

**MTH-418: Inference-I**  
**2023-2024: II Semester**  
**End Semester Examination**

Time Allowed: 3 Hours

Maximum Marks: 100

NOTE: (i) This question paper has 8 questions, to be attempted in 3 hours. Attempt all the 8 questions.

(ii) Start answer of every question on a new page. Moreover, attempt all the parts of a question at one place.

(iii) Answer each question legibly, clearly and concisely. Illegible answers will not be graded.

(iv) For each question, provide details of all the steps involved in arriving at the final answer(s)/conclusion(s) and box your final answer(s)/conclusion(s).

1. Let  $X_1, X_2$  be a random sample from a population having a pdf

$$f_{\theta}(x) = \begin{cases} \frac{2x}{\theta^2}, & \text{if } 0 < x < \theta \\ 0, & \text{otherwise} \end{cases},$$

where  $\theta \in [1, \infty) = \Theta$ , say, is an unknown parameter (**Note:** The parameter space is restricted).

(i) Show that  $X_{(2)} = \max\{X_1, X_2\}$  is **NOT** a complete statistic;

(ii) Find the UMVUE of  $\theta$ .

6+6=12 Marks

2. Let  $X_1, \dots, X_5$  be a random sample from a  $\text{Bin}(1, \theta)$  distribution, where  $\theta \in (0, 1) = \Theta$ . Let  $\delta_0(\underline{X}) = X_1 X_2$ , so that  $E_{\theta}(\delta_0(\underline{X})) = \theta^2, \theta \in \Theta$ .

(i) Show that  $T_1 = \sum_{i=1}^5 X_i$  is a minimal sufficient statistic and hence show that  $T_2 = (X_1 + X_2 + X_3, X_4 + X_5)$  is a sufficient statistic;

(ii) Using the Rao-Blackwell Theorem on  $\delta_0$  with sufficient statistic  $T_2$ , find an unbiased estimator  $\delta_1$  of  $\theta^2$  that has uniformly smaller variance than the variance of the estimator  $\delta_0$ . Is  $\delta_1$  the UMVUE?;

(iii) Using the Rao-Blackwell Theorem on  $\delta_1$  (obtained in (i) above) with minimal sufficient statistic  $T_1$ , find an unbiased estimator  $\delta_2$  of  $\theta^2$  that has uniformly smaller variance than the variance of the estimator  $\delta_1$ . Is  $\delta_2$  the UMVUE?

4+5+5=14 Marks

3. Let  $X_1, X_2, X_3$  be a random sample from an exponential distribution having the p.d.f.

$$f_{\theta}(x) = \begin{cases} \theta e^{-\theta x}, & \text{if } x > 0 \\ 0, & \text{otherwise} \end{cases},$$

where  $\theta \in (0, \infty) = \Theta$ .

6+6=12 Marks

- (i) Find the UMVUEs  $\delta_0, \delta_1$  and  $\delta_2$  of the estimands  $\psi_1(\theta) = \frac{1}{\theta}, \psi_2(\theta) = \theta$  and  $\psi_3(\theta) = e^{-\theta}$  (**Hint:**  $\delta(\underline{X}) = I(X_1 > 1)$  is an unbiased estimator of  $e^{-\theta}$ );
- (ii) Using the generalized Rao-Cramer bound, find sharp bounds on the variances of unbiased estimators of  $\psi_1(\theta), \psi_2(\theta)$  and  $\psi_3(\theta)$ .
4. (a) Let  $X_1, X_2$  be a random sample from a population having p.d.f.

$$f_{\theta}(x) = \begin{cases} \frac{2}{\theta^2}(\theta - x), & \text{if } 0 < x < \theta \\ 0, & \text{otherwise} \end{cases},$$

where  $\theta \in (0, \infty) = \Theta$ . Find the MME and the MLE of  $\theta$ .

4+8=12 Marks

5. Let  $X_1, X_2$  be a random sample from  $U(0, \theta)$ , where  $\theta \in \Theta = \{1, \frac{3}{2}, 2\}$ . For testing  $H_0 : \theta \in \{1, \frac{3}{2}\}$  against  $H_1 : \theta = 2$ , consider the test function

$$\phi(\underline{X}) = \begin{cases} 1, & \text{if } X_1 + X_2 > \frac{7}{4} \\ 0, & \text{otherwise} \end{cases}.$$

Find the size and the power of the test.

8+4=12 Marks

6. Let  $X_1, \dots, X_n$  ( $n \geq 2$ ) be a random sample from an  $U(\theta, \theta + 1)$  distribution, where  $\theta \in \mathbb{R} = \Theta$ . Find an  $UMP(\alpha)$  test for testing  $H_0 : \theta \leq 0$  against  $H_1 : \theta > 0$ ; here  $\alpha \in (0, 1)$ . .
7. Let  $X_1, \dots, X_n$  be a random sample from  $N(\theta, \sigma^2)$ , where  $\theta \in \mathbb{R}$  and  $\sigma > 0$  are both unknown. For testing  $H_0 : \theta = 0$  against  $H_1 : \theta \neq 0$ , find an  $UMPU(\alpha)$  test ( $\alpha \in (0, 1)$ ).
8. Let  $X_1, \dots, X_m$  and  $Y_1, \dots, Y_n$  be independent random samples from populations having p.d.f.s

