

Midsession test - MATH 3811/3911
 S1 2014 - Solutions

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1.) a) $a(\theta) = \frac{1}{\theta}$, $b(x) = 2x$, $c(\theta) = -\frac{1}{\theta}$, $d(x) = x^2$

imply that $f(x, \theta)$ is one-parameter exponential family.
 Hence $T = \sum_{i=1}^n d(X_i) = \sum_{i=1}^n x_i^2$ is minimal sufficient

b) $EX_1 = \int_0^{\infty} \frac{2x^2}{\theta} \exp(-\frac{x^2}{\theta}) dx = \frac{\sqrt{2}}{\sqrt{2\pi}\sqrt{\theta}} \int_{-\infty}^{\infty} x^2 \exp(-\frac{x^2}{2(\frac{\theta}{2})}) dx \cdot \frac{\sqrt{\pi\theta}}{\theta}$

Hence $EX_1 = \frac{\theta}{2} \frac{\sqrt{\pi\theta}}{\theta} = \frac{\sqrt{\pi\theta}}{2}$ as the variance of $N(0, \frac{\theta}{2})$

$EX_1^2 = \int_0^{\infty} \frac{2x^3}{\theta} \exp(-\frac{x^2}{\theta}) dx = -\int_0^{\infty} x^2 d \exp(-\frac{x^2}{\theta}) = \theta \int_0^{\infty} \exp(-\frac{x^2}{\theta}) d \frac{x^2}{\theta} = \theta$

c) Using (b), we have $E(T) = n\theta$ Therefore $\frac{T}{n}$ is unbiased for θ and is a function of complete and sufficient statistic. Hence, by Lehmann-Scheffe theorem, $\frac{T}{n}$ is the UMVUE

For CR bound: $\frac{\partial}{\partial \theta} \ln f(x, \theta) = -\frac{1}{\theta} + \frac{x^2}{\theta^2} \Rightarrow V = -\frac{n}{\theta} + \frac{T^2}{\theta^2}$

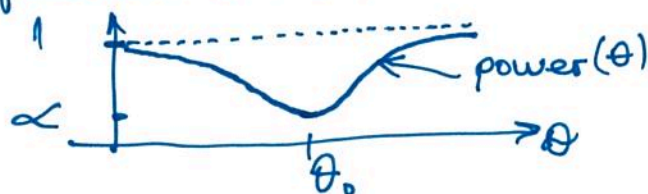
$\frac{\partial^2}{\partial \theta^2} \ln f(x, \theta) = \frac{1}{\theta^2} - \frac{2x^2}{\theta^3} \Rightarrow E(-\frac{\partial^2}{\partial \theta^2} \ln f(x, \theta)) = \frac{1}{\theta^2}$

Hence CR Bound is $\frac{1}{n I_{X_1}(\theta)} = \frac{\theta^2}{n}$ and it is attainable
 Since $V = \frac{\theta^2}{n} (T - \theta)^2$.

d) Since $f(x, \theta)$ is one-parameter exponential \Rightarrow uniformly most powerful unbiased α -size test φ^* of $H_0: \theta = \theta_0$ vs $H_1: \theta \neq \theta_0$ exists. Its structure is:

$\varphi^* = \begin{cases} 1 & \text{if } \sum_{i=1}^n x_i^2 < C_1 \text{ or } \sum_{i=1}^n x_i^2 > C_2 \\ 0 & \text{else} \end{cases}$

The powerfunction should look like:



e) Note: $\text{Var}(X_1^2) = 2\theta^2 - (EX_1^2)^2 = 2\theta^2 - \theta^2 = \theta^2$

Hence
$$\sqrt{n} \frac{\left(\frac{1}{n} \sum_{i=1}^n X_i^2 - \theta_0\right)}{\theta_0} \xrightarrow{d} N(0,1) \quad (\text{CLT})$$

So, a monotone transformation of T is asymptotically standard normal, or "equivalently" $T \approx N(n\theta_0, n\theta_0^2)$

