

QUIZ 6 SOLUTIONS

EXERCISE 1 :

$$P(D_2 | H_3) = \frac{P(D_2 \cap H_3)}{P(H_3)} = \frac{P(D_2 \cap H_3)}{P(H_3 \cap D_1) + P(H_3 \cap D_2)}$$

$$= \frac{P(H_3 | D_2) P(D_2)}{P(H_3 | D_1) P(D_1) + P(H_3 | D_2) P(D_2)}$$

$$= \frac{1 \cdot \frac{1}{3}}{\frac{1}{2} \cdot \frac{1}{3} + 1 \cdot \frac{1}{3}} = \frac{2}{5}$$

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$$\text{And } P(D_1 | H_3) = \frac{\frac{1}{2} \cdot \frac{1}{3}}{\frac{1}{2} \cdot \frac{1}{3} + 1 \cdot \frac{1}{3}} = \frac{1}{5}$$

EXERCISE 2 :

$$X_i \sim \exp\left(\frac{1}{i\theta}\right) \rightarrow M_{X_i}(t) = (1 - i\theta t)^{-1}$$

$$\text{Let } \hat{\theta} = \sum_{i=1}^n \frac{X_i}{ni}$$

$$M_{\hat{\theta}}(t) = M_{\frac{X_1}{n}}(t) \cdot M_{\frac{X_2}{2n}}(t) \cdots M_{\frac{X_n}{n^2}}(t)$$

$$= \Gamma_{x_1} \left(\frac{t}{n} \right) \cdot \Gamma_{x_2} \left(\frac{t}{n^2} \right) \dots \Gamma_{x_n} \left(\frac{t}{n^n} \right)$$

$$= \left(1 - \theta \frac{t}{n} \right)' \left(1 - 2\theta \frac{t}{n^2} \right)' \dots \left(1 - n\theta \frac{t}{n^n} \right)'$$

$$= \left(1 - \frac{\theta}{n} t \right)^n$$

$$\therefore \hat{\theta} \sim T \left(n, \frac{\theta}{n} \right)$$

(b) $E \hat{\theta}^{-1} = \frac{\Gamma(n) \left(\frac{\theta}{n} \right)^{-1}}{\Gamma(n)} \approx \frac{1}{n-1} \theta$

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(c) $MSE(\hat{\theta}^{-1}) = \text{Var}(c \hat{\theta}^{-1}) + B^2 = *$

$$B = E \hat{\theta}^{-1} - \frac{1}{\theta} = \frac{1}{\theta(n-1)}$$

$$* = c^2 \left\{ E(\hat{\theta}^{-2}) - (E \hat{\theta}^{-1})^2 \right\} + \frac{1}{\theta^2(n-1)^2}$$

$$= c^2 \left(\frac{\Gamma(n-2) \left(\frac{\theta}{n} \right)^{-2}}{\Gamma(n)} - \left(\frac{n}{n-1} \frac{1}{\theta} \right)^2 \right) + \frac{1}{\theta^2(n-1)^2}$$

Then $\frac{\partial MSE}{\partial c} = 0$ AND SOLVE FOR C.

EXERCISE 3 :

$$(a), E\hat{\theta} = \frac{1}{n} E \sum (X_i^2 - X_i)$$

$$= \frac{1}{n} \left(\sum E X_i^2 - \sum E X_i \right)$$

$$E X_i = 1$$

$$\text{var}(X_i) = 1$$

$$= \frac{1}{n} \left(n(\sigma^2 + \mu^2) - n\mu \right)$$

$$= \frac{1}{n} (n + n - n) = 1.$$

$$(b) f(X_i) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2\sigma^2} X_i^2} = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2\sigma^2} X_i^2}$$

$$\ln f(X_i) = -\frac{1}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} X_i^2$$

$$S = \sum \ln f(X_i) = -\frac{1}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \sum X_i^2$$

$$\text{var}(S) = \text{var} \left(-\frac{1}{2} \ln 2\pi\sigma^2 - \frac{1}{2\sigma^2} \sum X_i^2 \right)$$

$$= \text{var} \left(\frac{1}{2\sigma^2} \sum X_i^2 \right) = \frac{1}{4\sigma^4} \sum \text{var}(X_i^2) = \frac{1}{4\sigma^4} \cdot 2 = \frac{1}{2\sigma^4}.$$

USING SECOND DERIVATIVE:

$$\frac{\partial^2 \ln f(X)}{\partial \sigma^2 \partial \sigma^2} = +\frac{1}{2\sigma^4} - \frac{1}{\sigma^6} X_i^2$$

$$\begin{aligned}
 I(\theta) &= -E \left(\frac{1}{2\sigma^4} - \frac{1}{\sigma^6} X_i^2 \right) \\
 &= - \left(\frac{1}{2\sigma^4} - \frac{E X_i^2}{\sigma^6} \right) = - \left(\frac{1}{2\sigma^4} - \frac{\sigma^2}{\sigma^6} \right) \\
 &= \frac{1}{2\sigma^4}
 \end{aligned}$$

THUS THE CRAMER-RAO LOWER BOUND IS $\frac{1}{n I(\theta)} = \frac{2\sigma^4}{n}$.

Assignment Project Exam Help
 OUR ESTIMATOR (1) $\hat{\sigma}^2 = \frac{1}{n} \sum X_i^2$

$$E \hat{\sigma}^2 = E \frac{1}{n} \sum X_i^2 = \frac{1}{n} \sum E X_i^2 = \sigma^2.$$

$$\text{AND } \text{VAR}(\hat{\sigma}^2) = \text{VAR} \left(\frac{1}{n} \sum \frac{X_i^2}{\sigma^2} \right) = \frac{2\sigma^4}{n}.$$

\therefore YES, $\hat{\sigma}^2$ IS EFFICIENT ESTIMATOR OF σ^2 .

NOTE: HERE WE USE $\frac{\sum X_i^2}{\sigma^2} \sim \chi_n^2$
 AND THEREFORE $\text{VAR} \left(\frac{\sum X_i^2}{\sigma^2} \right) = 2n$.

(c). X_1, \dots, X_n iid $N(\mu, \sigma)$

$$E \bar{X} = \mu, \quad \text{VAR}(\bar{X}) = \frac{\sigma^2}{n}$$

$$\text{As } n \rightarrow \infty \quad \text{VAR}(\bar{X}) \rightarrow 0$$

$\therefore \bar{X}$ is consistent estimator of μ .

X_1, \dots, X_n are not independent:

$$E X_i = \mu \quad \text{AND} \quad E \bar{X} = \mu.$$

$$\text{BUT } \text{VAR}(\bar{X}) = \text{VAR}\left(\frac{\sum X_i}{n}\right)$$

$$= \frac{1}{n^2} \text{VAR}\left(\sum X_i\right) = \frac{1}{n^2} \left(n\sigma^2 + \left(\frac{n}{2}\right) 2\rho\sigma^2 \right)$$

$$= \frac{1}{n^2} \left(n\sigma^2 + n(n-1)\rho\sigma^2 \right)$$

$$= \frac{\sigma^2}{n} \left[1 + (n-1)\rho \right]$$

$$\text{As } n \rightarrow \infty \quad \text{VAR}(\bar{X}) \rightarrow \rho\sigma^2$$

$\therefore \bar{X}$ is not consistent estimator of μ .