

## CHAPTER 1

### 1. INTRODUCTION

#### **Definition and classifications of statistics**

**Definition:** We can define statistics in two ways.

1. Plural sense (lay man definition).

It is an aggregate or collection of numerical facts.

2. Singular sense (formal definition)

Statistics is defined as the science of collecting, organizing, presenting, analyzing and interpreting numerical data for the purpose of assisting in making a more effective decision.

#### **Classifications:**

Depending on how data can be used statistics is sometimes divided into two main areas or branches.

1. **Descriptive Statistics:** is concerned with summary calculations, graphs, charts and tables.

2. **Inferential Statistics:** is a method used to generalize from a sample to a population. For example, the average income of all families (the population) in Ethiopia can be estimated from figures obtained from a few hundred (the sample) families.

- It is important because statistical data usually arises from sample.
- Statistical techniques based on probability theory are required.

#### **Stages in Statistical Investigation**

There are five stages or steps in any statistical investigation.

1. Collection of data: the process of measuring, gathering, assembling the raw data upon which the statistical investigation is to be based.

- Data can be collected in a variety of ways; one of the most common methods is through the use of survey. Survey can also be done in different methods, three of the most common methods are:

- Telephone survey
- Mailed questionnaire
- Personal interview.

**Exercise:** discuss the advantage and disadvantage of the above three methods with respect to each other.

2. Organization of data: Summarization of data in some meaningful way, e.g table form

3. Presentation of the data: The process of re-organization, classification, compilation, and summarization of data to present it in a meaningful form.
4. Analysis of data: The process of extracting relevant information from the summarized data, mainly through the use of elementary mathematical operation.
5. Inference of data: The interpretation and further observation of the various statistical measures through the analysis of the data by implementing those methods by which conclusions are formed and inferences made.
  - Statistical techniques based on probability theory are required.

### **Definitions of some terms**

- a. **Statistical Population:** It is the collection of all possible observations of a specified characteristic of interest (possessing certain common property) and being under study. An example is all of the students in AAU 3101 course in this term.
- b. **Sample:** It is a subset of the population, selected using some sampling technique in such a way that they represent the population.
- c. **Sampling:** The process or method of sample selection from the population.
- d. **Sample size:** The number of elements or observation to be included in the sample.
- e. **Census:** Complete enumeration or observation of the elements of the population. Or it is the collection of data from every element in a population
- f. **Parameter:** Characteristic or measure obtained from a population.
- g. **Statistic:** Characteristic or measure obtained from a sample.
- h. **Variable:** It is an item of interest that can take on many different numerical values.

### **Types of Variables or Data:**

1. **Qualitative Variables** are nonnumeric variables and can't be measured. Examples include gender, religious affiliation, and state of birth.
2. **Quantitative Variables** are numerical variables and can be measured. Examples include balance in checking account, number of children in family. Note that quantitative variables are either discrete (which can assume only certain values, and there are usually "gaps" between the values, such as

the number of bedrooms in your house) or continuous (which can assume any value within a specific range, such as the air pressure in a tire.)

## **Applications, Uses and Limitations of statistics**

### **Applications of statistics:**

- In almost all fields of human endeavor.
- Almost all human beings in their daily life are subjected to obtaining numerical facts e.g. about price.
- Applicable in some process e.g. invention of certain drugs, extent of environmental pollution.
- In industries especially in quality control area.

### **Uses of statistics:**

The main function of statistics is to enlarge our knowledge of complex phenomena. The following are some uses of statistics:

1. It presents facts in a definite and precise form.
2. Data reduction.
3. Measuring the magnitude of variations in data.
4. Furnishes a technique of comparison
5. Estimating unknown population characteristics.
6. Testing and formulating of hypothesis.
7. Studying the relationship between two or more variables.
8. Forecasting future events.

### **Limitations of statistics**

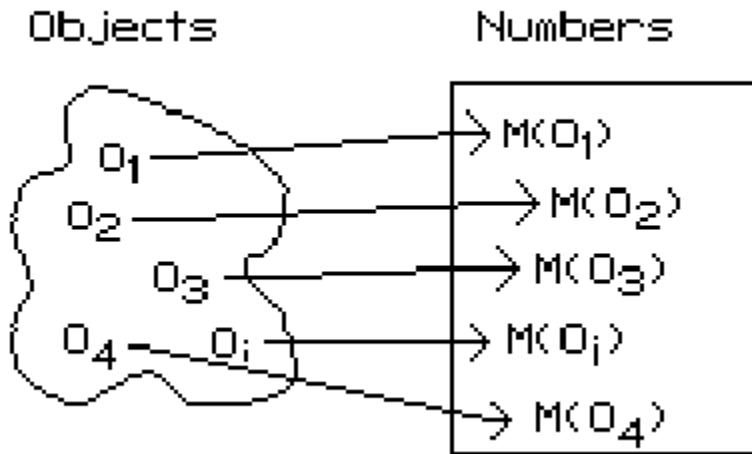
As a science statistics has its own limitations. The following are some of the limitations:

- Deals with only quantitative information.
- Deals with only aggregate of facts and not with individual data items.
- Statistical data are only approximately and not mathematically correct.
- Statistics can be easily misused and therefore should be used by experts.

## **Scales of measurement**

Proper knowledge about the nature and type of data to be dealt with is essential in order to specify and apply the proper statistical method for their analysis and inferences. Measurement scale refers to the property of value assigned to the data based on the properties of order, distance and fixed zero.

In mathematical terms measurement is a functional mapping from the set of objects  $\{O_i\}$  to the set of real numbers  $\{M(O_i)\}$ .



The goal of measurement systems is to structure the rule for assigning numbers to objects in such a way that the relationship between the objects is preserved in the numbers assigned to the objects. The different kinds of relationships preserved are called properties of the measurement system.

## Order

The property of order exists when an object that has more of the attribute than another object, is given a bigger number by the rule system. This relationship must hold for all objects in the "real world".

The property of ORDER exists

When for all  $i, j$  if  $O_i > O_j$ , then  $M(O_i) > M(O_j)$ .

## Distance

The property of distance is concerned with the relationship of differences between objects. If a measurement system possesses the property of distance it means that the unit of measurement means the same thing throughout the scale of numbers. That is, an inch is an inch, no matter where it falls - immediately ahead or a mile down the road.

More precisely, an equal difference between two numbers reflects an equal difference in the "real world" between the objects that were assigned the numbers. In order to define the property of distance in the mathematical

notation, four objects are required:  $O_i$ ,  $O_j$ ,  $O_k$ , and  $O_l$ . The difference between objects is represented by the "-" sign;  $O_i - O_j$  refers to the actual "real world" difference between object  $i$  and object  $j$ , while  $M(O_i) - M(O_j)$  refers to differences between numbers.

The property of DISTANCE exists, for all  $i, j, k, l$

If  $O_i - O_j \geq O_k - O_l$  then  $M(O_i) - M(O_j) \geq M(O_k) - M(O_l)$ .

### **Fixed Zero**

A measurement system possesses a rational zero (fixed zero) if an object that has none of the attribute in question is assigned the number zero by the system of rules. The object does not need to really exist in the "real world", as it is somewhat difficult to visualize a "man with no height". The requirement for a rational zero is this: if objects with none of the attribute did exist would they be given the value zero. Defining  $O_0$  as the object with none of the attribute in question, the definition of a rational zero becomes:

The property of FIXED ZERO exists if  $M(O_0) = 0$ .

The property of fixed zero is necessary for ratios between numbers to be meaningful.

## **SCALE TYPES**

Measurement is the assignment of numbers to objects or events in a systematic fashion. Four levels of measurement scales are commonly distinguished: nominal, ordinal, interval, and ratio and each possessed different properties of measurement systems.

### **Nominal Scales**

Nominal scales are measurement systems that possess none of the three properties stated above.

- Level of measurement which classifies data into mutually exclusive, all inclusive categories in which no order or ranking can be imposed on the data.
- No arithmetic and relational operation can be applied.

Examples:

- Political party preference (Republican, Democrat, or Other,)
- Sex (Male or Female.)
- Marital status(married, single, widow, divorce)
- Country code
- Regional differentiation of Ethiopia.

## **Ordinal Scales**

Ordinal Scales are measurement systems that possess the property of order, but not the property of distance. The property of fixed zero is not important if the property of distance is not satisfied.

- Level of measurement which classifies data into categories that can be ranked. Differences between the ranks do not exist.
- Arithmetic operations are not applicable but relational operations are applicable.
- Ordering is the sole property of ordinal scale.

Examples:

- Letter grades (A, B, C, D, F).
- Rating scales (Excellent, Very good, Good, Fair, poor).
- Military status.

## **Interval Scales**

Interval scales are measurement systems that possess the properties of Order and distance, but not the property of fixed zero.

- Level of measurement which classifies data that can be ranked and differences are meaningful. However, there is no meaningful zero, so ratios are meaningless.
- All arithmetic operations except division are applicable.
- Relational operations are also possible.

Examples:

- IQ
- Temperature in °F.

## **Ratio Scales**

Ratio scales are measurement systems that possess all three properties: order, distance, and fixed zero. The added power of a fixed zero allows ratios of numbers to be meaningfully interpreted; i.e. the ratio of Bekele's height to Martha's height is 1.32, whereas this is not possible with interval scales.

- Level of measurement which classifies data that can be ranked, differences are meaningful, and there is a true zero. True ratios exist between the different units of measure.
- All arithmetic and relational operations are applicable.

Examples:

- Weight
- Height
- Number of students
- Age

The following present a list of different attributes and rules for assigning numbers to objects. Try to classify the different measurement systems into one of the four types of scales. (Exercise)

1. Your checking account number as a name for your account.
2. Your checking account balance as a measure of the amount of money you have in that account.
3. The order in which you were eliminated in a spelling bee as a measure of your spelling ability.
4. Your score on the first statistics test as a measure of your knowledge of statistics.
5. Your score on an individual intelligence test as a measure of your intelligence.
6. The distance around your forehead measured with a tape measure as a measure of your intelligence.
7. A response to the statement "Abortion is a woman's right" where "Strongly Disagree" = 1, "Disagree" = 2, "No Opinion" = 3, "Agree" = 4, and "Strongly Agree" = 5, as a measure of attitude toward abortion.
8. Times for swimmers to complete a 50-meter race
9. Months of the year Meskerm, Tikimit...
10. Socioeconomic status of a family when classified as low, middle and upper classes.
11. Blood type of individuals, A, B, AB and O.

- 12. Pollen counts provided as numbers between 1 and 10 where 1 implies there is almost no pollen and 10 that it is rampant, but for which the values do not represent an actual counts of grains of pollen.
- 13. Regions numbers of Ethiopia (1, 2, 3 etc.)
- 14. The number of students in a college;
- 15. the net wages of a group of workers;
- 16. the height of the men in the same town;

## **INTRODUCTION TO METHODS OF DATA COLLECTION**

There are two sources of data:

- 1. Primary Data
  - Data measured or collect by the investigator or the user directly from the source.
  - Two activities involved: planning and measuring.
    - a) Planning:
      - Identify source and elements of the data.
      - Decide whether to consider sample or census.
      - If sampling is preferred, decide on sample size, selection method,... etc
      - Decide measurement procedure.
      - Set up the necessary organizational structure.
    - b) Measuring: there are different options.
      - Focus Group
      - Telephone Interview
      - Mail Questionnaires
      - Door-to-Door Survey
      - Mall Intercept
      - New Product Registration
      - Personal Interview and
      - Experiments are some of the sources for collecting the primary data.
- 2. Secondary Data
  - Data gathered or compiled from published and unpublished sources or files.
  - When our source is secondary data check that:
    - The type and objective of the situations.
    - The purpose for which the data are collected and compatible with the present problem.

- The nature and classification of data is appropriate to our problem.
- There are no biases and misreporting in the published data.

Note: Data which are primary for one may be secondary for the other.

## CHAPTER 2

### 2. METHODS OF DATA PRESENTATION

Having collected and edited the data, the next important step is to organize it. That is to present it in a readily comprehensible condensed form that aids in order to draw inferences from it. It is also necessary that the like be separated from the unlike ones.

The presentation of data is broadly classified into the following two categories:

- Tabular presentation
- Diagrammatic and Graphic presentation.

The process of arranging data into classes or categories according to similarities technically is called *classification*.

Classification is a preliminary and it prepares the ground for proper presentation of data.

Definitions:

- Raw data: recorded information in its original collected form, whether it be counts or measurements, is referred to as raw data.
- Frequency: is the number of values in a specific class of the distribution.
- Frequency distribution: is the organization of raw data in table form using classes and frequencies.

There are three basic types of frequency distributions

- Categorical frequency distribution

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- Ungrouped frequency distribution
- Grouped frequency distribution

There are specific procedures for constructing each type.

### 1) Categorical frequency Distribution:

Used for data that can be placed in specific categories such as nominal, or ordinal. e.g. marital status.

Example: a social worker collected the following data on marital status for 25 persons.(M=married, S=single, W=widowed, D=divorced)

M	S	D	W	D
S	S	M	M	M
W	D	S	M	M
W	D	D	S	S
S	W	W	D	D

Solution:

Since the data are categorical, discrete classes can be used. There are four types of marital status M, S, D, and W. These types will be used as class for the distribution. We follow procedure to construct the frequency distribution.

Step 1: Make a table as shown.

Class	Tally	Frequency	Percent
(1)	(2)	(3)	(4)
M			
S			
D			

Step 2: Tally the data and place the result in column (2).

Step 3: Count the tally and place the result in column (3).

Step 4: Find the percentages of values in each class by using;

$$\% = \frac{f}{n} * 100 \quad \text{Where } f = \text{frequency of the class, } n = \text{total number of value.}$$

Percentages are not normally a part of frequency distribution but they can be added since they are used in certain types diagrammatic such as pie charts.

Step 5: Find the total for column (3) and (4).

Combining all the steps one can construct the following frequency distribution.

Class	Tally	Frequency	Percent
(1)	(2)	(3)	(4)
M	/\	5	20
S	/\  //	7	28
D	/\  //	7	28
W	/\	6	24

## 2) Ungrouped frequency Distribution:

-Is a table of all the potential raw score values that could possibly occur in the data along with the number of times each actually occurred.

-Is often constructed for small set or data on discrete variable.

### Constructing ungrouped frequency distribution:

- First find the smallest and largest raw score in the collected data.
- Arrange the data in order of magnitude and count the frequency.
- To facilitate counting one may include a column of tallies.

Example:

The following data represent the mark of 20 students.

80	76	90	85	80
70	60	62	70	85
65	60	63	74	75
76	70	70	80	85

Construct a frequency distribution, which is ungrouped.

Solution:

Step 1: Find the range, Range=Max-Min=90-60=30.

Step 2: Make a table as shown

Step 3: Tally the data.

Step 4: Compute the frequency.

Mark	Tally	Frequency
60	//	2
62	/	1
63	/	1
65	/	1
70	////	4
74	/	1
75	//	2
76	/	1
80	///	3
85	///	3
90	/	1

Each individual value is presented separately, that is why it is named ungrouped frequency distribution.

### 3) Grouped frequency Distribution:

-When the range of the data is large, the data must be grouped in to classes that are more than one unit in width.

#### Definitions:

- **Grouped Frequency Distribution:** a frequency distribution when several numbers are grouped in one class.
- **Class limits:** Separates one class in a grouped frequency distribution from another. The limits could actually appear in the data and have gaps between the upper limits of one class and lower limit of the next.
- **Units of measurement (U):** the distance between two possible consecutive measures. It is usually taken as 1, 0.1, 0.01, 0.001, -----.
- **Class boundaries:** Separates one class in a grouped frequency distribution from another. The boundaries have one more decimal places than the raw data and therefore do not appear in the data. There is no gap between the upper boundary of one class and lower boundary of the next class. The lower class boundary is found by subtracting  $U/2$  from the corresponding lower class limit and the upper class boundary is found by adding  $U/2$  to the corresponding upper class limit.
- **Class width:** the difference between the upper and lower class boundaries of any class. It is also the difference between the lower limits of any two consecutive classes or the difference between any two consecutive class marks.
- **Class mark (Mid points):** it is the average of the lower and upper class limits or the average of upper and lower class boundary.
- **Cumulative frequency:** is the number of observations less than/more than or equal to a specific value.
- **Cumulative frequency above:** it is the total frequency of all values greater than or equal to the lower class boundary of a given class.
- **Cumulative frequency below:** it is the total frequency of all values less than or equal to the upper class boundary of a given class.
- **Cumulative Frequency Distribution (CFD):** it is the tabular arrangement of class interval together with their corresponding cumulative frequencies. It can be more than or less than type, depending on the type of cumulative frequency used.
- **Relative frequency (rf):** it is the frequency divided by the total frequency.

- **Relative cumulative frequency (rcf):** it is the cumulative frequency divided by the total frequency.

### **Guidelines for classes**

1. There should be between 5 and 20 classes.
2. The classes must be mutually exclusive. This means that no data value can fall into two different classes
3. The classes must be all inclusive or exhaustive. This means that all data values must be included.
4. The classes must be continuous. There are no gaps in a frequency distribution.
5. The classes must be equal in width. The exception here is the first or last class. It is possible to have an "below ..." or "... and above" class. This is often used with ages.

### **Steps for constructing Grouped frequency Distribution**

1. Find the largest and smallest values
2. Compute the Range( $R$ ) = Maximum - Minimum
3. Select the number of classes desired, usually between 5 and 20 or use Sturges rule  $k = 1 + 3.32 \log n$  where  $k$  is *number of classes desired* and  $n$  is *total number of observation*.
4. Find the class width by dividing the range by the number of classes and rounding up, not off.  $w = \frac{R}{k}$ .
5. Pick a suitable starting point less than or equal to the minimum value. The starting point is called the lower limit of the first class. Continue to add the class width to this lower limit to get the rest of the lower limits.
6. To find the upper limit of the first class, subtract U from the lower limit of the second class. Then continue to add the class width to this upper limit to find the rest of the upper limits.
7. Find the boundaries by subtracting  $U/2$  units from the lower limits and adding  $U/2$  units from the upper limits. The boundaries are also half-way between the upper limit of one class and the lower limit of the next class. !may not be necessary to find the boundaries.
8. Tally the data.
9. Find the frequencies.

10. Find the cumulative frequencies. Depending on what you're trying to accomplish, it may not be necessary to find the cumulative frequencies.
11. If necessary, find the relative frequencies and/or relative cumulative frequencies

Example\*:

Construct a frequency distribution for the following data.

11	29	6	33	14	31	22	27	19	20
18	17	22	38	23	21	26	34	39	27

Solutions:

Step 1: Find the highest and the lowest value H=39, L=6

Step 2: Find the range; R=H-L=39-6=33

Step 3: Select the number of classes desired using Sturges formula;

$$k = 1 + 3.32 \log n = 1 + 3.32 \log (20) = 5.32 = 6 \text{ (rounding up)}$$

Step 4: Find the class width; w=R/k=33/6=5.5=6 (rounding up)

Step 5: Select the starting point, let it be the minimum observation.

- 6, 12, 18, 24, 30, 36 are the lower class limits.

Step 6: Find the upper class limit; e.g. the first upper class=12-U=12-1=11

- 11, 17, 23, 29, 35, 41 are the upper class limits.

So combining step 5 and step 6, one can construct the following classes.

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Class limits

6 – 11

12 – 17

18 – 23

24 – 29

30 – 35

36 – 41

Step 7: Find the class boundaries;

E.g. for class 1 Lower class boundary=6-U/2=5.5

Upper class boundary =11+U/2=11.5

- Then continue adding w on both boundaries to obtain the rest boundaries. By doing so one can obtain the following classes.

Class boundary

5.5 – 11.5

11.5 – 17.5

17.5 – 23.5

23.5 – 29.5

29.5 – 35.5

35.5 – 41.5

Step 8: tally the data.

Step 9: Write the numeric values for the tallies in the frequency column.

Step 10: Find cumulative frequency.

Step 11: Find relative frequency or/and relative cumulative frequency.

The complete frequency distribution follows:

Class limit	Class boundary	Class Mark	Tally	Freq.	Cf (less than type)	Cf (more than type)	rf.	rcf (less than type)
6 – 11	5.5 – 11.5	8.5	//	2	2	20	0.10	0.10
12 – 17	11.5 – 17.5	14.5	//	2	4	18	0.10	0.20

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18 – 23	17.5 – 23.5	20.5	XXXX	7	11	16	0.35	0.55
24 – 29	23.5 – 29.5	26.5	///	4	15	9	0.20	0.75
30 – 35	29.5 – 35.5	32.5	///	3	18	5	0.15	0.90
36 – 41	35.5 – 41.5	38.5	//	2	20	2	0.10	1.00

### **Diagrammatic and Graphic presentation of data.**

These are techniques for presenting data in visual displays using geometric and pictures.

Importance:

- They have greater attraction.
- They facilitate comparison.
- They are easily understandable.

-Diagrams are appropriate for presenting discrete data.

-The three most commonly used diagrammatic presentation for discrete as well as qualitative data are:

- Pie charts
- Pictogram
- Bar charts

### **Pie chart**

- A pie chart is a circle that is divided in to sections or wedges according to the percentage of frequencies in each category of the distribution. The angle of the sector is obtained using:

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$$\text{Angle of sector} = \frac{\text{Value of the part}}{\text{the whole quantity}} * 100$$

Example: Draw a suitable diagram to represent the following population in a town.

Men	Women	Girls	Boys
2500	2000	4000	1500

Solutions:

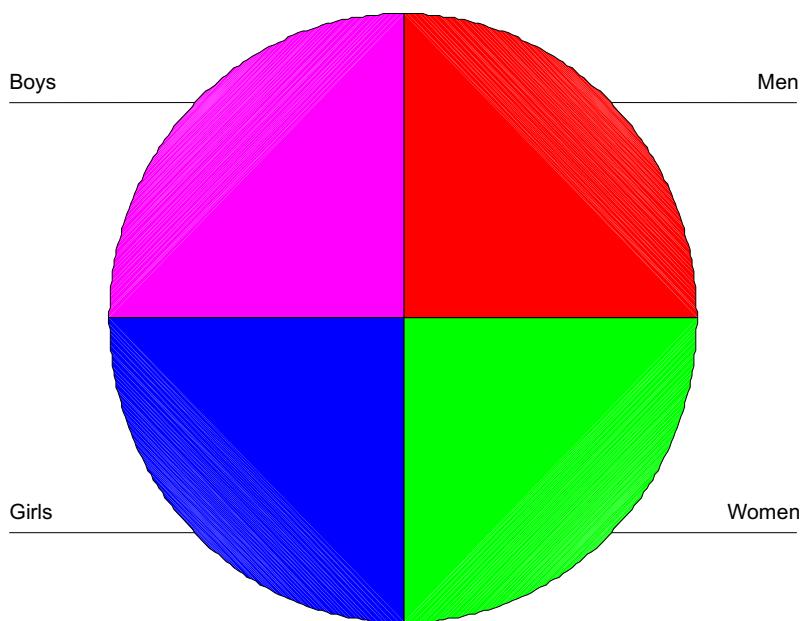
Step 1: Find the percentage.

Step 2: Find the number of degrees for each class.

Step 3: Using a protractor and compass, graph each section and write its name corresponding percentage.

Class	Frequency	Percent	Degree
Men	2500	25	90
Women	2000	20	72
Girls	4000	40	144
Boys	1500	15	54

## CLASS



### Pictogram

-In these diagram, we represent data by means of some picture symbols. We decide about a suitable picture to represent a definite number of units in which the variable is measured.

**Example:** draw a pictogram to represent the following population of a town.

Year	1989	1990	1991	1992
Population	2000	3000	5000	7000

### Bar Charts:

- A set of bars (thick lines or narrow rectangles) representing some magnitude over time space.
- They are useful for comparing aggregate over time space.
- Bars can be drawn either vertically or horizontally.
- There are different types of bar charts. The most common being :
  - Simple bar chart
  - Deviation or two way bar chart
  - Broken bar chart
  - Component or sub divided bar chart.

- Multiple bar charts.

### Simple Bar Chart

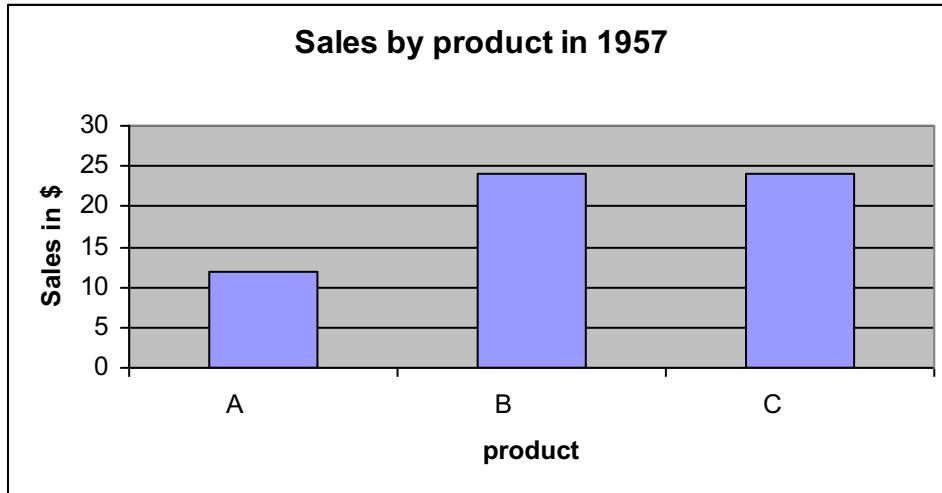
-Are used to display data on one variable.

-They are thick lines (narrow rectangles) having the same breadth. The magnitude of a quantity is represented by the height /length of the bar.

Example: The following data represent sale by product, 1957- 1959 of a given company for three products A, B, C.

