

- 1) Suppose that Y_1, \dots, Y_n are iid $\text{gamma}(\nu, \lambda)$ random variables. Find a minimal sufficient statistic for (ν, λ) . Hint: write as a 2 parameter REE.

$$f(y) = \frac{y^{\nu-1}}{\lambda^\nu \Gamma(\nu)} e^{-y/\lambda}, y > 0$$

$$= \underbrace{\frac{1}{\lambda^\nu \Gamma(\nu)}}_{c(\nu, \lambda)} \exp \left[(\nu-1) \log(y) - \frac{1}{\lambda} y \right] \underbrace{I(y > 0)}_{h(y)}$$

$w_1(\nu, \lambda) \quad t_1(y) \quad w_2(\nu, \lambda) \quad t_2(y) \quad h(y)$

\rightarrow 2PREF so $T = (\sum \log(Y_i), \sum Y_i)$ is min suff

always sums for 2PREF

- 40 → 2) Let Y be a uniform $U(-\theta, \theta)$ random variable where $\theta > 0$ (and the sample size $n = 1$). Show that $T(Y) = |Y| = \max(-Y, Y)$ is a minimal sufficient statistic.

$$f(y|\theta) = \frac{1}{2\theta} I(-\theta \leq y \leq \theta) = \frac{1}{2\theta} I(-\theta \leq y) I(y \leq \theta)$$

$$= \frac{1}{2\theta} I(\max(-y, y) \leq \theta) = \frac{1}{2\theta} I(|Y| \leq \theta).$$

$$\text{so } R_{xy}^{(0)} = \frac{I(|x| \leq \theta)}{I(|y| \leq \theta)} = c \neq \theta$$

iff $|x| = |y|$ so

$T(Y) = |Y|$ is min suff by LSM

$$f(y) = \theta^n \sigma^{n\theta} \exp[-(\theta+1)\sum \log(y_i)] I(y_{(1)} > \sigma)$$

3) Let Y_1, \dots, Y_n be iid with pdf

$$f(y) = \theta \sigma^\theta \frac{1}{y^{\theta+1}}$$

$$g(I(Y_{(1)}) | \theta, \sigma)$$

$$g_0\left(\sum_{i=1}^n \log Y_i, Y_{(1)}\right)$$

is suff by factorization

or

$$f(y) = \prod f(y_i) = \theta^n \sigma^{n\theta} \left(\frac{1}{\prod y_i}\right)^{\theta+1} I(y_{(1)} > \sigma)$$

$$g(I(Y_{(1)}) | \theta, \sigma)$$

$g_0\left(\prod_{i=1}^n Y_i, Y_{(1)}\right)$ is suff by Factorization

$$\left(= \prod_{i=1}^n \frac{1}{y_i} I(y_i > 0) \theta^n \sigma^{n\theta} \exp[-\theta \sum \log y_i] I(y_{(1)} > \sigma) \right)$$

hard way

2) Let Y_1, \dots, Y_n be iid with pdf $f(y) = \theta y^{\theta-1}$ where $\theta > 0$ and $0 < y < 1$. Find a minimal sufficient statistic for θ using the Lehmann Scheffé LSM theorem.

$$f(x) = \theta \exp[-(\theta-1) \log(x_i)] I(0 < x_i < 1)$$

$$R_{\bar{x}}(x) = \frac{f(x)}{f(\bar{x})} = \frac{\exp[-(\theta-1) \sum \log(x_i)]}{\exp[-(\theta-1) \sum \log(\bar{x}_i)]} \frac{I(0 < x_{(1)} < x_{(n)} < 1)}{I(0 < \bar{x}_{(1)} < \bar{x}_{(n)} < 1)} = c \propto \theta$$

iff $\exp((\theta-1)(\sum \log(x_i) - \sum \log(\bar{x}_i))) = d \propto \theta$

iff $\sum \log(x_i) = \sum \log(\bar{x}_i)$ so $\sum \log(Y_i)$ is min suff by LSM

$$\text{or } \frac{f(x)}{f(\bar{x})} = \frac{I(0 < x_{(1)} < x_{(n)} < 1)}{I(0 < \bar{x}_{(1)} < \bar{x}_{(n)} < 1)} \frac{(\prod x_i)^{\theta-1}}{(\prod \bar{x}_i)^{\theta-1}} = c \propto \theta$$

θ is not a possibility

constant w.r.t θ

iff $(\frac{\prod x_i}{\prod \bar{x}_i})^{\theta-1} = d \propto \theta$ iff $\frac{\prod x_i}{\prod \bar{x}_i} \equiv 1$ iff $\prod x_i = \prod \bar{x}_i$ (if $\prod x_i = k \prod \bar{x}_i$, $R = k^{\theta-1} \neq$ constant unless $k=1$)

so $\prod Y_i$ is min suff by LSM