Hence the UMP α test (unbiased) is asymptotically equivalent to a Z test

$$\hat{\phi}^* = \begin{cases} 1 & \text{if } \frac{|T - n\theta_0|}{\sqrt{n}\theta_0} > z_{\frac{\alpha}{2}} \\ 0 & \text{else} \end{cases}$$

f) We see that $\tau(\theta)$ represents $\text{Invgamma}(4, 2)$

Also: $f(\mathbf{X}, \theta) \tau(\theta) \propto \frac{1}{\theta^n} \exp\left(-\frac{\sum X_i^2}{\theta}\right) \frac{1}{\theta^5} \exp\left(-\frac{2}{\theta}\right)$

i.e., $f(\mathbf{X}, \theta) \tau(\theta) \propto \theta^{-n-5} e^{-(\sum_{i=1}^n X_i^2 + 2)/\theta}$ which implies that $h(\theta|\mathbf{X})$ is $\text{Invgamma}(n+4, \sum_{i=1}^n X_i^2 + 2)$

In our case: $h(\theta|\mathbf{X})$ is $\text{Invgamma}(14, 18)$

Hence
$$0.1426 = \frac{18^{14}}{\Gamma(14)} \int_0^1 x^{-15} \exp\left(-\frac{18}{x}\right) dx$$

The Bayes estimate is $\frac{18}{14-1} \approx 1.385$

The Frequentist estimate (the UMVUE) is $16/10 = 1.6$

Comment: For the prior, the expected value is $\frac{2}{3}$, i.e.

using b): $(2EX_1^2 = \theta\pi \rightarrow \theta \approx \frac{16}{9\pi} \approx 0.566$ is indicated

as the value of θ under the prior. This drags down the data-supported 1.6 to just 1.385 but 1.385 is still much larger than the borderline value $\theta_0 = 1$

and we reject H_0 as a consequence.

(That is, even the largest θ -value under H_0 ($\theta_0 = 1$) is not large enough to be supported by the combined information in data \oplus prior)

Q2 | a) $L(\bar{X}; \theta) = \frac{3^n}{\theta^{3n}} \prod_{i=1}^n x_i^2 \prod_{i=1}^n I_{(0, \theta)}(x_i) = \left(\frac{3}{\theta^3}\right)^n \prod_{i=1}^n x_i^2 I_{(0, \theta)}(X_{(n)})$

Then the ratio

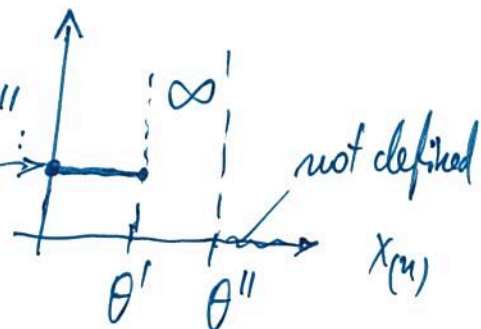
$$\frac{L(\bar{X}; \theta'')}{L(\bar{X}; \theta')} = \left(\frac{\theta'}{\theta''}\right)^{3n} \frac{I_{(0, \theta'')}(X_{(n)})}{I_{(0, \theta')}(X_{(n)})}$$

is monotone non-

decreasing in $X_{(n)}$ for fixed $0 < \theta' < \theta''$:

Hence we have a family with monotone LR in $X_{(n)}$.

$$\left(\frac{\theta'}{\theta''}\right)^{3n}$$



$$G/F_{X_1}(x; \theta) = \begin{cases} 0 & x < 0 \\ x^3/\theta^3 & 0 < x < \theta \\ 1 & x > \theta \end{cases}$$

Hence $F_{X_{(n)}}(x) = P(X_1 \leq x \cap X_2 \leq x \cap \dots \cap X_n \leq x) = \begin{cases} 0 & x < 0 \\ \left(\frac{x}{\theta}\right)^{3n} & 0 < x < \theta \\ 1 & x > \theta \end{cases}$

Then density: $f_{X_{(n)}}(x) = \begin{cases} 3n x^{3n-1} / \theta^{3n} & 0 < x < \theta \\ 0 & \text{else} \end{cases}$

c) $\theta_0 = 2$ BG theorem says that a UMP α -test φ^* exists with $\varphi^* = \begin{cases} 1 & X_{(n)} > c \\ 0 & X_{(n)} \leq c \end{cases}$

To find the threshold c , we

"exhaust the level":

$$\alpha = E_{\theta=2} \varphi^* = P_{\theta=2}(X_{(n)} > c) = 1 - \left(\frac{c}{2}\right)^{3n} = \alpha$$

Hence $\left(\frac{c}{2}\right)^{3n} = 1 - \alpha \Rightarrow c = \left[2(1 - \alpha)^{\frac{1}{3n}}\right]$

and the test is completely determined