Product	Sales(\$)	Sales(\$)	Sales(\$)
	In 1957	In 1958	In 1959
A	12	14	18
B	24	21	18
C	24	35	54

Solutions:



### Component Bar chart

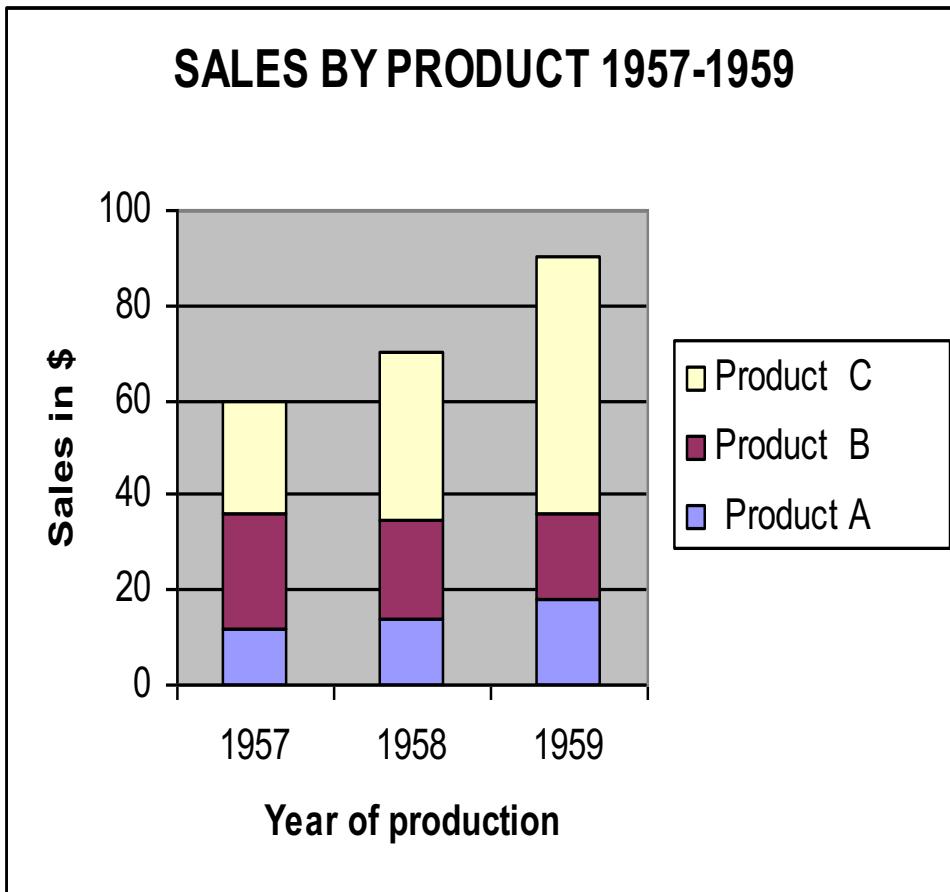
-When there is a desire to show how a total (or aggregate) is divided in to its component parts, we use component bar chart.

-The bars represent total value of a variable with each total broken in to its component parts and different colours or designs are used for identifications

Example:

Draw a component bar chart to represent the sales by product from 1957 to 1959.

Solutions:



### Multiple Bar charts

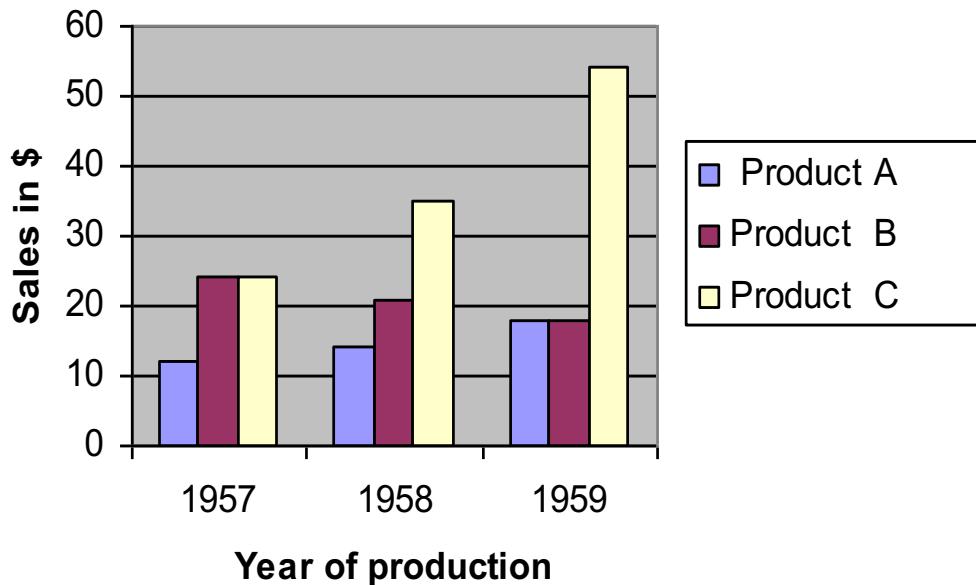
- These are used to display data on more than one variable.
- They are used for comparing different variables at the same time.

Example:

Draw a component bar chart to represent the sales by product from 1957 to 1959.

Solutions:

## Sales by product 1957-1959



### Graphical Presentation of data

- The histogram, frequency polygon and cumulative frequency graph or ogive are most commonly applied graphical representation for continuous data.

### **Procedures for constructing statistical graphs:**

- Draw and label the X and Y axes.
- Choose a suitable scale for the frequencies or cumulative frequencies and label it on the Y axes.
- Represent the class boundaries for the histogram or ogive or the mid points for the frequency polygon on the X axes.
- Plot the points.
- Draw the bars or lines to connect the points.

### Histogram

A graph which displays the data by using vertical bars of various height to represent frequencies. Class boundaries are placed along the horizontal axes. Class marks and class limits are some times used as quantity on the X axes.

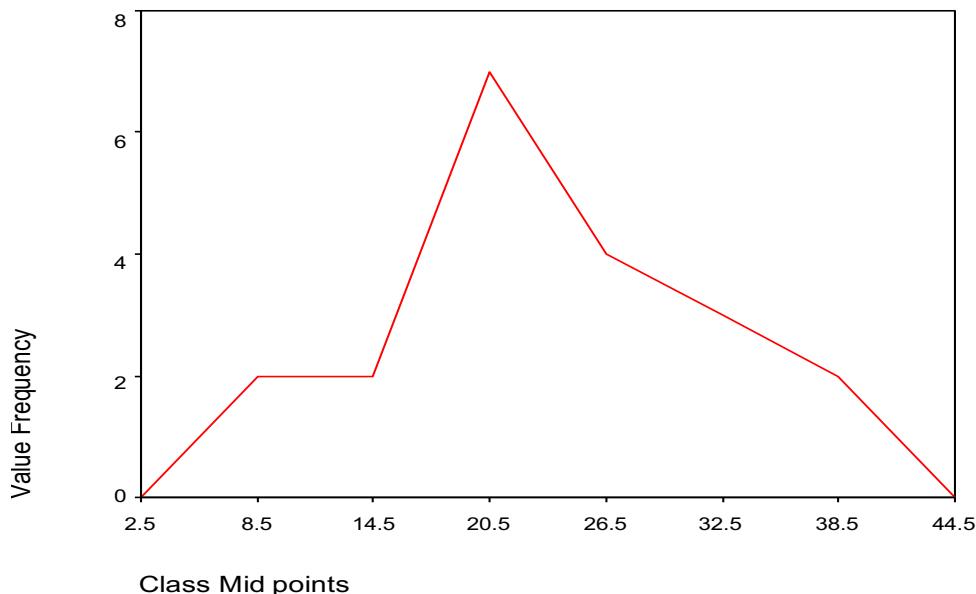
Example: Construct a histogram to represent the previous data (example \*).

### Frequency Polygon:

- A line graph. The frequency is placed along the vertical axis and classes mid points are placed along the horizontal axis. It is customer to the next higher and lower class interval with corresponding frequency of zero, this is to make it a complete polygon.

Example: Draw a frequency polygon for the above data (example \*).

Solutions:



### Ogive (cumulative frequency polygon)

- A graph showing the cumulative frequency (less than or more than type) plotted against upper or lower class boundaries respectively. That is class boundaries are plotted along the horizontal axis and the corresponding cumulative frequencies are plotted along the vertical axis. The points are joined by a free hand curve.

Example: Draw an ogive curve(less than type) for the above data.

(Example \*)

## **CHAPTER 3**

### **3. MEASURES OF CENTERAL TENDENCY**

#### **Introduction**

- When we want to make comparison between groups of numbers it is good to have a single value that is considered to be a good representative of each group. This single value is called the **average** of the group. Averages are also called measures of central tendency.
- An average which is representative is called typical average and an average which is not representative and has only a theoretical value is called a descriptive average. A typical average should posses the following:
  - It should be rigidly defined.
  - It should be based on all observation under investigation.
  - It should be as little as affected by extreme observations.
  - It should be capable of further algebraic treatment.

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- It should be as little as affected by fluctuations of sampling.
- It should be ease to calculate and simple to understand.

### Objectives:

- ☞ To comprehend the data easily.
- ☞ To facilitate comparison.
- ☞ To make further statistical analysis.

### The Summation Notation:

- Let  $X_1, X_2, X_3 \dots X_N$  be a number of measurements where  $N$  is the total number of observation and  $X_i$  is  $i^{\text{th}}$  observation.
- Very often in statistics an algebraic expression of the form  $X_1+X_2+X_3+\dots+X_N$  is used in a formula to compute a statistic. It is tedious to write an expression like this very often, so mathematicians have developed a shorthand notation to represent a sum of scores, called the summation notation.
- The symbol  $\sum_{i=1}^N X_i$  is a mathematical shorthand for  $X_1+X_2+X_3+\dots+X_N$

$$\sum_{i=1}^N X_i = X_1 + X_2 + \dots + X_N$$

The expression is read, "the sum of  $X$  sub  $i$  from  $i$  equals 1 to  $N$ ." It means "add up all the numbers."

**Example:** Suppose the following were scores made on the first homework assignment for five students in the class: 5, 7, 7, 6, and 8. In this example set of five numbers, where  $N=5$ , the summation could be written:

$$\sum_{i=1}^5 X_i = X_1 + X_2 + X_3 + X_4 + X_5 = 5 + 7 + 7 + 6 + 8 = 33$$

The " $i=1$ " in the bottom of the summation notation tells where to begin the sequence of summation. If the expression were written with " $i=3$ ", the summation would start with the third number in the set. For example:

$$\sum_{i=3}^N X_i = X_3 + X_4 + \dots + X_N$$

In the example set of numbers, this would give the following result:

$$\sum_{i=3}^5 X_i = X_3 + X_4 + X_5 = 7 + 6 + 8 = 21$$

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The "N" in the upper part of the summation notation tells where to end the sequence of summation. If there were only three scores then the summation and example would be:

$$\sum_{i=1}^3 X_i = X_1 + X_2 + X_3 = 5 + 7 + 7 = 21$$

Sometimes if the summation notation is used in an expression and the expression must be written a number of times, as in a proof, then a shorthand notation for the shorthand notation is employed. When the summation sign "" is used without additional notation, then "i=1" and "N" are assumed.

For example:

$$\sum X = \sum_{i=1}^N X_i = X_1 + X_2 + \dots + X_N$$

### PROPERTIES OF SUMMATION

1.  $\sum_{i=1}^n k = nk$  where k is any constant
2.  $\sum_{i=1}^n kX_i = k \sum_{i=1}^n X_i$  where k is any constant
3.  $\sum_{i=1}^n (a + bX_i) = na + b \sum_{i=1}^n X_i$  where a and b are any constant
4.  $\sum_{i=1}^n (X_i + Y_i) = \sum_{i=1}^n X_i + \sum_{i=1}^n Y_i$

The sum of the product of the two variables could be written:

$$\sum_{i=1}^N (X_i * Y_i) = (X_1 * Y_1) + (X_2 * Y_2) + \dots + (X_N * Y_N)$$

Example: considering the following data determine

X	Y
5	6
7	7
7	8
6	7
8	8

- a)  $\sum_{i=1}^5 X_i$
- b)  $\sum_{i=1}^5 Y_i$

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c)  $\sum_{i=1}^5 10$

d)  $\sum_{i=1}^5 (X_i + Y_i)$

e)  $\sum_{i=1}^5 (X_i - Y_i)$

f)  $\sum_{i=1}^5 X_i Y_i$

g)  $\sum_{i=1}^5 X_i^2$

h)  $(\sum_{i=1}^5 X_i)(\sum_{i=1}^5 Y_i)$

Solutions:

a)  $\sum_{i=1}^5 X_i = 5 + 7 + 7 + 6 + 8 = 33$

b)  $\sum_{i=1}^5 Y_i = 6 + 7 + 8 + 7 + 8 = 36$

c)  $\sum_{i=1}^5 10 = 5 * 10 = 50$

d)  $\sum_{i=1}^5 (X_i + Y_i) = (5 + 6) + (7 + 7) + (7 + 8) + (6 + 7) + (8 + 8) = 69 = 33 + 36$

e)  $\sum_{i=1}^5 (X_i - Y_i) = (5 - 6) + (7 - 7) + (7 - 8) + (6 - 7) + (8 - 8) = -3 = 33 - 36$

f)  $\sum_{i=1}^5 X_i Y_i = 5 * 6 + 7 * 7 + 7 * 8 + 6 * 7 + 8 * 8 = 241$

g)  $\sum_{i=1}^5 X_i^2 = 5^2 + 7^2 + 7^2 + 6^2 + 8^2 = 223$

h)  $(\sum_{i=1}^5 X_i)(\sum_{i=1}^5 Y_i) = 33 * 36 = 1188$

### Types of measures of central tendency

There are several different measures of central tendency; each has its advantage and disadvantage.

- The Mean (Arithmetic, Geometric and Harmonic)
- The Mode
- The Median
- Quantiles (Quartiles, Deciles and Percentiles)

The choice of these averages depends up on which best fit the property under discussion.

### **The Arithmetic Mean**

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- Is defined as the sum of the magnitude of the items divided by the number of items.
- The mean of  $X_1, X_2, X_3 \dots X_n$  is denoted by A.M ,m or  $\bar{X}$  and is given by:

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n}$$

$$\Rightarrow \bar{X} = \frac{\sum_{i=1}^n X_i}{n}$$

- If  $X_1$  occurs  $f_1$  times
- If  $X_2$  occurs  $f_2$  times
- .
- .
- If  $X_n$  occurs  $f_n$  times

Then the mean will be  $\bar{X} = \frac{\sum_{i=1}^k f_i X_i}{\sum_{i=1}^k f_i}$ , where k is the number of classes and

$$\sum_{i=1}^k f_i = n$$

Example: Obtain the mean of the following number

2, 7, 8, 2, 7, 3, 7

Solution:

$X_i$	$f_i$	$X_i f_i$
2	2	4
3	1	3
7	3	21
8	1	8
Total	7	36

$$\bar{X} = \frac{\sum_{i=1}^4 f_i X_i}{\sum_{i=1}^4 f_i} = \frac{36}{7} = 5.15$$

### Arithmetic Mean for Grouped Data

If data are given in the shape of a continuous frequency distribution, then the mean is obtained as follows:

$$\bar{X} = \frac{\sum_{i=1}^k f_i X_i}{\sum_{i=1}^k f_i}, \text{Where } X_i = \text{the class mark of the } i^{\text{th}} \text{ class and } f_i = \text{the frequency of the } i^{\text{th}}$$

class

Example: calculate the mean for the following age distribution.

Class	frequency
6- 10	35

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11- 15	23
16- 20	15
21- 25	12
26- 30	9
31- 35	6

Solutions:

- First find the class marks
- Find the product of frequency and class marks
- Find mean using the formula.

Class	f <sub>i</sub>	X <sub>i</sub>	X <sub>i</sub> f <sub>i</sub>
6- 10	35	8	280
11- 15	23	13	299
16- 20	15	18	270
21- 25	12	23	276
26- 30	9	28	252
31- 35	6	33	198
Total	100		1575

$$\bar{X} = \frac{\sum_{i=1}^6 f_i X_i}{\sum_{i=1}^6 f_i} = \frac{1575}{100} = 15.75$$

Exercises:

1. Marks of 75 students are summarized in the following frequency distribution:

Marks	No. of students
40-44	7
45-49	10
50-54	22
55-59	f <sub>4</sub>
60-64	f <sub>5</sub>
65-69	6
70-74	3

If 20% of the students have marks between 55 and 59

- Find the missing frequencies f<sub>4</sub> and f<sub>5</sub>.
  - Find the mean.
- If the values in a series or mid values of a class are large enough, coding of values is a good device to simplify the calculations.
  - For raw data suppose we have used the following coding system.

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$$\begin{aligned}
 d_i &= X_i - A \\
 \Rightarrow X_i &= d_i + A \\
 \bar{X} &= \frac{\sum_{i=1}^n X_i}{n} = \frac{\sum_{i=1}^n (d_i + A)}{n} \\
 &= A + \frac{\sum_{i=1}^n d_i}{n} \\
 \Rightarrow \bar{X} &= A + \bar{d}
 \end{aligned}$$

Where A is an assumed mean and  $\bar{d}$  is the mean of the coded data.

➤ If the data are expressed in terms of ungrouped frequency distribution

$$\begin{aligned}
 d_i &= X_i - A \\
 \Rightarrow X_i &= d_i + A \\
 \bar{X} &= \frac{\sum_{i=1}^k f_i X_i}{n} = \frac{\sum_{i=1}^k f_i (d_i + A)}{n} \\
 &= A + \frac{\sum_{i=1}^k f_i d_i}{n} \\
 \Rightarrow \bar{X} &= A + \bar{d}
 \end{aligned}$$

- In both cases the true mean is the assumed mean plus the average of the deviations from the assumed mean.
- Suppose the data is given in the shape of continuous frequency distribution with a constant class size of w then the following coding is appropriate.

$$\begin{aligned}
 d_i &= \frac{X_i - A}{w} \\
 \Rightarrow X_i &= wd_i + A \\
 \bar{X} &= \frac{\sum_{i=1}^k f_i X_i}{n} = \frac{\sum_{i=1}^k f_i (wd_i + A)}{n} \\
 &= A + \frac{\sum_{i=1}^k f_i wd_i}{n} \\
 \Rightarrow \bar{X} &= A + w\bar{d}
 \end{aligned}$$

Where:  $X_i$  is the original class mark for the  $i^{\text{th}}$  class.

$d_i$  is the transformed class mark for the  $i^{\text{th}}$  class.

A is an assumed mean usually the mean of the class marks.

$(i = 1, 2, \dots, k)$

Example:

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- 
1. Suppose the deviations of the observations from an assumed mean of 7 are: 1, -1, -2, -2, 0, -3, -2, 2, 0, -3.
    - a) Find the true mean
    - b) Find the original observation.

Solutions:

$$A = 7, \sum_{i=1}^{10} d_i = -10$$

$$\text{a)} \Rightarrow \bar{d} = \frac{-10}{10} = -1 \\ \Rightarrow \bar{X} = A + \bar{d} = 7 - 1 = 6$$

The true mean is 6.

- b) Using  $X_i = A + d_i$  we obtain the following original observations:  
8, 6, 5, 5, 7, 4, 5, 9, 7, 4.

### Special properties of Arithmetic mean

1. The sum of the deviations of a set of items from their mean is always zero.

$$\text{i.e. } \sum_{i=1}^n (X_i - \bar{X}) = 0.$$

2. The sum of the squared deviations of a set of items from their mean is the minimum. i.e.  $\sum_{i=1}^n (X_i - \bar{X})^2 < \sum_{i=1}^n (X_i - A)^2, A \neq \bar{X}$

3. If  $\bar{X}_1$  is the mean of  $n_1$  observations

If  $\bar{X}_2$  is the mean of  $n_2$  observations

.

.

If  $\bar{X}_k$  is the mean of  $n_k$  observations

Then the mean of all the observation in all groups often called the combined mean is given by:

$$\bar{X}_c = \frac{\bar{X}_1 n_1 + \bar{X}_2 n_2 + \dots + \bar{X}_k n_k}{n_1 + n_2 + \dots + n_k}$$

Example: In a class there are 30 females and 70 males. If females averaged 60 in an examination and boys averaged 72, find the mean for the entire class.

Solutions:

*Females*

$$\bar{X}_1 = 60$$

$$n_1 = 30$$

*Males*

$$\bar{X}_2 = 72$$

$$n_2 = 70$$

$$\bar{X}_c = \frac{\bar{X}_1 n_1 + \bar{X}_2 n_2}{n_1 + n_2} = \frac{\sum_{i=1}^2 \bar{X}_i n_i}{\sum_{i=1}^2 n_i}$$

$$\Rightarrow \bar{X}_c = \frac{30(60) + 70(72)}{30 + 70} = \frac{6840}{100} = 68.40$$

4. If a wrong figure has been used when calculating the mean the correct mean can be obtained without repeating the whole process using:

$$\text{CorrectMean} - \text{WrongMean} \cdot \frac{(\text{CorrectValue} - \text{WrongValue})}{n}$$

Where n is total number of observations.

Example: An average weight of 10 students was calculated to be 65. Later it was discovered that one weight was misread as 40 instead of 80 k.g. Calculate the correct average weight.

Solutions:

$$\text{CorrectMean} - \text{WrongMean} \cdot \frac{(\text{CorrectValue} - \text{WrongValue})}{n}$$

$$\text{CorrectMean} - 65 \cdot \frac{(80 - 40)}{10} = 65 - 4 = 69 \text{ k.g.}$$

5. The effect of transforming original series on the mean.

- a) If a constant  $k$  is added/ subtracted to/from every observation then the new mean will be  $\text{the old mean} \pm k$  respectively.
- b) If every observations are multiplied by a constant  $k$  then the new mean will be  $k * \text{old mean}$

Example:

1. The mean of  $n$  Tetracycline Capsules  $X_1, X_2, \dots, X_n$  are known to be 12 gm. New set of capsules of another drug are obtained by the linear transformation  $Y_i = 2X_i - 0.5$  ( $i = 1, 2, \dots, n$ ) then what will be the mean of the new set of capsules

Solutions:

$$\text{NewMean} = 2 * \text{OldMean} - 0.5 = 2 * 12 - 0.5 = 23.5$$

2. The mean of a set of numbers is 500.

- a) If 10 is added to each of the numbers in the set, then what will be the mean of the new set?
- b) If each of the numbers in the set are multiplied by -5, then what will be the mean of the new set?

Solutions:

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a). NewMean = OldMean +  $\frac{10 - 50}{10} \times 510$

b). NewMean =  $\frac{1}{2} * \text{OldMean} + \frac{1}{2} * 500 = 500$

### **Weighted Mean**

- ☞ When a proper importance is desired to be given to different data a weighted mean is appropriate.
- ☞ Weights are assigned to each item in proportion to its relative importance.
- ☞ Let  $X_1, X_2, \dots, X_n$  be the value of items of a series and  $W_1, W_2, \dots, W_n$  their corresponding weights , then the weighted mean denoted  $\bar{X}_w$  is defined as:

$$\bar{X}_w = \frac{\sum X_i W_i}{\sum W_i}$$

Example:

A student obtained the following percentage in an examination:

English 60, Biology 75, Mathematics 63, Physics 59, and chemistry 55.Find the students weighted arithmetic mean if weights 1, 2, 1, 3, 3 respectively are allotted to the subjects.

Solutions:

$$\bar{X}_w = \frac{\sum X_i W_i}{\sum W_i} = \frac{60 * 1 + 75 * 2 + 63 * 1 + 59 * 3 + 55 * 3}{1+2+1+3+3} = \frac{615}{10} = 61.5$$

### **Merits and Demerits of Arithmetic Mean**

#### **Merits:**

- It is rigidly defined.
- It is based on all observation.
- It is suitable for further mathematical treatment.
- It is stable average, i.e. it is not affected by fluctuations of sampling to some extent.
- It is easy to calculate and simple to understand.

#### **Demerits:**

- It is affected by extreme observations.
- It can not be used in the case of open end classes.
- It can not be determined by the method of inspection.
- It can not be used when dealing with qualitative characteristics, such as intelligence, honesty, beauty.
- It can be a number which does not exist in a series.
- Some times it leads to wrong conclusion if the details of the data from which it is obtained are not available.
- It gives high weight to high extreme values and less weight to low extreme values.

### **The Geometric Mean**

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- ⇒ The geometric mean of a set of n observation is the  $n^{\text{th}}$  root of their product.
- ⇒ The geometric mean of  $X_1, X_2, X_3 \dots X_n$  is denoted by G.M and given by:

$$\text{G.M} = \sqrt[n]{X_1 * X_2 * \dots * X_n}$$

- ⇒ Taking the logarithms of both sides

$$\begin{aligned}\log(\text{G.M}) &= \log(\sqrt[n]{X_1 * X_2 * \dots * X_n}) = \log(X_1 * X_2 * \dots * X_n)^{\frac{1}{n}} \\ &= \log(\text{G.M}) = \frac{1}{n} \log(X_1 * X_2 * \dots * X_n) = \frac{1}{n} (\log X_1 + \log X_2 + \dots + \log X_n) \\ &\Rightarrow \log(\text{G.M}) = \frac{1}{n} \sum \log X_i\end{aligned}$$

⇒ The logarithm of the G.M of a set of observation is the arithmetic mean of their logarithm.

$$\text{G.M} = \text{Anti log} \left( \frac{1}{n} \sum \log X_i \right)$$

### Example:

Find the G.M of the numbers 2, 4, 8.

Solutions:

$$\text{G.M} = \sqrt[3]{X_1 * X_2 * \dots * X_n} = \sqrt[3]{2 * 4 * 8} = \sqrt[3]{64} = 4$$

Remark: The Geometric Mean is useful and appropriate for finding averages of ratios.

## The Harmonic Mean

The harmonic mean of  $X_1, X_2, X_3 \dots X_n$  is denoted by H.M and given by:

$$\boxed{\text{H.M} = \frac{n}{\frac{1}{X_1} + \frac{1}{X_2} + \dots + \frac{1}{X_n}}}, \text{ This is called simple harmonic mean.}$$

In a case of frequency distribution:

$$\boxed{\text{H.M} = \frac{n}{\frac{f_1}{X_1} + \frac{f_2}{X_2} + \dots + \frac{f_n}{X_n}}}, \text{ n} = f_1 + f_2 + \dots + f_n$$

If observations  $X_1, X_2, \dots, X_n$  have weights  $W_1, W_2, \dots, W_n$  respectively, then their harmonic mean is given by

$$\boxed{\text{H.M} = \frac{W_1 + W_2 + \dots + W_n}{\frac{W_1}{X_1} + \frac{W_2}{X_2} + \dots + \frac{W_n}{X_n}}}, \text{ This is called Weighted Harmonic Mean.}$$

Remark: The Harmonic Mean is useful and appropriate in finding average speeds and average rates.