$$f_{\theta_1}(x) = \begin{cases} \frac{1}{\theta_1} e^{-\frac{x}{\theta_1}}, & \text{if } x > 0 \\ 0, & \text{otherwise} \end{cases} \quad \text{and} \quad f_{\theta_2}(x) = \begin{cases} \frac{1}{\theta_2} e^{-\frac{x}{\theta_2}}, & \text{if } x > 0 \\ 0, & \text{otherwise} \end{cases},$$

respectively, where  $(\theta_1, \theta_2) \in (0, \infty) \times (0, \infty) = \Theta$ . For testing  $H_0 : \theta_1 = \theta_2$  against  $\theta_1 \neq \theta_2$  at  $\alpha \in (0, 1)$  of significance, find an  $UMPU(\alpha)$  test.

12 Marks

Course No. MTH 418 : Inference - I  
2023-2024 - II Semester  
End Semester Examination  
Model Solutions

**Problem No. 1** (i) The pdf of  $X_{(2)} = \max\{X_1, X_2\}$  is

$$g_{\theta}(x) = \frac{4x^3}{\theta^4}, \quad 0 < x < \theta$$

Let the function  $h(\cdot)$  be such that

$$E_{\theta}(h(X_{(2)})) = 0, \quad \forall \theta \geq 1$$

$$\Leftrightarrow \int_0^{\theta} h(x) x^3 dx = 0, \quad \forall \theta \geq 1$$

$$\Leftrightarrow \int_0^1 h(x) x^3 dx = 0, \quad \text{and} \quad \int_0^{\theta} h(x) x^3 dx = 0, \quad \forall \theta \geq 1$$

$$\Leftrightarrow \int_0^1 h(x) x^3 dx = 0, \quad \text{and} \quad h(x) = 0, \quad \forall x \geq 1$$

3 MARKS

Consider

$$h(x) = \begin{cases} x - \frac{4}{5}, & 0 < x < 1 \\ 0 & x \geq 1 \end{cases}$$

Then  $E_{\theta}(h(X_{(2)})) = 0, \quad \forall \theta \geq 1$

But  $P_{\theta}(h(X_{(2)}) > 0) = P_{\theta}(X_{(2)} > 1)$

$$= \int_1^{\theta} \frac{4x^3}{\theta^4} dx = \theta^{-4} \neq 1, \quad \forall \theta \geq 1$$

$\Rightarrow X_{(2)}$  is NOT Complete

3 MARKS

(15)  $T = X_{(2)}$  is a minimal sufficient statistic and therefore, it suffices to consider only those estimators that depend on  $X_1$  or  $X_2$  only through  $T = X_{(2)}$ .

$U \in \mathcal{U} \Leftrightarrow \int_0^1 u(t) t^3 dt = 0$  and  $u(t) \geq 0 \quad \forall t \geq 1$  (by (11)) — (\*)

Let  $h(\cdot)$  be a continuous function  $E_0(h(X_{(2)})) = 0, \quad \forall \theta \geq 1$

$\Leftrightarrow \int_0^1 h(t) \frac{4t^3}{\theta^4} dt = 0, \quad \forall \theta \geq 1$

$\Leftrightarrow 4 \int_0^1 h(t) t^3 dt + 4 \int_1^\theta h(t) t^3 dt = 0, \quad \forall \theta \geq 1$

$\Leftrightarrow 4 \int_0^1 h(t) t^3 dt = 1$  and  $4h(t) t^3 = 5t^4, \quad \forall t \geq 1$   
 $\Leftrightarrow h(t) = \frac{5}{4} t, \quad \forall t \geq 1$

Then, for  $U \in \mathcal{U}$ , we must have

$$E_0(h(t) U(t)) = \int_0^1 h(t) u(t) \frac{4t^3}{\theta^4} dt + \int_1^\theta h(t) u(t) \frac{4t^3}{\theta^4} dt$$

$$= \frac{4}{\theta^4} \int_0^1 h(t) u(t) t^3 dt = 0 \dots \dots \dots \boxed{3 \text{ MARKS}}$$

$\Leftrightarrow$  This suggests taking 
$$g(X_{(2)}) = \begin{cases} a, & X_{(2)} < 1 \\ \frac{5}{4} X_{(2)}, & X_{(2)} \geq 1 \end{cases}$$

$4a \int_0^1 t^3 dt = 1 \Rightarrow a = 1$

Thus the UMVUE of  $\theta$  is

$$g(X_{(2)}) = \begin{cases} 1, & X_{(2)} < 1 \\ \frac{5}{4} X_{(2)}, & X_{(2)} \geq 1 \end{cases} \dots \dots \dots \boxed{3 \text{ MARKS}}$$



**Problem No. 2**

(1) For  $x, y \in \mathcal{X}$

$$\frac{h_0(x)}{h_0(y)} = \frac{(\theta/1-\theta)^{T_1(x)}}{(\theta/1-\theta)^{T_1(y)}} \text{ is independent of } \theta \in (0,1)$$

$$\Leftrightarrow \eta^{T_1(x) - T_1(y)} \text{ is independent of } \eta \in (0,1)$$

$$\Leftrightarrow T_1(x) = T_1(y)$$

Thus  $T_1 \equiv T_1(x) = \sum_{i=1}^5 x_i$  is a minimal sufficient statistic ... **2 MARKS**

