

## Uniqueness

3- To find unbiased estimator?

$$T = \sum_{i=1}^n x_i \rightarrow \text{From Point (1)}$$

$$\rightarrow \text{We have } \hat{\theta} = \frac{T}{n} = \frac{\sum_{i=1}^n x_i}{n} = \bar{x}$$

$$\rightarrow E(\hat{\theta}) = E(\bar{x}) = \frac{\sum_{i=1}^n E(x_i)}{n} = \frac{n\theta}{n} = \theta$$

$\therefore \hat{\theta} = \bar{x}$  is the best Unique estimator for  $\theta$ .

H.W

Q- let  $X_i \sim \text{Ber}(\theta), i=1, 2, \dots, n$   
is  $\bar{x}$  unique estimator for  $\theta$ ?

## ∴ Uniqueness ∴

To find the Unique estimator for the parameter  $\theta$ , we found the following steps:-

- 1- Find  $T(x)$  the comp. Suff for  $\theta$ .
- 2- Find the expected value of  $T$ .
- 3- Find the unbiased estimator  $\hat{\theta}$  as a function of  $T$ .

⇒ Then  $\hat{\theta}$  is the best unique estimator for  $\theta$ .

Ex 18- let  $X_i \sim \text{Exp}(\theta)$ ,  $i=1, 2, \dots, n$ . is  $\bar{X}$  Unique estimator for  $\theta$ ?

Solution:- ① Find  $T(x)$  the comp S.S for  $\theta$ .

$$\therefore x \sim \text{Exp}(\theta)$$

$$\Rightarrow f(x, \theta) = \exp[\ln f(x, \theta)] = \exp\left[\ln\left(\frac{1}{\theta} \cdot e^{-\frac{x}{\theta}}\right)\right]$$

$$= \exp\left[\ln\left(\frac{1}{\theta}\right) - \frac{x}{\theta}\right] = \exp\left[\ln(1) - \ln(\theta) - \frac{x}{\theta}\right]$$

$$a(\theta) = \ln(1) - \ln(\theta), \quad b(x) = 0, \quad c(\theta) = \frac{-1}{\theta}$$

$$d(x) = x \quad \Rightarrow \quad T = \sum_{i=1}^n d(x_i) = \sum_{i=1}^n x_i \text{ is a Comp S.S. for } \theta.$$

2- Find the expected value of  $T$ .

$$\Rightarrow E(T) = E\left(\sum_{i=1}^n x_i\right) = \sum_{i=1}^n E(x_i) = \sum_{i=1}^n \theta = n\theta.$$

3- To find unbiased estimator for  $T$ ?

$$\text{from Point (1)} \quad \Rightarrow \quad T = \sum_{i=1}^n x_i$$