Example: A cyclist pedals from his house to his college at speed of 10 km/hr and back from the college to his house at 15 km/hr. Find the average speed.

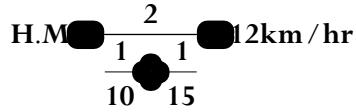
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Solution: Here the distance is constant

→ The simple H.M is appropriate for this problem.

$$X_1 = 10 \text{ km/hr}$$

$$X_2 = 15 \text{ km/hr}$$



### The Mode

- Mode is a value which occurs most frequently in a set of values
- The mode may not exist and even if it does exist, it may not be unique.
- In case of discrete distribution the value having the maximum frequency is the model value.

Examples:

1. Find the mode of 5, 3, 5, 8, 9

$$\text{Mode} = 5$$

2. Find the mode of 8, 9, 9, 7, 8, 2, and 5.

It is a bimodal Data: 8 and 9

3. Find the mode of 4, 12, 3, 6, and 7.

No mode for this data.

- The mode of a set of numbers  $X_1, X_2, \dots, X_n$  is usually denoted by  $\hat{X}$ .

### Mode for Grouped data

If data are given in the shape of continuous frequency distribution, the mode is defined as:



Where:

$\hat{x}$  the mode of the distribution

w the size of the modal class



$f_{mo}$  frequency of the modal class

$f_1$  frequency of the class preceding the modal class

$f_2$  frequency of the class following the modal class

**Note:** The modal class is a class with the highest frequency.

Example: Following is the distribution of the size of certain farms selected at random from a district. Calculate the mode of the distribution.

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Size of farms	No. of farms
5-15	8
15-25	12
25-35	17
35-45	29
45-55	31
55-65	5
65-75	3

Solutions:

45-55 is the modal class, since it is a class with the highest frequency.

$$L_{mo} = 45$$

$$w = 10$$

$$f_{mo} = f_1 = 2$$

$$f_{mo} = f_2 = 26$$

$$f_{mo} = 31$$

$$f_1 = 29$$

$$f_2 = 5$$

$$\hat{X} = L_{mo} + \left( \frac{f_{mo} - f_1}{2f_1} \right) w$$
$$\hat{X} = 45 + \left( \frac{2}{2 \times 26} \right) 10$$
$$\hat{X} = 45.71$$

### Merits and Demerits of Mode

#### Merits:

- It is not affected by extreme observations.
- Easy to calculate and simple to understand.
- It can be calculated for distribution with open end class

#### Demerits:

- It is not rigidly defined.
- It is not based on all observations
- It is not suitable for further mathematical treatment.
- It is not stable average, i.e. it is affected by fluctuations of sampling to some extent.
- Often its value is not unique.

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**Note:** being the point of maximum density, mode is especially useful in finding the most popular size in studies relating to marketing, trade, business, and industry. It is the appropriate average to be used to find the ideal size.

### **The Median**

- In a distribution, median is the value of the variable which divides it in to two equal halves.
- In an ordered series of data median is an observation lying exactly in the middle of the series. It is the middle most value in the sense that the number of values less than the median is equal to the number of values greater than it.
- If  $X_1, X_2, \dots, X_n$  be the observations, then the numbers arranged in ascending order will be  $X_{[1]}, X_{[2]}, \dots, X_{[n]}$ , where  $X_{[i]}$  is  $i^{\text{th}}$  smallest value.  
 $\Rightarrow X_{[1]} < X_{[2]} < \dots < X_{[n]}$
- Median is denoted by  $\hat{X}$ .

### **Median for ungrouped data**

$$\tilde{X} = \begin{cases} X_{[(n+1)/2]} & \text{If } n \text{ is odd.} \\ \frac{1}{2}(X_{[(n/2)]} + X_{[(n/2)+1]}) & \text{If } n \text{ is even} \end{cases}$$

Example: Find the median of the following numbers.

- 6, 5, 2, 8, 9, 4.
- 2, 1, 8, 3, 5, 8.

Solutions:

- First order the data: 2, 4, 5, 6, 8, 9

Here  $n=6$

$$\begin{aligned} \tilde{X} &= \frac{1}{2}(X_{[\frac{n}{2}]} + X_{[\frac{n}{2}+1]}) \\ &= \frac{1}{2}(X_{[3]} + X_{[4]}) \\ &= \frac{1}{2}(5 + 6) = 5.5 \end{aligned}$$

- Order the data :1, 2, 3, 5, 8

Here  $n=5$

$$\begin{aligned} \tilde{X} &= X_{[\frac{n+1}{2}]} \\ &= X_{[3]} \\ &= 3 \end{aligned}$$

### **Median for grouped data**

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If data are given in the shape of continuous frequency distribution, the median is defined

$$\tilde{x} = L_{\text{med}} + \frac{w}{f_{\text{med}}} \left( \frac{n}{2} - c \right)$$

Where :

$L_{\text{med}}$  lower class boundary of the median class.

as:  $w$  the size of the median class

$n$  total number of observations.

$c$  the cumulative frequency (less than type) preceding the median class.

$f_{\text{med}}$  the frequency of the median class.

Remark:

The median class is the class with the smallest cumulative frequency (less than type) greater

than or equal to  $\frac{n}{2}$ .

**Example:** Find the median of the following distribution.

Class	Frequency
40-44	7
45-49	10
50-54	22
55-59	15
60-64	12
65-69	6
70-74	3

Solutions:

- First find the less than cumulative frequency.
- Identify the median class.
- Find median using formula.

Class	Frequency	Cumu.Freq(less than type)
40-44	7	7
45-49	10	17
50-54	22	39
55-59	15	54
60-64	12	66
65-69	6	72
70-74	3	75

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$$\frac{n}{2} = 37.5$$

39 is the first cumulative frequency to be greater than or equal to 37.5

50-54 is the median class.

$$L_{med} = 49.5, w = 5$$

$$n = 75, c = 7, f_{med} = 2$$

$$\text{Median} = L_{med} + \frac{w}{2} \left( \frac{n}{2} - c \right)$$
$$= 49.5 + \frac{5}{2} (37.5 - 7)$$
$$= 54.16$$

### Merits and Demerits of Median

Merits:

- Median is a positional average and hence not influenced by extreme observations.
- Can be calculated in the case of open end intervals.
- Median can be located even if the data are incomplete.

Demerits:

- It is not a good representative of data if the number of items is small.
- It is not amenable to further algebraic treatment.
- It is susceptible to sampling fluctuations.

### Quantiles

When a distribution is arranged in order of magnitude of items, the median is the value of the middle term. Their measures that depend up on their positions in distribution quartiles, deciles, and percentiles are collectively called quantiles.

#### Quartiles:

- Quartiles are measures that divide the frequency distribution in to four equal parts.
- The value of the variables corresponding to these divisions are denoted  $Q_1$ ,  $Q_2$ , and  $Q_3$  often called the first, the second and the third quartile respectively.
- $Q_1$  is a value which has 25% items which are less than or equal to it. Similarly  $Q_2$  has 50% items with value less than or equal to it and  $Q_3$  has 75% items whose values are less than or equal to it.
- To find  $Q_i$  ( $i=1, 2, 3$ ) we count  $\frac{iN}{4}$  of the classes beginning from the lowest class.

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- For grouped data: we have the following formula

$$Q_i = L_{Q_i} + \frac{w}{4} \left( \frac{iN}{4} - f_{Q_i} \right), i=1,2,3$$

Where :

$L_{Q_i}$  lower class boundary of the quartile class.

$w$  the size of the quartile class

$N$  total number of observations.

$c$  the cumulative frequency (less than type) preceding the quartile class.

$f_{Q_i}$  the frequency of the quartile class.

### Remark:

The quartile class (class containing  $Q_i$ ) is the class with the smallest cumulative frequency (less than type) greater than or equal to  $\frac{iN}{4}$ .

### Deciles:

- Deciles are measures that divide the frequency distribution into ten equal parts.
- The values of the variables corresponding to these divisions are denoted  $D_1, D_2, \dots, D_9$  often called the first, the second,..., the ninth decile respectively.
- To find  $D_i$  ( $i=1, 2, \dots, 9$ ) we count  $\frac{iN}{10}$  of the classes beginning from the lowest class.
- For grouped data: we have the following formula

$$D_i = L_{D_i} + \frac{w}{10} \left( \frac{iN}{10} - c \right), i=1,2,\dots,9$$

Where :

$L_{D_i}$  lower class boundary of the decile class.

$w$  the size of the decile class

$N$  total number of observations.

$c$  the cumulative frequency (less than type) preceding the decile class.

$f_{D_i}$  the frequency of the decile class.

### Remark:

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The decile class (class containing  $D_i$ ) is the class with the smallest cumulative frequency (less than type) greater than or equal to  $\frac{iN}{10}$ .

### Percentiles:

- Percentiles are measures that divide the frequency distribution into hundred equal parts.
  - The values of the variables corresponding to these divisions are denoted  $P_1, P_2, \dots, P_{99}$  often called the first, the second, ..., the ninety-ninth percentile respectively.
  - To find  $P_i$  ( $i=1, 2, \dots, 99$ ) we count  $\frac{iN}{100}$  of the classes beginning from the lowest class.
  - For grouped data: we have the following formula  
$$P_i = L_{P_i} + \frac{w}{f_{P_i}} \left( \frac{iN}{100} - c \right), i=1, 2, \dots, 99$$
- Where :
- $L_{P_i}$  lower class boundary of the percentile class.
  - $w$  the size of the percentile class
  - $N$  total number of observations.
  - $c$  the cumulative frequency (less than type) preceding the percentile class.
  - $f_{P_i}$  the frequency of the percentile class.

### Remark:

The percentile class (class containing  $P_i$ ) is the class with the smallest cumulative frequency (less than type) greater than or equal to  $\frac{iN}{100}$ .

**Example:** Considering the following distribution

Calculate:

- All quartiles.
- The 7<sup>th</sup> decile.
- The 90<sup>th</sup> percentile.

Values	Frequency
140- 150	17
150- 160	29
160- 170	42
170- 180	72
180- 190	84
190- 200	107
200- 210	49
210- 220	34
220- 230	31
230- 240	16

Solutions:

- First find the less than cumulative frequency.
- Use the formula to calculate the required quantile.

Values	Frequency	Cum.Freq(less than type)
140- 150	17	17
150- 160	29	46
160- 170	42	88
170- 180	72	160
180- 190	84	244
190- 200	107	351
200- 210	49	400
210- 220	34	434
220- 230	31	465
230- 240	16	481
240- 250	12	493

a) Quartiles:

$$\text{i. } Q_1$$

- determine the class containing the first quartile.

$$\frac{N}{4} = 123.25$$

170 - 180 is the class containing the first quartile.

$$L_{Q_1} = 170, \quad w = 10$$

$$N = 493, \quad c = 88, \quad f_{Q_1} = 72$$

$$\begin{aligned} \Rightarrow Q_1 &= L_{Q_1} + \frac{w}{f_{Q_1}} \left( \frac{N}{4} - c \right) \\ &= 170 + \frac{10}{72} (123.25 - 88) \\ &= \underline{\underline{174.90}} \end{aligned}$$

$$\text{ii. } Q_2$$

- determine the class containing the second quartile.

$$\frac{2 * N}{4} = 246.5$$

$\Rightarrow 190 - 200$  is the class containing the second quartile.

$$L_{Q_2} = 190, \quad w = 10$$

$$N = 493, \quad c = 244, \quad f_{Q_2} = 107$$

$$\begin{aligned}\Rightarrow Q_2 &= L_{Q_2} + \frac{w}{f_{Q_2}} \left( \frac{2 * N}{4} - c \right) \\ &= 170 + \frac{10}{72} (246.5 - 244) \\ &= \underline{\underline{190.23}}\end{aligned}$$

iii.  $Q_3$

- determine the class containing the third quartile.

$$\frac{3 * N}{4} = 369.75$$

$\Rightarrow 200 - 210$  is the class containing the third quartile.

$$L_{Q_3} = 200, \quad w = 10$$

$$N = 493, \quad c = 351, \quad f_{Q_3} = 49$$

$$\begin{aligned}\Rightarrow Q_3 &= L_{Q_3} + \frac{w}{f_{Q_3}} \left( \frac{3 * N}{4} - c \right) \\ &= 200 + \frac{10}{49} (369.75 - 351) \\ &= \underline{\underline{203.83}}\end{aligned}$$

b)  $D_7$

- determine the class containing the 7<sup>th</sup> decile.

$$\frac{7 * N}{10} = 345.1$$

$\Rightarrow 190 - 200$  is the class containing the seventh decile.

$$L_{D_7} = 190, \quad w = 10$$

$$N = 493, \quad c = 244, \quad f_{D_7} = 107$$

$$\begin{aligned}\Rightarrow D_7 &= L_{D_7} + \frac{w}{f_{D_7}} \left( \frac{7 * N}{10} - c \right) \\ &= 190 + \frac{10}{107} (345.1 - 244) \\ &= \underline{\underline{199.45}}\end{aligned}$$

c)  $P_{90}$

- determine the class containing the 90<sup>th</sup> percentile.

$$\frac{90 * N}{100} = 443.7$$

$\Rightarrow 220 - 230$  is the class containing the 90<sup>th</sup> percentile.

$$L_{P_{90}} = 220, \quad w = 10$$

$$N = 493, \quad c = 434, \quad f_{P_{90}} = 3107$$

$$\begin{aligned}\Rightarrow P_{90} &= L_{P_{90}} + \frac{w}{f_{P_{90}}} \left( \frac{90 * N}{100} - c \right) \\ &= 220 + \frac{10}{31} (443.7 - 434) \\ &= \underline{\underline{223.13}}\end{aligned}$$

## CHAPTER 4

### 4. Measures of Dispersion (Variation)

#### Introduction and objectives of measuring Variation

-The scatter or spread of items of a distribution is known as dispersion or variation. In other words the degree to which numerical data tend to spread about an average value is called dispersion or variation of the data.

-Measures of dispersions are statistical measures which provide ways of measuring the extent in which data are dispersed or spread out.

### Objectives of measuring Variation:

- To judge the reliability of measures of central tendency
- To control variability itself.
- To compare two or more groups of numbers in terms of their variability.
- To make further statistical analysis.

### Absolute and Relative Measures of Dispersion

The measures of dispersion which are expressed in terms of the original unit of a series are termed as absolute measures. Such measures are not suitable for comparing the variability of two distributions which are expressed in different units of measurement and different average size. Relative measures of dispersions are a ratio or percentage of a measure of absolute dispersion to an appropriate measure of central tendency and are thus pure numbers independent of the units of measurement. For comparing the variability of two distributions (even if they are measured in the same unit), we compute the relative measure of dispersion instead of absolute measures of dispersion.

### Types of Measures of Dispersion

Various measures of dispersions are in use. The most commonly used measures of dispersions are:

- 1) Range and relative range
- 2) Quartile deviation and coefficient of Quartile deviation
- 3) Mean deviation and coefficient of Mean deviation
- 4) Standard deviation and coefficient of variation.

### **The Range (R)**

The range is the largest score minus the smallest score. It is a quick and dirty measure of variability, although when a test is given back to students they very often wish to know the range of scores. Because the range is greatly affected by extreme scores, it may give a distorted picture of the scores. The following two distributions have the same range, 13, yet appear to differ greatly in the amount of variability.

Distribution 1: 32 35 36 36 37 38 40 42 42 43 43 45

Distribution 2: 32 32 33 33 33 34 34 34 34 35 35 45

For this reason, among others, the range is not the most important measure of variability.

$$R = L - S \quad , L = \text{largest observation}$$
$$S = \text{smallest observation}$$

### Range for grouped data:

If data are given in the shape of continuous frequency distribution, the range is computed as:

$$R = UCL_k - LCL_1, \quad UCL_k \text{ is upperclass limit of the last class.}$$
$$UCL_1 \text{ is lower class limit of the first class.}$$

This is sometimes expressed as:

$$R = X_k - X_1, \quad X_k \text{ is class mark of the last class.}$$
$$X_1 \text{ is classmark of the first class.}$$

## Merits and Demerits of range

### Merits:

- It is rigidly defined.
- It is easy to calculate and simple to understand.

### Demerits:

- It is not based on all observations.
- It is highly affected by extreme observations.
- It is affected by fluctuation in sampling.
- It is not liable to further algebraic treatment.
- It can not be computed in the case of open end distribution.
- It is very sensitive to the size of the sample.

### **Relative Range (RR)**

-it is also sometimes called coefficient of range and given by:

$$RR = \frac{L - S}{L + S} = \frac{R}{L + S}$$

Example:

1. Find the relative range of the above two distributions. (exercise!)
2. If the range and relative range of a series are 4 and 0.25 respectively. Then what is the value of:
  - a) Smallest observation
  - b) Largest observation

Solutions : (2)

$$R = 4 \Rightarrow L - S = 4 \quad (1)$$

$$RR = 0.25 \Rightarrow L + S = 16 \quad (2)$$

*Solving (1) and (2) at the same time, one can obtain the following values*

$$L = 10 \text{ and } S = 6$$

### **The Quartile Deviation (Semi-inter quartile range), Q.D**

The inter quartile range is the difference between the third and the first quartiles of a set of items and semi-inter quartile range is half of the inter quartile range.

$$Q.D = \frac{Q_3 - Q_1}{2}$$

### **Coefficient of Quartile Deviation (C.Q.D)**

$$C.Q.D = \frac{(Q_3 - Q_1)/2}{(Q_3 + Q_1)/2} = \frac{2 * Q.D}{Q_3 + Q_1} = \frac{Q_3 - Q_1}{Q_3 + Q_1}$$

- It gives the average amount by which the two quartiles differ from the median.

**Example:** Compute Q.D and its coefficient for the following distribution.

Values	Frequency
140- 150	17
150- 160	29
160- 170	42

170- 180	72
180- 190	84
190- 200	107
200- 210	49
210- 220	34
220- 230	31
230- 240	16
240- 250	12

**Solutions:**

In the previous chapter we have obtained the values of all quartiles as:

$$Q_1 = 174.90, \quad Q_2 = 190.23, \quad Q_3 = 203.83$$

$$\Rightarrow Q.D = \frac{Q_3 - Q_1}{2} = \frac{203.83 - 174.90}{2} = 14.47$$

$$C.Q.D = \frac{2 * Q.D}{Q_3 + Q_1} = \frac{2 * 14.47}{203.83 + 174.90} = 0.076$$

**Remark:** Q.D or C.Q.D includes only the middle 50% of the observation.

**The Mean Deviation (M.D):**

The mean deviation of a set of items is defined as the arithmetic mean of the values of the absolute deviations from a given average. Depending up on the type of averages used we have different mean deviations.

## a) Mean Deviation about the mean

- Denoted by M.D( $\bar{X}$ ) and given by

$$M.D(\bar{X}) = \frac{\sum_{i=1}^n |X_i - \bar{X}|}{n}$$

- For the case of frequency distribution it is given as:

$$M.D(\bar{X}) = \frac{\sum_{i=1}^k f_i |X_i - \bar{X}|}{n}$$

**Steps to calculate M.D ( $\bar{X}$ ):**

1. Find the arithmetic mean,  $\bar{X}$
2. Find the deviations of each reading from  $\bar{X}$ .
3. Find the arithmetic mean of the deviations, ignoring sign.

b) Mean Deviation about the median.

- Denoted by  $M.D(\tilde{X})$  and given by

$$M.D(\tilde{X}) = \frac{\sum_{i=1}^n |X_i - \tilde{X}|}{n}$$

- For the case of frequency distribution it is given as:

$$M.D(\tilde{X}) = \frac{\sum_{i=1}^k f_i |X_i - \tilde{X}|}{n}$$

**Steps to calculate M.D ( $\tilde{X}$ ):**

1. Find the median,  $\tilde{X}$
2. Find the deviations of each reading from  $\tilde{X}$ .
3. Find the arithmetic mean of the deviations, ignoring sign.

c) Mean Deviation about the mode.

- Denoted by  $M.D(\hat{X})$  and given by

$$M.D(\hat{X}) = \frac{\sum |x_i - \hat{X}|}{n}$$

- For the case of frequency distribution it is given as:

$$M.D(\hat{X}) = \frac{\sum_{i=1}^k f_i |X_i - \hat{X}|}{n}$$

**Steps to calculate M.D ( $\hat{X}$ ):**

1. Find the mode,  $\hat{X}$
2. Find the deviations of each reading from  $\hat{X}$ .
3. Find the arithmetic mean of the deviations, ignoring sign.

**Examples:**

1. The following are the number of visit made by ten mothers to the local doctor's surgery. 8, 6, 5, 5, 7, 4, 5, 9, 7, 4

Find mean deviation about mean, median and mode.

**Solutions:**

First calculate the three averages

$$\bar{X} = 6, \tilde{X} = 5.5, \hat{X} = 5$$

Then take the deviations of each observation from these averages.

$X_i$	4	4	5	5	5	6	7	7	8	9	total
$ X_i - 6 $	2	2	1	1	1	0	1	1	2	3	14
$ X_i - 5.5 $	1.5	1.5	0.5	0.5	0.5	0.5	1.5	1.5	2.5	3.5	14
$ X_i - 5 $	1	1	0	0	0	1	2	2	3	4	14

$$\Rightarrow M.D(\bar{X}) = \frac{\sum_{i=1}^{10} |X_i - 6|}{10} = \frac{14}{10} = 1.4$$

$$M.D(\tilde{X}) = \frac{\sum_{i=1}^{10} |X_i - 5.5|}{10} = \frac{14}{10} = 1.4$$

$$M.D(\hat{X}) = \frac{\sum_{i=1}^{10} |X_i - 5|}{10} = \frac{14}{10} = 1.4$$

2. Find mean deviation about mean, median and mode for the following distributions.(exercise)

Class	Frequency
40-44	7
45-49	10
50-54	22
55-59	15
60-64	12
65-69	6
70-74	3

**Remark:** Mean deviation is always minimum about the median.

### Coefficient of Mean Deviation (C.M.D)

$$C.M.D = \frac{M.D}{\text{Average about which deviations are taken}}$$

$$\Rightarrow C.M.D(\bar{X}) = \frac{M.D(\bar{X})}{\bar{X}}$$

$$C.M.D(\tilde{X}) = \frac{M.D(\tilde{X})}{\tilde{X}}$$

$$C.M.D(\hat{X}) = \frac{M.D(\hat{X})}{\hat{X}}$$

Example: calculate the C.M.D about the mean, median and mode for the data in example 1 above.

### Solutions:

$$C.M.D = \frac{M.D}{\text{Average about which deviations are taken}}$$

$$\Rightarrow C.M.D(\bar{X}) = \frac{M.D(\bar{X})}{\bar{X}} = \frac{1.4}{6} = 0.233$$

$$C.M.D(\tilde{X}) = \frac{M.D(\tilde{X})}{\tilde{X}} = \frac{1.4}{5.5} = 0.255$$

$$C.M.D(\hat{X}) = \frac{M.D(\hat{X})}{\hat{X}} = \frac{1.4}{5} = 0.28$$

Exercise: Identify the merits and demerits of Mean Deviation

## The Variance

### Population Variance

If we divide the variation by the number of values in the population, we get something called the population variance. This variance is the "average squared deviation from the mean".

$$\text{Population Varince} = \sigma^2 = \frac{1}{N} \sum (X_i - \mu)^2, \quad i = 1, 2, \dots, N$$

For the case of frequency distribution it is expressed as:

$$\text{Population Varince} = \sigma^2 = \frac{1}{N} \sum f_i (X_i - \mu)^2, \quad i = 1, 2, \dots, k$$

### Sample Variance

One would expect the sample variance to simply be the population variance with the population mean replaced by the sample mean. However, one of the major uses of statistics is to estimate the corresponding parameter. This formula has the problem that the estimated value isn't the same as the parameter. To counteract this, the sum of the squares of the deviations is divided by one less than the sample size.

$$\text{Sample Varince} = S^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2, \quad i = 1, 2, \dots, n$$

For the case of frequency distribution it is expressed as:

$$\text{Sample Variance } S^2 = \frac{1}{n-1} \sum f_i (X_i - \bar{X})^2, \quad i=1,2,\dots,k$$

We usually use the following short cut formula.

$$S^2 = \frac{\sum_{i=1}^n X_i^2 - n\bar{X}^2}{n-1}, \text{ for raw data.}$$

$$\text{Population standard deviation } \sigma = \sqrt{\sigma^2} \quad S^2 = \frac{\sum_{i=1}^k f_i X_i^2 - n\bar{X}^2}{n-1}, \text{ for frequency distribution.}$$

$$\text{Sample standard deviation } s = \sqrt{S^2}$$

### Standard Deviation

There is a problem with variances. Recall that the deviations were squared. That means that the units were also squared. To get the units back the same as the original data values, the square root must be taken.

The following steps are used to calculate the sample variance:

1. Find the arithmetic mean.
2. Find the difference between each observation and the mean.
3. Square these differences.
4. Sum the squared differences.
5. Since the data is a sample, divide the number (from step 4 above) by the number of observations minus one, i.e.,  $n-1$  (where  $n$  is equal to the number of observations in the data set).

**Examples:** Find the variance and standard deviation of the following sample data

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1. 5, 17, 12, 10.
2. The data is given in the form of frequency distribution.

