

Unbiased Estimator

1. Definition

Let $X_1, X_2, \dots, X_n$ be a random sample from a population depending on an unknown parameter θ . An estimator $T = T(X_1, \dots, X_n)$ is called an **unbiased estimator** of θ if its expected value is equal to θ .

$$E_{\theta}[T] = \theta$$

Equivalently, the bias of T is zero:

$$\text{Bias}(T) = E_{\theta}[T] - \theta = 0$$

2. Bias of an Estimator

For an estimator T of a parameter θ , the bias is defined by

$$\text{Bias}(T) = E_{\theta}[T] - \theta$$

Therefore:

$$T \text{ is unbiased} \Leftrightarrow E_{\theta}[T] = \theta$$

3. Example: Sample Mean

Let $X_1, X_2, \dots, X_n$ be a random sample with population mean μ . The sample mean is

$$\bar{X} = (1/n) \sum_{i=1}^n X_i$$

Using linearity of expectation:

$$E[\bar{X}] = (1/n) \sum E[X_i] = (1/n)(n\mu) = \mu$$

Hence,

$\bar{X}$ is an unbiased estimator of μ .

4. Example: Sample Variance

Let the population variance be σ^2 . The usual sample variance is

$$S^2 = 1/(n-1) \sum_{i=1}^n (X_i - \bar{X})^2$$

For a random sample with finite variance σ^2 ,

$$E[S^2] = \sigma^2$$

Therefore, S^2 is an unbiased estimator of σ^2 .

Important: If we instead use the maximum-likelihood form

$$V = 1/n \sum_{i=1}^n (X_i - \bar{X})^2$$

then $E[V] = ((n-1)/n)\sigma^2$, so V is biased downward for σ^2 .

5. Example: Estimating a Function of a Parameter

If T is an unbiased estimator of θ , it does not generally follow that $g(T)$ is an unbiased estimator of $g(\theta)$.

For example, if $\bar{X}$ is an unbiased estimator of μ , then generally

$$E[(\bar{X})^2] \neq \mu^2$$

In fact,

$$E[(X_{\blacksquare})^2] = \text{Var}(X_{\blacksquare}) + \mu^2$$

Thus, $X_{\blacksquare}^2$ is generally a biased estimator of μ^2 .

6. How to Check Whether an Estimator Is Unbiased

Step	What to do
1	Write down the proposed estimator T .
2	Find $E[T]$.
3	Compare $E[T]$ with the parameter being estimated.
4	If $E[T] = \theta$, the estimator is unbiased; otherwise it is biased.

7. Worked Example

Suppose $X_1, \dots, X_n$ have mean μ . Consider the estimator

$$T = (2X_{\blacksquare} + 1)/2$$

Since $E[X_{\blacksquare}] = \mu$,

$$E[T] = (2E[X_{\blacksquare}] + 1)/2 = \mu + 1/2$$

Since $E[T] \neq \mu$, T is a biased estimator of μ . Its bias is

$$\text{Bias}(T) = 1/2$$

8. Important Properties for CSIR-NET

- The expectation of an unbiased estimator equals the parameter it estimates.
- An estimator may be unbiased even though it is not the only unbiased estimator.
- A biased estimator can sometimes have smaller variance than an unbiased estimator.
- Unbiasedness concerns the expectation of the estimator, not the value obtained from one sample.
- If T is unbiased for θ and a, b are constants, then $aT + b$ is unbiased for $a\theta + b$.

9. Linear Transformation Property

If $E[T] = \theta$, then for constants a and b ,

$$E[aT + b] = aE[T] + b = a\theta + b$$

Hence, $aT + b$ is an unbiased estimator of $a\theta + b$.

10. Quick Revision

Definition:

$$T \text{ is unbiased for } \theta \Leftrightarrow E[T] = \theta$$

Bias:

$$\text{Bias}(T) = E[T] - \theta$$

Unbiasedness:

$$\text{Bias}(T) = 0$$

Key examples:

$\bar{X}$ is unbiased for μ .

$S^2 = 1/(n-1) \sum (X_i - \bar{X})^2$ is unbiased for σ^2 .

$1/n \sum (X_i - \bar{X})^2$ is biased for σ^2 .

CSIR-NET Shortcut

Whenever a question asks whether an estimator is unbiased, calculate its expectation first.