Since the sufficient statistic (in fact minimal sufficient statistic)  $T_1$  is a function of the statistic  $T_2$ , it follows that  $T_2$  is also a sufficient statistic. ... **2 MARKS**

(ii) The derived estimator is

$$\delta_1(T_2) = \delta_1(z_1, z_2) = E_\theta [\delta_0(x) | (z_1, z_2)] \dots \text{2 MARKS}$$

where  $z_1 = x_1 + x_2 + x_3$  and  $z_2 = x_4 + x_5$  and  $T_2 = (z_1, z_2)$

for fixed  $z_1, z_2 \in \{0, 1, 2, 3\} \times \{0, 1, 2\}$

$$\delta_1(z_1, z_2) = P_\theta (x_1 x_2 = 1 | x_1 + x_2 + x_3 = z_1, x_4 + x_5 = z_2)$$

$$= \frac{P(x_1=1, x_2=1, x_3 = z_1 - 2, x_4 + x_5 = z_2)}{P(x_1 + x_2 + x_3 = z_1, x_4 + x_5 = z_2)}$$

~~$$= \frac{\binom{z_1}{2} \theta^2 (1-\theta)^{z_1-2} \binom{z_2}{2} \theta^2 (1-\theta)^{z_2-2}}{\binom{z_1}{2} \theta^2 (1-\theta)^{z_1-2} \binom{z_2}{2} \theta^2 (1-\theta)^{z_2-2}}$$~~

$z_1 = 0, 1$   
 $z_1 = 2, 3, 4, 5$   
 $z_2 \in \{0, 1, 2\}$

$$= \begin{cases} 0, & z_1 = 0, \\ \frac{\theta^2(1-\theta)}{3\theta^2(1-\theta)} = \frac{1}{3}, & z_1 = 2 \\ \frac{\theta^3}{\theta^3} = 1, & z_1 = 3 \end{cases}$$

Thus the derived estimator is

$$\delta_1(T_2) = \frac{Z_1(Z_1-1)}{6}, \quad \text{where } Z_1 = X_1 + X_2 + X_3$$

The estimator  $\delta_1$  can be further improved by using Rao-Blackwell theorem on the minimal sufficient statistic  $T_1$ .  
So  $\delta_1$  is NOT the UMVUE.

3 MARKS

(iii) The derived estimator is

$$\delta_2(T_1) = E\theta[\delta_1(T_1) | T_1]$$

For  $t_2 \in \{0, 1, \dots, 5\}$

$$\delta_2(t_1) = \sum_{t_2=0}^3 \delta_1(t_2) P_\theta(\delta_1(T_2) = \delta_1(t_2) | T_1 = t_1)$$

$$= \frac{1}{3} P_\theta(X_1 + X_2 + X_3 = 2 | \sum_{i=1}^5 X_i = t_1) + P_\theta(X_1 + X_2 + X_3 = 3 | \sum_{i=1}^5 X_i = t_1)$$

$$= \frac{1}{3} \frac{P_\theta(X_1 + X_2 + X_3 = 2) P_\theta(X_4 + X_5 = t_1 - 2)}{P_\theta(\sum_{i=1}^5 X_i = t_1)}$$

$$+ \frac{P_\theta(X_1 + X_2 + X_3 = 3) P_\theta(X_4 + X_5 = t_1 - 3)}{P_\theta(\sum_{i=1}^5 X_i = t_1)}$$

2 MARKS

$$t_2 = 0, 1 \rightarrow \frac{1}{3} \frac{\binom{3}{2} \theta^2(1-\theta)(1-\theta)}{\binom{5}{2} \theta^2(1-\theta)^2}, t_1 = 2$$

$$= \begin{cases} 0, \\ \frac{1}{3} \frac{\binom{3}{2} \theta^2(1-\theta) \binom{2}{t_1-2} \theta^{t_1-2} (1-\theta)^{4-t_1}}{\binom{5}{t_1} \theta^{t_1} (1-\theta)^{5-t_1}}, & t_1 = 3, 4 \\ \frac{\theta^3 \binom{2}{t_1-3} \theta^{t_1-3} (1-\theta)^{5-t_1}}{\binom{5}{t_1} \theta^{t_1} (1-\theta)^{5-t_1}}, \\ \frac{\theta^3 \times \theta}{\theta^5}, & t_1 = 5 \end{cases}$$

$$= \left\{ \begin{array}{l} 0, \\ \frac{1}{10}, \\ \frac{\binom{2}{t_1-2}}{\binom{5}{t_1}} + \frac{\binom{2}{t_1-3}}{\binom{5}{t_1}}, \\ 1 \end{array} \right. , \quad \begin{array}{l} t_1=0, 1 \\ t_1=2 \\ t_1=3, 4 \\ t_1=5 \end{array}$$

$$= \left\{ \begin{array}{l} 0, \\ \frac{1}{10}, \\ \frac{t_2(t_2-1)(5-t_2)}{60} + \frac{t_2(t_2-1)(t_2-2)}{60}, \\ 1 \end{array} \right. , \quad \begin{array}{l} t_1=0, 1 \\ t_1=2 \\ t_1=3, 4 \\ t_1=5 \end{array}$$

$$= \left\{ \begin{array}{l} 0, \\ \frac{1}{10}, \\ \frac{t_2(t_2-1)}{20}, \\ 1 \end{array} \right. , \quad \begin{array}{l} t_1=0, 1 \\ t_1=2 \\ t_1=3, 4 \\ t_1=5 \end{array}$$