Class	Frequency
40-44	7
45-49	10
50-54	22
55-59	15
60-64	12
65-69	6
70-74	3

**Solutions:**

$$1. \quad \bar{X} = 11$$

X <sub>i</sub>	5	10	12	17	Total
(X <sub>i</sub> - $\bar{X}$ ) <sup>2</sup>	36	1	1	36	74

$$\Rightarrow S^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1} = \frac{74}{3} = 24.67.$$

$$\Rightarrow S = \sqrt{S^2} = \sqrt{24.67} = 4.97.$$

2.  $\bar{X} = 55$

X <sub>i</sub> (C.M)	42	47	52	57	62	67	72	Total
f <sub>i</sub> (X <sub>i</sub> - $\bar{X}$ ) <sup>2</sup>	1183	640	198	60	588	864	867	4400

$$\Rightarrow S^2 = \frac{\sum_{i=1}^n f_i (X_i - \bar{X})^2}{n-1} = \frac{4400}{74} = 59.46.$$

$$\Rightarrow S = \sqrt{S^2} = \sqrt{59.46} = 7.71.$$

### **Special properties of Standard deviations**

1.  $\sqrt{\frac{\sum(X_i - \bar{X})^2}{n-1}} < \sqrt{\frac{\sum(X_i - A)^2}{n-1}}, A \neq \bar{X}$

2. For normal (symmetric distribution the following holds.

- Approximately 68.27% of the data values fall within one standard deviation of the mean. i.e. with in  $(\bar{X} - S, \bar{X} + S)$
- Approximately 95.45% of the data values fall within two standard deviations of the mean. i.e. with in  $(\bar{X} - 2S, \bar{X} + 2S)$
- Approximately 99.73% of the data values fall within three standard deviations of the mean. i.e. with in  $(\bar{X} - 3S, \bar{X} + 3S)$

### 3. Chebyshev's Theorem

For any data set ,no matter what the pattern of variation, the proportion of the values that fall with in k standard deviations of the mean or

$(\bar{X} - kS, \bar{X} + kS)$  will be at least  $1 - \frac{1}{k^2}$ , where k is an number greater

than 1. i.e. the proportion of items falling beyond k standard deviations of the mean is at most  $\frac{1}{k^2}$

**Example:** Suppose a distribution has mean 50 and standard deviation

6.What percent of the numbers are:

- a) Between 38 and 62
- b) Between 32 and 68
- c) Less than 38 or more than 62.
- d) Less than 32 or more than 68.

### Solutions:

a) 38 and 62 are at equal distance from the mean,50 and this distance is 12  
 $\Rightarrow ks = 12$

$$\Rightarrow k = \frac{12}{S} = \frac{12}{6} = 2$$

→ Applying the above theorem at least  $(1 - \frac{1}{k^2}) * 100\% = 75\%$  of the numbers lie between 38 and 62.

- b) Similarly done.
- c) It is just the complement of a) i.e. at most  $\frac{1}{k^2} * 100\% = 25\%$  of the numbers lie less than 32 or more than 62.
- d) Similarly done.

**Example 2:**

The average score of a special test of knowledge of wood refinishing has a mean of 53 and standard deviation of 6. Find the range of values in which at least 75% the scores will lie. (Exercise)

4. If the standard deviation of  $X_1, X_2, \dots, X_n$  is  $S$ , then the standard deviation of
- a)  $X_1 + k, X_2 + k, \dots, X_n + k$  will also be  $S$
  - b)  $kX_1, kX_2, \dots, kX_n$  would be  $|k|S$
  - c)  $a + kX_1, a + kX_2, \dots, a + kX_n$  would be  $|k|S$

**Exercise:** Verify each of the above relationship, considering  $k$  and  $a$  as constants.

**Examples:**

1. The mean and standard deviation of  $n$  Tetracycline Capsules  $X_1, X_2, \dots, X_n$  are known to be 12 gm and 3 gm respectively. New set of capsules of another drug are obtained by the linear transformation  $Y_i = 2X_i - 0.5$  ( $i = 1, 2, \dots, n$ ) then what will be the standard deviation of the new set of capsules
2. The mean and the standard deviation of a set of numbers are respectively 500 and 10.
  - a. If 10 is added to each of the numbers in the set, then what will be the variance and standard deviation of the new set?
  - b. If each of the numbers in the set are multiplied by -5, then what will be the variance and standard deviation of the new set?

**Solutions:**

1. Using c) above the new standard deviation =  $|k|S = 2 * 3 = 6$
2. a. They will remain the same.
- b. New standard deviation =  $|k|S = 5 * 10 = 50$

### **Coefficient of Variation (C.V)**

- Is defined as the ratio of standard deviation to the mean usually expressed as percents.

$$C.V = \frac{S}{\bar{X}} * 100$$

- The distribution having less C.V is said to be less variable or more consistent.

#### **Examples:**

1. An analysis of the monthly wages paid (in Birr) to workers in two firms A and B belonging to the same industry gives the following results

Value	Firm A	Firm B
Mean wage	52.5	47.5
Median wage	50.5	45.5
Variance	100	121

In which firm A or B is there greater variability in individual wages?

#### **Solutions:**

Calculate coefficient of variation for both firms.

$$C.V_A = \frac{S_A}{\bar{X}_A} * 100 = \frac{10}{52.5} * 100 = 19.05\%$$

$$C.V_B = \frac{S_B}{\bar{X}_B} * 100 = \frac{11}{47.5} * 100 = 23.16\%$$

Since  $C.V_A < C.V_B$ , in firm B there is greater variability in individual wages.

2. A meteorologist interested in the consistency of temperatures in three cities during a given week collected the following data. The temperatures for the five days of the week in the three cities were

City 1	25	24	23	26	17
City2	22	21	24	22	20
City3	32	27	35	24	28

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Which city have the most consistent temperature, based on these data?  
(Exercise)

### **Standard Scores (Z-scores)**

- If  $X$  is a measurement from a distribution with mean  $\bar{X}$  and standard deviation  $S$ , then its value in standard units is

$$Z = \frac{X - \mu}{\sigma}, \text{ for population.}$$

$$Z = \frac{X - \bar{X}}{S}, \text{ for sample}$$

- $Z$  gives the deviations from the mean in units of standard deviation
- $Z$  gives the number of standard deviation a particular observation lie above or below the mean.
- It is used to compare two observations coming from different groups.

#### **Examples:**

1. Two sections were given introduction to statistics examinations. The following information was given.

Value	Section 1	Section 2
Mean	78	90
Stan.deviation	6	5

Student A from section 1 scored 90 and student B from section 2 scored 95. Relatively speaking who performed better?

#### **Solutions:**

Calculate the standard score of both students.

$$Z_A = \frac{X_A - \bar{X}_1}{S_1} = \frac{90 - 78}{6} = 2$$

$$Z_B = \frac{X_B - \bar{X}_2}{S_2} = \frac{95 - 90}{5} = 1$$

→ Student A performed better relative to his section because the score of student A is two standard deviation above the mean score of his section

while, the score of student B is only one standard deviation above the mean score of his section.

2. Two groups of people were trained to perform a certain task and tested to find out which group is faster to learn the task. For the two groups the following information was given:

Value	Group one	Group two
Mean	10.4 min	11.9 min
Stan.dev.	1.2 min	1.3 min

Relatively speaking:

- a) Which group is more consistent in its performance
- b) Suppose a person A from group one take 9.2 minutes while person B from Group two take 9.3 minutes, who was faster in performing the task? Why?

### **Solutions:**

- a) Use coefficient of variation.

$$C.V_1 = \frac{S_1}{\bar{X}_1} * 100 = \frac{1.2}{10.4} * 100 = 11.54\%$$

$$C.V_2 = \frac{S_2}{\bar{X}_2} * 100 = \frac{1.3}{11.9} * 100 = 10.92\%$$

Since  $C.V_2 < C.V_1$ , group 2 is more consistent.

- b) Calculate the standard score of A and B

$$Z_A = \frac{X_A - \bar{X}_1}{S_1} = \frac{9.2 - 10.4}{1.2} = -1$$

$$Z_B = \frac{X_B - \bar{X}_2}{S_2} = \frac{9.3 - 11.9}{1.3} = -2$$

→ Child B is faster because the time taken by child B is two standard deviation shorter than the average time taken by group 2 while, the time taken by child A is only one standard deviation shorter than the average time taken by group 1.

### **Moments**

- If X is a variable that assume the values  $X_1, X_2, \dots, X_n$  then
1. The  $r^{\text{th}}$  moment is defined as:

$$\bar{X}^r = \frac{X_1^r + X_2^r + \dots + X_n^r}{n}$$

$$= \frac{\sum_{i=1}^n X_i^r}{n}$$

- For the case of frequency distribution this is expressed as:

$$\bar{X}^r = \frac{\sum_{i=1}^k f_i X_i^r}{n}$$

- If  $r = 1$ , it is the simple arithmetic mean, this is called the first moment.
2. The  $r^{\text{th}}$  moment about the mean (the  $r^{\text{th}}$  central moment)

- Denoted by  $M_r$  and defined as:

$$M_r = \frac{\sum_{i=1}^n (X_i - \bar{X})^r}{n} = \frac{(n-1)}{n} \frac{\sum_{i=1}^n (X_i - \bar{X})^r}{n-1}$$

- For the case of frequency distribution this is expressed as:

$$M_r = \frac{\sum_{i=1}^k f_i (X_i - \bar{X})^r}{n}$$

- If  $r = 2$ , it is population variance, this is called the second central moment.  
If we assume  $n-1 \approx n$ , it is also the sample variance.

3. The  $r^{\text{th}}$  moment about any number A is defined as:

- Denoted by  $M_r'$  and

$$M_r' = \frac{\sum_{i=1}^n (X_i - A)^r}{n} = \frac{(n-1)}{n} \frac{\sum_{i=1}^n (X_i - A)^r}{n-1}$$

- For the case of frequency distribution this is expressed as:

$$M_r' = \frac{\sum_{i=1}^k f_i (X_i - A)^r}{n}$$

**Example:**

1. Find the first two moments for the following set of numbers 2, 3, 7
2. Find the first three central moments of the numbers in problem 1
3. Find the third moment about the number 3 of the numbers in problem 1.

**Solutions:**

1. Use the  $r^{\text{th}}$  moment formula.

$$\bar{X}^r = \frac{\sum_{i=1}^n X_i^r}{n}$$

$$\Rightarrow \bar{X}^1 = \frac{2+3+7}{3} = 4 = \bar{X}$$

$$\bar{X}^2 = \frac{2^2 + 3^2 + 7^2}{3} = 20.67$$

2. Use the  $r^{\text{th}}$  central moment formula.

$$M_r = \frac{\sum_{i=1}^n (X_i - \bar{X})^r}{n}$$

$$\Rightarrow M_1 = \frac{(2-4) + (3-4) + (7-4)}{3} = 0$$

$$M_2 = \frac{(2-4)^2 + (3-4)^2 + (7-4)^2}{3} = 4.67$$

$$M_3 = \frac{(2-4)^3 + (3-4)^3 + (7-4)^3}{3} = 6$$

3. Use the  $r^{\text{th}}$  moment about A.

$$M_r = \frac{\sum_{i=1}^n (X_i - A)^r}{n}$$

$$\Rightarrow M_3 = \frac{(2-3)^3 + (3-3)^3 + (7-3)^3}{3} = 21$$

### Skewness

- Skewness is the degree of asymmetry or departure from symmetry of a distribution.
- A skewed frequency distribution is one that is not symmetrical.
- Skewness is concerned with the shape of the curve not size.
- If the frequency curve (smoothed frequency polygon) of a distribution has a longer tail to the right of the central maximum than to the left, the distribution is said to be skewed to the right or said to have positive skewness. If it has a longer tail to the left of the central maximum than to

the right, it is said to be skewed to the left or said to have negative skewness.

- For moderately skewed distribution, the following relation holds among the three commonly used measures of central tendency.

$$Mean - Mode = 3 * (Mean - Median)$$

### **Measures of Skewness**

- Denoted by  $\alpha_3$
- There are various measures of skewness.

1. The Pearsonian coefficient of skewness

$$\alpha_3 = \frac{Mean - Mode}{S \tan dard deviation} = \frac{\bar{X} - \hat{X}}{S}$$

2. The Bowley's coefficient of skewness ( coefficient of skewness based on quartiles)

$$\alpha_3 = \frac{(Q_3 - Q_2) - (Q_2 - Q_1)}{Q_3 - Q_1} = \frac{Q_3 + Q_1 - 2Q_2}{Q_3 - Q_1}$$

3. The moment coefficient of skewness

$$\alpha_3 = \frac{M_3}{M_2^{3/2}} = \frac{M_3}{(\sigma^2)^{3/2}} = \frac{M_3}{\sigma^3}, \text{ Where } \sigma \text{ is the population standard deviation.}$$

The shape of the curve is determined by the value of  $\alpha_3$

- If  $\alpha_3 > 0$  then the distribution is positively skewed.
- If  $\alpha_3 = 0$  then the distribution is symmetric.
- If  $\alpha_3 < 0$  then the distribution is negatively skewed.

### **Remark:**

- In a positively skewed distribution, smaller observations are more frequent than larger observations. i.e. the majority of the observations have a value below an average.
- In a negatively skewed distribution, smaller observations are less frequent than larger observations. i.e. the majority of the observations have a value above an average.

**Examples:**

- Suppose the mean, the mode, and the standard deviation of a certain distribution are 32, 30.5 and 10 respectively. What is the shape of the curve representing the distribution?

**Solutions:**

Use the Pearsonian coefficient of skewness

$$\alpha_3 = \frac{\text{Mean} - \text{Mode}}{\text{Standard deviation}} = \frac{32 - 30.5}{10} = 0.15$$

$\alpha_3 > 0 \Rightarrow \text{The distribution is positively skewed.}$

- In a frequency distribution, the coefficient of skewness based on the quartiles is given to be 0.5. If the sum of the upper and lower quartile is 28 and the median is 11, find the values of the upper and lower quartiles.

**Solutions:**

Given:  $\alpha_3 = 0.5$ ,  $\tilde{X} = Q_2 = 11$       Required:  $Q_1, Q_3$

$$Q_1 + Q_3 = 28 \dots \dots \dots (*)$$

$$\alpha_3 = \frac{(Q_3 - Q_2) - (Q_2 - Q_1)}{Q_3 - Q_1} = \frac{Q_3 + Q_1 - 2Q_2}{Q_3 - Q_1} = 0.5$$

Substituting the given values, one can obtain the following

$$Q_3 - Q_1 = 12 \dots \dots \dots (**)$$

Solving (\*) and (\*\*) at the same time we obtain the following values

$$Q_1 = 8 \quad \text{and} \quad Q_3 = 20$$

- Some characteristics of annually family income distribution (in Birr) in two regions is as follows:

Region	Mean	Median	Standard Deviation
A	6250	5100	960
B	6980	5500	940

- Calculate coefficient of skewness for each region
- For which region is, the income distribution more skewed. Give your interpretation for this Region
- For which region is the income more consistent?

Solutions: (exercise)

4. For a moderately skewed frequency distribution, the mean is 10 and the median is 8.5. If the coefficient of variation is 20%, find the Pearsonian coefficient of skewness and the probable mode of the distribution. (exercise)
5. The sum of fifteen observations, whose mode is 8, was found to be 150 with coefficient of variation of 20%
  - (a) Calculate the pearsonian coefficient of skewness and give appropriate conclusion.
  - (b) Are smaller values more or less frequent than bigger values for this distribution?
  - (c) If a constant  $k$  was added on each observation, what will be the new pearsonian coefficient of skewness? Show your steps. What do you conclude from this?

Solutions: (exercise)

## Kurtosis

Kurtosis is the degree of peakness of a distribution, usually taken relative to a normal distribution. A distribution having relatively high peak is called leptokurtic. If a curve representing a distribution is flat topped, it is called platykurtic. The normal distribution which is not very high peaked or flat topped is called mesokurtic.

### Measures of kurtosis

#### The moment coefficient of kurtosis:

- Denoted by  $\alpha_4$  and given by

$$\alpha_4 = \frac{M_4}{M_2^2} = \frac{M_4}{\sigma^4}$$

Where :  $M_4$  is the fourth moment about the mean.

$M_2$  is the second moment about the mean.

$\sigma$  is the population standard deviation.

The peakness depends on the value of  $\alpha_4$ .

- If  $\alpha_4 > 3$  then the curve is leptokurtic.

- If  $\alpha_4 = 3$  then the curve is mesokurtic.
- If  $\alpha_4 < 3$  then the curve is platykurtic.

**Examples:**

1. If the first four central moments of a distribution are:

$$M_1 = 0, M_2 = 16, M_3 = -60, M_4 = 162$$

- a) Compute a measure of skewness
- b) Compute a measure of kurtosis and give your interpretation.

**Solutions:**

$$\text{a)} \quad \alpha_3 = \frac{M_3}{M_2^{3/2}} = \frac{-60}{16^{3/2}} = -0.94 < 0$$

$\Rightarrow$  The distribution is negatively skewed.

$$\text{b)} \quad \alpha_4 = \frac{M_4}{M_2^2} = \frac{162}{16^2} = 0.6 < 3$$

$\Rightarrow$  The curve is platykurtic.

2. The median and the mode of a mesokurtic distribution are 32 and 34 respectively. The 4<sup>th</sup> moment about the mean is 243. Compute the Pearsonian coefficient of skewness and identify the type of skewness. Assume (n-1 = n).
  3. If the standard deviation of a symmetric distribution is 10, what should be the value of the fourth moment so that the distribution is mesokurtic?
- Solutions (exercise).**

## CHAPTER 5

### 5. ELEMENTARY PROBABILITY

#### Introduction

- Probability theory is the foundation upon which the logic of inference is built.
- It helps us to cope up with uncertainty.
- In general, probability is the chance of an outcome of an experiment. It is the measure of how likely an outcome is to occur.

#### Definitions of some probability terms

1. **Experiment:** Any process of observation or measurement or any process which generates well defined outcome.

2. **Probability Experiment:** It is an experiment that can be repeated any number of times under similar conditions and it is possible to enumerate the total number of outcomes without predicting an individual outcome. It is also called random experiment.

*Example:* If a fair die is rolled once it is possible to list all the possible outcomes i.e. 1, 2, 3, 4, 5, 6 but it is not possible to predict which outcome will occur.

3. **Outcome :**The result of a single trial of a random experiment  
4. **Sample Space:** Set of all possible outcomes of a probability experiment  
5. **Event:** It is a subset of sample space. It is a statement about one or more outcomes of a random experiment .They are denoted by capital letters.  
*Example:* Considering the above experiment let A be the event of odd numbers, B be the event of even numbers, and C be the event of number 8.

$$\Rightarrow A = \{1,3,5\}$$

$$B = \{2,4,6\}$$

$$C = \{ \} \text{ or empty space or impossible event}$$

Remark:

If S (sample space) has n members then there are exactly  $2^n$  subsets or events.

6. **Equally Likely Events:** Events which have the same chance of occurring.  
7. **Complement of an Event:** the complement of an event A means non-occurrence of A and is denoted by  $A'$ , or  $A^c$ , or  $\bar{A}$  contains those points of the sample space which don't belong to A.  
8. **Elementary Event:** an event having only a single element or sample point.  
9. **Mutually Exclusive Events:** Two events which cannot happen at the same time.  
10. **Independent Events:** Two events are independent if the occurrence of one does not affect the probability of the other occurring.  
11. **Dependent Events:** Two events are dependent if the first event affects the outcome or occurrence of the second event in a way the probability is changed.

**Example:** .What is the sample space for the following experiment

- Toss a die one time.
- Toss a coin two times.
- A light bulb is manufactured. It is tested for its life length by time.

**Solution**

- $S=\{1,2,3,4,5,6\}$

b)  $S=\{(HH),(HT),(TH),(TT)\}$

c)  $S=\{t / t \geq 0\}$

➤ Sample space can be

- Countable ( finite or infinite)
- Uncountable.

### **Counting Rules**

In order to calculate probabilities, we have to know

- The number of elements of an event
- The number of elements of the sample space.

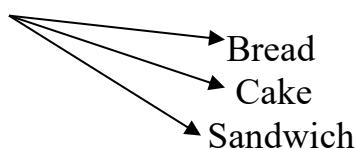
That is in order to judge what is **probable**, we have to know what is **possible**.

- In order to determine the number of outcomes, one can use several rules of counting.
  - The addition rule
  - The multiplication rule
  - Permutation rule
  - Combination rule
- To list the outcomes of the sequence of events, a useful device called **tree diagram** is used.

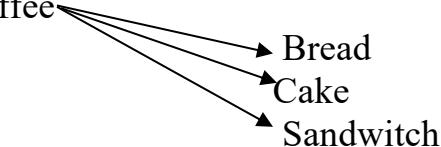
**Example:** A student goes to the nearest snack to have a breakfast. He can take tea, coffee, or milk with bread, cake and sandwich. How many possibilities does he have?

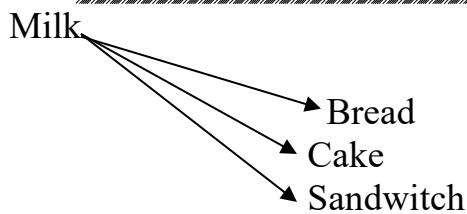
### **Solutions:**

Tea



Coffee





→ There are nine possibilities.

### **The Multiplication Rule:**

If a choice consists of  $k$  steps of which the first can be made in  $n_1$  ways, the second can be made in  $n_2$  ways..., the  $k^{\text{th}}$  can be made in  $n_k$  ways, then the whole choice can be made in  $(n_1 * n_2 * ..... * n_k)$  ways.

Example: The digits 0, 1, 2, 3, and 4 are to be used in 4 digit identification card. How many different cards are possible if

- Repetitions are permitted.
- Repetitions are not permitted.

### **Solutions**

a)

1 <sup>st</sup> digit	2 <sup>nd</sup> digit	3 <sup>rd</sup> digit	4 <sup>th</sup> digit
5	5	5	5

There are four steps

- Selecting the 1<sup>st</sup> digit, this can be made in 5 ways.
- Selecting the 2<sup>nd</sup> digit, this can be made in 5 ways.
- Selecting the 3<sup>rd</sup> digit, this can be made in 5 ways.
- Selecting the 4<sup>th</sup> digit, this can be made in 5 ways.

$$\Rightarrow 5 * 5 * 5 * 5 = 625 \text{ different cards are possible.}$$

b)

1 <sup>st</sup> digit	2 <sup>nd</sup> digit	3 <sup>rd</sup> digit	4 <sup>th</sup> digit
5	4	3	2

There are four steps

5. Selecting the 1<sup>st</sup> digit, this can be made in 5 ways.
6. Selecting the 2<sup>nd</sup> digit, this can be made in 4 ways.
7. Selecting the 3<sup>rd</sup> digit, this can be made in 3 ways.
8. Selecting the 4<sup>th</sup> digit, this can be made in 2 ways.

$\Rightarrow 5 * 4 * 3 * 2 = 120$  different cards are possible.

## **Permutation**

An arrangement of  $n$  objects in a specified order is called permutation of the objects.

### **Permutation Rules:**

1. The number of permutations of  $n$  distinct objects taken all together is  $n!$

$$\text{Where } n! = n * (n - 1) * (n - 2) * \dots * 3 * 2 * 1$$

2. The arrangement of  $n$  objects in a specified order using  $r$  objects at a time is called the permutation of  $n$  objects taken  $r$  objects at a time. It is written as  ${}_n P_r$  and the formula is

$${}_n P_r = \frac{n!}{(n - r)!}$$

3. The number of permutations of  $n$  objects in which  $k_1$  are alike  $k_2$  are alike ---- etc is

$${}_n P_r = \frac{n!}{k_1! * k_2! * \dots * k_n!}$$

### **Example:**

1. Suppose we have 4 letters A, B, C, D
  - a) How many permutations are there taking all the four?
  - b) How many permutations are there two letters at a time?
2. How many different permutations can be made from the letters in the word "CORRECTION"?

### Solutions:

1.

a)

Here  $n = 4$ , there are four distinct objects

$\Rightarrow$  There are  $4! = 24$  permutations.

b)

Here  $n = 4$ ,  $r = 2$

$\Rightarrow$  There are  ${}_4P_2 = \frac{4!}{(4-2)!} = \frac{24}{2} = 12$  permutations.

2.

Here  $n = 10$

Of which 2 are C, 2 are O, 2 are R, 1 E, 1 T, 1 I, 1 N

$\Rightarrow K_1 = 2, k_2 = 2, k_3 = 2, k_4 = k_5 = k_6 = k_7 = 1$

Using the 3<sup>rd</sup> rule of permutation, there are

$$\frac{10!}{2!*2!*2!*1!*1!*1!*1!} = 453600 \text{ permutations.}$$

### Exercises:

1. Six different statistics books, seven different physics books, and 3 different Economics books are arranged on a shelf. How many different arrangements are possible if;
  - i. The books in each particular subject must all stand together
  - ii. Only the statistics books must stand together
2. If the permutation of the word WHITE is selected at random, how many of the permutations
  - i. Begins with a consonant?
  - ii. Ends with a vowel?
  - iii. Has a consonant and vowels alternating?

### Combination

A selection of objects with out regard to order is called combination.

Example: Given the letters A, B, C, and D list the permutation and combination for selecting two letters.

Solutions:

Permutation	AC	BD	Combination
	AD	DC	
AB	BA	CA	DA
AC	BC	CB	DB
AD	BD	CD	DC

Note that in permutation AB is different from BA. But in combination AB is the same as BA.

### **Combination Rule**

The number of combinations of r objects selected from n objects is denoted by

$_n C_r$  or  $\binom{n}{r}$  and is given by the formula:

$$\binom{n}{r} = \frac{n!}{(n-r)! * r!}$$

Examples:

1. In how many ways a committee of 5 people be chosen out of 9 people?

Solutions:

$$n = 9, r = 5$$

$$\binom{n}{r} = \frac{n!}{(n-r)! * r!} = \frac{9!}{4! * 5!} = 126 \text{ ways}$$

2. Among 15 clocks there are two defectives .In how many ways can an inspector chose three of the clocks for inspection so that:

- a) There is no restriction.
- b) None of the defective clock is included.
- c) Only one of the defective clocks is included.
- d) Two of the defective clock is included.

Solutions:

$n = 15$  of which 2 are defective and 13 are non-defective.

$$r = 3$$

- a) If there is no restriction select three clocks from 15 clocks and this can be done in :

$$n = 15, r = 3$$

$$\binom{n}{r} = \frac{n!}{(n-r)! * r!} = \frac{15!}{12! * 3!} = 455 \text{ ways}$$

- b) None of the defective clocks is included.

This is equivalent to zero defective and three non defective, which can be done in:

$$\binom{2}{0} * \binom{13}{3} = 286 \text{ ways.}$$

- c) Only one of the defective clocks is included.

