

Lecture # 1*

→ Statistics:

Definition, Data, variables, Types of variables, Data Distribution, Frequency Distribution, Exploratory data analysis tools, Measure of Central Tendency & Dispersion, Moments, Skewness and Kurtosis.

→ Probability:

Definition, Random variables, Probability Distribution, Types of Probability Distribution, Commutative Probability Distribution, Laws of Probability, Independent event, Conditional Probability, Baye's Theorem,

Discrete Probability Distribution, Binomial, Poisson distributions,

Continuous Probability distribution, Normal Distribution, Moment generating function,

Chesyshain Inequality, Time Series, Regression Correlation (Simple, Partial & multiple)



→ Books :-

- * Introductory Statistics (3rd ed.)
by Sheldon M. Ross.
- * Introductory Statistics (9th ed.)
by Wadpole, Myers, R.H.

Definitions :-

Statistics :-

is defined as study of collecting information, presenting, analysing and drawing conclusions with the aid of mathematics and computers.

The most essential activity is perhaps the development of methods based on probability theory. This area include estimation of population parameter(s) and test of hypothesis based on probability theory.

Probability :-

A phenomenon is called random if

The exact outcome is uncertain. The mathematical study of such randomness is called Theory of Probability.

A probability model has two essential pieces of its description:

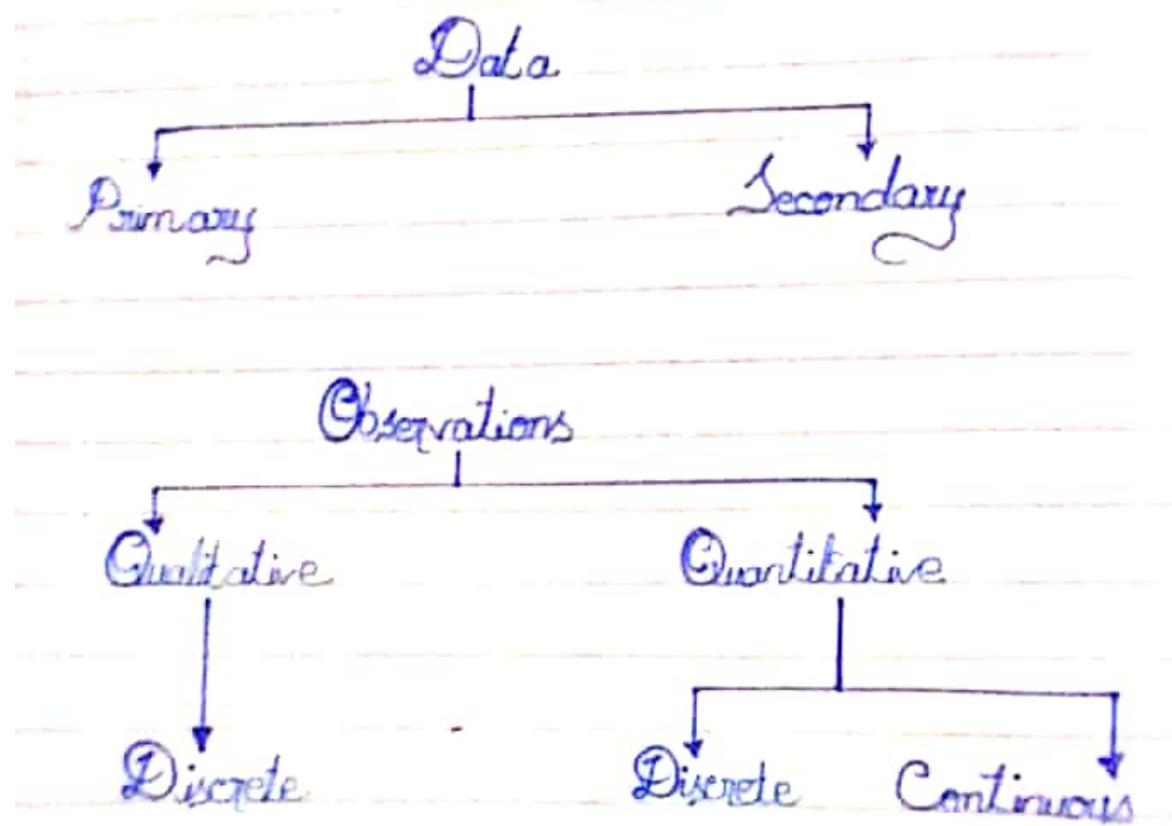
a. the sample space, the set of possible

An event is a collection of outcomes of subsets of the sample space.

Statistics

- P. It probability assigns a numerical value to each event.

Data



Data analysis is:

Frequency Order

Frequency analysis

Frequency distribution

• Frequency Distribution :-

Important characteristics of large mass of data can be readily assessed by grouping data into different classes and determining no. of observations that fall in each of classes. Such an arrangement in tabular form is called frequency distribution.

Some important terms related to frequency distribution are:-

1. No of Classes (m) :-

* Sturge's Rule.

$$\therefore m = 1 + 3 \cdot 3 \log_{10} N \rightarrow \text{arbitrary}$$

* Square root Rule.

$$\therefore m = \sqrt{N}.$$

2. Range :-

: Range = Max. value - Min. value.