3 MARKS

$$\Rightarrow S_2(T_1) = \frac{T_1(T_1-1)}{20}$$

$S_2$  is the unique complete sufficient statistic based on the unique unbiased estimator  $T_1$  (Lehmann-Scheffé Theorem).

**Problem No. 3**

(1)  $T = x_1 + x_2 + x_3 \sim \text{Gamma}(3, \frac{1}{\theta})$   
 $= 3\bar{x}$

$T$  is a Complete-Sufficient Statistic.

$$E_{\theta}(T^r) = \frac{\theta^3}{\Gamma(3)} \int_0^{\infty} t^r e^{-\theta t} t^2 dt$$

$$= \frac{\Gamma(r+3)}{2} \theta^{-r}, \quad r > -3$$

$$E_{\theta}(T) = \frac{\Gamma(4)}{2} \psi_1(\theta) = 3 \psi_1(\theta), \quad \theta > 0$$

$$\Rightarrow E_{\theta}\left(\frac{T}{3}\right) = \psi_1(\theta), \quad \theta > 0$$

$\Rightarrow \bar{x}$  is UMVUE of  $\psi_1(\theta)$

**1-MARK**

$$E_{\theta}(T^{-1}) = \frac{\Gamma(2)}{2} \cdot \frac{1}{\theta} \Rightarrow E_{\theta}\left(\frac{2}{3\bar{x}}\right) = \psi_2(\theta), \quad \theta > 0$$

$\Rightarrow \frac{2}{3\bar{x}}$  is the UMVUE of  $\psi_2(\theta)$

**1-MARK**

The UMVUE of  $\psi_3(\theta)$  is

$$\delta_0(T) = E_{\theta}(\delta | T)$$

$$= P_{\theta}(x_1 > 1 | T)$$

For  $t > 0$

$$\delta_0(t) = P_{\theta}(x_1 > 1 | T=t)$$

$$= P_{\theta}\left(\frac{x_1}{x_1 + x_2 + x_3} > \frac{1}{t} \mid T=t\right)$$

$$= P_{\theta}\left(\frac{x_1}{x_1 + x_2 + x_3} > \frac{1}{t}\right)$$

$$= P_{\theta}\left(\text{Beta}(1, 2) > \frac{1}{t}\right) = \begin{cases} 2 \int_{1/t}^1 (1-x) dx & \text{if } \frac{1}{t} < 1 \\ 0 & \text{if } \frac{1}{t} > 1 \end{cases}$$

(Beta's Thm)

$$\frac{1}{t} < 1$$

$$\frac{1}{t} > 1$$



$$= \begin{cases} (1 - \frac{1}{t})^2, & \text{if } t > 1 \\ 0 & \text{if } t < 1 \end{cases}$$

$$\Rightarrow \delta_0(\tau) = \begin{cases} (1 - \frac{1}{3\bar{x}})^2, & \text{if } \bar{x} > \frac{1}{3} \\ 0 & \text{if } \bar{x} < \frac{1}{3} \end{cases}$$

4 MARKS

$$(ii) J_{3 \times 1} = \begin{pmatrix} \psi_1'(\theta) \\ \psi_2'(\theta) \\ \psi_3'(\theta) \end{pmatrix} = \begin{pmatrix} -\frac{1}{\theta^2} \\ 1 \\ -e^{-\theta} \end{pmatrix}$$

$$I(\theta) = \frac{3}{\text{Var}(X_1)} = \frac{3}{\theta^2}$$

For unbiased estimator  $\delta_i$  of  $\psi_i(\theta)$  is p.d.

3 MARKS

$$\text{Var}(\delta) = \frac{\theta^2}{3} \begin{pmatrix} -\frac{1}{\theta^2} \\ 1 \\ -e^{-\theta} \end{pmatrix} \begin{pmatrix} -\frac{1}{\theta^2}, 1, -e^{-\theta} \end{pmatrix} \text{ is p.d.}$$

$$\text{or } \text{Var}(\delta) = \frac{\theta^2}{3} \begin{pmatrix} \frac{1}{\theta^4} & -\frac{1}{\theta^2} & \frac{e^{-\theta}}{\theta^2} \\ -\frac{1}{\theta^2} & 1 & -e^{-\theta} \\ \frac{e^{-\theta}}{\theta^2} & -e^{-\theta} & e^{-2\theta} \end{pmatrix}$$

$$\Rightarrow \text{Var}(\delta_1) \geq \frac{1}{3\theta^2}, \quad \text{Var}(\delta_2) \geq \frac{\theta^2}{3}$$

$$\text{Var}(\delta_3) \geq \frac{\theta^2 e^{-2\theta}}{3}$$

3 MARKS

**Problem No. 4**

$$Eg(x) = \frac{2}{\theta^2} \int_0^{\theta} x(\theta-x) dx$$

$$= \frac{2}{\theta} \int_0^{\theta} x dx$$

$$\Rightarrow \hat{\theta}_{MLE} = 3\bar{x}$$

4 MARKS

For any fixed  $\lambda = (\lambda_1, \lambda_2) \in (0, \infty)^2$

$$L_{\lambda}(\theta) = \frac{4(\theta - \lambda_{(1)})(\theta - \lambda_{(2)})}{\theta^4}, \quad \theta \geq \lambda_{(2)}$$

