

# →Unbiased Estimator →UMVUE



# **Unbiased Estimator**

An estimator T is said to be an unbiased estimator of  $\tau(\theta)$  if  $E(T) = \tau(\theta)$  for all  $\theta \in \Omega$ . Otherwise we say that T is a biased estimator of  $\tau(\theta)$ 

Example 1

If  $x_1, x_2, ..., x_n$  denote a random variable of size *n* from f(x) with  $E[X] = \mu$  and  $V(X) = \sigma^2$  then  $E[s^2] = \sigma^2$ 

Example 2

 $X_i \sim EXP(\theta)$ proof that  $\hat{\theta} = \overline{x}$ 



### Obvious question....?

Which estimators are "best" in some sense?  $\rightarrow$  Idea:

We've select the estimator that tend to be closest or "most" concentrated around the true value of the parameter

Say that

 $T_1$  is more concentrated than  $T_2$  about  $\tau(\theta)$  if  $P[\tau(\theta) - \varepsilon < T_1 < \tau(\theta) + \varepsilon], \quad \forall \varepsilon > 0$ 

An estimator is most concentrated if it is more concentrated than any other estimator



# So, it follows from the Chebychev inequality...

T is an unbiased estimator of  $\tau(\theta)$ ,

$$\mathbb{P}[\tau(\theta) - \varepsilon < T < \tau(\theta) + \varepsilon] \ge 1 - \frac{\operatorname{var}(T)}{\varepsilon^2}, \quad \forall \varepsilon > 0$$

#### Example 3

If  $\hat{\theta}_1 = \overline{X}$  and  $\hat{\theta}_2 = n\overline{X}_{1:n}$  then both estimators are biased for  $\theta$ but  $V(\hat{\theta}_1) = \frac{\theta^2}{n}$  and  $V(\hat{\theta}_2) = \theta^2$ 

Which better estimator?





# UMVUE

If  $X_1, X_2, ..., X_n$  denote a random variable of size *n* from  $f(x; \theta)$ . An estimator  $T^*$  of  $\tau(\theta)$  is called a

Uniformly Minimum Variance Unbiased Estimator (UMVUE) of  $\tau(\theta)$  if :

1.  $T^*$  is unbiased for  $\tau(\theta)$ 

2. for any unbiased estimator  $T^*$  of  $\tau(\theta)$ ,  $V(T^*) \leq V(T)$ ,  $\forall \theta \in \Omega$ 



*T* is an unbiased estimator of  $\tau(\theta)$ , then CRLB based on a sample random is

$$V(T) \ge \frac{[\tau'(\theta)]^2}{nE\left[\frac{\partial}{\partial \theta} \ln f(X;\theta)\right]^2}$$





#### Example 4

If  $X_i \sim EXP(\theta)$ 

then find CRLB for  $\theta$  !





# Efficiency

The relative efficiency of an unbiased estimator T of  $\tau(\theta)$ to another unbiased estimator  $T^*$  of  $\tau(\theta)$  is given by  $\operatorname{re}(T,T^*) = \frac{V(T^*)}{V(T)}$ 

An unbiased estimator  $T^*$  of  $\tau(\theta)$  is said to be efficient if  $\operatorname{re}(T,T^*) \leq 1$  for all unbiased estimators T of  $\tau(\theta), \forall \theta \in \Omega$ The efficiency of an unbiased estimator T of  $\tau(\theta)$  is given by:  $\operatorname{e}(T) = \operatorname{re}(T,T^*)$ if  $T^*$  is an efficient estimator of  $\tau(\theta)$ 

### Example 5

consider  $\hat{\theta}_1 = \bar{X}$  and  $\hat{\theta}_2 = n\bar{X}_{1:n}$  are both estimators are biased for  $\theta$ with  $V(\hat{\theta}_1) = \frac{\theta^2}{n}$  and  $V(\hat{\theta}_2) = \theta^2$ 

Which one estimator that an efficient estimator for  $\theta?$ 





# Definition

If *T* is an estimator of  $\tau(\theta)$  then the bias is given by  $b(T) = E[T] - \tau(\theta)$ 

the Mean Squared Error (MSE) of *T* is given by:  $MSE(T) = E[T - \tau(\theta)]^{2}$ 

#### Theorem

If *T* is an estimator of  $\tau(\theta)$  then  $MSE(T) = V(T) + [b(T)]^2$ 