3. Class Intervals :-

$$: C.I. = \frac{\text{Range}}{\text{No of Classes}} \Rightarrow \frac{R}{m}.$$

4. The starting point:-

e.g:-

Frequency Distribution for the
Weights of 50 pcs of luggage:

Weight (in Kg.):	No. of Pcs.
7 - 9	2
10 - 12	8
13 - 15	14
16 - 18	19
19 - 21	7

* Class Limits:-

The smallest and largest values that can fall in a given class interval are referred to as its class limits.

e.g. in interval 10 - 12.

10 → smaller no. : Lower class limit.

12 → larger no. : Upper class limit.

* Class Boundary :-

* Class Frequency :-

No. of observations falling in a particular class is called class frequency (f).

* Class width :-

Numerical difference between the upper and lower class boundaries of a class interval is called class width (C).

e.g. $c=3$ in above data.

* Class mark or midpoint :-

The midpoint between upper and lower class boundaries or limits is called class mark or midpoint.

e.g.

Class Interval:	Class Boundries:	Class (x)	Frequency: $f \sim$
7 - 9	6.5 - 9.5	8	2
10 - 12	9.5 - 12.5	11	8
13 - 15	12.5 - 15.5	14	14
16 - 18	15.5 - 18.5	17	19
19 - 21	18.5 - 21.5	20	7

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Lecture # 2 ^e

• Problem :-

Consider the data of lives of 40 similar car batteries.

2.2	4.1	3.5	4.5	3.2	3.7	3.0	2.6
3.4	1.6	3.1	3.3	3.8	3.1	4.7	3.7
2.5	4.3	3.4	3.6	2.9	3.3	3.9	3.1
3.3	3.1	3.7	4.4	3.2	4.1	1.9	3.4
4.7	3.8	3.2	2.6	3.9	3.0	4.2	3.5

Construct frequency distribution and comment on shape of distribution graphically:

Solution :-

• no. of classes:

$$\therefore m = 1 + 3.3 \log_{10} N \quad n = 40 \text{ (Given)}$$

$$m = 6.226$$

$$m \approx 7.$$

• Range:

$$\therefore R = \text{Max. - Min.}$$

$$\text{max} = 4.7$$

$$R = 4.7 - 1.6$$

$$\text{min} = 1.6$$

$$R = 3.1$$

→ Class Interval :

$$C.I = \frac{R}{m}$$

$$C.I = \frac{3.1}{7}$$

$$c/C.I = 0.442 \Rightarrow 0.5 = c$$

→ The starting point :

Lowest C.I is = 1.5.

so, Lowest C.B will be = 1.45 ; Lower C.L = 1.5

$$\therefore c = 0.5 \quad ; \text{Upper C.L} = 1.5 + 0.4.$$

Highest C.B will be = 1.45 + 0.5 = 1.95.

→ midpoints :

$$(\text{Upper C.L} + \text{Lower C.L}) \div 2$$

$$\Rightarrow \frac{1.5 + 1.9}{2} = 1.7$$

* C.F : Cumulative Frequency.

* R.F : Relative Frequencies or percentages.

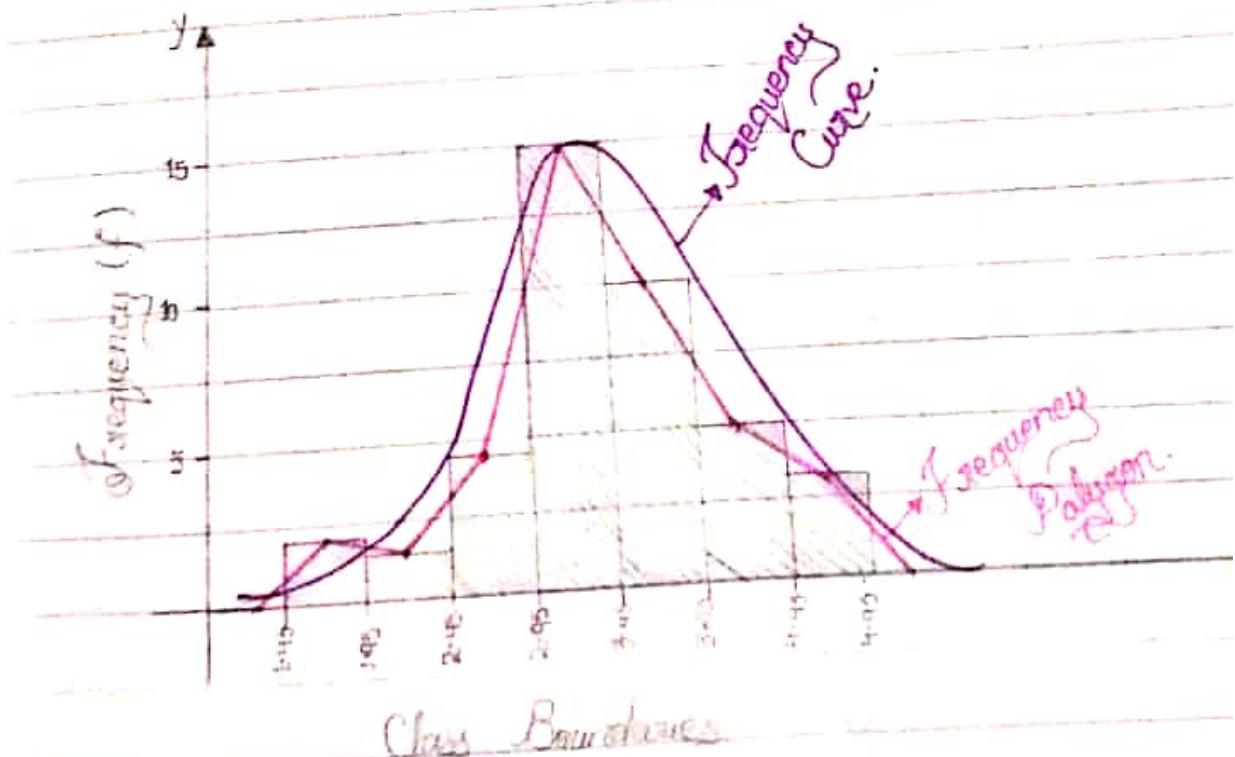
$$\therefore \frac{f}{T.f} \times 100\% \Rightarrow R.F \times 100$$

