

ESTIMATING THE MEAN OF A SAMPLE

random variables : X_1, X_2, \dots, X_n

→ assume all follow distribution F_X

→ all independent (i.i.d.)

→ Let mean of F_X be $\theta \in \mathbb{R}$
where θ is unknown

Q: How to estimate θ ?

Prop 1.2.6 $\bar{X} = \frac{1}{n} \sum_{i=1}^n x_i \Rightarrow E[\bar{X}] = \theta$

Now suppose we observe $x_1, x_2, \dots, x_n \in \mathbb{R}$

→ observations of X_1, X_2, \dots, X_n , respectively

To estimate θ , we compute

$$\hat{\theta} = \frac{1}{n} \sum_{i=1}^n x_i = \bar{x} \quad \hat{\theta} = x_1 + 2x_2 + x_3^5$$

estimate of θ

Point estimators ✓

$$\hat{\theta} = \hat{\theta}(x_1, x_2, \dots, x_n) = r(x_1, \dots, x_n)$$

example $\hat{\theta} = \frac{1}{n} \sum_{i=1}^n x_i$

Definition 1.5.18

The ESTIMATION ERROR of the estimator $\hat{\theta}$ of a parameter θ is defined as $\hat{\theta} - \theta$

Definition 1.5.19 The BIAS of the

estimator $\hat{\theta}$ of a parameter θ is denoted $b_\theta(\hat{\theta})$ and is defined as $b_\theta(\hat{\theta}) = E[\hat{\theta}] - \theta$

$$E[\hat{\theta} - \theta] = E[\hat{\theta}] - \theta = b_\theta(\hat{\theta})$$

Definition: If $E[\hat{\theta}] = \theta$
the $\hat{\theta}$ is UNBIASED

because $b_\theta(\hat{\theta}) = E[\hat{\theta}] - \theta = 0$

$\hat{\theta} = \bar{X}$ as an estimator of
mean θ of F_X (in Prop 1.2.6)

$$E[\hat{\theta}] = E[\bar{X}] = \theta$$

\Rightarrow unbiased

$$\hat{\Gamma} = s^2 \text{ estimator of variance } \sigma^2 \text{ of } F_X$$
$$E[\hat{\Gamma}] = E[s^2] = \sigma^2 \quad (\text{in 1.2.6})$$

\Rightarrow unbiased

Definition 1.5.23

The MEAN SQUARED ERROR of
the estimator $\hat{\theta}$ of parameter θ
is defined as $E[(\hat{\theta} - \theta)^2]$

expected squared deviation of $\hat{\theta}$ from θ

Theorem 1.5-24

The mean squared error of an estimator $\hat{\theta}$ of a parameter θ can be expressed in terms of its bias and variance

$$E[(\hat{\theta} - \theta)^2] = [b_{\theta}(\hat{\theta})]^2 + \text{Var}[\hat{\theta}]$$

Proof:

$$\begin{aligned} E[x^2] &= (E[x])^2 + \text{Var}[x] \\ E[(\hat{\theta} - \theta)^2] &= (E[\hat{\theta} - \theta])^2 + \text{Var}[\hat{\theta} - \theta] \end{aligned}$$

(Exercise 1.1.5)

$$\text{bias: } b_{\theta}[\hat{\theta}] = E[\hat{\theta}] - \theta = E[\hat{\theta} - \theta]$$

$$\text{Var}[\hat{\theta} - \theta] = \text{Var}[\hat{\theta}]$$

$$\Rightarrow E[(\hat{\theta} - \theta)^2] = (b_{\theta}[\hat{\theta}])^2 + \text{Var}[\hat{\theta}]$$