

STAT 721

Solutions to Midterm #1

marks

(10) $\left. \begin{array}{l} y = \sqrt{x_1/x_2} \\ z = x_2 \end{array} \right\}$ is a 2-1 map from $(\mathbb{R}^+)^2 \times \mathbb{R}^+$ to $\mathbb{R}^+ \times \mathbb{R}^+$.

The inverse map is

$$\begin{array}{l} x_1 = zy^2 \\ x_2 = z \end{array} \quad \leftarrow \textcircled{2}$$

$$J = \begin{vmatrix} 2zy & y^2 \\ 0 & 1 \end{vmatrix} = 2zy \quad \leftarrow \textcircled{2}$$

$$f(x_1, x_2) = \frac{1}{2\pi} (x_1 x_2)^{-3/2} e^{-(x_1 + x_2)/2}, \quad x_1 > 0, x_2 > 0.$$

So the joint pdf of Y and Z is:

$$\begin{aligned} g(y, z) &= \frac{1}{2\pi} (z^2 y^2)^{-3/2} e^{-(zy^2 + z)/2} \cdot 2zy \\ &= \frac{1}{\pi} e^{-z(y^2 + 1)/2}, \quad 0 < z < \infty, 0 < y < \infty \quad \leftarrow \textcircled{2} \end{aligned}$$

$$\text{So } g(y) = \int_0^\infty g(y, z) dz = \int_0^\infty e^{-z(y^2 + 1)/2} dz \quad \leftarrow \textcircled{2}$$

$$= \frac{1}{\pi} \int_0^\infty \frac{z}{y^2 + 1} d(-e^{-z(y^2 + 1)/2})$$

$$= \frac{2}{\pi(y^2 + 1)}, \quad 0 < y < \infty \quad \leftarrow \textcircled{2}$$

[This is a "half-Cauchy" distribution.]

(10) ② $\prod_{i=1}^n f(x_i, \theta) = e^{-\sum x_i + n\theta}$ if $x_i \geq 0, x_i > 0, \dots, x_n > 0$
 ④ $= 0$ elsewhere

Since $\{x_i > 0, \dots, x_n > 0\} \Leftrightarrow \{X_{(1)} > 0\}$, we can write the joint pdf as:

$$\prod_{i=1}^n f(x_i, \theta) = e^{-\sum x_i} e^{n\theta} I_{[X_{(1)} > 0]}$$

By the factorization theorem, $X_{(1)}$ is sufficient for θ .

③ \rightarrow (b) For $x \in [0, \infty)$, the cdf. function of $X_{(1)}$ at x is:

$$F_{\theta}(x) = 1 - P_{\theta}(X_{(1)} \geq x) = 1 - \prod_{i=1}^n P_{\theta}[X_i \geq x]$$

$$\begin{aligned} &= 1 - [P_{\theta}(X_{(1)} - \theta \geq x - \theta)]^n = 1 - \left[\int_{x-\theta}^{\infty} e^{-u} du \right]^n \\ &= 1 - e^{-n(x-\theta)}. \end{aligned}$$

∴ the pdf of $X_{(1)}$ is $f_{\theta}(x) = f'_{\theta}(x) = n e^{-n(x-\theta)}$, $x > \theta$
 $= 0$ elsewhere.

③ \rightarrow (c) By the definition of completeness, we need

to show that if $E_{\theta} g(X_{(1)}) = 0$ for all θ ,

then it follows that $g(x) = 0$ for all x .