For  $\theta \geq \lambda_{(2)}$

$$\frac{\partial}{\partial \theta} L_{\lambda}(\theta) > (<) 0$$

$$\Leftrightarrow \frac{d}{d\theta} \left[ 2\theta^2 - 3(\lambda_{(1)} + \lambda_{(2)})\theta + 4\lambda_{(1)}\lambda_{(2)} \right] < (>) 0$$

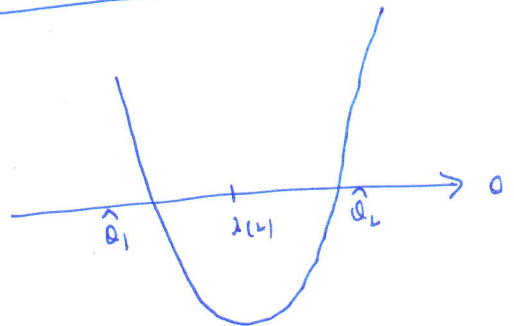
4 MARKS

The quadratic equation  $2\theta^2 - 3(\lambda_{(1)} + \lambda_{(2)})\theta + 4\lambda_{(1)}\lambda_{(2)} = 0$  has 2 real roots (discriminant  $\Delta = 9\lambda_{(1)}^2 + 9\lambda_{(2)}^2 - 14\lambda_{(1)}\lambda_{(2)} = 4\lambda_{(1)}^2 + 9\lambda_{(2)}^2 - 14\lambda_{(1)}\lambda_{(2)} > 0$ )

$$\hat{\theta}_1 = \frac{3(\lambda_{(1)} + \lambda_{(2)}) - \sqrt{9(\lambda_{(1)} + \lambda_{(2)})^2 - 32\lambda_{(1)}\lambda_{(2)}}}{4}$$

$$\text{and } \hat{\theta}_2 = \frac{3(\lambda_{(1)} + \lambda_{(2)}) + \sqrt{9(\lambda_{(1)} + \lambda_{(2)})^2 - 32\lambda_{(1)}\lambda_{(2)}}}{4}$$

with  $\hat{\theta}_1 < \lambda_{(2)} < \hat{\theta}_2$



Thus  $L_{\lambda}(\theta) \uparrow$  in  $\theta \in (\lambda_{(2)}, \hat{\theta}_2)$  and

$\downarrow$  in  $\theta \in (\hat{\theta}_2, \infty)$

$$\Rightarrow \hat{\theta}_{MLE} = \hat{\theta}_2 = \frac{3(\lambda_{(1)} + \lambda_{(2)}) + \sqrt{9(\lambda_{(1)} + \lambda_{(2)})^2 - 32\lambda_{(1)}\lambda_{(2)}}}{4}$$

4 MARKS

**Problem No. 5**

For  $\alpha \in \Theta = \{1, \frac{3}{2}, 2\}$

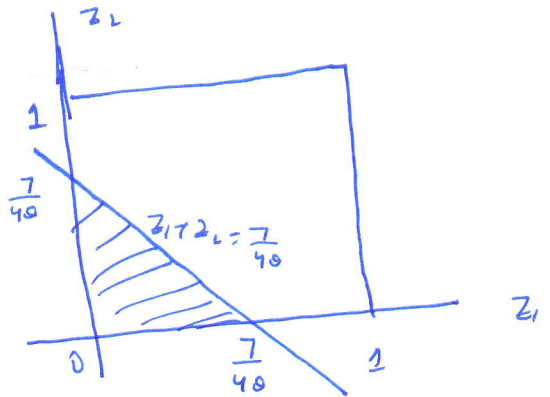
$$E_{\alpha}(Q(x)) = P_{\alpha}(X_1 + X_2 > \frac{7}{4})$$

$$= P_0(z_1 + z_2 > \frac{7}{40}),$$

Where  $z_1$  and  $z_2$  are iid  $U(0,1)$

For  $\frac{7}{40} \in (0,1)$ , i.e.  $0 < \frac{7}{40}$

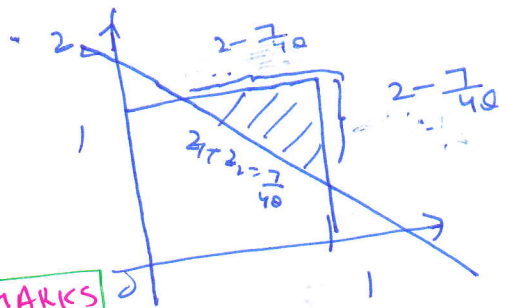
$$E_{\alpha}(Q(x)) = 1 - \frac{1}{2} \left(\frac{7}{40}\right)^2 \\ = 1 - \frac{49}{3200}$$



