

**Question 1.** (20 pts.) Let  $X_1, X_2, \dots, X_n$  be independently and identically distributed as  $\text{Unif}(a\theta, b\theta)$  for known constants  $a, b$  satisfying  $0 < a < b$  and  $\theta > 0$  is an unknown parameter. Let  $X_{(1)} < X_{(2)} < \dots < X_{(n)}$  denote the order statistics.

a) (9 points) Find a minimal sufficient statistic for  $\theta$ . No need to show that it is actually minimal sufficient.

b) (7 points) Is the statistic found in a) complete? Justify your answer.

$$a) \quad f(x_i | \theta) = \frac{1}{(b-a)\theta} \mathbb{1}_{(a\theta, b\theta)}(x_i) \quad g(T(x), \theta)$$

$$\Rightarrow f(x_1, \dots, x_n | \theta) = \left[ \frac{1}{(b-a)\theta} \right]^n \mathbb{1}_{(X_{(n)} < b\theta)} \mathbb{1}_{(X_{(1)} > a\theta)}$$

By Factorization Criterion,  $T = (X_{(1)}, X_{(n)})$  is a SS

One can use Lehmann-Scheffé Thm to show  $T$  is MSS

b) This is a scale family, so  $T_1 = \frac{X_{(1)}}{X_{(n)}}$  is ancillary.

If  $T$  were complete, Basu's Lemma would have implied  $T$  independent of  $T_1$ . But  $T_1 = (X_{(1)}/X_{(n)})$  is a function of  $T = (X_{(1)}, X_{(n)})$ . So  $T$  cannot be independent of  $T_1$ . Hence  $T$  cannot be a complete statistic

**Question 2.** (20 pts.) Suppose we observe a single observation  $X$  from Binomial(2,  $p$ ) distribution where  $p$  can take only two values 0.25 and 0.5.

a) (10 points) Find a minimal sufficient statistic for  $p$ . Call it  $T$ . No need to show that it is actually minimal sufficient.

b) (10 points) Show that the statistic  $T$  found in a) is not complete.

(Hint: Try to find a function  $g$  which is not identically zero but  $E_{p=0.25}\{g(T)\} = E_{p=0.5}\{g(T)\} = 0$ ).

$$a) \quad f(x|p) = \binom{2}{x} p^x (1-p)^{2-x}, \quad x=0, 1, 2$$

We only have 1 observation  $X$ .

So joint density is same as above.

$$f(x|p) = \underbrace{\binom{2}{x}}_{h(x)} \underbrace{(1-p)^2 e^{x \log \frac{p}{1-p}}}_{g(T(x), p)}$$

$T = X$  is a sufficient statistic by F.C.

It can be shown to be minimal using L-S Theorem

b) Need to show for some  $g$

$$E_{p=0.25} \{g(T)\} = g(0) \binom{2}{0} (0.25)^0 (0.75)^{2-0} + g(1) \binom{2}{1} (0.25)^1 (0.75)^1 + g(2) \binom{2}{2} (0.25)^2 (0.75)^0 = 0 \rightarrow \textcircled{1}$$

$$\& E_{p=0.5} \{g(T)\} = g(0) \binom{2}{0} (0.5)^0 (0.5)^{2-0} + g(1) \binom{2}{1} (0.5)^1 (0.5)^1 + g(2) \binom{2}{2} (0.5)^2 (0.5)^0 = 0 \rightarrow \textcircled{2}$$

$$\textcircled{1} \Rightarrow g(0) \frac{9}{16} + 2g(1) \frac{3}{16} + g(2) \frac{1}{16} = 0 \Rightarrow 9g(0) + 6g(1) + g(2) = 0$$

$$\textcircled{2} \Rightarrow g(0) \frac{1}{4} + 2g(1) \frac{1}{4} + g(2) \frac{1}{4} = 0 \Rightarrow g(0) + 2g(1) + g(2) = 0$$

$$\text{Choose } g(0) = 1 \Rightarrow \left. \begin{aligned} 6g(1) + g(2) &= -9 \\ 2g(1) + g(2) &= -1 \end{aligned} \right\} \Rightarrow \begin{aligned} 4g(1) &= -8 \\ g(1) &= -2 \\ g(2) &= 3 \end{aligned}$$

$$\boxed{g(0) = 1, g(1) = -2, g(2) = 3}$$

**Question 3.** (15 pts.) Consider the family of distributions with pdf

$$f(x | \theta) = \frac{3x^2}{\theta^3}, \quad 0 \leq x \leq \theta.$$

and cdf

$$F(x | \theta) = \begin{cases} \frac{x^3}{\theta^3}, & 0 \leq x \leq \theta \\ 1, & x > \theta \\ 0, & x < 0. \end{cases}$$

Let  $X_1, \dots, X_n$  be iid from  $f(x | \theta)$ . Show that  $T = \max\{X_i, i = 1, \dots, n\}$  is a complete statistic.

