

# Point Estimation

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# Statistical Inference

With the knowledge of **statistic**, **sampling distribution** and **central limit theorem**, what should we do next? **Statistical Inference!**

- 1 **Estimation**: Point Estimation and Interval Estimation
- 2 **Hypothesis Test**

# Point Estimator

Let  $X_1, \dots, X_n$  be sampled from a population described by a pdf or pmf  $f(x|\theta)$ . The parameter  $\theta$  is unknown. The goal is to find a good estimator for  $\theta$ .

## Definition (Point Estimator)

A point estimator is any function  $W(X_1, \dots, X_n; C)$  of a sample; that is, any statistic a point estimator.

Two jobs in point estimation:

- 1 how to find a point estimator?
- 2 how to evaluate a candidate point estimator?

# Example 1

**Question:** We have a bent coin and are interested in  $p$ , the probability of heads coming up on a single toss. How can we estimate  $p$ ?

**Analysis:**

- ①  $X_i$  denotes the  $i$ th toss result, i.e. 1 if head and 0 if tail.  $i = 1, \dots, n$ .
- ②  $\hat{p} = \frac{1}{n} \sum_{i=1}^n X_i$ .

## Remark

Some terminology is needed to be attention.

- ①  $p$ : target parameter or estimand;
- ②  $\hat{p}$ : an estimator or a point estimator for  $p$ ;
- ③ the value of  $\hat{p}$ : an estimate or a point estimate.

# Criteria to Judge an Estimator

Two important concepts

- ➊ **Bias:** The bias of an estimator  $\hat{\theta}$  of  $\theta$  is  $B(\hat{\theta}) = \mathbb{E}(\hat{\theta}) - \theta$ .
- ➋ **Mean Square Error (MSE):** The MSE of an estimator  $\hat{\theta}$  of  $\theta$  is  $MSE(\hat{\theta}) = \mathbb{E}(\hat{\theta} - \theta)^2$ .

## Remark

Attention to these two criteria from the following points.

- ➊ If  $B(\hat{\theta}) = 0$ ,  $\hat{\theta}$  is **unbiased** for  $\theta$ .
- ➋  $MSE(\hat{\theta}) = Var(\hat{\theta}) + B(\hat{\theta})^2$ .
- ➌ **Bias** is used to assess accuracy while **MSE** can also measure precision.

## Example 2

**Question:** Two numbers are randomly chosen between 0 and  $c$ . They are  $x = 3.6$  and  $y = 5.4$ . Consider the two following estimates of  $c$ :

$$u = x + y = 3.6 + 5.4 = 9.0, v = \max(x, y) = \max(3.6, 5.4) = 5.4.$$

Which estimate should we choose? Why?

**Analysis:**

- ①  $X, Y \sim i.i.d.U(0, c)$ .
- ②  $U = X + Y$  and  $V = \max(X, Y)$ .
- ③  $\mathbb{E}(U) = \mathbb{E}(X + Y) = c$ ;  $B(U) = \mathbb{E}(U) - c = 0$ ;  
 $MSE(U) = Var(U) = Var(X) + Var(Y) = c^2/6$ .
- ④ CDF for  $V$ :  $F(v) = P(V < v) = P(X < v, Y < v) = P(X < v)P(Y < v) = (v/c)^2$  and then its pdf is  $f(v) = F'(v) = 2v/c^2$ . So  
 $\mathbb{E}V = \int_0^c v \frac{2v}{c^2} dv = \frac{2c}{3}$ ,  $\mathbb{E}V^2 = \int_0^c v^2 \frac{2v}{c^2} dv = \frac{c^2}{2}$ ,  $B(V) = -\frac{c}{3}$ ,  
 $MSE(V) = \frac{c^2}{6}$ .

# Modification for an Estimator

**Conclusion:** the estimator  $U$  is unbiased but has larger MSE than the estimator  $V$ .

A modified estimator  $W = \frac{3V}{2}$  is unbiased since  $\mathbb{E}(V) = \frac{2c}{3}$ . For this new estimator  $W$ ,  $MSE(W) = \frac{c^2}{8}$ . In view of this,  $W$  is a better estimator than  $V$  and  $U$ .



# Assess Two Important Estimators

## Unbiased Specific Estimators:

Suppose that  $X_1, \dots, X_n \sim i.i.d.(\mu, \sigma^2)$ . Consider the sample mean  $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$  and the sample variance  $S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$ . Then  $\mathbb{E}(\bar{X}) = \mu$  and  $\mathbb{E}(S^2) = \sigma^2$ .

- ①  $\mathbb{E}(\bar{X}) = \frac{1}{n} \sum_{i=1}^n \mathbb{E}(X_i) = \mu.$
- ②  $\mathbb{E}(S^2) = \frac{1}{n-1} \sum_{i=1}^n \mathbb{E}(X_i - \bar{X})^2 = \frac{n}{n-1} \mathbb{E}(X_1 - \bar{X})^2 =$   
 $\frac{n}{n-1} (Var(X_1) + Var(\bar{X}) - 2Cov(X_1, \bar{X})) =$   
 $\frac{n}{n-1} \left( \sigma^2 + \frac{\sigma^2}{n} - 2\frac{\sigma^2}{n} \right) = \sigma^2.$

## Example 3

**Question:** A bottling machine dispenses volumes independently with mean  $\mu$  and variance  $\sigma^2$ .  $n = 3$  bottles are randomly sampled from the output of the machine, and their volumes are 1.9, 1.4, 1.8 litres. Find unbiased estimates of  $\mu$  and  $\sigma^2$ .

**Analysis:**

- ①  $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i = 1.7;$
- ②  $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = 0.07.$

# Summary

- 1 What is a point estimator?
- 2 How to find a point estimator?
- 3 How to evaluate a point estimator? Unbiased, smaller MSE.