

Seat No. : \_\_\_\_\_

**NS-124**  
**December-2015**  
**M.Sc., Sem.-I**  
**403 : Statistics**  
**(Estimation Theory)**

**Time : 3 Hours]**

**[Max. Marks : 70**

**Instruction : All questions carry equal marks.**

1. (a) Define minimal sufficient statistic. State & prove Lehmann-Scheffe theorem on minimal sufficient statistic. Show that for Cauchy distribution with location parameter  $\theta$ , no nontrivial minimal sufficient statistics exist.

**OR**

Define complete sufficient statistic. Discuss complete sufficient statistic in exponential family of distributions. Hence deduce complete sufficient statistic for  $(\mu, \sigma^2)$  in case of normal distribution  $N(\mu, \sigma^2)$  based on a random sample of size  $n$ .

- (b) Let  $X_1, X_2, \dots, X_n$  be a random sample from negative binomial distribution with parameters  $1$  and  $p$ ,  $0 < p < 1$ . Obtain minimal sufficient statistics for  $p$ . Check the completeness property of the statistic obtained by you.

**OR**

Let  $X_1$  and  $X_2$  be iid with discrete uniform distribution given by  $P(X=x) = 1/N$ ,  $x = 1, 2, \dots, N$ ,  $N \geq 1$ . Obtain minimal sufficient statistic for  $N$ . Check whether the statistic obtained is complete.

2. (a) State and prove Cramer-Rao inequality for the variance of unbiased estimator. Hence deduce it for MSE of biased estimator.

**OR**

State and prove Rao-Blackwell theorem. Let  $X_1, X_2, \dots, X_n$  be a random sample from Poisson distribution with mean  $\lambda$ ,  $\lambda > 0$ . Obtain UMVUE for  $e^{-\lambda}(1 + \lambda)$ .

- (b) State and prove necessary and sufficient condition for unbiased estimator  $T$  for  $g(\theta)$  to have a minimum variance at the value  $\theta = \theta_0$ . Hence deduce that the correlation coefficient between MVUE and any unbiased estimator is non-negative.

**OR**

Let  $X_1, X_2, \dots, X_n$  be a random sample from uniform distribution  $U(\alpha, \beta)$ ,  $\alpha < \beta$ . Obtain UMVUE for  $\alpha$ ,  $\beta$  and  $\beta - \alpha$ .

3. (a) Prove or disprove: MLE is consistent. Check the property in case of normal distribution with mean  $\mu$ , and variance 1 based on a random sample of size  $n$ .

**OR**

Let  $X_1, X_2, \dots, X_n$  be a random sample from the distribution given below. Obtain MLE of  $\alpha$ . What do you conclude from your answer ?

$$f(x; \alpha, \beta) = \begin{cases} \frac{2x}{\alpha\theta}, & 0 \leq x \leq \theta \\ \frac{2(\alpha - x)}{\alpha(\alpha - \theta)}, & \theta \leq x \leq \alpha \end{cases}$$

- (b) Show that MLE is asymptotically normally distributed statistic. Let  $X_1, X_2, \dots, X_n$  be a random sample from exponential distribution with mean  $\theta > 0$ . Obtain mle of  $\theta$ . Obtain its asymptotic distribution.

**OR**

Let  $X_1, X_2, \dots, X_n$  be a random sample from the exponential distribution with location parameter  $\mu$  and scale parameter  $\theta$ . Obtain MLE of the parameters. Hence obtain MLE of  $\mu + 3\theta$ ,  $\theta > 0$ .

4. (a) What is  $(1 - \alpha)100\%$  confidence interval estimation ? State its interpretation. How it differs from the point estimation ? Discuss with illustration.

**OR**

Discuss pivotal method of obtaining confidence interval with illustration.

- (b) Let  $X_1, X_2, \dots, X_n$  be a random sample from the exponential distribution with mean  $1/\theta$ ,  $\theta > 0$ . Construct 95% confidence interval for  $\theta$ . Obtain expected length of the confidence interval.

**OR**

Discuss a general method of obtaining confidence interval with illustration.