$$\Rightarrow \frac{2}{40} = 0.05$$

$$\Rightarrow 5\%$$

Class Limit	Class Boundary Points	Mid Marks	Tally	Freq. f.	C.F.	R.F.
1.5 - 1.9	1.45 - 1.95	1.7		2	2	0.05
2.0 - 2.4	1.95 - 2.45	2.2		1	2+1=3	0.075
2.5 - 2.9	2.45 - 2.95	2.7		4	3+4=7	0.175
3.0 - 3.4	2.95 - 3.45	3.2	()	15	7+15=22	0.35
3.5 - 3.9	3.45 - 3.95	3.7		10	22+10=32	0.8
4.0 - 4.4	3.95 - 4.45	4.2		5	32+5=37	0.925
4.5 - 4.9	4.45 - 4.95	4.7		3	37+3=40	1.
				40		

Graph :-



* Assumptions:

- Class frequency concentrated at midpoint.
- Class Interval evenly distributed among whole class.

or Measure of Central Location/Tendency :-

* Average :-

The average is the measure of center of a set of data when data are arranged in an increasing or decreasing order of magnitude. A single value representing whole data.

Any measure indicating the centre of the set of data, arranged in an increasing or decreasing order of magnitude is called a measure of central location/tendency.

and,

The most commonly used measures of central location are :

- mean
- median
- mode

* Types of Means :-

- Arithmetic Mean, Median, Mode } mostly used.
- Geometric mean } for %ages
- Harmonic mean
- Weighted mean
- Trim mean

or Arithmetic Mean (AM) :-

For Data:-

* Ungrouped :-

* Mean :-

$$\bar{x} = \frac{\sum x_i}{n}$$

n → no. of observations

x_i → data values

Σ → summation (sum)

* Grouped :-

$$(\sum_{i=1}^k f_i x_i) \div (\sum_{i=1}^k f_i)$$

x_i → mid-points

f_i → frequency

* Mode :-

most repeated value.

data may be :

uni-modal → 1 mode

bi-modal → 2 modes

tri-modal → 3 modes

no modes

$$x_m + \frac{f_m - f_1}{2f_m - f_1 - f_2} \times h.$$

x_m → median class

f_m → max. frequency

f₁ → frequency above median class.

f₂ → frequency below median class.

h → class interval.

Median :-

if n is:
odd:

$\left(\frac{n+1}{2}\right)^{\text{th}}$ value of
ordered data.

even:

$\left(\frac{n}{2}\right)^{\text{th}} + \left(\frac{n+2}{2}\right)^{\text{th}}$ value of
ordered data.