This is equivalent to one defective and two non defective, which can be done in:

$$\binom{2}{1} * \binom{13}{2} = 156 \text{ ways.}$$

- d) Two of the defective clock is included.

This is equivalent to two defective and one non defective, which can be done in:

$$\binom{2}{2} * \binom{13}{1} = 13 \text{ ways.}$$

### Exercises:

- Out of 5 Mathematician and 7 Statistician a committee consisting of 2 Mathematician and 3 Statistician is to be formed. In how many ways this can be done if
  - There is no restriction

- b) One particular Statistician should be included
  - c) Two particular Mathematicians can not be included on the committee.
2. If 3 books are picked at random from a shelf containing 5 novels, 3 books of poems, and a dictionary, in how many ways this can be done if
- a) There is no restriction.
  - b) The dictionary is selected?
  - c) 2 novels and 1 book of poems are selected?

### **Approaches to measuring Probability**

There are four different conceptual approaches to the study of probability theory. These are:

- The classical approach.
- The frequentist approach.
- The axiomatic approach.
- The subjective approach.

### **The classical approach**

This approach is used when:

- All outcomes are equally likely.
- Total number of outcome is finite, say N.

Definition: If a random experiment with N equally likely outcomes is conducted and out of these  $N_A$  outcomes are favourable to the event A, then the probability that event A occur denoted  $P(A)$  is defined as:

$$P(A) = \frac{N_A}{N} = \frac{\text{No.of outcomes favourable to } A}{\text{Total number of outcomes}} = \frac{n(A)}{n(S)}$$

Examples:

1. A fair die is tossed once. What is the probability of getting
  - a) Number 4?
  - b) An odd number?
  - c) An even number?
  - d) Number 8?

**Solutions:**

First identify the sample space, say S

$$S = \{1, 2, 3, 4, 5, 6\}$$

$$\Rightarrow N = n(S) = 6$$

a) Let A be the event of number 4

$$A = \{4\}$$

$$\Rightarrow N_A = n(A) = 1$$

$$P(A) = \frac{n(A)}{n(S)} = 1/6$$

b) Let A be the event of odd numbers

$$A = \{1, 3, 5\}$$

$$\Rightarrow N_A = n(A) = 3$$

$$P(A) = \frac{n(A)}{n(S)} = 3/6 = 0.5$$

c) Let A be the event of even numbers

$$A = \{2, 4, 6\}$$

$$\Rightarrow N_A = n(A) = 3$$

$$P(A) = \frac{n(A)}{n(S)} = 3/6 = 0.5$$

d) Let A be the event of number 8

$$A = \emptyset$$

$$\Rightarrow N_A = n(A) = 0$$

$$P(A) = \frac{n(A)}{n(S)} = 0/6 = 0$$

2. A box of 80 candles consists of 30 defective and 50 non defective candles. If 10 of this candles are selected at random, what is the probability

a) All will be defective.

- b) 6 will be non defective
- c) All will be non defective

**Solutions:**

$$\text{Total selection} = \binom{80}{10} = N = n(S)$$

- a) Let A be the event that all will be defective.

$$\text{Total way in which } A \text{ occur} = \binom{30}{10} * \binom{50}{0} = N_A = n(A)$$

$$\Rightarrow P(A) = \frac{n(A)}{n(S)} = \frac{\binom{30}{10} * \binom{50}{0}}{\binom{80}{10}} = 0.00001825$$

- b) Let A be the event that 6 will be non defective.

$$\text{Total way in which } A \text{ occur} = \binom{30}{4} * \binom{50}{6} = N_A = n(A)$$

$$\Rightarrow P(A) = \frac{n(A)}{n(S)} = \frac{\binom{30}{4} * \binom{50}{6}}{\binom{80}{10}} = 0.265$$

- c) Let A be the event that all will be non defective.

$$\text{Total way in which } A \text{ occur} = \binom{30}{0} * \binom{50}{10} = N_A = n(A)$$

$$\Rightarrow P(A) = \frac{n(A)}{n(S)} = \frac{\binom{30}{0} * \binom{50}{10}}{\binom{80}{10}} = 0.00624$$

**Exercises:**

1. What is the probability that a waitress will refuse to serve alcoholic beverages to only three minors if she randomly checks the I.D's of five students from among ten students of which four are not of legal age?
2. If 3 books are picked at random from a shelf containing 5 novels, 3 books of poems, and a dictionary, what is the probability that
  - a) The dictionary is selected?
  - b) 2 novels and 1 book of poems are selected?

**Short coming of the classical approach:**

This approach is not applicable when:

- The total number of outcomes is infinite.
- Outcomes are not equally likely.

**The Frequentist Approach**

This is based on the relative frequencies of outcomes belonging to an event.

Definition: The probability of an event A is the proportion of outcomes favourable to A in the long run when the experiment is repeated under same condition.

$$P(A) = \lim_{N \rightarrow \infty} \frac{N_A}{N}$$

**Example:** If records show that 60 out of 100,000 bulbs produced are defective.

What is the probability of a newly produced bulb to be defective?

Solution:

Let A be the event that the newly produced bulb is defective.

$$P(A) = \lim_{N \rightarrow \infty} \frac{N_A}{N} = \frac{60}{100,000} = 0.0006$$

**Axiomatic Approach:**

Let E be a random experiment and S be a sample space associated with E. With each event A a real number called the probability of A satisfies the following properties called axioms of probability or postulates of probability.

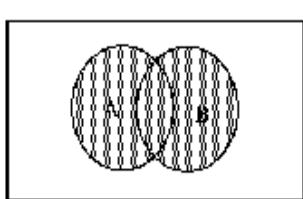
1.  $P(A) \geq 0$

2.  $P(S) = 1$ ,  $S$  is the sure event.
3. If A and B are mutually exclusive events, the probability that one or the other occur equals the sum of the two probabilities. i. e.

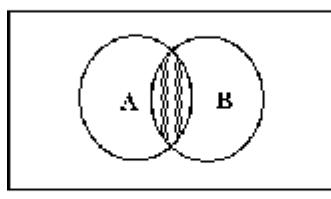
$$P(A \cup B) = P(A) + P(B)$$

4.  $P(A') = 1 - P(A)$
5.  $0 \leq P(A) \leq 1$
6.  $P(\emptyset) = 0$ ,  $\emptyset$  is the impossible event.

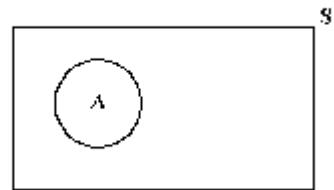
Remark: Venn-diagrams can be used to solve probability problems.



$A \cup B$



$A \cap B$



$A$

In general  $p(A \cup B) = p(A) + p(B) - p(A \cap B)$

### **Conditional probability and Independency**

Conditional Events: If the occurrence of one event has an effect on the next occurrence of the other event then the two events are conditional or dependant events.

**Example:** Suppose we have two red and three white balls in a bag

1. Draw a ball with replacement

Let A= the event that the first draw is red  $\rightarrow p(A) = \frac{2}{5}$

B= the event that the second draw is red  $\rightarrow p(B) = \frac{2}{5}$

A and B are independent.

2. Draw a ball with out replacement

Let A= the event that the first draw is red  $\rightarrow p(A) = \frac{2}{5}$

B= the event that the second draw is red  $\rightarrow p(B) = ?$

This is conditional.

Let B= the event that the second draw is red given that the first draw is red  $\rightarrow p(B) = 1/4$

### Conditional probability of an event

The conditional probability of an event A given that B has already occurred, denoted  $p(A/B)$  is

$$p(A/B) = \frac{p(A \cap B)}{p(B)}, \quad p(B) \neq 0$$

**Remark:** (1)  $p(A'/B) = 1 - p(A/B)$   
(2)  $p(B'/A) = 1 - p(B/A)$

### Examples

1. For a student enrolling at freshman at certain university the probability is 0.25 that he/she will get scholarship and 0.75 that he/she will graduate. If the probability is 0.2 that he/she will get scholarship and will also graduate. What is the probability that a student who get a scholarship graduate?

**Solution:** Let A= the event that a student will get a scholarship

B= the event that a student will graduate

$$\text{given } p(A) = 0.25, \quad p(B) = 0.75, \quad p(A \cap B) = 0.20$$

Required  $p(B/A)$

$$p(B/A) = \frac{p(A \cap B)}{p(A)} = \frac{0.20}{0.25} = 0.80$$

2. If the probability that a research project will be well planned is 0.60 and the probability that it will be well planned and well executed is 0.54, what is the probability that it will be well executed given that it is well planned?

**Solution;** Let A= the event that a research project will be well Planned

B= the event that a research project will be well

Executed

$$\text{given } p(A) = 0.60, \quad p(A \cap B) = 0.54$$

Required  $p(B/A)$

$$p(B/A) = \frac{p(A \cap B)}{p(A)} = \frac{0.54}{0.60} = 0.90$$

3. A lot consists of 20 defective and 80 non-defective items from which two items are chosen without replacement. Events A & B are defined as A = {the first item chosen is defective}, B = {the second item chosen is defective}
- What is the probability that both items are defective?
  - What is the probability that the second item is defective?

**Solution;** Exercise

**Note;** for any two events A and B the following relation holds.

$$p(B) = p(B/A)p(A) + p(B/A')p(A')$$

### Probability of Independent Events

Two events A and B are independent if and only if  $p(A \cap B) = p(A)p(B)$

Here  $p(A/B) = p(A)$ ,  $P(B/A) = p(B)$

**Example;** A box contains four black and six white balls. What is the probability of getting two black balls in drawing one after the other under the following conditions?

- The first ball drawn is not replaced
- The first ball drawn is replaced

**Solution;** Let A= first drawn ball is black

B= second drawn is black

Required  $p(A \cap B)$

$$\text{a. } p(A \cap B) = p(B/A)p(A) = (4/10)(3/9) = 2/15$$

$$\text{b. } p(A \cap B) = p(A)p(B) = (4/10)(4/10) = 4/25$$

## CHAPTER 6

### 6. RANDOM VARIABLES AND PROBABILITY DISTRIBUTIONS

Definition: A random variable is a numerical description of the outcomes of the experiment or a numerical valued function defined on sample space, usually denoted by capital letters.

Example: If  $X$  is a random variable, then it is a function from the elements of the sample space to the set of real numbers. i.e.

$X$  is a function  $X: S \rightarrow R$

→ A random variable takes a possible outcome and assigns a number to it.

Example: Flip a coin three times, let  $X$  be the number of heads in three tosses.

$$\begin{aligned}\Rightarrow S &= \{(HHH), (HHT), (HTH), (HTT), (THH), (THT), (TTH), (TTT)\} \\ \Rightarrow X(HHH) &= 3, \quad X(HHT) = X(HTH) = X(THH) = 2, \\ X(HTT) &= X(THT) = X(TTH) = 1 \\ X(TTT) &= 0\end{aligned}$$

$$X = \{0, 1, 2, 3, 4, 5\}$$

→  $X$  assumes a specific number of values with some probabilities.

Random variables are of two types:

1. Discrete random variable: are variables which can assume only a specific number of values. They have values that can be counted

Examples:

- Toss coin  $n$  times and count the number of heads.
- Number of children in a family.
- Number of car accidents per week.
- Number of defective items in a given company.
- Number of bacteria per two cubic centimeter of water.

2. Continuous random variable: are variables that can assume all values between any two give values.

Examples:

- Height of students at certain college.

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- Mark of a student.
- Life time of light bulbs.
- Length of time required to complete a given training.

Definition: a probability distribution consists of a value a random variable can assume and the corresponding probabilities of the values.

Example: Consider the experiment of tossing a coin three times. Let X is the number of heads. Construct the probability distribution of X.

Solution:

- First identify the possible value that X can assume.
- Calculate the probability of each possible distinct value of X and express X in the form of frequency distribution.

$X = x$	0	1	2	3
$P(X = x)$	1/8	3/8	3/8	1/8

Probability distribution is denoted by P for discrete and by f for continuous random variable.

### **Properties of Probability Distribution:**

1.

$$P(x) \geq 0, \text{ if } X \text{ is discrete.}$$

$$f(x) \geq 0, \text{ if } X \text{ is continuous.}$$

2.

$$\sum_x P(X = x) = 1, \text{ if } X \text{ is discrete.}$$

$$\int_x f(x) dx = 1, \text{ if } X \text{ is continuous.}$$

Note:

1. If X is a continuous random variable then

$$P(a < X < b) = \int_a^b f(x)dx$$

2. Probability of a fixed value of a continuous random variable is zero.

$$\Rightarrow P(a < X < b) = P(a \leq X < b) = P(a < X \leq b) = P(a \leq X \leq b)$$

3. If X is discrete random variable the

$$P(a < X < b) = \sum_{x=a+1}^{b-1} P(x)$$

$$P(a \leq X < b) = \sum_{x=a}^{b-1} p(x)$$

$$P(a < X \leq b) = \sum_{x=a+1}^b P(x)$$

$$P(a \leq X \leq b) = \sum_{x=a}^b P(x)$$

4. Probability means area for continuous random variable.

### **Introduction to expectation**

Definition:

1. Let a discrete random variable X assume the values  $X_1, X_2, \dots, X_n$  with the probabilities  $P(X_1), P(X_2), \dots, P(X_n)$  respectively. Then the expected value of X ,denoted as  $E(X)$  is defined as:

$$\begin{aligned} E(X) &= X_1 P(X_1) + X_2 P(X_2) + \dots + X_n P(X_n) \\ &= \sum_{i=1}^n X_i P(X_i) \end{aligned}$$

2. Let X be a continuous random variable assuming the values in the

interval  $(a, b)$  such that  $\int_a^b f(x)dx = 1$ ,then

$$E(X) = \int_a^b x f(x)dx$$

**Examples:**

- What is the expected value of a random variable X obtained by tossing a coin three times where is the number of heads

**Solution:**

First construct the probability distribution of X

$X = x$	0	1	2	3
$P(X = x)$	• 1/8	• 3/8	• 3/8	• 1/8

$$\begin{aligned} \Rightarrow E(X) &= X_1 P(X_1) + X_2 P(X_2) + \dots + X_n P(X_n) \\ &= 0 * 1/8 + 1 * 3/8 + \dots + 2 * 1/8 \\ &= 1.5 \end{aligned}$$

- Suppose a charity organization is mailing printed return-address stickers to over one million homes in the Ethiopia. Each recipient is asked to donate \$1, \$2, \$5, \$10, \$15, or \$20. Based on past experience, the amount a person donates is believed to follow the following probability distribution:

$X = x$	\$1	\$2	\$5	\$10	\$15	\$20
$P(X = x)$	0.1	0.2	0.3	0.2	0.15	0.05

What is expected that an average donor to contribute?

**Solution:**

$X = x$	\$1	\$2	\$5	\$10	\$15	\$20	Total
$P(X = x)$	0.1	0.2	0.3	0.2	0.15	0.05	1
$xP(X = x)$	0.1	0.4	1.5	2	2.25	1	7.25

$$\Rightarrow E(X) = \sum_{i=1}^6 x_i P(X = x_i) = \$7.25$$

## Mean and Variance of a random variable

Let X is given random variable.

1. The expected value of X is its mean  $\Rightarrow \text{Mean of } X = E(X)$
2. The variance of X is given by:

$$\text{Variance of } X = \text{var}(X) = E(X^2) - [E(X)]^2$$

Where:

$$\begin{aligned} E(X^2) &= \sum_{i=1}^n x_i^2 P(X = x_i) , \quad \text{if } X \text{ is discrete} \\ &= \int_x x^2 f(x) dx , \quad \text{if } X \text{ is continuous.} \end{aligned}$$

Examples:

1. Find the mean and the variance of a random variable X in example 2 above.

**Solutions:**

$X = x$	\$1	\$2	\$5	\$10	\$15	\$20	Total
$P(X = x)$	0.1	0.2	0.3	0.2	0.15	0.05	1
$xP(X = x)$	0.1	0.4	1.5	2	2.25	1	7.25
$x^2P(X = x)$	0.1	0.8	7.5	20	33.75	20	82.15

$$\Rightarrow E(X) = 7.25$$

$$\text{Var}(X) = E(X^2) - [E(X)]^2 = 82.15 - 7.25^2 = 29.59$$

2. Two dice are rolled. Let X be a random variable denoting the sum of the numbers on the two dice.
  - i) Give the probability distribution of X
  - ii) Compute the expected value of X and its variance

There are some general rules for mathematical expectation.

Let X and Y are random variables and k is a constant.

**RULE 1**  $E(k) = k$ , **RULE 2**  $Var(k) = k$ , **RULE 3**  $E(kX) = kE(X)$

**RULE 4**  $Var(kX) = k^2 Var(X)$

**RULE 5**  $E(X + Y) = E(X) + E(Y)$

## CHAPTER 7 COMMON PROBABILITY DISTRIBUTIONS

### Common Discrete Probability Distributions

## 1. Binomial Distribution

A binomial experiment is a probability experiment that satisfies the following four requirements called assumptions of a binomial distribution.

1. The experiment consists of  $n$  identical trials.
2. Each trial has only one of the two possible mutually exclusive outcomes, success or a failure.
3. The probability of each outcome does not change from trial to trial, and
4. The trials are independent, thus we must sample with replacement.

### Examples of binomial experiments

- Tossing a coin 20 times to see how many tails occur.
- Asking 200 people if they watch BBC news.
- Registering a newly produced product as defective or non defective.
- Asking 100 people if they favor the ruling party.
- Rolling a die to see if a 5 appears.

Definition: The outcomes of the binomial experiment and the corresponding probabilities of these outcomes are called **Binomial Distribution**.

Let  $P = \text{the probability of success}$

$q = 1 - p = \text{the probability of failure on any given trial}$

Then the probability of getting  $x$  successes in  $n$  trials becomes:

$$P(X = x) = \binom{n}{x} p^x q^{n-x}, \quad x = 0, 1, 2, \dots, n$$

And this is sometimes written as:

$$X \sim \text{Bin}(n, p)$$

When using the binomial formula to solve problems, we have to identify three things:

- The number of trials ( $n$ )
- The probability of a success on any one trial ( $p$ ) and
- The number of successes desired ( $X$ ).

**Examples:**

1. What is the probability of getting three heads by tossing a fair coin four times?

**Solution:**

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Let X be the number of heads in tossing a fair coin four times

$$X \sim Bin(n = 4, p = 0.50)$$

$$\Rightarrow P(X = x) = \binom{n}{x} p^x q^{n-x}, \quad x = 0,1,2,3,4.$$

$$\Rightarrow P(X = 3) = \binom{4}{3} 0.5^4 = 0.25$$

2. Suppose that an examination consists of six true and false questions, and assume that a student has no knowledge of the subject matter. The probability that the student will guess the correct answer to the first question is 30%. Likewise, the probability of guessing each of the remaining questions correctly is also 30%.
  - a) What is the probability of getting more than three correct answers?
  - b) What is the probability of getting at least two correct answers?
  - c) What is the probability of getting at most three correct answers?
  - d) What is the probability of getting less than five correct answers?

### Solution

Let X = the number of correct answers that the student gets.

$$X \sim Bin(n = 6, p = 0.30)$$

a)  $P(X > 3) = ?$

$$\Rightarrow P(X = x) = \binom{n}{x} p^x q^{n-x}, \quad x = 0,1,2,..6$$

$$= \binom{6}{x} 0.3^x 0.7^{6-x}$$

$$\Rightarrow P(X > 3) = P(X = 4) + P(X = 5) + P(X = 6)$$

$$= 0.060 + 0.010 + 0.001$$

$$= 0.071$$

Thus, we may conclude that

if 30% of the exam questions are answered by guessing, the probability is 0.071 (or 7.1%) that more than four of the questions are answered correctly by the student.

b)  $P(X \geq 2) = ?$

$$\begin{aligned}P(X \geq 2) &= P(X = 2) + P(X = 3) + P(X = 4) + P(X = 5) + P(X = 6) \\&= 0.324 + 0.185 + 0.060 + 0.010 + 0.001 \\&= 0.58\end{aligned}$$

c)  $P(X \leq 3) = ?$

$$\begin{aligned}P(X \leq 3) &= P(X = 0) + P(X = 1) + P(X = 2) + P(X = 3) \\&= 0.118 + 0.303 + 0.324 + 0.185 \\&= 0.93\end{aligned}$$

d)  $P(X < 5) = ?$

$$\begin{aligned}P(X < 5) &= 1 - P(X \geq 5) \\&= 1 - \{P(X = 5) + P(X = 6)\} \\&= 1 - (0.010 + 0.001) \\&= 0.989\end{aligned}$$

### Exercises:

1. Suppose that 4% of all TVs made by A&B Company in 2000 are defective. If eight of these TVs are randomly selected from across the country and tested, what is the probability that *exactly* three of them are defective? Assume that each TV is made independently of the others.
2. An allergist claims that 45% of the patients she tests are allergic to some type of weed. What is the probability that
  - a) Exactly 3 of her next 4 patients are allergic to weeds?
  - b) None of her next 4 patients are allergic to weeds?
3. Explain why the following experiments are not Binomial
  - Rolling a die until a 6 appears.
  - Asking 20 people how old they are.
  - Drawing 5 cards from a deck for a poker hand.

**Remark:** If X is a binomial random variable with parameters n and p then

$E(X) = np$	, $Var(X) = npq$
-------------	------------------

## 2. Poisson Distribution

- A random variable X is said to have a Poisson distribution if its probability distribution is given by:

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}, \quad x = 0, 1, 2, \dots$$

*Where  $\lambda$  = the average number.*

- The Poisson distribution depends *only* on the average number of occurrences per unit time or space.
- The Poisson distribution is used as a distribution of rare events, such as:
  - Number of misprints.
  - Natural disasters like earth quake.
  - Accidents.
  - Hereditary.
  - Arrivals
- The process that gives rise to such events are called Poisson process.

### Examples:

1. If 1.6 accidents can be expected at an intersection on any given day, what is the probability that there will be 3 accidents on any given day?

Solution; Let X = the number of accidents,  $\lambda = 1.6$

$$X = \text{poisson}(1.6) \Rightarrow p(X = x) = \frac{1.6^x e^{-1.6}}{x!}$$

$$p(X = 3) = \frac{1.6^3 e^{-1.6}}{3!} = 0.1380$$

2. On the average, five smokers pass a certain street corners every ten minutes, what is the probability that during a given 10 minutes the number of smokers passing will be
  - a. 6 or fewer

- b. 7 or more
- c. Exactly 8..... (Exercise)

If X is a Poisson random variable with parameters  $\lambda$  then

$E(X) = \lambda$	,	$Var(X) = \lambda$
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**Note:**

The Poisson probability distribution provides a close approximation to the binomial probability distribution when n is large and p is quite small or quite large with  $\lambda = np$ .

$$P(X = x) = \frac{(np)^x e^{-(np)}}{x!}, \quad x = 0, 1, 2, \dots$$

Where  $\lambda = np = \text{the average number}$ .

Usually we use this approximation if  $np \leq 5$ . In other words, if  $n > 20$  and  $np \leq 5$  [or  $n(1 - p) \leq 5$ ], then we may use Poisson distribution as an approximation to binomial distribution.

**Example:**

1. Find the binomial probability  $P(X=3)$  by using the Poisson distribution if  $p = 0.01$  and  $n = 200$

**Solution:**

$$\text{Using Poisson, } \lambda = np = 0.01 * 200 = 2$$

$$\Rightarrow P(X = 3) = \frac{2^3 e^{-2}}{3!} = 0.1804$$

$$\text{Using Binomial, } n = 200, p = 0.01$$

$$\Rightarrow P(X = 3) = \binom{200}{3} (0.01)^3 (0.99)^{99} = 0.1814$$

**Common Continuous Probability Distributions**

## 1. Normal Distribution

A random variable X is said to have a normal distribution if its probability density function is given by

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}, \quad -\infty < x < \infty, \quad -\infty < \mu < \infty, \quad \sigma > 0$$

Where  $\mu = E(X)$ ,  $\sigma^2 = \text{Variance}(X)$

$\mu$  and  $\sigma^2$  are the Parameters of the Normal Distribution.

### Properties of Normal Distribution:

1. It is bell shaped and is symmetrical about its mean and it is mesokurtic. The maximum ordinate is at  $x = \mu$  and is given by

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}}$$

2. It is asymptotic to the axis, i.e., it extends indefinitely in either direction from the mean.
3. It is a continuous distribution.
4. It is a family of curves, i.e., every unique pair of mean and standard deviation defines a different normal distribution. Thus, the normal distribution is completely described by two parameters: mean and standard deviation.
5. Total area under the curve sums to 1, i.e., the area of the distribution on each

$$\text{side of the mean is } 0.5. \Rightarrow \int_{-\infty}^{\infty} f(x)dx = 1$$

6. It is unimodal, i.e., values mound up only in the center of the curve.
7.  $\text{Mean} = \text{Median} = \text{mode} = \mu$
8. The probability that a random variable will have a value between any two points is equal to the area under the curve between those points.

**Note:** To facilitate the use of normal distribution, the following distribution known as the standard normal distribution was derived by using the transformation

$$Z = \frac{X - \mu}{\sigma}$$

$$\Rightarrow f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$$

### Properties of the Standard Normal Distribution:

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Same as a normal distribution, but also...

- Mean is zero
- Variance is one
- Standard Deviation is one

- Areas under the standard normal distribution curve have been tabulated in various ways. The most common ones are the areas between  $Z = 0$  and a positive value of  $Z$ .