5. Suggest correct single answer in the following :

(i) If  $T$  be any sufficient statistic for parameter  $\theta$  then

$$L(x, \theta) = \underline{\hspace{2cm}}.$$

- (a)  $g(t, \theta)h(x)$  (b)  $g(\theta)h(t, x)$   
 (c)  $g(t)h(x, \theta)$  (d) none of these

(ii) Let  $X_1, X_2, \dots, X_n$  be a random sample from  $U(0, \theta)$ ,  $\theta > 0$  distribution. And  $T = X_{(n)}$ . Then UMVUE of  $\theta^2$  is

- (a)  $\frac{3T^2}{n}$  (b)  $\frac{n+2}{n} T^2$   
 (c)  $\frac{n}{n+2} T^2$  (d)  $\frac{n}{n+1} T^2$

(iii) Let  $X_1, X_2, \dots, X_n$  be a random sample from the pdf  $f(x; \theta) = (\pi[1 + (x - \theta)^2])^{-1}$ ,  $-\infty < x < \infty$ . Define  $T_1 = X_{(n)} - X_{(1)}$  and  $T_2 = X_{(n)} + X_{(1)}$ . Then

- (a)  $T_1$  is ancillary and  $T_2$  is not a sufficient for  $\theta$   
 (b)  $T_2$  is ancillary and  $T_1$  is sufficient for  $\theta$   
 (c)  $T_1$  is minimal sufficient and  $T_2$  is complete statistic for  $\theta$   
 (d)  $T_1$  and  $T_2$  both are sufficient statistic for  $\theta$

(iv) Let  $X_1, X_2, \dots, X_n$  be a random sample from  $N(\theta, \theta)$  normal distribution and

$$S = \sum_{i=1}^n (x_i - \bar{x})^2. \text{ Then } S \text{ is unbiased estimator of}$$

- (a)  $n\theta$  (b)  $n\theta^2$   
 (c)  $(n-1)\theta$  (d)  $(n-1)\theta^2$

(v) Let  $X_1, X_2, \dots, X_n$  be a random sample from the pdf  $f(x; \theta) = \theta^2 x e^{-\theta x^2}$ ,  $x > 0$ . An unbiased estimator of  $\theta$  is

- (a)  $\frac{X_1 + X_2 + X_3}{3}$  (b)  $\frac{1}{3} \left( \frac{1}{X_1} + \frac{1}{X_2} + \frac{1}{X_3} \right)$   
 (c)  $\left( \frac{X_1 + X_2 + X_3}{3} \right)^{-1}$  (d)  $3 \left( \frac{1}{X_1} + \frac{1}{X_2} + \frac{1}{X_3} \right)^{-1}$

- (vi) If  $T$  is an unbiased and sufficient for  $\theta$  then  $\text{Log}T$  for  $\text{Log}\theta$  is
- unbiased and sufficient both
  - unbiased only
  - Neither unbiased nor sufficient
  - sufficient only
- (vii) Let  $X_1, X_2, \dots, X_n$  be a random sample from  $U(0, \theta)$ ,  $\theta > 0$  uniform distribution  
Let  $T = 2\bar{X}$  be an estimator of  $\theta$ . Then mean square error of  $T$  is
- $\theta^2/12$
  - $\theta^2/3$
  - $\theta^2/12n$
  - $\theta^2/3n$
- (viii) Say True or False : MLE is always better than any other estimator.
- (ix) Let  $X_1, X_2, \dots, X_n$  be a random sample from the  $U(0, \theta)$ ,  $\theta > 0$  uniform distribution. State the UMVUE of  $\log\theta$ .
- (x) Let  $X_1, X_2, \dots, X_n$  be a random sample from  $N(\mu, \sigma^2)$  distribution,  $\mu$  and  $\sigma^2$  both are unknown. Which one is not a statistic ? Here  $S^2 = \sum_{i=1}^n (x_{(i)} - \bar{x})^2$ .
- $\sum_{i=1}^n (x_i - \mu)^2$
  - $\sum_{i=1}^n (x_i - \bar{x})^2 / n$
  - $\sum_{i=1}^n (x_i - S)^2 / n$
  - $\bar{x} / S^2$
- (xi) Define Fisher information contained in statistic.
- (xii) Define ancillary statistic.
- (xiii) State Basu's theorem.
- (xiv) Give an example of non unique MLE.
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