

## Sufficiency.



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Q1 Let  $x_1, x_2, \dots, x_n$  be a random sample from  $N(\mu, \sigma^2)$  population. Find the sufficient estimators for  $\mu$  and  $\sigma^2$ .

Solution: Here  $\theta = (\mu, \sigma^2)$ ;  $-\infty < \mu < \infty$   
 $0 < \sigma^2 < \infty$

We have  $L = f(x_1, \theta_1) f(x_2, \theta_2) \dots f(x_n, \theta_n)$

$$= \prod_{i=1}^n f(x_i, \theta_i) = \prod_{i=1}^n f_{\theta}(x_i)$$

(Since  $x_1, x_2, \dots, x_n \sim N(\mu, \sigma^2)$ )

$$= \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x_1-\mu)^2}{2\sigma^2}} \cdot \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x_2-\mu)^2}{2\sigma^2}} \dots \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x_n-\mu)^2}{2\sigma^2}}$$

$$= \left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n \cdot e^{-\frac{(x_1-\mu)^2}{2\sigma^2}} \cdot e^{-\frac{(x_2-\mu)^2}{2\sigma^2}} \dots e^{-\frac{(x_n-\mu)^2}{2\sigma^2}}$$

$$= \left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n \cdot e^{-\frac{1}{2\sigma^2} \left\{ \sum_{i=1}^n (x_i - \mu)^2 \right\}}$$

$$= \left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n \cdot e^{-\frac{1}{2\sigma^2} \left\{ \sum_i (x_i^2 - 2\mu x_i + \mu^2) \right\}}$$

$$= \left(\frac{1}{\sigma\sqrt{2\pi}}\right)^n \cdot e^{-\frac{1}{2\sigma^2} \left\{ \sum_i x_i^2 - 2\mu \sum_i x_i + n\mu^2 \right\}}$$

$$= g_{\theta}[t(x)] \cdot h(x)$$





where  $g_0[t(x)] = \left( \frac{1}{\sigma \sqrt{2\pi}} \right)^n e^{-\frac{1}{2\sigma^2} \{t_0(x) - 2\mu t_1(x) + n\mu^2\}}$

$$t(x) = [t_1(x), t_0(x)] = [\sum x_i, \sum x_i^2]$$

and  $h(x) = 1$ .

Thus  $\sum x_i$  is sufficient for  $\mu$  and  $\sum x_i^2$  is sufficient for  $\sigma^2$ .

Q2 Let  $x_1, x_2, \dots, x_n$  be a random sample from a ~~continuous~~ population with p.d.f

$$f(x; \theta) = \theta x^{\theta-1}; \quad 0 < x < 1, \quad \theta > 0.$$

Show that ~~the sample~~  $t_1 = \sum_{i=1}^n x_i$  is sufficient for  $\theta$ .

Solution: We have  $L = f(x_1, \theta) \cdot f(x_2, \theta) \cdots f(x_n, \theta)$ .

$$L = \theta x_1^{\theta-1} \cdot \theta x_2^{\theta-1} \cdots \theta x_n^{\theta-1}$$

$$= \theta^n \cdot x_1^{\theta-1} \cdot x_2^{\theta-1} \cdots x_n^{\theta-1}$$

$$= \theta^n \cdot \prod_{i=1}^n x_i^{\theta-1}$$

$$= \theta^n \cdot \frac{\prod_{i=1}^n x_i^{\theta}}{\prod_{i=1}^n x_i}$$





$$L = \theta^n \cdot \prod_{i=1}^n x_i \cdot \frac{1}{\prod_{i=1}^n x_i!}$$

$$= g(t_1, \theta) \cdot h(x_1, x_2, \dots, x_n).$$

Hence by Factorization Theorem,

$$t_1 = \sum_{i=1}^n x_i \text{ is sufficient for } \theta.$$

Q3. Let  $x_1, x_2, \dots, x_n$  be a random sample of 'n' observations from Poisson population having parameter  $\lambda$ . Find a sufficient statistic for  $\lambda$ ?

Solution:  $P(X=x) = \frac{e^{-\lambda} \cdot \lambda^x}{x!} \quad x=0,1,2,\dots$   
 $\lambda > 0.$

$$L = P(x_1, \lambda) \cdot P(x_2, \lambda) \dots P(x_n, \lambda)$$

$$= \frac{e^{-\lambda} \cdot \lambda^{x_1}}{x_1!} \cdot \frac{e^{-\lambda} \cdot \lambda^{x_2}}{x_2!} \dots \frac{e^{-\lambda} \cdot \lambda^{x_n}}{x_n!}$$

$$= \frac{e^{-n\lambda} \cdot \lambda^{x_1+x_2+\dots+x_n}}{x_1! \cdot x_2! \dots x_n!}$$

$$= e^{-n\lambda} \cdot \frac{\lambda^{\sum x_i}}{\prod_{i=1}^n x_i!} = g(t_1, \lambda) \cdot h(x_1, x_2, \dots, x_n)$$

where  $g(t_1, \lambda) = e^{-n\lambda} \cdot \lambda^{\sum x_i}$  and  $h(x_1, x_2, \dots, x_n) = \frac{1}{\prod_{i=1}^n x_i!}$



Thus  $t_1 = \sum x_i = n\bar{x}$  is a sufficient statistic for  $\lambda$ .



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Q4. Let  $x_1, x_2, \dots, x_n$  be a random sample of  $n$  observations from a distribution with p.d.f

$$f(x; \theta) = \frac{1}{\Gamma(\theta)} e^{-x} x^{\theta-1}, \quad 0 < x < \infty$$

Find a sufficient statistic for  $\theta$ ?

Solution:  $L = f(x_1, \theta) \cdot f(x_2, \theta) \cdots f(x_n, \theta)$

$$= \frac{1}{\Gamma(\theta)} e^{-x_1} x_1^{\theta-1} \cdot \frac{1}{\Gamma(\theta)} e^{-x_2} x_2^{\theta-1} \cdots \frac{1}{\Gamma(\theta)} e^{-x_n} x_n^{\theta-1}$$

$$= \left( \frac{1}{\Gamma(\theta)} \right)^n \cdot e^{-\sum x_i} \prod_{i=1}^n x_i^{\theta-1}$$

$$= \left( \frac{1}{\Gamma(\theta)} \right)^n \cdot e^{-\sum x_i} \frac{\prod_{i=1}^n x_i^{\theta}}{\prod_{i=1}^n x_i}$$

$$= g(t_1, \theta) \cdot h(x_1, x_2, \dots, x_n)$$

where  $g(t_1, \theta) = \left( \frac{1}{\Gamma(\theta)} \right)^n \cdot \prod_{i=1}^n x_i^{\theta}$

$$\text{and } h(x_1, x_2, \dots, x_n) = \frac{e^{-\sum x_i}}{\prod_{i=1}^n x_i}$$

$\therefore \prod_{i=1}^n x_i$  is sufficient for  $\theta$ .