- Given a normal distributed random variable  $X$  with

Mean  $\mu$  and standard deviation  $\sigma$

$$P(a < X < b) = P\left(\frac{a - \mu}{\sigma} < \frac{X - \mu}{\sigma} < \frac{b - \mu}{\sigma}\right)$$

$$\Rightarrow P(a < X < b) = P\left(\frac{a - \mu}{\sigma} < Z < \frac{b - \mu}{\sigma}\right)$$

**Note:**

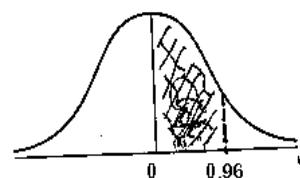
$$\begin{aligned} P(a < X < b) &= P(a \leq X < b) \\ &= P(a < X \leq b) \\ &= P(a \leq X \leq b) \end{aligned}$$

**Examples:**

1. Find the area under the standard normal distribution which lies
  - a) Between  $Z = 0$  and  $Z = 0.96$

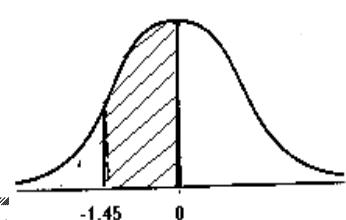
Solution:

$$Area = P(0 < Z < 0.96) = 0.3315$$



1. Find the area under the standard normal distribution which lies
  - b) Between  $Z = -1.45$  and  $Z = 0$

Solution:



$$\text{Area} = P(-1.45 < Z < 0)$$

$$= P(0 < Z < 1.45)$$

$$= 0.4265$$

c) To the right of  $Z = -0.35$

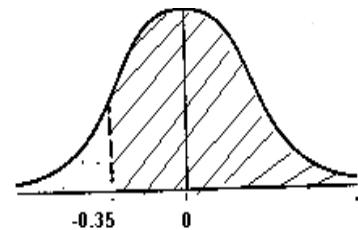
Solution:

$$\text{Area} = P(Z > -0.35)$$

$$= P(-0.35 < Z < 0) + P(Z > 0)$$

$$= P(0 < Z < 0.35) + P(Z > 0)$$

$$= 0.1368 + 0.50 = 0.6368$$



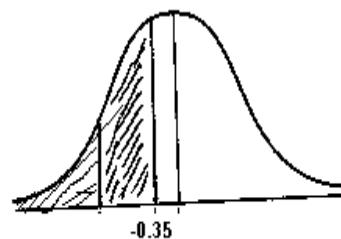
d) To the left of  $Z = -0.35$

Solution:

$$\text{Area} = P(Z < -0.35)$$

$$= 1 - P(Z > -0.35)$$

$$= 1 - 0.6368 = 0.3632$$



e) Between

$$Z = -0.67 \text{ and } Z = 0.75$$

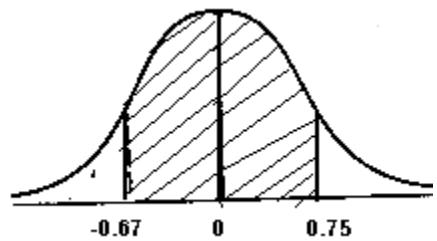
Solution:

$$\text{Area} = P(-0.67 < Z < 0.75)$$

$$= P(-0.67 < Z < 0) + P(0 < Z < 0.75)$$

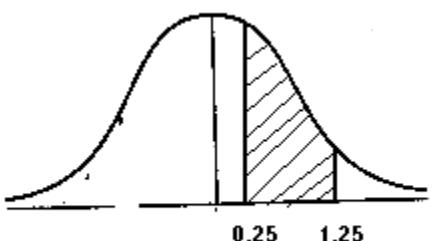
$$= P(0 < Z < 0.67) + P(0 < Z < 0.75)$$

$$= 0.2486 + 0.2734 = 0.5220$$



f) Between  $Z = 0.25$  and  $Z = 1.25$

Solution:



$$\begin{aligned}
 \text{Area} &= P(0.25 < Z < 1.25) \\
 &= P(0 < Z < 1.25) - P(0 < Z < 0.25) \\
 &= 0.3934 - 0.0987 = 0.2957
 \end{aligned}$$

**2.** Find the value of Z if

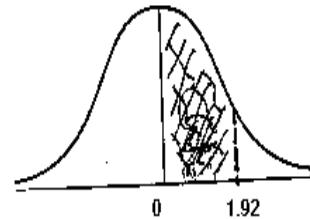
- a) The normal curve area between 0 and z(positive) is 0.4726

Solution

$$P(0 < Z < z) = 0.4726 \text{ and from table}$$

$$P(0 < Z < 1.92) = 0.4726$$

$$\Leftrightarrow z = 1.92 \dots \text{uniqueness of Areaa.}$$



- b) The area to the left of z is 0.9868

Solution

$$P(Z < z) = 0.9868$$

$$= P(Z < 0) + P(0 < Z < z)$$

$$= 0.50 + P(0 < Z < z)$$

$$\Rightarrow P(0 < Z < z) = 0.9868 - 0.50 = 0.4868$$

and from table

$$P(0 < Z < 2.2) = 0.4868$$

$$\Leftrightarrow z = 2.2$$

**3.** A random variable X has a normal distribution with mean 80 and standard deviation 4.8. What is the probability that it will take a value

- a) Less than 87.2
- b) Greater than 76.4
- c) Between 81.2 and 86.0

**Solution**

$X$  is normal with mean,  $\mu = 80$ , standard deviation,  $\sigma = 4.8$

a)

$$\begin{aligned}
 P(X < 87.2) &= P\left(\frac{X - \mu}{\sigma} < \frac{87.2 - \mu}{\sigma}\right) \\
 &= P\left(Z < \frac{87.2 - 80}{4.8}\right) \\
 &= P(Z < 1.5) \\
 &= P(Z < 0) + P(0 < Z < 1.5) \\
 &= 0.50 + 0.4332 = \underline{\underline{0.9332}}
 \end{aligned}$$

**b)**

$$\begin{aligned}
 P(X > 76.4) &= P\left(\frac{X - \mu}{\sigma} > \frac{76.4 - \mu}{\sigma}\right) \\
 &= P\left(Z > \frac{76.4 - 80}{4.8}\right) \\
 &= P(Z > -0.75) \\
 &= P(Z > 0) + P(0 < Z < 0.75) \\
 &= 0.50 + 0.2734 = \underline{\underline{0.7734}}
 \end{aligned}$$

**c)**

$$\begin{aligned}
 P(81.2 < X < 86.0) &= P\left(\frac{81.2 - \mu}{\sigma} < \frac{X - \mu}{\sigma} < \frac{86.0 - \mu}{\sigma}\right) \\
 &= P\left(\frac{81.2 - 80}{4.8} < Z < \frac{86.0 - 80}{4.8}\right) \\
 &= P(0.25 < Z < 1.25) \\
 &= P(0 < Z < 1.25) - P(0 < Z < 0.25) \\
 &= 0.3934 - 0.0987 = \underline{\underline{0.2957}}
 \end{aligned}$$

4. A normal distribution has mean 62.4. Find its standard deviation if 20.0% of the area under the normal curve lies to the right of 72.9

### Solution

$$\begin{aligned}
 P(X > 72.9) = 0.2005 &\Rightarrow P\left(\frac{X - \mu}{\sigma} > \frac{72.9 - \mu}{\sigma}\right) = 0.2005 \\
 &\Rightarrow P\left(Z > \frac{72.9 - 62.4}{\sigma}\right) = 0.2005 \\
 &\Rightarrow P\left(Z > \frac{10.5}{\sigma}\right) = 0.2005 \\
 &\Rightarrow P\left(0 < Z < \frac{10.5}{\sigma}\right) = 0.50 - 0.2005 = 0.2995 \\
 &\text{And from table } P(0 < Z < 0.84) = 0.2995 \\
 &\Leftrightarrow \frac{10.5}{\sigma} = 0.84 \\
 &\Rightarrow \sigma = \underline{\underline{12.5}}
 \end{aligned}$$

5. A random variable has a normal distribution with  $\sigma = 5$ . Find its mean if the probability that the random variable will assume a value less than 52.5 is 0.6915.

Solution

$$\begin{aligned}
 P(Z < z) = P\left(Z < \frac{52.5 - \mu}{5}\right) = 0.6915 \\
 \Rightarrow P(0 < Z < z) = 0.6915 - 0.50 = 0.1915.
 \end{aligned}$$

*But from the table*

$$\begin{aligned}
 &\Rightarrow P(0 < Z < 0.5) = 0.1915 \\
 &\Leftrightarrow z = \frac{52.5 - \mu}{5} = 0.5 \\
 &\Rightarrow \mu = \underline{\underline{50}}
 \end{aligned}$$

6. Of a large group of men, 5% are less than 60 inches in height and 40% are between 60 & 65 inches. Assuming a normal distribution, find the mean and standard deviation of heights.

Solution (Exercise)

## CHAPTER 7

### 7. Sampling and Sampling Distribution

#### Introduction

Given a variable  $X$ , if we arrange its values in ascending order and assign probability to each of the values or if we present  $X_i$  in a form of relative frequency distribution the result is called *Sampling Distribution of X*.

#### Definitions:

1. Parameter: Characteristic or measure obtained from a population.
2. Statistic: Characteristic or measure obtained from a sample.
3. Sampling: The process or method of sample selection from the population.
4. Sampling unit: the ultimate unit to be sampled or elements of the population to be sampled.

#### Examples:

- If some body studies Scio-economic status of the households, households are the sampling unit.
  - If one studies performance of freshman students in some college, the student is the sampling unit.
5. Sampling frame: is the list of all elements in a population.

#### Examples:

- List of households.
- List of students in the registrar office.

#### 6. Errors in sample survey:

There are two types of errors

- a) Sampling error:
  - Is the discrepancy between the population value and sample value.
  - May arise due to in appropriate sampling techniques applied
- b) Non sampling errors: are errors due to procedure bias such as:
  - Due to incorrect responses
  - Measurement
  - Errors at different stages in processing the data.

#### The Need for Sampling

- Reduced cost
- Greater speed
- Greater accuracy
- Greater scope
- More detailed information can be obtained.

- There are two types of sampling.

### **1. Random Sampling or probability sampling.**

- Is a method of sampling in which all elements in the population have a pre-assigned non-zero probability to be included in to the sample.

Examples:

- Simple random sampling
- Stratified random sampling
- Cluster sampling
- Systematic sampling

#### **1. Simple Random Sampling:**

- Is a method of selecting items from a population such that every possible sample of specific size has an equal chance of being selected. In this case, sampling may be with or without replacement. Or
- All elements in the population have the same pre-assigned non-zero probability to be included in to the sample.
- Simple random sampling can be done either using the lottery method or table of random numbers.

#### **2. Stratified Random Sampling:**

- The population will be divided in to non-overlapping but exhaustive groups called strata.
- Simple random samples will be chosen from each stratum.
- Elements in the same strata should be more or less homogeneous while different in different strata.
- It is applied if the population is heterogeneous.
- Some of the criteria for dividing a population into strata are: Sex (male, female); Age (under 18, 18 to 28, 29 to 39); Occupation (blue-collar, professional, other).

#### **3. Cluster Sampling:**

- The population is divided in to non-overlapping groups called clusters.
- A simple random sample of groups or cluster of elements is chosen and all the sampling units in the selected clusters will be surveyed.
- Clusters are formed in a way that elements with in a cluster are heterogeneous, i.e. observations in each cluster should be more or less dissimilar.
- Cluster sampling is useful when it is difficult or costly to generate a simple random sample. For example, to estimate the average annual household income in a large city we use cluster sampling, because to use simple random sampling we need a complete list of households in the city from which to sample. To use stratified random sampling, we

would again need the list of households. A less expensive way is to let each block within the city represent a cluster. A sample of clusters could then be randomly selected, and every household within these clusters could be interviewed to find the average annual household income.

#### **4. Systematic Sampling:**

- A complete list of all elements within the population (sampling frame) is required.
- The procedure starts in determining the first element to be included in the sample.
- Then the technique is to take the  $k^{\text{th}}$  item from the sampling frame.
- Let

$$N = \text{population size}, n = \text{sample size}, k = \frac{N}{n} = \text{sampling interval}.$$

- Choose any number between 1 and  $k$ . Suppose it is  $j (1 \leq j \leq k)$ .
- The  $j^{\text{th}}$  unit is selected at first and then  $(j+k)^{\text{th}}, (j+2k)^{\text{th}}, \dots, \text{etc}$  until the required sample size is reached.

#### **2. Non Random Sampling or non-probability sampling.**

- It is a sampling technique in which the choice of individuals for a sample depends on the basis of convenience, personal choice or interest.

Examples:

- Judgment sampling.
- Convenience sampling
- Quota Sampling.

#### **1. Judgment Sampling**

- In this case, the person taking the sample has direct or indirect control over which items are selected for the sample.

#### **2. Convenience Sampling**

- In this method, the decision maker selects a sample from the population in a manner that is relatively easy and convenient.

#### **3. Quota Sampling**

- In this method, the decision maker requires the sample to contain a certain number of items with a given characteristic. Many political polls are, in part, quota sampling.

**Note:**

let  $N = \text{population size}$ ,  $n = \text{sample size}$ .

1. Suppose simple random sampling is used
  - We have  $N^n$  possible samples if sampling is with replacement.
  - We have  $\binom{N}{n}$  possible samples if sampling is with out replacement.
2. After this on wards we consider that samples are drawn from a given population using simple random sampling.

### **Sampling Distribution of the sample mean**

- Sampling distribution of the sample mean is a theoretical probability distribution that shows the functional relation ship between the possible values of a given sample mean based on samples of size  $n$  and the probability associated with each value, for all possible samples of size  $n$  drawn from that particular population.
- There are commonly three properties of interest of a given sampling distribution.
  - Its Mean
  - Its Variance
  - Its Functional form.

### **Steps for the construction of Sampling Distribution of the mean**

1. From a finite population of size  $N$  , randomly draw all possible samples of size  $n$  .
2. Calculate the mean for each sample.
3. Summarize the mean obtained in step 2 in terms of frequency distribution or relative frequency distribution.

### **Example:**

Suppose we have a population of size  $N = 5$ , consisting of the age of five children: 6, 8, 10, 12, and 14

$$\Rightarrow \text{Population mean} = \mu = 10$$

$$\text{population Variance} = \sigma^2 = 8$$

Take samples of size 2 with replacement and construct sampling distribution of the sample mean.

**Solution:**

$$N = 5, \quad n = 2$$

→ We have  $N^n = 5^2 = 25$  possible samples since sampling is with replacement.

**Step 1:** Draw all possible samples:

	6	8	10	12	14
6	(6, 6)	(6, 8)	(6, 10)	(6, 12)	(6, 14)
8	(8, 6)	(8, 8)	(8, 10)	(8, 12)	(8, 14)
10	(10, 6)	(10, 8)	(10, 10)	(10, 12)	(10, 14)
12	(12, 6)	(12, 8)	(12, 10)	(12, 12)	(12, 14)
14	(12, 6)	(14, 8)	(12, 10)	(12, 12)	(12, 14)

**Step 2:** Calculate the mean for each sample:

	6	8	10	12	14
6	6	7	8	9	10
8	7	8	9	10	11
10	8	9	10	11	12
12	9	10	11	12	13
14	10	11	12	13	14

**Step 3:** Summarize the mean obtained in step 2 in terms of frequency distribution.

$\bar{X}$	Frequency
-----------	-----------

## Introduction to Statistics Hand Out

6	1
7	2
8	3
9	4
10	5
11	4
12	3
13	2
14	1

- a) Find the mean of  $\bar{X}$ , say  $\mu_{\bar{X}}$

$$\mu_{\bar{X}} = \frac{\sum \bar{X}_i f_i}{\sum f_i} = \frac{250}{25} = 10 = \mu$$

- b) Find the variance of  $\bar{X}$ , say  $\sigma_{\bar{X}}^2$

$$\sigma_{\bar{X}}^2 = \frac{\sum (\bar{X}_i - \mu_{\bar{X}})^2 f_i}{\sum f_i} = \frac{100}{25} = 4 \neq \sigma^2$$

**Remark:**

1. In general if sampling is with replacement

$$\sigma_{\bar{X}}^2 = \frac{\sigma^2}{n}$$

2. If sampling is with out replacement

$$\sigma_{\bar{X}}^2 = \frac{\sigma^2}{n} \left( \frac{N-n}{N-1} \right)$$

3. In any case the sample mean is unbiased estimator of the population mean. i.e.  $\mu_{\bar{X}} = \mu \Rightarrow E(\bar{X}) = \mu$  (Show!)

- Sampling may be from a normally distributed population or from a non - normally distributed population.
- When sampling is from a normally distributed population, the distribution of  $\bar{X}$  will posses the following property.

1. The distribution of  $\bar{X}$  will be normal
2. The mean of  $\bar{X}$  is equal to the population mean , i.e.  $\mu_{\bar{X}} = \mu$

3. The variance of  $\bar{X}$  is equal to the population variance divided by the

$$\text{sample size, i.e. } \sigma_{\bar{X}}^2 = \frac{\sigma^2}{n}$$

$$\Rightarrow \bar{X} \sim N(\mu, \frac{\sigma^2}{n})$$

$$\Rightarrow Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0,1)$$

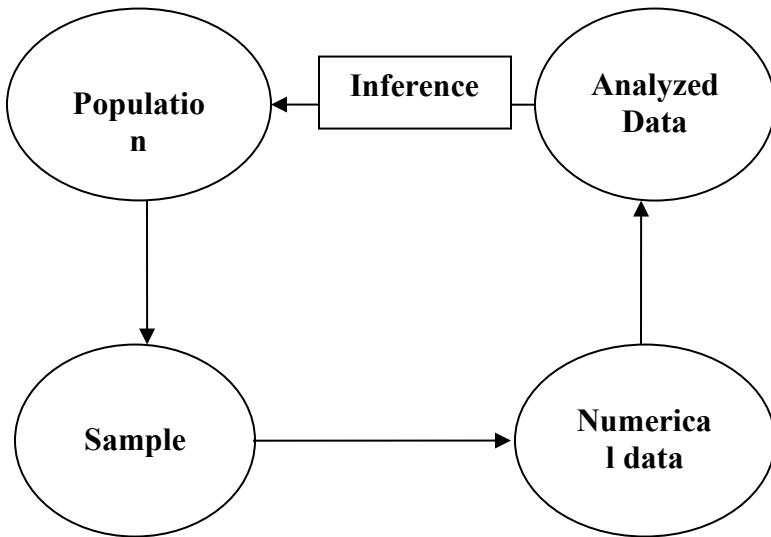
### Central Limit Theorem

Given a population of any functional form with mean  $\mu$  and finite variance  $\sigma^2$ , the sampling distribution of  $\bar{X}$ , computed from samples of size  $n$  from the population will be approximately normally distributed with mean  $\mu$  and variance  $\frac{\sigma^2}{n}$ , when the sample size is large.

## CHAPTER 8

### 8. ESTIMATION AND HYPOTHESIS TESTING

- Inference is the process of making interpretations or conclusions from sample data for the totality of the population.
- It is only the sample data that is ready for inference.
- In statistics there are two ways through which inference can be made.
  - ❖ Statistical estimation
  - ❖ Statistical hypothesis testing.



Data analysis is the process of extracting relevant information from the summarized data.

## Statistical Estimation

This is one way of making inference about the population parameter where the investigator does not have any prior notion about values or characteristics of the population parameter.

There are two ways estimation.

### 1) Point Estimation

It is a procedure that results in a single value as an estimate for a parameter.

### 2) Interval estimation

It is the procedure that results in the interval of values as an estimate for a parameter, which is interval that contains the likely values of a parameter. It deals with identifying the upper and lower limits of a parameter. The limits by themselves are random variable.

## Definitions

**Confidence Interval:** An interval estimate with a specific level of confidence

**Confidence Level:** The percent of the time the true value will lie in the interval estimate given.

**Consistent Estimator:** An estimator which gets closer to the value of the parameter as the sample size increases.

**Degrees of Freedom:** The number of data values which are allowed to vary once a statistic has been determined.

**Estimator:** A sample statistic which is used to estimate a population parameter. It must be unbiased, consistent, and relatively efficient.

**Estimate:** Is the different possible values which an estimator can assumes.

**Interval Estimate:** A range of values used to estimate a parameter.

**Point Estimate:** A single value used to estimate a parameter.

**Relatively Efficient Estimator:** The estimator for a parameter with the smallest variance.

**Unbiased Estimator:** An estimator whose expected value is the value of the parameter being estimated.

## Point and Interval estimation of the population mean: $\mu$

### ☞ Point Estimation

Another term for statistic is **point estimate**, since we are estimating the parameter value. A **point estimator** is the mathematical way we compute the point estimate. For instance, sum of  $x_i$  over n is the point estimator used to compute the estimate of the population

means,  $\mu$ . That is  $\bar{X} = \frac{\sum x_i}{n}$  is a point estimator of the population mean.

### ☞ Confidence interval estimation of the population mean

Although  $\bar{X}$  possesses nearly all the qualities of a good estimator, because of sampling error, we know that it's not likely that our sample statistic will be equal to the population parameter, but instead will fall into an interval of values. We will have to be satisfied knowing that the statistic is "close to" the parameter. That leads to the obvious question, what is "close"?

We can phrase the latter question differently: How confident can we be that the value of the statistic falls within a certain "distance" of the parameter? Or, what is the probability that the parameter's value is within a certain range of the statistic's value? This range is the confidence interval.

The **confidence level** is the *probability* that the value of the parameter falls within the range specified by the confidence interval surrounding the statistic.

There are different cases to be considered to construct confidence intervals.

### Case 1:

#### **If sample size is large or if the population is normal with known variance**

Recall the *Central Limit Theorem*, which applies to the sampling distribution of the mean of a sample. Consider samples of size n drawn from a population, whose mean is  $\mu$  and standard deviation is  $\sigma$  with replacement and order important. The population can have any frequency distribution. The sampling distribution of  $\bar{X}$  will have a mean

$\mu_{\bar{x}} = \mu$  and a standard deviation  $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$ , and approaches a normal distribution as

$n$  gets large. This allows us to use the normal distribution curve for computing confidence intervals.

$$\Rightarrow Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \text{ has a normal distribution with mean } = 0 \text{ and variance } = 1$$

$$\Rightarrow \mu = \bar{X} \pm Z \sigma/\sqrt{n}$$

$$= \bar{X} \pm \varepsilon, \quad \text{where } \varepsilon \text{ is a measure of error.}$$

$$\Rightarrow \varepsilon = Z \sigma/\sqrt{n}$$

- For the interval estimator to be good the error should be small. How it be small?
  - By making n large
  - Small variability
  - Taking Z small
- To obtain the value of Z, we have to attach this to a theory of chance. That is, there is an area of size  $1 - \alpha$  such that  $P(-Z_{\alpha/2} < Z < Z_{\alpha/2}) = 1 - \alpha$

Where  $\alpha$  is the probability that the parameter lies outside the interval

$Z_{\alpha/2}$  stands for the standard normal variable to the right of which

$\alpha/2$  probability lies, i.e  $P(Z > Z_{\alpha/2}) = \alpha/2$

$$\Rightarrow P(-Z_{\alpha/2} < \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} < Z_{\alpha/2}) = 1 - \alpha$$

$$\Rightarrow P(\bar{X} - Z_{\alpha/2} \sigma/\sqrt{n} < \mu < \bar{X} + Z_{\alpha/2} \sigma/\sqrt{n}) = 1 - \alpha$$

$\Rightarrow (\bar{X} - Z_{\alpha/2} \sigma/\sqrt{n}, \bar{X} + Z_{\alpha/2} \sigma/\sqrt{n})$  is a  $100(1 - \alpha)\%$  confidence interval for  $\mu$

But usually  $\sigma^2$  is not known, in that case we estimate by its point estimator  $S^2$

$\Rightarrow (\bar{X} - Z_{\alpha/2} S/\sqrt{n}, \bar{X} + Z_{\alpha/2} S/\sqrt{n})$  is a  $100(1 - \alpha)\%$  confidence interval for  $\mu$

Here are the z values corresponding to the most commonly used confidence levels.

$100(1 - \alpha)\%$	$\alpha$	$\alpha/2$	$Z_{\alpha/2}$
90	0.10	0.05	1.645
95	0.05	0.025	1.96

**Case 2:**

**If sample size is small and the population variance,  $\sigma^2$  is not known.**

$$t = \frac{\bar{X} - \mu}{S/\sqrt{n}} \quad \text{has } t \text{ distribution with } n-1 \text{ degrees of freedom.}$$

$\Rightarrow (\bar{X} - t_{\alpha/2} S/\sqrt{n}, \quad \bar{X} + t_{\alpha/2} S/\sqrt{n})$  is a  $100(1-\alpha)\%$  confidence interval for  $\mu$

The unit of measurement of the confidence interval is the standard error. This is just the standard deviation of the sampling distribution of the statistic.

Examples:

1. From a normal sample of size 25 a mean of 32 was found. Given that the population standard deviation is 4.2. Find
  - a) A 95% confidence interval for the population mean.
  - b) A 99% confidence interval for the population mean.

Solution:

a)

$$\bar{X} = 32, \quad \sigma = 4.2, \quad 1 - \alpha = 0.95 \Rightarrow \alpha = 0.05, \alpha/2 = 0.025$$

$$\Rightarrow Z_{\alpha/2} = 1.96 \text{ from table.}$$

$$\Rightarrow \text{The required interval will be } \bar{X} \pm Z_{\alpha/2} \sigma / \sqrt{n}$$

$$= 32 \pm 1.96 * 4.2 / \sqrt{25}$$

$$= 32 \pm 1.65$$

$$= \underline{\underline{(30.35, 33.65)}}$$

b)

$$\bar{X} = 32, \quad \sigma = 4.2, \quad 1 - \alpha = 0.99 \Rightarrow \alpha = 0.01, \quad \alpha/2 = 0.005$$

$$\Rightarrow Z_{\alpha/2} = 2.58 \text{ from table.}$$

$$\Rightarrow \text{The required interval will be } \bar{X} \pm Z_{\alpha/2} \sigma / \sqrt{n}$$

$$= 32 \pm 2.58 * 4.2 / \sqrt{25}$$

$$= 32 \pm 2.17$$

$$= \underline{\underline{(29.83, 34.17)}}$$

2. A drug company is testing a new drug which is supposed to reduce blood pressure. From the six people who are used as subjects, it is found that the average drop in blood pressure is 2.28 points, with a standard deviation of .95 points. What is the 95% confidence interval for the mean change in pressure?