**4 MARKS**

For  $\frac{7}{40} > 1$ , i.e.  $0 < \frac{7}{4}$

$$E_{\alpha}(Q(x)) = \frac{1}{2} \left(2 - \frac{7}{40}\right)^2$$



**2 MARKS**

$$E_{\alpha=1}(Q(x)) = \frac{1}{2} \left(2 - \frac{7}{4}\right)^2 = \frac{1}{32}$$

$$E_{\alpha=\frac{3}{2}}(Q(x)) = \frac{1}{2} \times \frac{25}{36} = \frac{25}{72}$$

**2 MARKS**

$$\text{Size} = \max \left\{ E_{\alpha=1}(Q(x)), E_{\alpha=\frac{3}{2}}(Q(x)) \right\}$$

$$= \max \left\{ \frac{1}{32}, \frac{25}{72} \right\} = \frac{25}{72}$$

**4 MARKS**

**Problem No. 6**

First, consider test  $H_0: \theta = 0$  vs  $H_1: \theta > \theta_1$ , where  $\theta_1 > 0$  is a fixed constant.

Then  $P_0(x(1) > 0 \text{ and } x(n) < 0, \dots, 1) = 1$   $\forall \theta \in [0, 1]$ .

$$\frac{b_{\theta_1}(x)}{b_0(x)} = \begin{cases} 0 & x(1) < \theta_1 \text{ and } x(n) < 1 \\ 1 & x(1) > \theta_1 \text{ and } x(n) < 1 \\ \infty & x(1) > \theta_1 \text{ and } x(n) > 1. \end{cases}$$

For  $\theta = 0$ , aUMP test is

$$Q_0(x) = \begin{cases} 0 & \text{if } x(1) < \theta_1 \text{ and } x(n) < 1 \\ 1 & \text{if } x(1) > \theta_1 \end{cases}$$

where  $\theta \in [0, 1]$  is A.T.

$$E_0(Q_0(x)) = \alpha$$

$$\Rightarrow 0 \cdot P_0(x(1) < \theta_1, x(n) < 1) + 1 \cdot P_0(x(1) > \theta_1) = \alpha$$

$$\Rightarrow 0 \cdot P_0(x(1) < \theta_1) + P_0(x(1) > \theta_1) = \alpha$$

$$\Rightarrow 0 = \frac{\alpha - P_0(x(1) > \theta_1)}{1 - P_0(x(1) > \theta_1)} = \begin{cases} \frac{\alpha - (1-\theta_1)^n}{1 - (1-\theta_1)^n} & 0 < \theta_1 < 1 \\ \alpha & \theta_1 > 1 \end{cases}$$

For  $\theta_1 \in (0, 1)$  and  $\alpha \geq (1-\theta_1)^n$  UMP( $\alpha$ ) test is

$$Q_0(x) = \begin{cases} \frac{\alpha - (1-\theta_1)^n}{1 - (1-\theta_1)^n} & x(1) < \theta_1, x(n) < 1 \\ 1 & x(1) > \theta_1 \end{cases}$$

with power =  $E_{\theta_1}(Q(x)) = P_{\theta_1}(x(1) > \theta_1) = 1$  ... (A)

For  $\theta_1 > 1$  and  $\alpha \in (0, 1)$ , the UMP( $\alpha$ ) test is

$$Q_0(x) = \begin{cases} \alpha & x(1) < \theta_1, x(n) < 1 \\ 1 & x(1) > \theta_1 \end{cases}$$

with power  $E_{\theta_1}(Q(x)) = P_{\theta_1}(x(1) > \theta_1) = 1$  ... (B)

10/16

3 MARKS

For  $\beta=1$ , UMP( $\alpha$ ) test is

$$Q_{\beta}(x) = \begin{cases} 0 & x_{(1)} < \theta_1, \quad x_{(n)} < 1 \\ \alpha & x_{(1)} > \theta_1, \quad x_{(n)} < 1 \\ 1 & x_{(1)} > \theta_1, \quad x_{(n)} > 1 \end{cases}$$

where  $\alpha \in (0, 1)$  is a constant.

$$E_{\theta_1}(Q_{\beta}(x)) = \alpha$$

$$\Leftrightarrow \alpha P_{\theta_1}(x_{(1)} > \theta_1) = \alpha$$

$$\Leftrightarrow \alpha = \begin{cases} \frac{\alpha}{(1-\theta_1)^n} & 0 < \theta_1 < 1 \\ 0 & \theta_1 > 1 \end{cases}$$

