

Ex 7.6:

Let  $X_1, \dots, X_n \stackrel{iid}{\sim} f(x_i|\theta) = \theta x_i^{-2}$ ;  $0 < \theta \leq x_i < \infty$ ,  $i=1, 2, \dots, n$

$$f(x|\theta) = \prod_{i=1}^n f(x_i|\theta) = \prod_{i=1}^n \theta x_i^{-2} \cdot I_{[\theta, \infty)}(x_i) = \theta^n \cdot I_{\{\min x_i \geq \theta, \forall i\}} \cdot \prod_{i=1}^n x_i^{-2} = g(T(x)|\theta) \cdot h(x)$$

$T(X) = \min X_i$ . By the Factorization theorem,  $T(X) = \min X_i$  is a sufficient statistic for  $\theta$ .

∴  $g(\theta) = \prod_{i=1}^n \theta x_i^{-2} \cdot I_{[\theta, \infty)}(x_i) = \theta^n \cdot I_{\{\min x_i \geq \theta, \forall i\}} \cdot \prod_{i=1}^n x_i^{-2}$

$g(\theta)$  is an increasing function in  $\{\min x_i > \theta > 0\}$

Therefore the MLE of  $\theta$  is  $\hat{\theta} = \min X_i$ .

∴  $E(X) = \int_{\theta}^{\infty} x \cdot \theta \cdot x^{-2} dx = \theta \cdot \int_{\theta}^{\infty} x^{-1} dx = \theta \cdot [\ln x]_{\theta}^{\infty}$

∴  $E(X)$  does not exist. And M.M.E does not exist either.

Ex 7.8:

$X \sim N(0, \sigma^2)$ ,  $\sigma^2 > 0$ ,  $-\infty < X < \infty$

∴

$\text{Var}(X) = E(X^2) - (E(X))^2 = E(X^2) - 0 = E(X^2) = \sigma^2$

Therefore  $X^2$  is an unbiased estimator of  $\sigma^2$

∴  $g(\sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{X^2}{2\sigma^2}\right)$

$\log g(\sigma) = \frac{1}{2} \log 2\pi - \log \sigma - \frac{X^2}{2\sigma^2}$

$\frac{d}{d\sigma} \log g(\sigma) = -\frac{1}{\sigma} + \frac{X^2}{\sigma^3} \stackrel{\text{let}}{=} 0$

$\Rightarrow \hat{\sigma}^2 = X^2 \Rightarrow \hat{\sigma} = |X| \quad \therefore \sigma > 0$

$\frac{d^2}{d\sigma^2} \log g(\sigma) = \frac{1}{\sigma^2} - \frac{3X^2}{\sigma^4} \Big|_{\sigma=|X|} = \frac{1}{X^2} - \frac{3X^2}{X^4} = \frac{-2}{X^2} < 0$  for all  $X$

Therefore the MLE of  $\sigma$  is  $\hat{\sigma} = |X|$

∴

$E(X) = 0$

$E(X^2) = 0 + \sigma^2$

$\Rightarrow \hat{\sigma}^2 = X^2 \Rightarrow \hat{\sigma} = |X| \quad (\because \sigma > 0)$

Therefore the M.M.E of  $\sigma$  is  $\hat{\sigma} = |X|$

