

**Answer Key 12**

1.

$$(1) \quad L(y_1, y_2, \dots, y_n : \theta) = \prod_{i=1}^n f(y_i; \theta) = \theta^{-n} \exp\left(-\frac{1}{\theta} \sum_{i=1}^n y_i\right)$$

$$\hat{\theta} = \operatorname{argmax}_{\theta} \ln L(y_1, y_2, \dots, y_n : \theta)$$

$$= \operatorname{argmax}_{\theta} \left( -n \ln \theta - \frac{1}{\theta} \sum_{i=1}^n y_i \right)$$

(2)

$$F.O.C.: \frac{d \ln L}{d\theta} = -\frac{n}{\theta} + \frac{\sum_{i=1}^n y_i}{\theta^2} = 0$$

$$\Rightarrow \hat{\theta} = \bar{y}$$

(3) First, find  $E(y_i)$  and  $E(y_i^2)$ .

$$\text{Since } \ln f(y_i; \theta) = -\ln \theta - \frac{y_i}{\theta}$$

$$\frac{d \ln f(y_i; \theta)}{d\theta} = -\frac{1}{\theta} + \frac{y_i}{\theta^2} \quad \text{and} \quad \frac{d^2 \ln f(y_i; \theta)}{d\theta^2} = \frac{1}{\theta^2} - 2 \frac{y_i}{\theta^3}.$$

$$E\left(\frac{d^2 \ln f(y_i; \theta)}{d\theta^2}\right) = \frac{1}{\theta^2} - 2 \frac{E(y_i)}{\theta^3} = -\frac{1}{\theta^2}$$

$$\text{Or } \left( \begin{aligned} E\left(\frac{d \ln f(y_i; \theta)}{d\theta}\right)^2 &= E\left(-\frac{1}{\theta} + \frac{y_i}{\theta^2}\right) = E\left(\frac{-\theta + y_i}{\theta^2}\right)^2 \\ &= \frac{1}{\theta^4} (\theta^2 - 2\theta E(y_i) + E(y_i^2)) \\ &= \frac{1}{\theta^2} \end{aligned} \right)$$

Therefore,  $\sqrt{n}(\hat{\theta} - \theta_0) \xrightarrow{d} N(0, \theta_0^2)$  or  $\hat{\theta} \overset{a}{\sim} N(\theta_0, \frac{1}{n} \theta_0^2)$

(4) From exponential distribution,  $E(y) = \theta_0$ ,  $E(\bar{y}) = E(\hat{\theta}) = \theta_0$ .

(5) From (2),  $\frac{d^2 \ln L}{d\theta^2} = \frac{n}{\theta^2} - 2 \frac{\sum_i y_i}{\theta^3},$

Therefore,  $\frac{d^2 \ln L(\hat{\theta})}{d\theta^2} = -\frac{n}{\bar{y}^2} < 0.$

2.

(1) Done in the class.  $\hat{\theta} = \bar{X}.$

(2) Since  $\bar{X} \xrightarrow{p} \theta$ ,  $E(\bar{X}) = \theta$ , and  $V(\bar{X}) = \theta(1-\theta)$ ,

$$\hat{\theta} \overset{a}{\sim} N\left(\theta, \frac{1}{n}\theta(1-\theta)\right).$$

(3) Natural estimator for  $\gamma$ ,  $\hat{\gamma} = \frac{1}{\hat{\theta}}.$

(4) By delta method and  $\frac{d\gamma}{d\theta} = -\theta^{-2}$ , therefore  $\hat{\gamma} \overset{a}{\sim} N\left(\gamma, \frac{1}{n}\theta(1-\theta)\theta^{-4}\right)$

Or  $\sqrt{n}(\hat{\gamma} - \gamma) \xrightarrow{d} N(0, \theta(1-\theta)\theta^{-4}).$

Note that consistent estimator of  $AV.(\hat{\gamma}) = \frac{1}{n}(1-\hat{\theta})\hat{\theta}^{-3}$