$$= x_b + \frac{(n/2 - C)}{f} \times h.$$

$x_b \rightarrow$ median class's upper
class limit

$C \rightarrow$ no. of classes

$f \rightarrow$ max frequency

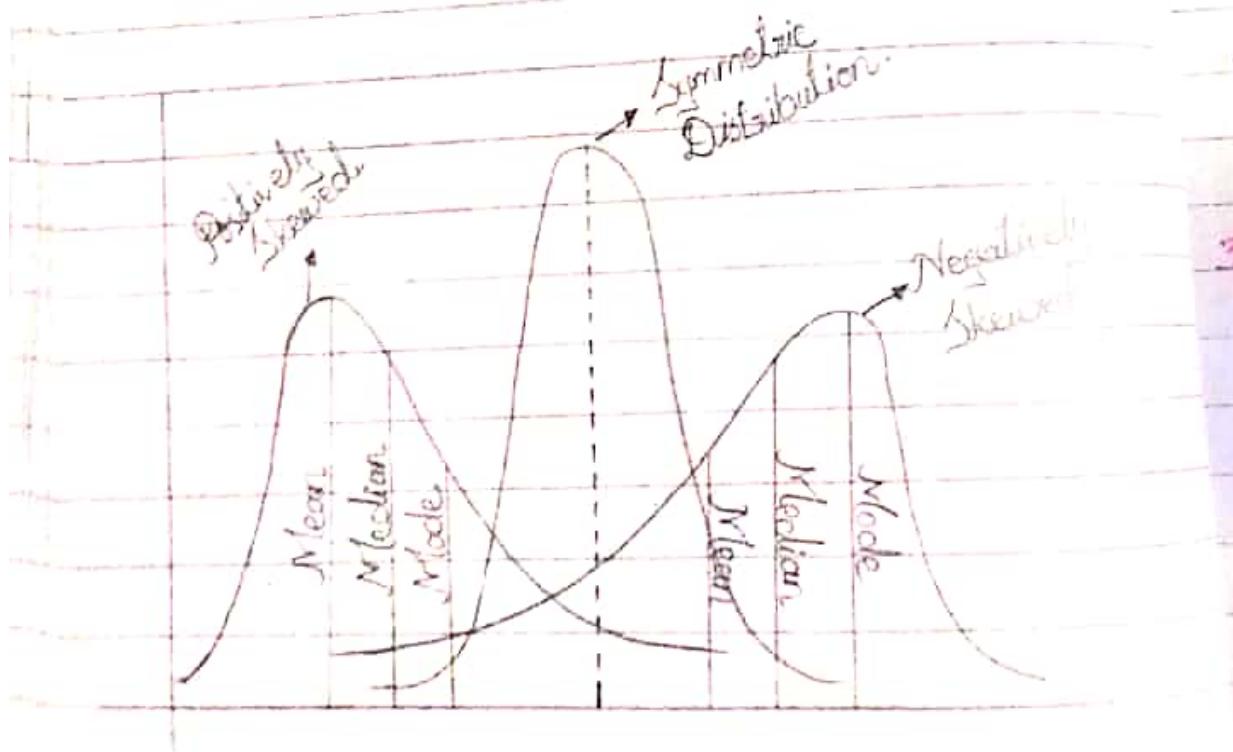
$h \rightarrow$ C.I.

or Frequency Curves nature by
Shape :-

Data Shapes can be of three types :-

e.g.

1. Symmetric Distribution.
2. Positively Skewed.
3. Negatively Skewed.



Frequency Curves

1. Symmetric Distribution :-

In S.D.

$\therefore \text{mean} = \text{mode} = \text{median}$.

eg:
height of children.

2. Negatively Skewed :-

In N.S (Long-tail towards left).

$\therefore \text{mode} > \text{median} > \text{mean}$.

eg:
result if exam is easy.

3. Positively Skewed :-

In P.S (Long-tail towards right).

$\therefore \text{mode} < \text{median} < \text{mean}$.

eg: result if exam is difficult.

* Measure of Dispersion :-

Dispersion is a word that measure the variability in values of units and is define as degree of scatterness or clusteress of observation of given data from its central value.

* Fractiles or Quantiles :-

Measures of location that describe or locate the position of certain non-central pieces of data relative to entire set of data; are referred to as fractiles or quantiles.

1. Percentiles.
2. Deciles.
3. Quartiles.

* Quartiles :-

Quartiles are values that divide data (a set of observations) into four equal parts.

$$\therefore Q_i = \left(\frac{i}{4} \right)^{\text{th}} \text{ value of ordered data}$$

where:

$$i = 1, 2, 3, 4$$

$$\Rightarrow Q_1 = \frac{n}{4}$$

$$\Rightarrow Q_2 = \frac{2n}{4} \Rightarrow \frac{n}{2}.$$

$$\Rightarrow Q_3 = \frac{3n}{4} \Rightarrow \frac{3n}{4}.$$

$$\Rightarrow Q_4 = \frac{4n}{4} \Rightarrow n.$$

* Deciles :-

Deciles are values that divide data (a set of observations) into ten equal parts.

$$\therefore D_i = \left(\frac{in}{10} \right)^{\text{th}} \text{ value of ordered data.}$$

where :

$$i = 1 - 10.$$

* Percentiles :-

Percentiles are values that divide data (a set of observations) into hundred equal parts.

$$\therefore P_i = \left(\frac{in}{100} \right)^{\text{th}} \text{ value of ordered data.}$$

• or Deviations or Variance :-

Types of deviations are:

1. Range Deviation.
2. Quartile Deviation.
3. Mean Deviation.
4. Standard Deviation.

1. Range Deviation :-

$$\therefore R.D = \frac{x_m - x_o}{2}$$

2. Quartile Deviation :-

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

where: $Q_3 = \frac{3n}{4}$ &
 $Q_1 = \frac{n}{4}$.

