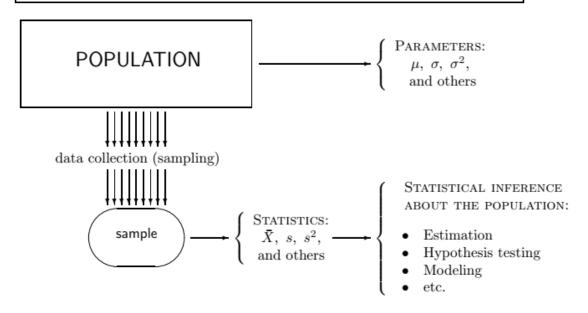
### DEFINITION 8.1 -

A **population** consists of all units of interest. Any numerical characteristic of a population is a **parameter**. A **sample** consists of observed units collected from the population. It is used to make statements about the population. Any function of a sample is called **statistic**.



## Sample Statistics (Used as estimators for Population Parameters)

### Mean

# DEFINITION 8.3

Sample mean  $\bar{X}$  is the arithmetic average,

$$\bar{X} = \frac{X_1 + \ldots + X_n}{n}$$

This is an unbiased, consistent estimator of Population Average.

## DEFINITION 8.4 -

An estimator  $\hat{\theta}$  is **unbiased** for a parameter  $\theta$  if its expectation equals the parameter,

$$\mathbf{E}(\hat{\theta}) = \theta$$

for all possible values of  $\theta$ .

**Bias** of  $\hat{\theta}$  is defined as  $Bias(\hat{\theta}) = \mathbf{E}(\hat{\theta} - \theta)$ .

#### DEFINITION 8.5 -

An estimator  $\hat{\theta}$  is **consistent** for a parameter  $\theta$  if the probability of its sampling error of any magnitude converges to 0 as the sample size increases to infinity. Stating it rigorously,

$$P\left\{|\hat{\theta} - \theta| > \varepsilon\right\} \to 0 \text{ as } n \to \infty$$

for any  $\varepsilon > 0$ . That is, when we estimate  $\theta$  from a large sample, the estimation error  $|\hat{\theta} - \theta|$  is unlikely to exceed  $\varepsilon$ , and it does it with smaller and smaller probabilities as we increase the sample size.

Asymptotic Normality of Sample Mean. By Central Limit Theorem, Sample mean follows a normal distribution as number of samples increases

i.e.

$$Z = \frac{\bar{X} - \mathbf{E}\bar{X}}{\mathrm{Std}\bar{X}} = \frac{\bar{X} - \mu}{\sigma \! \! \sqrt{n}}$$
 converges to a Normal(0,1) as n  $\rightarrow \infty$ 

Variance and Std. Deviation

DEFINITION 8.8 -

For a sample  $(X_1, X_2, \dots, X_n)$ , a sample variance is defined as

$$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (X_{i} - \bar{X})^{2}.$$
 (8.4)

It measures variability among observations and estimates the population variance  $\sigma^2 = \text{Var}(X)$ .

Sample standard deviation is a square root of a sample variance,

$$s = \sqrt{s^2}$$
.

It measures variability in the same units as X and estimates the population standard deviation  $\sigma = \text{Std}(X)$ .

Alt formula for Variance:

$$s^{2} = \frac{\sum_{i=1}^{n} X_{i}^{2} - n\bar{X}^{2}}{n-1}.$$

DEFINITION 8.6 -

Median means a "central" value.

Sample median  $\hat{M}$  is a number that is exceeded by at most a half of observations and is preceded by at most a half of observations.

**Population median** M is a number that is exceeded with probability no greater than 0.5 and is preceded with probability no greater than 0.5. That is, M is such that

$$\left\{ \begin{array}{lcl} {\bf P} \left\{ {X > M} \right\} & \le & 0.5 \\ {\bf P} \left\{ {X < M} \right\} & \le & 0.5 \end{array} \right.$$

Sample median If n is odd, the  $\left(\frac{n+1}{2}\right)$ -th smallest observation is a median. If n is even, any number between the  $\left(\frac{n}{2}\right)$ -th smallest and the  $\left(\frac{n+2}{2}\right)$ -th smallest observations is a median.

Shape of a distribution (comparing mean and median)

Symmetric distribution 
$$\Rightarrow M = \mu$$
  
Right-skewed distribution  $\Rightarrow M < \mu$   
Left-skewed distribution  $\Rightarrow M > \mu$ 

Quantiles, percentiles and quartiles

DEFINITION 8.7 -

A p-quantile of a population is such a number x that solves equations

$$\left\{ \begin{array}{lcl} \boldsymbol{P}\left\{X < x\right\} & \leq & p \\ \boldsymbol{P}\left\{X > x\right\} & \leq & 1-p \end{array} \right.$$

A sample p-quantile is any number that exceeds at most 100p% of the sample, and is exceeded by at most 100(1-p)% of the sample.

A  $\gamma$ -percentile is  $(0.01\gamma)$ -quantile.

First, second, and third **quartiles** are the 25th, 50th, and 75th percentiles. They split a population or a sample into four equal parts.

A median is at the same time a 0.5-quantile, 50th percentile, and 2nd quartile.

#### IQR and outliers.

#### DEFINITION 8.10 -

An **interquartile range** is defined as the difference between the first and the third quartiles,

$$IQR = Q_3 - Q_1.$$

It measures variability of data. Not much affected by outliers, it is often used to detect them. IQR is estimated by the sample interquartile range

$$\widehat{IQR} = \hat{Q}_3 - \hat{Q}_1.$$

Any samples that are less than  $Q_1 - 1.5(IQR)$  or more than  $Q_3 + 1.5(IQR)$  can be treated as potential outliers.

Standard error of any estimator is its std deviation.

$$\sigma(\hat{\theta}) = \text{standard error of estimator } \hat{\theta} \text{ of parameter } \theta$$
 $s(\hat{\theta}) = \text{estimated standard error } = \hat{\sigma}(\hat{\theta})$ 

## **Graphical Statistics:**

Can be used to visualize the given samples to make observations about the nature of the population

- Histograms are bar charts for columns for each bin
  - If height of bin is freq count: Frequency histogram
  - o If height of bin is proportion of data: Relative Frequency histogram
  - Each sample value can have its own bin, or you can have multiple nearby values in one bin.
  - o You can use histograms to guess shape of distribution.
- Stem and leaf plots
  - Choose the stem such that the values are not all limited to one stem.
  - o Distribute values to each stem and sort he leaf values.
  - You can use this to calculate mean, median and guess shape of distribution
  - It can also be used to compare two distributions
- Boxplots
  - Boxplots are based on 5-point summaries
    - $< \min(X_i)$ ,  $\widehat{Q_1}$ ,  $\widehat{M}$ ,  $\widehat{Q_3}$ ,  $\max(X_i) >$
  - Represent sample mean with small cross. Draw a box between sample Q1 and Q3 and draw a line for sample median. Draw whiskers to smallest sample and largest sample that is within the 1.5 IQR range. Draw dots for all samples outside the 1.5 IQR range.
  - Can be used to compare multiple distributions.