Solution:

$$\bar{X} = 2.28, \quad S = 0.95, \quad 1 - \alpha = 0.95 \Rightarrow \alpha = 0.05, \quad \alpha/2 = 0.025$$

$$\Rightarrow t_{\alpha/2} = 2.571 \text{ with } df = 5 \text{ from table.}$$

$$\Rightarrow \text{The required interval will be } \bar{X} \pm t_{\alpha/2} S / \sqrt{n}$$

$$= 2.28 \pm 2.571 * 0.95 / \sqrt{6}$$

$$= 2.28 \pm 1.008$$

$$= \underline{\underline{(1.28, 3.28)}}$$

That is, we can be 95% confident that the mean decrease in blood pressure is between 1.28 and 3.28 points.

## Hypothesis Testing

- This is also one way of making inference about population parameter, where the investigator has prior notion about the value of the parameter.

Definitions:

- **Statistical hypothesis:** is an assertion or statement about the population whose plausibility is to be evaluated on the basis of the sample data.
- **Test statistic:** is a statistics whose value serves to determine whether to reject or accept the hypothesis to be tested. It is a random variable.
- **Statistic test:** is a test or procedure used to evaluate a statistical hypothesis and its value depends on sample data.

There are two types of hypothesis:

### **Null hypothesis:**

- It is the hypothesis to be tested.
- It is the hypothesis of equality or the hypothesis of no difference.
- Usually denoted by  $H_0$ .

### **Alternative hypothesis:**

- It is the hypothesis available when the null hypothesis has to be rejected.
- It is the hypothesis of difference.
- Usually denoted by  $H_1$  or  $H_a$ .

### **Types and size of errors:**

- Testing hypothesis is based on sample data which may involve sampling and non sampling errors.
- The following table gives a summary of possible results of any hypothesis test:

		Decision	
		Reject $H_0$	Don't reject $H_0$
Truth	$H_0$	Type I Error	Right Decision
	$H_1$	Right Decision	Type II Error

- Type I error: Rejecting the null hypothesis when it is true.
- Type II error: Failing to reject the null hypothesis when it is false.

NOTE:

1. There are errors that are prevalent in any two choice decision making problems.
2. There is always a possibility of committing one or the other errors.
3. Type I error ( $\alpha$ ) and type II error ( $\beta$ ) have inverse relationship and therefore, can not be minimized at the same time.

- In practice we set  $\alpha$  at some value and design a test that minimize  $\beta$ . This is because a type I error is often considered to be more serious, and therefore more important to avoid, than a type II error.

### General steps in hypothesis testing:

1. The first step in hypothesis testing is to specify the null hypothesis ( $H_0$ ) and the alternative hypothesis ( $H_1$ ).
2. The next step is to select a significance level,  $\alpha$
3. Identify the sampling distribution of the estimator.
4. The fourth step is to calculate a statistic analogous to the parameter specified by the null hypothesis.
5. Identify the critical region.
6. Making decision.
7. Summarization of the result.

### Hypothesis testing about the population mean, $\mu$ :

Suppose the assumed or hypothesized value of  $\mu$  is denoted by  $\mu_0$ , then one can formulate two sided (1) and one sided (2 and 3) hypothesis as follows:

1.  $H_0 : \mu = \mu_0$       vs       $H_1 : \mu \neq \mu_0$
2.  $H_0 : \mu = \mu_0$       vs       $H_1 : \mu > \mu_0$
3.  $H_0 : \mu = \mu_0$       vs       $H_1 : \mu < \mu_0$

### CASES:

#### Case 1: When sampling is from a normal distribution with $\sigma^2$ known

- The relevant test statistic is

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

- After specifying  $\alpha$  we have the following regions (critical and acceptance) on the standard normal distribution corresponding to the above three hypothesis.

Summary table for decision rule.

$H_0$	Reject $H_0$ if	Accept $H_0$ if	Inconclusive if
$\mu \neq \mu_0$	$ Z_{cal}  > Z_{\alpha/2}$	$ Z_{cal}  < Z_{\alpha/2}$	$Z_{cal} = Z_{\alpha/2}$ or $Z_{cal} = -Z_{\alpha/2}$
$\mu < \mu_0$	$Z_{cal} < -Z_\alpha$	$Z_{cal} > -Z_\alpha$	$Z_{cal} = -Z_\alpha$
$\mu > \mu_0$	$Z_{cal} > Z_\alpha$	$Z_{cal} < Z_\alpha$	$Z_{cal} = Z_\alpha$

$$\text{Where: } Z_{cal} = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$$

**Case 2: When sampling is from a normal distribution with  $\sigma^2$  unknown and small sample size**

- The relevant test statistic is

$$t = \frac{\bar{X} - \mu}{S / \sqrt{n}} \sim t \text{ with } n-1 \text{ degrees of freedom.}$$

- After specifying  $\alpha$  we have the following regions on the student t-distribution corresponding to the above three hypothesis.

$H_0$	Reject $H_0$ if	Accept $H_0$ if	Inconclusive if
$\mu \neq \mu_0$	$ t_{cal}  > t_{\alpha/2}$	$ t_{cal}  < t_{\alpha/2}$	$t_{cal} = t_{\alpha/2}$ or $t_{cal} = -t_{\alpha/2}$
$\mu < \mu_0$	$t_{cal} < -t_\alpha$	$t_{cal} > -t_\alpha$	$t_{cal} = -t_\alpha$
$\mu > \mu_0$	$t_{cal} > t_\alpha$	$t_{cal} < t_\alpha$	$t_{cal} = t_\alpha$

$$\text{Where: } t_{cal} = \frac{\bar{X} - \mu_0}{S / \sqrt{n}}$$

**Case3: When sampling is from a non- normally distributed population or a population whose functional form is unknown.**

- If a sample size is large one can perform a test hypothesis about the mean by using:

$$Z_{cal} = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}, \quad \text{if } \sigma^2 \text{ is known.}$$

$$= \frac{\bar{X} - \mu_0}{S/\sqrt{n}}, \quad \text{if } \sigma^2 \text{ is unknown.}$$

- The decision rule is the same as **case I.**

**Examples:**

1. Test the hypotheses that the average height content of containers of certain lubricant is 10 liters if the contents of a random sample of 10 containers are 10.2, 9.7, 10.1, 10.3, 10.1, 9.8, 9.9, 10.4, 10.3, and 9.8 liters. Use the 0.01 level of significance and assume that the distribution of contents is normal.

**Solution:**

Let  $\mu = \text{Population mean.}, \quad \mu_0 = 10$

Step 1: Identify the appropriate hypothesis

$$H_0: \mu = 10 \quad \text{vs} \quad H_1: \mu \neq 10$$

Step 2: select the level of significance,  $\alpha = 0.01(\text{given})$

Step 3: Select an appropriate test statistics

t- Statistic is appropriate because population variance is not known and the sample size is also small.

Step 4: identify the critical region.

Here we have two critical regions since we have two tailed hypothesis

$$\text{The critical region is } |t_{cal}| > t_{0.005}(9) = 3.2498$$

$\Rightarrow (-3.2498, 3.2498)$  is acceptance region.

Step 5: Computations:

$$\bar{X} = 10.06, \quad S = 0.25$$

$$\Rightarrow t_{cal} = \frac{\bar{X} - \mu_0}{S/\sqrt{n}} = \frac{10.06 - 10}{0.25/\sqrt{10}} = 0.76$$

Step 6: Decision

Accept  $H_0$ , since  $t_{cal}$  is in the acceptance region.

### Step 7: Conclusion

At 1% level of significance, we have no evidence to say that the average height content of containers of the given lubricant is different from 10 litters, based on the given sample data.

2. The mean life time of a sample of 16 fluorescent light bulbs produced by a company is computed to be 1570 hours. The population standard deviation is 120 hours. Suppose the hypothesized value for the population mean is 1600 hours. Can we conclude that the life time of light bulbs is decreasing? (Use  $\alpha = 0.05$  and assume the normality of the population)

### Solution:

Let  $\mu = \text{Population mean. , } \mu_0 = 1600$

Step 1: Identify the appropriate hypothesis

$$H_0 : \mu = 1600 \quad \text{vs} \quad H_1 : \mu < 1600$$

Step 2: select the level of significance,  $\alpha = 0.05$  (*given*)

Step 3: Select an appropriate test statistics

Z- Statistic is appropriate because population variance is known.

Step 4: identify the critical region.

*The critical region is  $Z_{\text{cal}} < -Z_{0.05} = -1.645$*

$\Rightarrow (-1.645, \infty)$  is acceptance region.

Step 5: Computations:

$$Z_{\text{cal}} = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} = \frac{1570 - 1600}{120/\sqrt{16}} = -1.0$$

Step 6: Decision

Accept  $H_0$ , since  $Z_{\text{cal}}$  is in the acceptance region.

Step 7: Conclusion

At 5% level of significance, we have no evidence to say that the life time of light bulbs is decreasing, based on the given sample data.

3. It is known in a pharmacological experiment that rats fed with a particular diet over a certain period gain an average of 40 gms in weight. A new diet was tried on a sample of 20 rats yielding a weight gain of 43 gms with variance 7 gms<sup>2</sup>. Test the hypothesis that the new diet is an improvement assuming normality.

- a) State the appropriate hypothesis
- b) What is the appropriate test statistic? Why?

- c) Identify the critical region(s)
- d) On the basis of the given information test the hypothesis and make conclusion.

**Solution (exercise).**

### **Test of Association**

- Suppose we have a population consisting of observations having two attributes or qualitative characteristics say A and B.
- If the attributes are independent then the probability of possessing both A and B is  $P_A * P_B$

Where  $P_A$  is the probability that a number has attribute A.

$P_B$  is the probability that a number has attribute B.

- Suppose A has  $r$  mutually exclusive and exhaustive classes.  
B has  $c$  mutually exclusive and exhaustive classes
- The entire set of data can be represented using  $r * c$  contingency table.

		B							
A		B <sub>1</sub>	B <sub>2</sub>	.	.	B <sub>j</sub>	.	B <sub>c</sub>	Total
A <sub>1</sub>	O <sub>11</sub>	O <sub>12</sub>			O <sub>1j</sub>		O <sub>1c</sub>	R <sub>1</sub>	
A <sub>2</sub>	O <sub>21</sub>	O <sub>22</sub>			O <sub>2j</sub>		O <sub>2c</sub>	R <sub>2</sub>	
.									
.									
A <sub>i</sub>	O <sub>i1</sub>	O <sub>i2</sub>			O <sub>ij</sub>		O <sub>ic</sub>	R <sub>i</sub>	
.									
.									
A <sub>r</sub>	O <sub>r1</sub>	O <sub>r2</sub>			O <sub>rz</sub>		O <sub>rc</sub>		
Total	C <sub>1</sub>	C <sub>2</sub>			C <sub>j</sub>				n

- The chi-square procedure test is used to test the hypothesis of independency of two attributes .For instance we may be interested
  - Whether the presence or absence of hypertension is independent of smoking habit or not.
  - Whether the size of the family is independent of the level of education attained by the mothers.
  - Whether there is association between father and son regarding boldness.

- Whether there is association between stability of marriage and period of acquaintance ship prior to marriage.

- The  $\chi^2$  statistic is given by:

$$\chi^2_{cal} = \sum_{i=1}^r \sum_{j=1}^c \left[ \frac{(O_{ij} - e_{ij})^2}{e_{ij}} \right] \sim \chi^2_{(r-1)(c-1)}$$

Where  $O_{ij}$  = the number of units that belong to category  $i$  of  $A$  and  $j$  of  $B$ .

$e_{ij}$  = Expected frequency that belong to category  $i$  of  $A$  and  $j$  of  $B$ .

- The  $e_{ij}$  is given by :

$$e_{ij} = \frac{R_i * C_j}{n}$$

Where  $R_i$  = the  $i^{th}$  row total.

$C_j$  = the  $j^{th}$  column total.

$n$  = total number of observations

### **Remark:**

$$n = \sum_{i=1}^r \sum_{j=1}^c O_{ij} = \sum_{i=1}^r \sum_{j=1}^c e_{ij}$$

- The null and alternative hypothesis may be stated as:

$H_0$  : There is no association between  $A$  and  $B$ .

$H_1$  : not  $H_0$  (There is association between  $A$  and  $B$ ).

### **Decision Rule:**

- Reject  $H_0$  for independency at  $\alpha$  level of significance if the calculated value of  $\chi^2$  exceeds the tabulated value with degree of freedom equal to  $(r-1)(c-1)$ .

$$\Rightarrow \text{Reject } H_0 \text{ if } \chi^2_{cal} = \sum_{i=1}^r \sum_{j=1}^c \left[ \frac{(O_{ij} - e_{ij})^2}{e_{ij}} \right] > \chi^2_{(r-1)(c-1)} \text{ at } \alpha$$

**Examples:**

1. A geneticist took a random sample of 300 men to study whether there is association between father and son regarding boldness. He obtained the following results.

		Son	
		Father	Bold
Bold	85	59	
	65	91	

Using  $\alpha = 5\%$  test whether there is association between father and son regarding boldness.

**Solution:**

$H_0$  : There is no association between Father and Son regarding boldness.

$H_1$  : not  $H_0$

- First calculate the row and column totals

$$R_1 = 144, \quad R_2 = 156, \quad C_1 = 150, \quad C_2 = 150$$

- Then calculate the expected frequencies(  $e_{ij}$ 's)

$$e_{ij} = \frac{R_i * C_j}{n}$$

$$\Rightarrow e_{11} = \frac{R_1 * C_1}{n} = \frac{144 * 150}{300} = 72$$

$$e_{12} = \frac{R_1 * C_2}{n} = \frac{144 * 150}{300} = 72$$

$$e_{21} = \frac{R_2 * C_1}{n} = \frac{156 * 150}{300} = 78$$

$$e_{22} = \frac{R_2 * C_2}{n} = \frac{156 * 150}{300} = 78$$

- Obtain the calculated value of the chi-square.

$$\begin{aligned}\chi^2_{cal} &= \sum_{i=1}^2 \sum_{j=1}^2 \left[ \frac{(O_{ij} - e_{ij})^2}{e_{ij}} \right] \\ &= \frac{(85 - 72)^2}{72} + \frac{(59 - 72)^2}{72} + \frac{(65 - 78)^2}{78} + \frac{(91 - 78)^2}{78} = 9.028\end{aligned}$$

- Obtain the tabulated value of chi-square  
 $\alpha = 0.05$

$$Degrees\ of\ freedom = (r-1)(c-1) = 1*1 = 1$$

$$\chi^2_{0.05}(1) = 3.841 \ from\ table.$$

- The decision is to reject  $H_0$  since  $\chi^2_{cal} > \chi^2_{0.05}(1)$

**Conclusion:** At 5% level of significance we have evidence to say there is association between father and son regarding boldness, based on this sample data.

2. Random samples of 200 men, all retired were classified according to education and number of children is as shown below

<i>Education level</i>	<i>Number of children</i>		
	<i>0-1</i>	<i>2-3</i>	<i>Over 3</i>
<i>Elementary</i>	14	37	32
<i>Secondary and above</i>	31	59	27

Test the hypothesis that the size of the family is independent of the level of education attained by fathers. (Use 5% level of significance)

**Solution:**

$H_0$  : There is no association between the size of the family and the level of education attained by fathers.

$H_1$  : not  $H_0$ .

- First calculate the row and column totals

$$R_1 = 83, R_2 = 117, C_1 = 45, C_2 = 96, C_3 = 59$$

- Then calculate the expected frequencies(  $e_{ij}$ 's)

$$e_{ij} = \frac{R_i * C_j}{n}$$

$$\Rightarrow e_{11} = 18.675, e_{12} = 39.84, e_{13} = 24.485$$

$$e_{21} = 26.325, e_{22} = 56.16, e_{23} = 34.515$$

- Obtain the calculated value of the chi-square.

$$\chi^2_{cal} = \sum_{i=1}^2 \sum_{j=1}^3 \left[ \frac{(O_{ij} - e_{ij})^2}{e_{ij}} \right]$$

$$= \frac{(14 - 18.675)^2}{18.675} + \frac{(37 - 39.84)^2}{39.84} + \dots + \frac{(27 - 34.515)^2}{34.515} = 6.3$$

- Obtain the tabulated value of chi-square

$$\alpha = 0.05$$

$$Degrees\ of\ freedom = (r - 1)(c - 1) = 1 * 2 = 2$$

$$\chi^2_{0.05}(2) = 5.99 \ from\ table.$$

- The decision is to reject  $H_0$  since  $\chi^2_{cal} > \chi^2_{0.05}(2)$

**Conclusion:** At 5% level of significance we have evidence to say there is association between the size of the family and the level of education attained by fathers, based on this sample data.

## Inferential Statistics: Regression and Correlation

### Introduction

**Regression** and **correlation analysis** are statistical techniques used extensively in physical geography to examine causal relationships between variables. Regression and correlation measure the degree of **relationship** between two or more variables in two different but related ways. In regression analysis, a single **dependent** variable, Y, is considered to be a

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function of one or more independent variables,  $X_1$ ,  $X_2$ , and so on. The values of both the dependent and independent variables are assumed as being ascertained in an error-free random manner. Further, parametric forms of regression analysis assume that for any given value of the independent variable, values of the dependent variable are normally distributed about some mean. Application of this statistical procedure to dependent and independent variables produces an equation that "best" approximates the functional relationship between the data observations.

Correlation analysis measures the degree of association between two or more variables. Parametric methods of correlation analysis assume that for any pair or set of values taken under a given set of conditions, variation in each of the variables is random and follows a normal distribution pattern. Utilization of correlation analysis on dependent and independent variables produces a statistic called the correlation coefficient (r). The square of this statistical parameter (the coefficient of determination or  $r^2$ ) describes what proportion of the variation in the dependent variable is associated with the **regression** of an independent variable.

Analysis of variance is used to test the significance of the **variation** in the dependent variable that can be attributed to the **regression** of one or more independent variables. Employment of this statistical procedure produces a **calculated F-value** that is compared to a **critical F-values** for a particular level of statistical probability. Obtaining a significant calculated F-value indicates that the results of regression and correlation are indeed true and not the consequence of chance.

### Simple Linear Regression

In a simple regression analysis, one dependent variable is examined in relation to only one independent variable. The analysis is designed to derive an equation for the line that best models the relationship between

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the dependent and independent variables. This equation has the mathematical form:

$$Y = a + bX$$

where, **Y** is the value of the dependent variable, **X** is the value of the independent variable, **a** is the intercept of the regression line on the **Y** axis when **X** = 0, and **b** is the slope of the regression line.

The following **table** contains randomly collected data on growing season precipitation and cucumber yield (**Table 3h-1**). It is reasonable to suggest that the amount of water received on a field during the growing season will influence the yield of cucumbers growing on it. We can use this data to illustrate how regression analysis is carried out. In this table, precipitation is our independent variable and is not affected by variation in cucumber yield. However, cucumber yield is influenced by precipitation, and is therefore designated as the Y variable in the analysis.

**Table 3h-1:** Cucumber yield vs precipitation data for 62 observations.

Precipitation mm (X)	Cucumbers kilograms per m <sup>2</sup> (Y)	Precipitation mm (X)	Cucumbers kilograms per m <sup>2</sup> (Y)
22	.36	103	.74
6	.09	43	.64
93	.67	22	.50
62	.44	75	.39
84	.72	29	.30
14	.24	76	.61
52	.33	20	.29
69	.61	29	.38
104	.66	50	.53
100	.80	59	.58
41	.47	70	.62
85	.60	81	.66
90	.51	93	.69
27	.14	99	.71

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18	.32	14	.14
48	.21	51	.41
37	.54	75	.66
67	.70	6	.18
56	.67	20	.21
31	.42	36	.29
17	.39	50	.56
7	.25	9	.13
2	.06	2	.10
53	.47	21	.18
70	.55	17	.17
6	.07	87	.63
90	.69	97	.66
46	.42	33	.18
36	.39	20	.06
14	.09	96	.58
60	.54	61	.42

$$\square \square X = 3,050$$

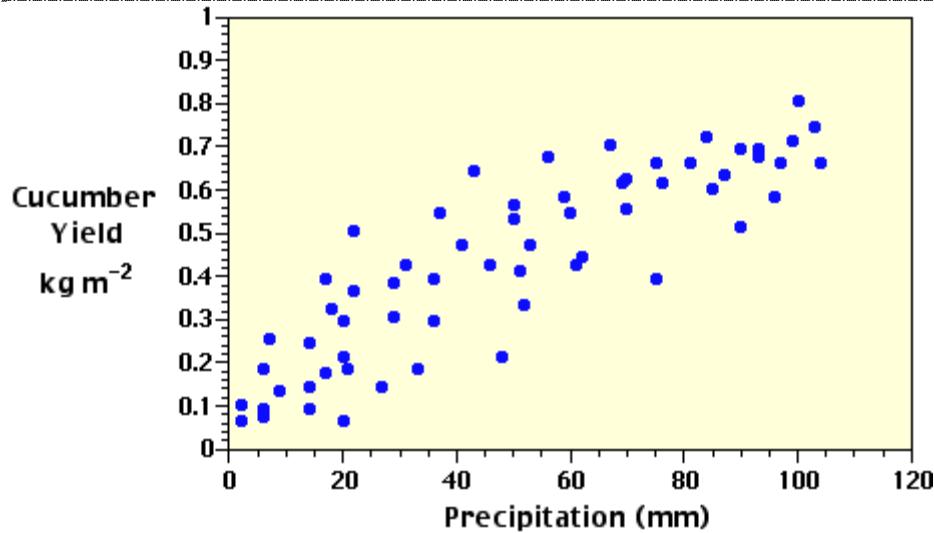
$$\bar{x} = 49.1935$$

$$\square Y = 26.62$$

$$\bar{y} = 0.4294$$

$$n = 62$$

Often the first step in regression analysis is to plot the **X** and **Y** data on a graph (**Figure 3h-1**). This is done to graphically visualize the relationship between the two variables. If there is a simple relationship, the plotted points will have a tendency to form a recognizable pattern (a straight line or curve). If the relationship is strong, the pattern will be very obvious. If the relationship is weak, the points will be more spread out and the pattern less distinct. If the points appear to fall pretty much at random, there may be no relationship between the two variables.



**Figure 3h-1:** Scatter gram plot of the precipitation and cucumber yield data found in **Table 3h-1**. The distribution of the data points indicates a possible positive linear relationship between the two variables.

The type of pattern (straight line, parabolic curve, exponential curve, etc.) will determine the type of regression model to be applied to the data. In this particular case, we will examine data that produces a simple straight-line relationship (see **Figure 3h-1**). After selecting the model to be used, the next step is to calculate the corrected sums of squares and products used in a bivariate linear regression analysis. In the following equations, capital letters indicate uncorrected values of the variables and lower-case letters are used for the corrected parameters in the analysis.

**The corrected sum of squares for Y:**

$$\begin{aligned}
 \square y^2 &= \square Y^2 - \frac{(\Sigma Y)^2}{n} \\
 &= (0.36^2 + 0.09^2 + \dots + 0.42^2) - (26.62^2) / 62 \\
 &= 2.7826
 \end{aligned}$$

The corrected sum of squares for X:

$$\begin{aligned}\square x^2 &= \square X^2 - \frac{(\sum x)^2}{n} \\ &= (22^2 + 6^2 + \dots + 61^2) - (3,050^2) / 62 \\ &= 59,397.6775\end{aligned}$$

The corrected sum of products:

$$\begin{aligned}\square xy &= \square (XY) - \frac{(\sum x)(\sum y)}{n} \\ &= ((22)(.36) + (6)(.09) + \dots + (61)(.42)) - ((26.62)(3,050)) / 62 \\ &= 354.1477\end{aligned}$$

As discussed earlier, the general form of the equation for a straight line is  $Y = a + bX$ . In this equation,  $a$  and  $b$  are constants or regression coefficients that are estimated from the data set. Based on the mathematical procedure of least squares, the best estimates of these coefficients are:

$$\begin{aligned}b &= \frac{(\sum xy)}{(\sum x^2)} \\ &= (354.1477) / (59,397.6775) = 0.0060 \\ a &= Y - bX = 0.42935 - (0.0060)(49.1935) = 0.1361\end{aligned}$$

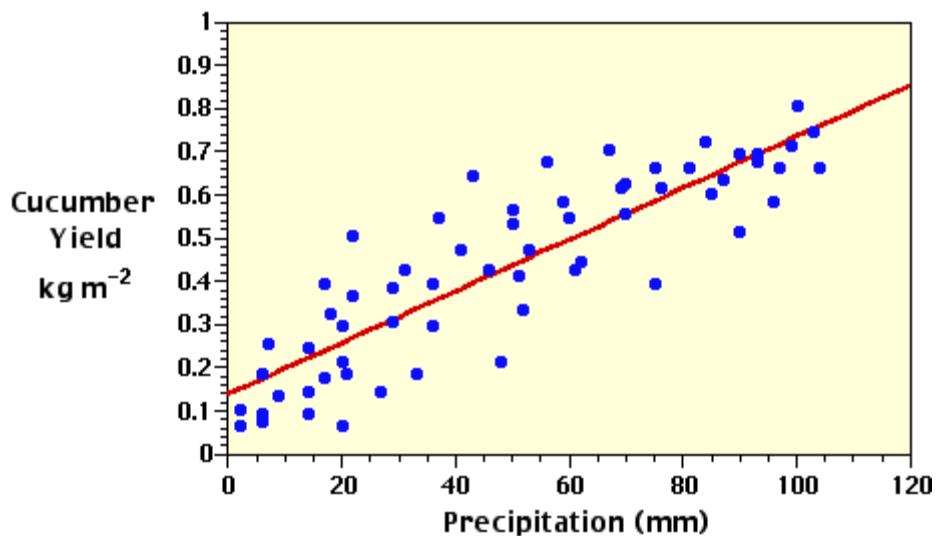
Substituting these estimates into the general linear equation suggests the following relationship between the Y and X variables:

$$\hat{Y} = 0.1361 + 0.0060X$$

where  $\hat{Y}$  indicates that we are using an estimated value of Y.

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With this equation, we can estimate the the number of cucumbers (**Y**) from the measurements of precipitation (**X**) and describe this relationship on our **scattergram** with a best fit straight-line (**Figure 3h-2**). Because **Y** is estimated from a known value of **X**, it is called the dependent variable and **X** the independent variable. In plotting the data in a graph, the values of **Y** are normally plotted along the vertical axis and the values of **X** along the horizontal axis.