3. Mean Deviation:-

• For Ungrouped data:-

$$\therefore M.D_c = \frac{1}{n} \sum |x_i - c|$$

where c = mean or median or mode.
[can put any of above].

if $c = \text{mean}(\bar{x})$.

$$\Rightarrow M.D_c = \frac{1}{n} \sum |x_i - \bar{x}|.$$

• for Grouped data:-

$$\therefore M.D_c = \frac{1}{n} \sum_{i=1}^k f_i |x_i - c|.$$

4. Standard Deviation:

positive square root of average of the square deviation taken from mean.

$$\therefore S.D = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2}$$

• Sample Standard Variance/Deviation:

$$\therefore S = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}. \Rightarrow S^2 = \frac{1}{n-1} \left[\sum x_i^2 - \frac{(\sum x_i)^2}{n} \right]$$

• Population Standard Variance/Deviation:

$$\therefore \sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}$$

$$; \mu_c = \frac{\sum_{i=1}^k N_i \bar{x}_i}{\sum_{i=1}^k N_i}$$

$$\Rightarrow \sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2,$$

; $\mu_c \rightarrow$ Combined mean
of population

27-1-20.

L # 3

* Moments: (3)

Moments of a distribution is designated by the power of deviation and it varies toward at end.

Mathematically:

→ μ^r moment about mean -

$$\therefore \mu^r = \frac{1}{n} \sum (x_i - \bar{x})^r \quad \text{or}$$

$$\therefore \mu^r = \frac{1}{n} \sum f_i (x_i - \bar{x})^r \quad (\text{Grouped data})$$

→ when $r=1$,

$$\mu = \frac{1}{n} \sum (x_i - \bar{x})$$

deviation about mean is always least.

i.e. = 0.

$$\therefore \frac{1}{n} \sum x_i - \frac{1}{n} \sum \bar{x} = \sum \bar{x} - n \bar{x}$$

$$\therefore \frac{1}{n} \sum x_i - \frac{1}{n} n \bar{x}$$

=

$$\therefore \sum x_i - n \bar{x} = 0$$

for $\sigma = 2$

$$\mu_2 = M_2' - \mu_1'^2$$

Moments about origin or
Raw Moments :-

"1st moment about origin :-

$$\mu_1' = \frac{1}{n} \sum x_i$$

$$i=1, \Rightarrow \mu_1' = \frac{1}{n} \sum x_i - \bar{x}$$

$$i=2, \Rightarrow \mu_2' = \frac{1}{n} \sum x_i^2$$

$$i=3, \Rightarrow \mu_3' = \frac{1}{n} \sum x_i^3$$

$$i=n \Rightarrow \mu_n' = \frac{1}{n} \sum x_i^{n/4}$$

when $\sigma = 3$,

$$\mu_3 = M_3' - 3\mu_2'\mu_1'^2 + 2\mu_1'^3$$

when $\sigma = 4$,

$$\mu_4 = \mu_4' - 4\mu_3'\mu_1' + 6\mu_2'\mu_1'^2 - 3\mu_1'^4$$

→ For Moment about Constant:-

$$\bar{x} = c$$

→ 7th moment about constant 'c' :-

$$\mu_7 = \frac{1}{n} \sum (x_i - c)^7$$

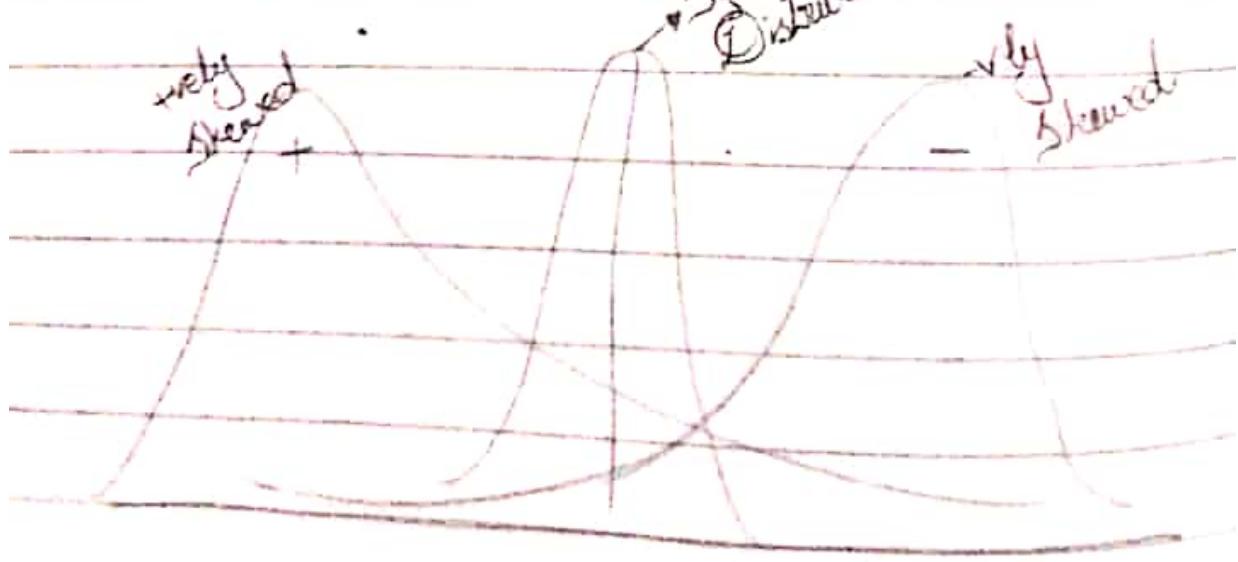
★ Skewness :-

The degree of departure of frequency curve of a data distribution from symmetric distribution is called as Skewness. The statistics used for measuring the skewness is:

$$[\text{is always } \leftarrow] \quad \beta_1 = \frac{\mu_3^2}{\mu_2^3}, \text{ measure (+ve)}$$

symmetric
distribution

+ very skewed
- very skewed



$$\mu_3 = \frac{1}{n} \sum (x_i - \bar{x})^3$$

so if μ_3 is -ve :

then it is -vely skewed.