First find distribution of  $T$ . for  $0 \leq t \leq \theta$

$$\begin{aligned} P(T \leq t) &= P(X_{(n)} \leq t) = P(X_1 < t)^n \\ &= \left[ \frac{t^3}{\theta^3} \right]^n \\ &= \frac{t^{3n}}{\theta^{3n}} \end{aligned}$$

Density of  $T$  is

$$f_T(t) = \frac{d}{dt} P(T \leq t) = \frac{3n t^{3n-1}}{\theta^{3n}}, \quad 0 \leq t \leq \theta$$

Suppose  $E_\theta [g(T)] = 0$  for some  $g$  and  $\forall \theta > 0$ .

$$\Rightarrow \int_0^\theta g(t) f_T(t) dt = 0 \quad \forall \theta > 0$$

$$\Rightarrow \frac{3n}{\theta^{3n}} \int_0^\theta t^{3n-1} g(t) dt = 0 \quad \forall \theta > 0$$

$$\Rightarrow \int_0^\theta t^{3n-1} g(t) dt = 0 \quad \forall \theta > 0$$

$\Rightarrow$  By differentiating w.r.t  $\theta$  and using fundamental Thm of calculus

$$\theta^{3n-1} g(\theta) = 0 \quad \forall \theta > 0$$

$$\Rightarrow g(\theta) = 0 \quad \forall \theta > 0$$

$$\Rightarrow g(t) = 0 \quad \forall t > 0$$

$$\Rightarrow g \equiv 0 \quad \Rightarrow T \text{ is complete.}$$

**Question 4.** (25 pts.) Suppose  $X_i$  are i.i.d. samples from  $f(x | \theta)$ , for  $i = 1, \dots, n$ , where  $\theta > 0$  is unknown, and the family of distributions  $f(x | \theta)$  is the Rayleigh family given by

$$f(x | \theta) = \frac{x}{\theta^2} e^{-x^2/(2\theta^2)}, \quad x \geq 0.$$

- a) (10 points) Show that this is an exponential as well as a scale family.  
 b) (7 points) Find a complete and sufficient statistic for this family. No need to prove this is actually complete and sufficient.  
 c) (8 points) Find the value of

$$E\left[\frac{(\sum X_i)^2}{\sum X_i^2}\right]$$

using Basu's theorem if we know  $EX_i = \theta\sqrt{\pi/2}$  and  $\text{Var}(X_i) = (4 - \pi)\theta^2/2$ .

$$a) \quad f(x_1, \dots, x_n) = \underbrace{\frac{1}{\theta^{2n}}}_{c(\theta)} \underbrace{\prod_{i=1}^n x_i}_{h(x)} \underbrace{e^{-\frac{\sum x_i^2}{2\theta^2}}}_{\exp\{T(x) \omega(\theta)\}}$$

$\Rightarrow$  It is exponential family.

$$\text{where } T(x) = \sum x_i^2 \\ \omega(\theta) = -\frac{1}{2\theta^2}$$

$$f(x|\theta) = \frac{1}{\theta} \psi\left(\frac{x}{\theta}\right) \quad \text{where } \psi(x) = x e^{-\frac{x^2}{2}}, \quad x \geq 0$$

$\Rightarrow$  It is scale family.

b) By Master Theorem,  $T(x) = \sum_{i=1}^n x_i^2$  is CSS.

c)  $S = \frac{(\sum x_i)^2}{\sum x_i^2}$  is ancillary ( $\because$  it is scale invariant)

By Basu,  $S$  &  $T$  are independent

$$\Rightarrow E(ST) = E(T) \cdot E(S)$$

$$\begin{aligned}
E(ST) &= E\left(\left(\sum X_i\right)^2\right) \\
&= \text{Var}\left(\sum X_i\right) + \left[E\left(\sum X_i\right)\right]^2 \\
&= n \text{Var}(X_1) + \left(n\theta\sqrt{\frac{\pi}{2}}\right)^2 \\
&= n \frac{(4-\pi)\theta^2}{2} + \frac{n^2\theta^2\pi}{2}
\end{aligned}$$

$$\begin{aligned}
E(T) &= E\left(\sum X_i^2\right) = \sum E(X_i^2) \\
&= \sum \left[E^2(X_i) + \text{Var}(X_i)\right] \\
&= n \cdot \frac{\theta^2\pi}{2} + n \cdot \frac{(4-\pi)\theta^2}{2}
\end{aligned}$$

$$\begin{aligned}
\text{So } E(S) &= \frac{E(ST)}{E(T)} \\
&= \frac{\cancel{n} \frac{(4-\pi)\theta^2}{2} + \frac{\cancel{n}^2 \theta^2 \pi}{2}}{\frac{\cancel{n} \theta^2 \pi}{2} + \frac{\cancel{n} (4-\pi)\theta^2}{2}} \\
&= \frac{4-\pi + n\pi}{\pi + 4-\pi}
\end{aligned}$$

$$\boxed{E(S) = 1 + \frac{\pi(n-1)}{4}}$$