Thus for  $\theta_1 \in (0, 1)$  and  $0 < \alpha \leq (1-\theta_1)^n$  UMP( $\alpha$ ) test is

$$Q_1(x) = \begin{cases} 0 & x_{(1)} < \theta_1, \quad x_{(n)} < 1 \\ \frac{\alpha}{(1-\theta_1)^n} & x_{(1)} > \theta_1, \quad x_{(n)} < 1 \\ 1 & x_{(1)} > \theta_1, \quad x_{(n)} > 1 \end{cases}$$

with power  $E_{\theta_1}(Q_1)$

$$= \frac{\alpha}{(1-\theta_1)^n} P_{\theta_1}(x_{(n)} < 1) + P_{\theta_1}(x_{(n)} > 1)$$

$$= \frac{\alpha}{(1-\theta_1)^n} (1-\theta_1)^n + 1 - (1-\theta_1)^n$$

$$= 1 + \alpha - (1-\theta_1)^n \quad \dots \quad (C)$$

3 MARKS

Consider the test

$$Q_2(x) = \begin{cases} 0 & x_{(1)} < c, \quad x_{(n)} < 1 \\ 1 & x_{(1)} > c \text{ or } x_{(n)} > 1 \end{cases}$$

where  $c$  is chosen so that

$$E_{\theta_1}(Q_2(x)) = \alpha$$



$$\Leftrightarrow P_0 (X_{(1)} < c, X_{(n)} < 1) = 1 - \alpha$$

$$\Leftrightarrow P_0 (X_{(1)} < c) = 1 - \alpha$$

$$\Leftrightarrow 1 - (1-c)^n = 1 - \alpha$$

$$\Leftrightarrow c = 1 - \alpha^{1/n}$$

The power of this test is

$$E_{\theta_1} (Q_2(X)) = P_{\theta_1} (X_{(1)} > 1 - \alpha^{1/n} \text{ or } X_{(n)} > 1)$$

Clearly for  $\theta_1 > 1$ ,  $P_{\theta_1} (X_{(n)} > 1) = 1$  and therefore

$$\text{power} = E_{\theta_1} (Q_2(X)) = 1 \quad \dots (D)$$

For  $\theta_1 \in (0, 1)$

$$\text{power} = E_{\theta_1} (Q_2(X))$$

$$= P_{\theta_1} (X_{(1)} > 1 - \alpha^{1/n} \text{ or } X_{(n)} > 1)$$

$$= P_{\theta_1} (X_{(1)} > 1 - \alpha^{1/n}, X_{(n)} < 1) + P_0 (X_{(n)} > 1)$$

$$= \begin{cases} (1 - \theta_1)^n + 1 - (1 - \theta_1)^n = 1 & \theta_1 > 1 - \alpha^{1/n} \\ \alpha + 1 - (1 - \theta_1)^n & \theta_1 < 1 - \alpha^{1/n} \end{cases}$$

$$\theta_1 < 1 - \alpha^{1/n}$$

... (E)

Comparing (A) - (E), we conclude that UMP( $\alpha$ ) test

is

$$Q_2(X) = \begin{cases} 0 & X_{(1)} < 1 - \alpha^{1/n}, X_{(n)} < 1 \\ 1 & X_{(1)} > 1 - \alpha^{1/n} \text{ or } X_{(n)} > 1 \end{cases}$$

As  $Q_2$  does not depend on  $\theta_1$  as long as  $\theta_1 > 0$ ,  $Q_2$  is also UMP( $\alpha$ ) for testing  $H_0: \theta = 0$  vs  $H_1: \theta > 0$ .

4 PAGES

For  $\theta \leq 0$ ,

$$E_{\theta}(Q_2(X)) = P_{\theta}(X_{(1)} > 1 - \alpha^{1/n} \text{ or } X_{(n)} > 1)$$

$$= P_{\theta}(X_{(1)} > 1 - \alpha^{1/n})$$

$$= \begin{cases} 0 \\ (1 - \alpha^{1/n})^n \end{cases}$$

$$\theta \leq -\alpha^{1/n}$$

$$-\alpha^{1/n} \leq \theta \leq 0$$

and

and  $E_{\theta}(Q_2(X)) = \alpha$

$\Rightarrow Q_2$  has size  $\alpha$  (level  $\alpha$ ) uniformly most powerful invariant unbiased test for  $H_0: \theta \leq 0$  vs  $H_1: \theta > 0$

2 MARKS

**Problem No. 7**

The joint pdf of  $\underline{x} = (x_1, \dots, x_n)$  is

$$f_{\underline{x}}(\underline{x}) = \left(\frac{1}{2\pi\sigma^2}\right)^{\frac{n}{2}} e^{-\frac{n\theta\bar{x} - \frac{1}{2\sigma^2} \sum x_i^2 - \frac{n\theta^2}{2}}$$

Let  $\eta = n\theta$ . Then  $H_0: \eta = 0$  and  $H_1: \eta \neq 0$ . We have

$$Y = \bar{x} \quad \text{and} \quad U = \sum_{i=1}^n x_i^2$$

$$T = \frac{\sqrt{h}\bar{x}}{\sqrt{\frac{\sum x_i^2}{n}}} \quad \uparrow \quad \text{in } \bar{x} \quad \text{for fixed } U.$$

Thus  $U \cap P(\alpha)$  test is  $T = \frac{\sqrt{h}\bar{x}}{\sqrt{\frac{\sum x_i^2}{n}}} < c_1(U)$  or  $T = \frac{\sqrt{h}\bar{x}}{\sqrt{\frac{\sum x_i^2}{n}}} > c_2(U)$

$$f_{\underline{x}}(\underline{x}) = \begin{cases} 1 \\ 0 \end{cases}$$

O.W

**4 MARKS**

where

$$E_0(D_0|U) = \alpha$$

$$\text{and } E_0(D_0 Y | U) = \alpha E_0(Y|U)$$

Under  $H_0$ ,  $V = \sum x_i^2$  is C-S and  $T$  is ancillary. So by Basu's Thm,  $\sum_{i=1}^n x_i^2$  and  $U$  are independent.