(continued)

② (c) (continued)

$$E_0 g(X_n) = 0 \text{ for all } \theta \Rightarrow \int_0^\infty n e^{-nx} e^{n\theta} g(x) dx = 0 \text{ for all } \theta$$

$$\Rightarrow \int_0^\infty e^{-nx} g(x) dx = 0 \text{ for all } \theta \text{ (since } n e^{n\theta} > 0 \text{ for all } \theta)$$

$$\Rightarrow 0 = \frac{d}{d\theta} \int_0^\infty e^{-nx} g(x) dx = -e^{-n\theta} g(\theta) = 0 \text{ for all } \theta$$

$$\Rightarrow g(x) = 0 \text{ for all } x, -\infty < x < \infty.$$

Hence X_n is complete

(10) \rightarrow ③

$$\textcircled{5} \rightarrow \textcircled{a) \quad \prod f(x_i; \theta) = 2^n \prod x_i / \theta^{2n} \quad 0 < x_i \leq \theta \quad \text{elsewhere} = 0$$

$$\text{or } \prod f(x_i; \theta) = [2^n (\prod x_i) / \theta^{2n}] I_{\{x_{(n)} \leq \theta\}} \leftarrow \textcircled{2}$$

By the Factorization Theorem, $X_{(n)}$ is sufficient.

$$P[X_i \leq \theta] = \frac{x_i}{\theta} \text{ for } 0 < x < \theta, \text{ as}$$

$$\cancel{P[X_i \leq \theta]} = \int_0^\theta \frac{2nx^{2n-1}}{\theta^{2n}} dx = \frac{x^{2n}}{\theta^{2n}} \Big|_0^\theta = \frac{x^{2n}}{\theta^{2n}}, \quad 0 < x < \theta.$$

So the pdf of $X_{(n)}$ is

$$f(x) = \frac{2nx^{2n-1}}{\theta^{2n}}, \quad 0 < x < \theta \leftarrow \textcircled{1}$$

(continued)

② (a) (continued)

if $E_{\theta}[g(X_{(n)})] = 0$ for all θ ,

then $\int_0^{\theta} \frac{2n x^{2n-1}}{\theta^{2n}} g(x) dx = 0$ for all θ ,

① then $\int_0^{\theta} x^{2n-1} g(x) dx = 0$ for all θ ,

then $0 = \frac{d}{d\theta} \int_0^{\theta} x^{2n-1} g(x) dx = 0 = \theta^{2n-1} g(\theta)$ for all $\theta > 0$,

so $g(\theta) = 0$ for all $\theta > 0$, or $g(x) = 0$ for all $x > 0$.

This shows that $X_{(n)}$ is complete.

⑤ ⇒ (b) The pdf of a sample X_i is

$$f(x_i) = \theta x_i^{\theta-1} = \theta e^{(\theta-1) \log x_i}, \quad 0 < x_i < 1$$

This is therefore a one-dimensional regular exponential family.

It follows by a Theorem about random

samples X_1, \dots, X_n from a regular exponential

family that $T = \sum_{i=1}^n \log X_i$

(or equivalently $T' = \prod_{i=1}^n X_i$) is a complete

sufficient statistic for θ .

$$(10) \quad \textcircled{4} \quad \prod f(x_i | \theta) = \left(\frac{1}{\sqrt{2\pi\theta^2}} \right)^n e^{-\frac{1}{2\theta^2} \sum (x_i - \theta)^2}$$

$$= \left(\frac{1}{\sqrt{2\pi\theta^2}} \right)^n e^{-w_2} e^{-\frac{1}{2\theta^2} \sum x_i^2 + \frac{1}{\theta} \sum x_i}$$

~~By the factorization theorem:~~

$\textcircled{1}$

Hence the family of distributions is a two-dimensional exponential family with sufficient statistic $(\sum X_i, \sum X_i^2)$.

Since (\bar{X}, S^2) is a 1-1 function of $(\sum X_i, \sum X_i^2)$,

(\bar{X}, S^2) is also a sufficient statistic for θ .

[Another solution is to rewrite the exponential family representation so that \bar{X}, S^2 are seen directly to be sufficient statistics.]

$\textcircled{2}$ [The family of distributions is not complete, because the set of parameter $\{(\theta, \theta^2) : \theta > 0\}$ does not contain a 2-dimensional rectangle.]

$\textcircled{3}$ [Another solution is to construct a function of \bar{X} and S^2 , $g(\bar{X}, S^2)$ which does not vanish but for which $E_\theta [g(\bar{X}, S^2)] = 0$, $\forall \theta > 0$.
There are many such functions g .]