if : $\beta_2 = +ve$	(+vely skewed)
$\beta_2 = -ve$	(-vely skewed)
$\beta_2 = 0$	(Symmetric)

★ Kurtosis:-

It measures the peakedness & flatness of data distribution.

Mathematically:

$$\beta_2 = \frac{\mu_4}{\mu_2^2}$$

$\rightarrow 3$ (Platykurtic)

$$\rightarrow = 3 \text{ (Mesokurtic)}$$

$$\rightarrow < 3 \text{ (Leptokurtic)}$$

→ β_2

$\beta_2 > 3 \rightarrow$ Platykurtic

$\beta_2 < 3$; Leptokurtic.

$\beta_2 = 3$; Mesokurtic

→ if :-

• $\beta_1 = 0$ & $\beta_2 = 3$ so;

Normal or Gaussian Distribution

Coefficient of Variation (CV) :-

$$CV = \frac{\sigma}{\bar{x}} \times 100$$

$$CV = \frac{s}{\bar{x}} \times 100$$

* CV < 10 are less large.

* Exploring Data Analysis Techniques -

Ansley (1977)

1- Stem-and-Leaf Plot.

2- Box-and-Whiskers Plot.

Stem & Leaf Plot Method:-

Eg.: Data (40 Observations)

2.2 4.1 3.0 4.5 3.2 3.7 3.0 2.6

2.4 1.6 3.1 3.3 3.8 3.1 7.7 3.7

2.5 4.3 3.4 3.6 2.9 3.3 3.9 3.1

3.3 3.1 3.7 4.4 3.2 4.1 3.9 3.4

4.7 3.8 3.2 2.6 3.9 3.0 4.2 3.5

2-digit data

so, left digit - stem

right digit - leaf.

Calculations:-

<u>Item</u>	<u>Leaf - [Unordered plot]</u>
1	6 9
2	2 4 5 6 9 6
3	3 1 8 5 1 4 7 2 5 6 2 8 2 9 7 1 3 0
4	7 1 3 5 4 1 7 2

<u>Item</u>	<u>Leaf - [Ordered]</u>
1	6 9
2	2 4 5 6 6 9
Median Class 3	0 0 1 1 1 1 2 2 3 3 3 4 4 5 5 6 7
4	1 1 2 3 4 5 7 7

$\frac{40}{2} = 20 \text{ Q2}$

gives idea of Histogram i.e.

Symmetric

Box & Whiskers Plot Method:-

2 Method:-

$$\text{Quartiles: } n = \frac{40}{4} = 10$$

$$Q_1 = \frac{n}{4} = 3.0$$

$$Q_3 = \frac{3n}{4} = 30^m = 3.9$$

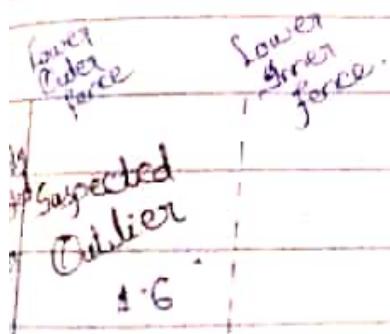


$$Q_2 = \frac{2n}{4} = 20^{\text{th}} = 3.3$$

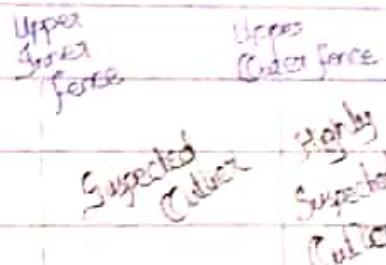
IQR (Inter Quartile Range) = $Q_3 - Q_1$ (IQR)

$$1, 2, 3, 4, 5 \Rightarrow \text{IQR} = Q_3 - Q_1$$

1 2 3 4 5
Q₁



Whisker



IQR	$Q_1 + 1.5(\text{IQR})$	$Q_3 - 1.5(\text{IQR})$	$Q_1 + 3.0(\text{IQR})$	$Q_3 - 3.0(\text{IQR})$
20.9	2.0 + 1.5(0.9)	3.0 - 1.5(0.9)	2.0 + 3.0(0.9)	3.0 - 3.0(0.9)

7.6 + 9.1 + 6.5

+ 5.25

= 6.65

* very skewed

$$Q_3 - Q_1 = \text{IQR}$$

Lec 4 :-

3 - Feb - 20

Probability :-

or Definition :-

- is defined as the study of randomness.
- depends on chances.
- is measure of belief

