta$

$$\hat{\theta} \pm 1.96 \text{ se}_{\hat{\theta}}$$

$$2.2532 \pm 1.96 (.513)$$

^ - - - ^

( 1.25 , 3.25 )

c. Exfra Credit.

$H_0: \alpha = 10$  Fail to Reject  $H_0$

$H_0: \theta = 2$  Fail to Reject  $H_0$

both CIs contain the  
null value, so Fail to Reject  
each null hypothesis.