$$\Leftrightarrow c_1(U) = c_1 \quad \text{and} \quad c_2(U) = c_2, \quad \text{Under } H_0$$

$$T = \frac{\sqrt{h}\bar{x}}{\sqrt{\frac{\sum x_i^2}{n}}} = \frac{N(0,1)}{\sqrt{\frac{\chi^2_n}{n}}} \text{ independent}$$

$\sim t$  dist with  $n$  d.f.

$$E_0(D_0|U) = \alpha \quad \Leftrightarrow \quad P_0(T < c_1) + P_0(T > c_2) = \alpha \quad \dots (I)$$

$$E_0(Y|U) - E_0((1-D_0)Y|U) = \alpha E_0(Y|U)$$

$$E_0(T|U) = E_0\left(\frac{1}{n} \sqrt{U} T | U\right) = \frac{\sqrt{U}}{n} E_0(T|U)$$

$$= \frac{\sqrt{U}}{n} E_0(T) = 0$$

$$\Rightarrow E_0((1-\phi)T|U) = 0$$

$$\Leftrightarrow P_0(c_1 < T < c_2 | U) = 0 \Leftrightarrow P_0(c_1 < T < c_2) = 0$$

$$\Leftrightarrow P_0(-c_2 < T < -c_1) = 0$$

Thus we may take  $c_1 = -c_2$ . Using in (F) we get

$$P_0(T < -c_2) + P_0(T > c_2) = \alpha$$

$$2(1 - P_0(T \leq c_2)) = \alpha$$

$$P_0(T \leq c_2) = 1 - \alpha/2$$

$$P_0(T > c_2) = \alpha/2$$

$$\Rightarrow c_2 = t_{n, \alpha/2} = -c_1$$

Thus the unbiased test is

$$\phi_0(X) = \begin{cases} 1 \\ 0 \end{cases}$$

$$= \begin{cases} 1 \\ 0 \end{cases}$$

$$T < -t_{n, \alpha/2} \text{ or } T > t_{n, \alpha/2}$$

0.w

$$|T| > t_{n, \alpha/2}$$

0.w

4 MARKS

**Problem No. 8**

The joint pdf of  $\underline{X} = (X_1, \dots, X_m, Y_1, \dots, Y_n)$  is

$$h_{\theta_1, \theta_2}(\underline{x}, \underline{y}) = \frac{1}{\theta_1^m \theta_2^n} e^{-\frac{\sum x_i}{\theta_1} - \frac{\sum y_i}{\theta_2}}$$

$$= \frac{1}{\theta_1^m \theta_2^n} e^{\frac{(\frac{1}{\theta_1} - \frac{1}{\theta_2}) \sum x_i}{\eta}} = \frac{1}{\theta_1} (\sum x_i + \sum y_i)$$

$H_0: \theta_1 = \theta_2 \Leftrightarrow \eta = 0$   
 $H_1: \theta_1 \neq \theta_2 \Leftrightarrow \eta \neq 0$   
 $Y = \sum Y_i, U = \sum X_i + \sum Y_i$

Under  $\bar{\Theta}_{\theta_1} = \{(\theta_1, \theta_2) : \theta_1 = \theta_2\}$   $U$  is Complete-Sufficient. For given  $Y, T = \frac{Y}{U} \uparrow \eta$ . Thus UMPU( $\alpha$ ) test is

$$d_0(\underline{x}) = \begin{cases} 1 & T < c_1 \text{ or } T > c_2 \\ 0 & \text{o.w.} \end{cases} \dots \text{6 MARKS}$$

What  
 (\*)  $E_{\theta_1 = \theta_2} (d_0(\underline{x}) | U) = \alpha$   
 (\*\*)  $E_{\theta_1 = \theta_2} (d_0(\underline{x}) Y | U) = \alpha E_{\theta_1 = \theta_2} (Y | U)$

Under  $\theta_1 = \theta_2$   
 $B = T \sim \text{Beta}(n, m)$  and  $U$  are independent

(\*)  $\Rightarrow P_0(c_1 < B(n, m) < c_2) = 1 - \alpha$   
 $\Rightarrow \int_{c_1}^{c_2} \frac{t^{n-1} (1-t)^{m-1}}{B(n, m)} dt = 1 - \alpha \dots \text{3 MARKS}$

(\*\*)  $\Rightarrow E_{\theta_1 = \theta_2} ((1-d_0) Y | U) = (1-\alpha) E_{\theta_1 = \theta_2} (Y | U)$   
 $\Rightarrow U P_{\theta_1 = \theta_2} (1 - (c_1 < T < c_2) | T) = (1-\alpha) U E_{\theta_1 = \theta_2} (T | U)$   
 $\Rightarrow \int_{c_1}^{c_2} t \frac{t^{n-1} (1-t)^{m-1}}{B(n, m)} dt = (1-\alpha) \frac{n}{n+m+1} \dots \text{3 MARKS}$