**Figure 3h-2:** Scattergram plot of the precipitation and cucumber yield data and the regression model best fit straight-line describing the linear relationship between the two variables.

## Regression Analysis and ANOVA

A regression model can be viewed of as a type of moving average. The regression equation attempts to explain the relationship between the **Y** and **X** variables through linear association. For a particular value of **X**, the regression model provides us with an estimated value of **Y**. Yet, **Figure 3h-2** indicates that many of the plotted values of the actual data are observed to be above the regression line while other values are found below it. These variations are caused either by sampling error or the fact that some other unexplained independent variable influences the individual values of the **Y** variable.

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The **corrected sum of squares** for **Y** (i.e.,  $\sum y^2$ ) determines the total amount of variation that occurs with the individual observations of **Y** about the mean estimate of  $\bar{y}$ . The amount of variation in **Y** that is directly related with the regression on **X** is called the **regression sum of squares**. This value is calculated accordingly:

$$\text{Regression SS} = \frac{(\sum xy)^2}{\sum x^2} = (354.1477)^2 / (59,397.6775) = 2.1115$$

As discussed above, the total variation in **Y** is determined by  $\sum y^2 = 2.7826$ . The amount of the total variation in **Y** that is not associated with the regression is termed the **residual sum of squares**. This statistical parameter is calculated by subtracting the **regression sum of squares** from the **corrected sum of squares** for **Y** ( $\sum y^2$ ):

$$\begin{aligned}\text{Residual SS} &= \sum y^2 - \text{Regression SS} \\ &= 2.7826 - 2.1115 \\ &= 0.6711\end{aligned}$$

The unexplained variation can now be used as a standard for testing the amount of variation attributable to the regression. Its significance can be tested with the **F test** from calculations performed in an *Analysis of Variance* table.

Source of variation	df 1	SS	MS 2
Due to regression	1	2.1115	2.1115
Residual (unexplained)	60	0.6711	0.0112
Total	61	2.7826	-

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1 There were 62 values of Y analyzed and therefore n = 62.

The total sum of squares ***degrees of freedom*** (df) is determined as n-1 or 61. The regression of Y on X has 1 ***degree of freedom***. The residual or unexplained ***degrees of freedom*** is determined by subtracting regression df (1) from total sum of squares df (61).

2 MS is calculated as **SS / df**.

Using the ***Analysis of Variance*** procedure, the regression is tested by determining the calculated ***F statistic***:

$$F = (\text{Regression MS}) / (\text{Residual SS}) = (2.1115) / (0.0112) = 188.86$$

To test this statistic we use a ***table of F*** to determine a ***critical test value*** for a probability of 0.01 or 1 % (this relationship can occur by chance only in 1 out 100 cases) and with **1,60 degrees of freedom**. According to the table the ***critical test value*** is **7.1**. In this test, the relationship is deemed significant if the calculated ***F statistic*** is greater than the ***critical test value***. This regression is statistically significant at the **0.01 level** because **188.86** is greater than **7.1**.

Caution must be taken when interpreting the results of regression. In our example, we found a significant relationship between precipitation and cucumber yield. However, this conclusion may not be the result of a causal relationship between the two variables. A third variable that is directly associated to both precipitation and cucumber yield may be ***confounding*** the interpretation of the analysis. Absolute verification of associations between variables can only be confirmed with experimental manipulation.

## Coefficient of Determination

To measure how strong the correlation is between the two variables, we can determine the amount of the total variation in **Y** that is associated with the regression model. This ratio is sometimes called the coefficient of determination and is represented by the symbol  $r^2$ . The value of the coefficient of determination ranges from 1.00 to 0.00. The calculated *coefficient of determination* from the data set above was 0.76 or 76 % (as calculated below). This value suggests that 76 % of the *variation* in **Y** was associated with the change seen **X** from the data set observations.

$$\text{Coefficient of determination} = (\text{Regression SS}) / (\text{Total SS})$$

$$= (2.1115) / (2.7826) = 0.7588$$

## Correlation Coefficient

Another useful regression statistic that measures the strength of the correlation between two variables is the **coefficient. This** statistic is often represented by the symbol "r" and is determined by taking the square-root of the *coefficient of determination*. The value of the correlation coefficient ranges from 1.00 to -1.00. A value of 0.0 indicates that there is absolutely no relationship between the **X** and **Y** variables. The strength of the relationship between the **X** and **Y** variables increases as the value of r approaches 1.00 and -1.00. Perfect correlation occurs if r equals either 1.00 (perfect positive) or -1.00 (perfect negative). Positive correlation coefficients indicate that an increase in the value of the **X** variable results in an increase in the value of the **Y** variable. Negative correlation coefficients indicate that an increase in the value of the **X** variable results in a decrease in the value of the **Y** variable.

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These techniques are used to investigate relationships between two variables (x and y). Is a change in one of these variables associated with a change in the other? For example, if we increase the temperature do we increase the growth rate of a culture or the rate of a chemical reaction?

If we find evidence of any of these relationships, we can use the technique of **correlation** to test the statistical significance of the association. In some cases we can then use **regression analysis** to describe the relationship precisely by means of an equation that has predictive value. We deal separately with these two types of analysis - correlation and regression - because they have different roles.

### Business Statistics

Today's good decisions are driven by data. In all aspects of our lives, and importantly in the business context, an amazing diversity of data is available for inspection and analytical insight. Business managers and professionals are increasingly **required to justify decisions on the basis of data**. They need statistical model-based decision support systems.

Statistical skills enable them to intelligently collect, analyze and interpret data relevant to their decision-making. Statistical concepts and statistical thinking enable them to:

- solve problems in a diversity of contexts.
- add substance to decisions.
- reduce guesswork

In competitive environment, business managers must design quality into products, and into the processes of making the products. They must facilitate a process of never-ending improvement at all stages of manufacturing and service. This is a strategy that employs statistical

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methods, particularly [statistically designed experiments](#), and produces processes that provide high yield and products that seldom fail. Moreover, it facilitates development of robust products that are insensitive to changes in the environment and internal component variation. Carefully planned statistical studies remove hindrances to high quality and productivity at every stage of production. This saves time and money. It is well recognized that quality must be engineered into products as early as possible in the design process. One must know how to use carefully planned, cost-effective [statistical experiments](#) to improve, optimize and make robust products and processes.

**Business Statistics is a science assisting you to make business decisions under uncertainties** based on some numerical and measurable scales. Decision making processes must be based on data, neither on personal opinion nor on belief.

### What is Business Statistics?

The main objective of Business Statistics is to make [inferences](#) (e.g., prediction, making decisions) about certain characteristics of a [population](#) based on information contained in a random sample from the entire population. The condition for **randomness** is essential to make sure the sample is representative of the population.

Business Statistics is the science of '**good**' decision making in the face of **uncertainty** and is used in many disciplines, such as financial analysis, econometrics, auditing, production and operations, and marketing research. It provides knowledge and skills to interpret and use statistical techniques in a variety of business applications. A typical Business Statistics course is intended for business majors, and covers statistical study, descriptive statistics (collection, description, analysis, and summary of data), probability, and the binomial and normal distributions, test of hypotheses and confidence intervals, linear regression, and correlation.

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Statistics is a science of making decisions with respect to the characteristics of a group of persons or objects on the basis of numerical information obtained from a randomly selected sample of the group. Statisticians refer to this numerical observation as **realization** of a random sample. However, notice that **one cannot see a random sample**. A random sample is only a sample of a finite **outcomes of a random process**.

At the planning stage of a statistical investigation, the question of sample size ( $n$ ) is critical. For example, sample size for sampling from a finite population of size  $N$ , is set at:  $N^{1/2}+1$ , rounded up to the nearest integer. Clearly, a larger sample provides more relevant information, and as a result a more accurate estimation and better statistical judgement regarding test of hypotheses.

Hypothesis testing is a procedure for reaching a probabilistic conclusive decision about a claimed value for a population's parameter based on a sample. To reduce this uncertainty and having high confidence that statistical inferences are correct, a sample must give equal chance to each member of population to be selected which can be achieved by sampling randomly and relatively large sample size  $n$ .

Statistics is a tool that enables us to impose order on the disorganized cacophony of the real world of modern society. The business world has grown both in size and competition. Corporate executive must take risk in business, hence the need for business statistics.

Business statistics has grown with the art of constructing charts and tables! It is a science of basing decisions on numerical data in the face of uncertainty.

Business statistics is a scientific approach to decision making under risk. In practicing business statistics, we search for an insight, not the solution. Our search is for the one solution that meets all the business's needs with

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the lowest level of risk. Business statistics can take a normal business situation, and with the proper data gathering, analysis, and re-search for a solution, turn it into an opportunity.

While business statistics cannot replace the knowledge and experience of the decision maker, it is a valuable tool that the manager can employ to assist in the decision making process in order to reduce the inherent risk, measured by, e.g., the standard deviation  $s$ .

Among other useful questions, you may ask why we are interested in estimating the population's expected value  $m$  and its Standard Deviation  $s$ ? Here are some applicable reasons. Business Statistics must provide justifiable answers to the following concerns for every consumer and producer:

1. What is your or your customer's, **Expectation** of the product/service you sell or that your customer buys? That is, what is a good estimate for  $m$ ?
2. Given the information about your, or your customer's, expectation, what is the **Quality** of the product/service you sell or that you customer buys. That is, what is a good estimate for  $s$ ?
3. Given the information about your or your customer's expectation, and the quality of the product/service you sell or you customer buy, how does the product/service **compare** with other existing similar types? That is, comparing several  $m$ 's, and several  $s$ 's.

### Common Statistical Terminology with Applications

Like all profession, also statisticians have their own keywords and phrases to ease a precise communication. However, one must interpret the results of any decision making in a language that is easy for the decision-maker to understand. Otherwise, he/she does not believe in what you recommend, and therefore does not go into the implementation phase. This lack of communication between statisticians and the managers is the major roadblock for using statistics.

**Population:** A population is any entire collection of people, animals, plants or things on which we may collect data. It is the entire group of interest, which we wish to describe or about which we wish to draw conclusions. In the above figure the life of the light bulbs manufactured say by GE, is the concerned population.

**Qualitative and Quantitative Variables:** Any object or event, which can vary in successive observations either in quantity or quality is called a "variable." Variables are classified accordingly as quantitative or qualitative. A qualitative variable, unlike a quantitative variable does not vary in magnitude in successive observations. The values of quantitative and qualitative variables are called "Variates" and "Attributes", respectively.

**Variable:** A characteristic or phenomenon, which may take different values, such as weight, gender since they are different from individual to individual.

**Randomness:** Randomness means unpredictability. The fascinating fact about inferential statistics is that, although each random observation may not be predictable when taken alone, collectively they follow a predictable pattern called its distribution function. For example, it is a fact that the distribution of a sample average follows a normal distribution for sample size over 30. In other words, an extreme value of the sample mean is less likely than an extreme value of a few raw data.

**Sample:** A subset of a population or universe.

**An Experiment:** An experiment is a process whose outcome is not known in advance with certainty.

**Statistical Experiment:** An experiment in general is an operation in which one chooses the values of some variables and measures the values of other variables, as in physics. A statistical experiment, in contrast is an operation in which one takes a random sample from a population and infers

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the values of some variables. For example, in a survey, we "survey" i.e. "look at" the situation without aiming to change it, such as in a survey of political opinions. A random sample from the relevant population provides information about the voting intentions.

In order to make any generalization about a population, a random sample from the entire population; that is meant to be representative of the population, is often studied. For each population, there are many possible samples. A sample statistic gives information about a corresponding population parameter. For example, the sample mean for a set of data would give information about the overall population mean  $m$ .

It is important that the investigator carefully and completely defines the population before collecting the sample, including a description of the members to be included.

**Design of experiments** is a key tool for increasing the rate of acquiring new knowledge. Knowledge in turn can be used to gain competitive advantage, shorten the product development cycle, and produce new products and processes which will meet and exceed your customer's expectations.

**Primary data and Secondary data sets:** If the data are from a planned experiment relevant to the objective(s) of the statistical investigation, collected by the analyst, it is called a Primary Data set. However, if some condensed records are given to the analyst, it is called a Secondary Data set.

**Random Variable:** A random variable is a real function (yes, it is called "variable", but in reality it is a function) that assigns a numerical value to each simple event. For example, in sampling for quality control an item could be defective or non-defective, therefore, one may assign  $X=1$ , and  $X = 0$  for a defective and non-defective item, respectively. You may assign any other two distinct real numbers, as you wish; however, non-negative integer random variables are easy to work with. Random variables are needed since one cannot do arithmetic operations on words; the random

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variable enables us to compute statistics, such as average and variance.

Any random variable has a distribution of probabilities associated with it.

**Probability:** Probability (i.e., probing for the unknown) is the tool used for anticipating what the distribution of data should look like under a given model. Random phenomena are not haphazard: they display an order that emerges only in the long run and is described by a [distribution](#).

The mathematical description of variation is central to statistics. The probability required for [statistical inference](#) is not primarily axiomatic or combinatorial, but is oriented toward describing data distributions.

**Sampling Unit:** A unit is a person, animal, plant or thing which is actually studied by a researcher; the basic objects upon which the study or experiment is executed. For example, a person; a sample of soil; a pot of seedlings; a zip code area; a doctor's practice.

**Parameter:** A parameter is an unknown value, and therefore it has to be estimated. Parameters are used to represent a certain population characteristic. For example, the population mean  $m$  is a parameter that is often used to indicate the average value of a quantity.

Within a population, a parameter is a fixed value that does not vary. Each sample drawn from the population has its own value of any statistic that is used to estimate this parameter. For example, the mean of the data in a sample is used to give information about the overall mean  $m$  in the population from which that sample was drawn.

**Statistic:** A statistic is a quantity that is calculated from a sample of data. It is used to give information about unknown values in the corresponding population. For example, the average of the data in a sample is used to give information about the overall average in the population from which that sample was drawn.

A statistic is a function of an observable random sample. It is therefore an observable [random variable](#). Notice that, while a statistic is a "function" of observations, unfortunately, it is commonly called a random "variable" not a function.

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It is possible to draw more than one sample from the same population, and the value of a statistic will in general vary from sample to sample. For example, the average value in a sample is a statistic. The average values in more than one sample, drawn from the same population, will not necessarily be equal.

Statistics are often assigned Roman letters (e.g.  $\bar{x}$  and  $s$ ), whereas the equivalent unknown values in the population (parameters) are assigned Greek letters (e.g.,  $\mu$ ,  $s$ ).

The word estimate means to esteem, that is giving a value to something. A statistical estimate is an indication of the value of an unknown quantity based on observed data.

More formally, an estimate is the particular value of an estimator that is obtained from a particular sample of data and used to indicate the value of a parameter.

There are two broad subdivisions of statistics: Descriptive Statistics and Inferential Statistics as described below.

**Descriptive Statistics:** The numerical statistical data should be presented clearly, concisely, and in such a way that the decision maker can quickly obtain the essential characteristics of the data in order to incorporate them into decision process.

The principal descriptive quantity derived from sample data is the mean ( $\bar{x}$ ), which is the arithmetic average of the sample data. It serves as the most reliable single measure of the value of a typical member of the sample. If the sample contains a few values that are so large or so small that they have an exaggerated effect on the value of the mean, the sample is more accurately represented by the median -- the value where half the sample values fall below and half above.

The quantities most commonly used to measure the dispersion of the values about their mean are the variance  $s^2$  and its square root, the standard deviation  $s$ . The variance is calculated by determining the mean, subtracting it from each of the sample values (yielding the deviation of the

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samples), and then averaging the squares of these deviations. The mean and standard deviation of the sample are used as estimates of the corresponding characteristics of the entire group from which the sample was drawn. They **do not**, in general, completely describe the distribution ( $F_x$ ) of values within either the sample or the parent group; indeed, different distributions may have the same mean and standard deviation. They do, however, provide a complete description of the normal distribution, in which positive and negative deviations from the mean are equally common, and small deviations are much more common than large ones. For a normally distributed set of values, a graph showing the dependence of the frequency of the deviations upon their magnitudes is a bell-shaped curve. About 68 percent of the values will differ from the mean by less than the standard deviation, and almost 100 percent will differ by less than three times the standard deviation.

**Inferential Statistics:** Inferential statistics is concerned with making inferences from samples about the populations from which they have been drawn. In other words, if we find a difference between two samples, we would like to know, is this a "real" difference (i.e., is it present in the population) or just a "chance" difference (i.e. it could just be the result of random sampling error). That's what tests of statistical significance are all about. Any inferred conclusion from a sample data to the population from which the sample is drawn must be expressed in a probabilistic term. **Probability is the language and a measuring tool for uncertainty in our statistical conclusions.**

Inferential statistics could be used for explaining a phenomenon or checking for validity of a claim. In these instances, inferential statistics is called **Exploratory Data Analysis** or **Confirmatory Data Analysis**, respectively.

**Statistical Inference:** Statistical inference refers to extending your knowledge obtained from a random sample from the entire population to the whole population. This is known in mathematics as **Inductive Reasoning**, that is, knowledge of the whole from a particular. Its main

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application is in hypotheses testing about a given population. Statistical inference guides the selection of appropriate statistical models. Models and data interact in statistical work. Inference from data can be thought of as the process of selecting a reasonable model, including a statement in probability language of how confident one can be about the selection.

**Normal Distribution Condition:** The normal or Gaussian distribution is a continuous symmetric distribution that follows the familiar bell-shaped curve. One of its nice features is that, the mean and variance uniquely and independently determines the distribution. It has been noted empirically that many measurement variables have distributions that are at least approximately normal. Even when a distribution is non-normal, the distribution of the mean of many independent observations from the same distribution becomes arbitrarily close to a normal distribution, as the number of observations grows large. Many frequently used statistical tests make the condition that the data come from a normal distribution.

**Estimation and Hypothesis Testing:** Inference in statistics are of two types. The first is **estimation**, which involves the determination, with a possible error due to sampling, of the unknown value of a population characteristic, such as the proportion having a specific attribute or the average value  $m$  of some numerical measurement. To express the accuracy of the estimates of population characteristics, one must also compute the **standard errors** of the estimates. The second type of inference is **hypothesis testing**. It involves the definitions of a **hypothesis** as one set of possible population values and **an alternative**, a different set. There are many statistical procedures for determining, on the basis of a sample, whether the true population characteristic belongs to the set of values in the hypothesis or the alternative.

**Statistical inference** is grounded in probability, idealized concepts of the group under study, called the population, and the sample. The statistician may view the population as a set of balls from which the sample is selected at random, that is, in such a way that each ball has the same chance as every other one for inclusion in the sample.

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Notice that to be able to estimate the [population parameters](#), the sample size  $n$  must be greater than one. For example, with a sample size of one, the variation ( $s^2$ ) within the sample is  $0/1 = 0$ . An estimate for the variation ( $s^2$ ) within the population would be  $0/0$ , which is indeterminate quantity, meaning impossible

### Type of Data and Levels of Measurement

Information can be collected in statistics using [qualitative or quantitative](#) data. Qualitative data, such as eye color of a group of individuals, is not computable by arithmetic relations. They are labels that advise in which category or class an individual, object, or process fall. They are called categorical variables.

[Quantitative data](#) sets consist of measures that take numerical values for which descriptions such as means and standard deviations are meaningful. They can be put into an order and further divided into two groups: discrete data or continuous data.

[Discrete data](#) are countable data and are collected by **counting**, for example, the number of defective items produced during a day's production.

[Continuous data](#) are collected by **measuring** and are expressed on a continuous scale. For example, measuring the height of a person.

Among the first activities in statistical analysis is to count or measure: Counting/measurement theory is concerned with the connection between data and reality. A set of data is a representation (i.e., a model) of the reality based on numerical and measurable scales. Data are called "primary type" data if the analyst has been involved in collecting the data relevant to his/her investigation. Otherwise, it is called "secondary type" data.

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Data come in the forms of Nominal, Ordinal, Interval, and Ratio (remember the French word NOIR for the color black). Data can be either continuous or discrete.

### Levels of Measurements

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	Nominal	Ordinal	Interval/Ratio
Ranking?	no	yes	yes
Numerical difference	no	no	yes

Both the zero point and the units of measurement are arbitrary on the Interval scale. While the unit of measurement is arbitrary on the Ratio scale, its zero point is a natural attribute. The categorical variable is measured on an ordinal or nominal scale.

Counting/measurement theory is concerned with the connection between data and reality. Both statistical theory and counting/measurement theory are necessary to make inferences about reality.

### Why Statistical Sampling?

Sampling is the selection of part of an aggregate or totality known as population, on the basis of which a decision concerning the population is made.

The following are the advantages and/or necessities for sampling in statistical decision making:

1. **Cost:** Cost is one of the main arguments in favor of sampling, because often a sample can furnish data of sufficient accuracy and at much lower cost than a census.
2. **Accuracy:** Much better control over data collection errors is possible with sampling than with a census, because a sample is a smaller-scale undertaking.
3. **Timeliness:** Another advantage of a sample over a census is that the sample produces information faster. This is important for timely decision making.

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4. **Amount of Information:** More detailed information can be obtained from a sample survey than from a census, because it takes less time, is less costly, and allows us to take more care in the data processing stage.
5. **Destructive Tests:** When a test involves the destruction of an item under study, sampling must be used. Statistical sampling determination can be used to find the optimal sample size within an acceptable cost.

### **Sampling Methods**

From the food you eat to the television you watch, from political elections to school board actions, much of your life is regulated by the results of sample surveys.

A sample is a group of units selected from a larger group (the population). By studying the sample, one hopes to draw valid conclusions about the larger group.

A sample is generally selected for study because the population is too large to study in its entirety. The sample should be representative of the general population. This is often best achieved by random sampling. Also, before collecting the sample, it is important that one carefully and completely defines the population, including a description of the members to be included.

A common problem in business statistical decision-making arises when we need information about a collection called a population but find that the cost of obtaining the information is prohibitive. For instance, suppose we need to know the average shelf life of current inventory. If the inventory is large, the cost of checking records for each item might be high enough to cancel the benefit of having the information. On the other hand, a hunch about the average shelf life might not be good enough for **decision-making** purposes. This means we must arrive at a compromise that involves selecting a small number of items and calculating an average shelf life as an estimate of the average shelf life of all items in inventory. This is a compromise, since the measurements for a sample from the inventory will produce only an estimate of the value we want, but at

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substantial savings. What we would like to know is how "good" the estimate is and how much more will it cost to make it "better". Information of this type is intimately related to **sampling techniques**. This section provides a short discussion on the common methods of business statistical sampling.

**Cluster sampling** can be used whenever the population is homogeneous but can be partitioned. In many applications the partitioning is a result of physical distance. For instance, in the insurance industry, there are small "clusters" of employees in field offices scattered about the country. In such a case, a random sampling of employee work habits might not require travel to many of the "clusters" or field offices in order to get the data. Totally sampling each one of a small number of clusters chosen at random can eliminate much of the cost associated with the data requirements of management.

**Stratified sampling** can be used whenever the population can be partitioned into smaller sub-populations, each of which is homogeneous according to the particular characteristic of interest. If there are  $k$  sub-populations and we let  $N_i$  denote the size of sub-population  $i$ , let  $N$  denote the overall population size, and let  $n$  denote the sample size, then we select a stratified sample whenever we choose:

$n_i = n(N_i/N)$  items at random from sub-population  $i$ ,  $i = 1, 2, \dots, k$ .

The estimates is:  $\bar{x}_s = \sum W_t \bar{x}_t$ , over  $t = 1, 2, \dots, L$  (strata), and  $\bar{x}_t$  is  $SX_{it}/n_t$ .

Its variance is:  $SW^2_t / (N_t - n_t) S^2_t / [n_t(N_t - 1)]$

Population total  $T$  is estimated by  $N_s \bar{x}_s$ ; its variance is

$SN^2_t (N_t - n_t) S^2_t / [n_t(N_t - 1)]$ .

**Random sampling** is probably the most popular sampling method used in decision making today. Many decisions are made, for instance, by choosing a number out of a hat or a numbered bead from a barrel, and both

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of these methods are attempts to achieve a random choice from a set of items. But true random sampling must be achieved with the aid of a computer or a random number table whose values are generated by computer random number generators.

A random sampling of size  $n$  is drawn from a population size  $N$ . The unbiased estimate for variance of  $\bar{x}$  is:

$\text{Var}(\bar{x}) = S^2(1-n/N)/n$ , where  $n/N$  is the sampling fraction. For sampling fraction less than 10% the finite population correction factor  $(N-n)/(N-1)$  is almost 1.

The total  $T$  is estimated by  $N \cdot \bar{x}$ , its variance is  $N^2\text{Var}(\bar{x})$ .

For 0, 1, (binary) type variables, variation in estimated proportion  $p$  is:

$$S^2 = p(1-p) \cdot (1-n/N)/(n-1).$$

For ratio  $r = Sx_i/Sy_i = \bar{x} / \bar{y}$ , the variation for  $r$  is:

$$[(N-n)(r^2S_x^2 + S_y^2 - 2r \text{Cov}(x, y)]/[n(N-1)\bar{x}^2].$$

Determination of sample sizes ( $n$ ) with regard to binary data: Smallest integer greater than or equal to:

$$[t^2 N p(1-p)] / [t^2 p(1-p) + a^2 (N-1)],$$

with  $N$  being the size of the total number of cases,  $n$  being the sample size,  $a$  the expected error,  $t$  being the value taken from the t-distribution corresponding to a certain confidence interval, and  $p$  being the probability of an event.

**Cross-Sectional Sampling:** Cross-Sectional study the observation of a defined population at a single point in time or time interval. Exposure and outcome are determined simultaneously.

**What is a statistical instrument?** A statistical instrument is any process that aim at describing a phenomena by using any instrument or device, however the results may be used as a control tool. Examples of statistical instruments are questionnaire and surveys sampling.

**What is grab sampling technique?** The grab sampling technique is to take a relatively small sample over a very short period of time, the result

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obtained are usually instantaneous. However, the **Passive Sampling** is a technique where a sampling device is used for an extended time under similar conditions. Depending on the desirable statistical investigation, the passive sampling may be a useful alternative or even more appropriate than grab sampling. However, a passive sampling technique needs to be developed and tested in the field.