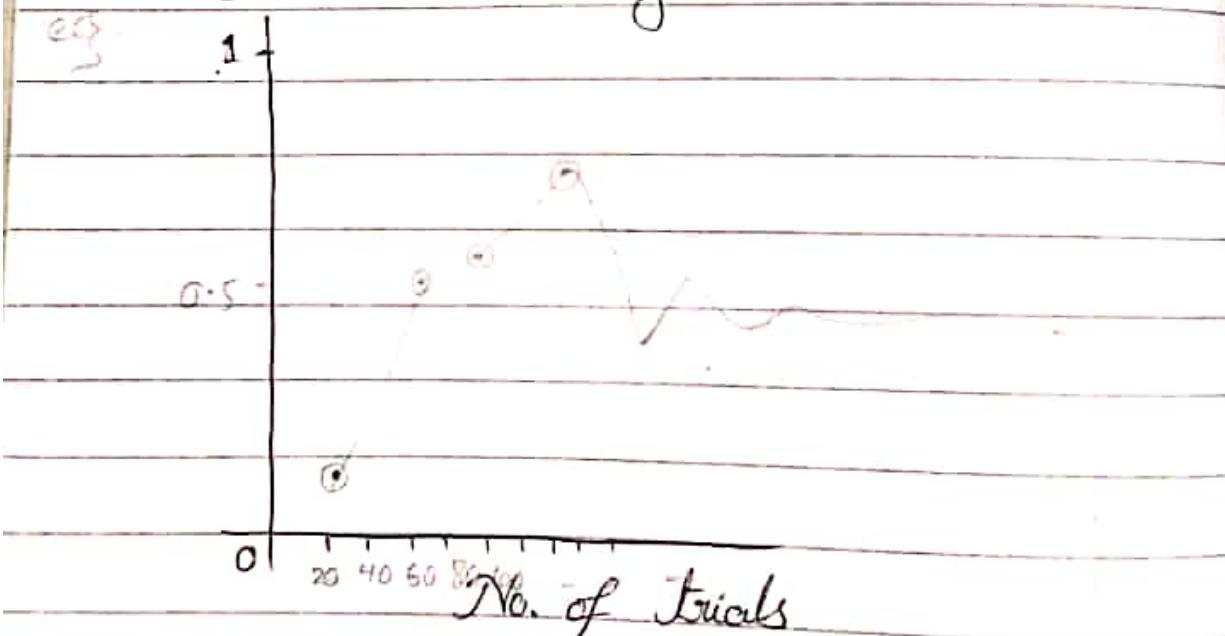
Cumulative

Probability

$$P_7(E) = \frac{\text{Cum no. of Heads}}{\text{Total no. of Trials}} \approx 0.5$$

Graph :-

Cumulative Probability .



* Probability lies b/w 0 & 1

i.e we have $\frac{1}{2}$ chance.

for Probability,
we should know:-

Sets:

a well defined collection of objects or
items under consideration.

Subset: $\rightarrow E \subset U$ \rightarrow Universal set.

a part of Universal (larger) set
 $E \subset U$.

Superset:

$U \supset E$.

larger set in which all subsets are
defined (lies).

Event:

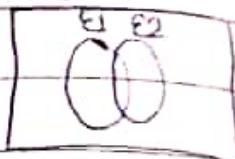
A sample point or a group of
sample point of some particular interests is
called and event.

Mutually Exclusive Events:

Two events are said to be
mutually exclusive if the occurrence
of one precludes the occurrence
of other.

In case of overlapping regions, occurrence of two events
if the simultaneous occurrence is not possible

* Non-Mutually Exclusive Events :-



* Independent Events :-

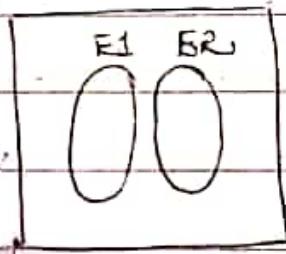
means dependent on each other.

e.g.:-

3 e.g.:

Venn

Diagram:



→ Disjoint events (e.g.: heads or tails (only one can occur at a time))

* Venn Diagrams :-

is a diagram related to set theory in mathematics by which the occurrence of events can be visualised or portrayed.

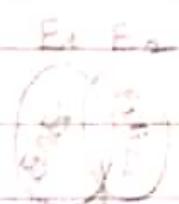
→ Theorem :-

If E_1 & E_2 be the two events define on the sample space S , Then the probability that E_1 occurs or E_2 occurs or atleast one of them occur is given by:

$$P(E_1 \cup E_2) = P(E_1) + P(E_2) - P(E_1 \cap E_2)$$

Proof :-

$$P(E_1 \cup E_2) = P(E_1 \cup (E_1' \cap E_2))$$

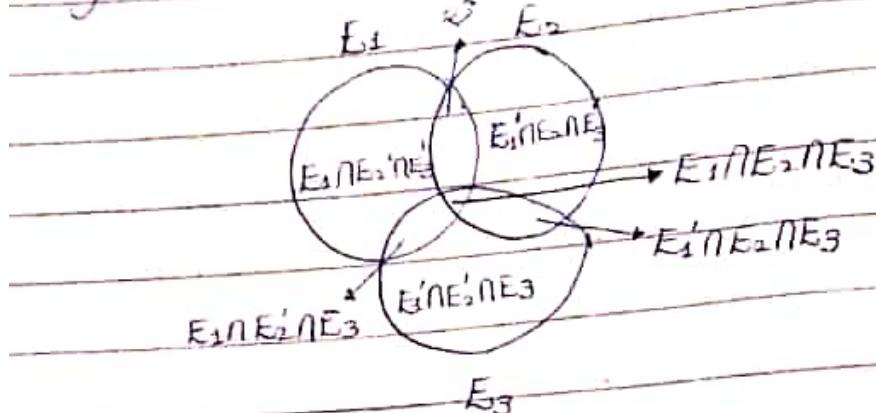


E_1 & E_2 are mutually exclusive.

$$\begin{aligned} &= P(E_1) + P(E_1' \cap E_2) \\ &= P(E_1) + P(E_1' \cap E_2) + P(E_1 \cap E_2) - P(E_1 \cap E_2) \\ &= P(E_1) + P(E_2) - P(E_1 \cap E_2). \end{aligned}$$

Proved.]

• For 3 Events :-



Q#2

So; same theorem for 3 events:-

If E₁, E₂ and E₃ be three events define on the sample space S, then the probability that E₁ occur or E₂ occur or E₃ occur or at least one of them occur is given by :-

$$\begin{aligned} P(E_1 \cup E_2 \cup E_3) &= P(E_1) + P(E_2) + P(E_3) - P(E_1 \cap E_2) - \\ &\quad P(E_1 \cap E_3) - P(E_2 \cap E_3) + P(E_1 \cap E_2 \cap E_3). \end{aligned}$$

$$\Rightarrow P(E_1 \cup E_2) = P[E_1 \cup (E_1' \cap E_2)]$$

$$\begin{aligned} &= P(E_1) + P(E_1' \cap E_2) + P(E_1 \cap E_2) - P(E_1 \cap E_2) \\ &\quad P(E_2) \end{aligned}$$

$$\begin{aligned} &= P(E_1) + P(E_2) - P(E_1 \cap E_2) \quad (2) \end{aligned}$$

Let

$$E_1 = E_1 \cup E_2 \cup E_3. \quad \text{From eqn (2)}$$
$$\Rightarrow P(L) = P(E_1 \cup E_2) = P(E_1) + P(E_2) - P(E_1 \cap E_2)$$

So;

$$\Rightarrow P(E_1 \cup E_2) = P(E_1) + P(E_2) - P(E_1 \cap E_2) \quad (3)$$

Substituting value from eqn (2) in (3).

$$P(E_1 \cup E_2) = P(E_1) + P(E_2) + P(E_3) - P(E_1 \cap E_2) - P(E_1 \cap E_3) \quad (4)$$

For $E_1 \cap E_3$:-

$$E_1 \cap E_3 = (E_1 \cup E_2) \cap E_3.$$

$$= E_1 \cap E_2 \cup E_2 \cap E_3 \quad [\text{By Distributive Law}]$$

So:

$$\begin{aligned} P(E_1 \cap E_3) &= P(E_1 \cap E_2 \cup E_2 \cap E_3). \\ &= P(E_1 \cap E_2) + P(E_2 \cap E_3) - P(E_1 \cap E_2 \cap E_3) \end{aligned}$$

Substitute (5) in (4) we have :-

$$\Rightarrow P(E_1 \cup E_2 \cup E_3) = P(E_1) + P(E_2) + P(E_3) - P(E_1 \cap E_2)$$
$$- [P(E_1 \cap E_3) + P(E_2 \cap E_3) - P(E_1 \cap E_2 \cap E_3)]$$

$$\Rightarrow P(E_1 \cup E_2 \cup E_3) = P(E_1) + P(E_2) + P(E_3) - P(E_1 \cap E_2)$$
$$- P(E_1 \cap E_3) - P(E_2 \cap E_3) + P(E_1 \cap E_2 \cap E_3) \quad \text{Proved},$$



* if N events:

Q3

If $E_1, E_2, E_3, \dots, E_N$ be N events

then -

$$S_1 = \sum_{i=1}^N P(E_i)$$

$$S_2 = \sum_{i < j} P(E_i \cap E_j)$$

$$S_3 = \sum_{i < j < k} P(E_i \cap E_j \cap E_k)$$

$$\vdots$$

$$S_N = \sum_{i < j < \dots < N} P(E_1 \cap E_2 \cap \dots \cap E_N)$$

so,

$$P(E_1 \cup E_2 \cup \dots \cup E_N)$$

$$= S_1 - S_2 + S_3 - \dots + (-1)^{N+1} S_N.$$

• Independent Events :-

Two events are said to be independent if the occurrence of one doesn't effect on the occurrence of other.

i.e.

$$P(E_1 \cap E_2) = P(E_1) \cdot P(E_2).$$

if the intersection of both events is the product of events.

• Theorem :- OR E_1 and E_2 , OR, $E_1' \& E_2'$

If E_1 & E_2 are independent then show that E_1' and E_2' are also independent.

Proof :-

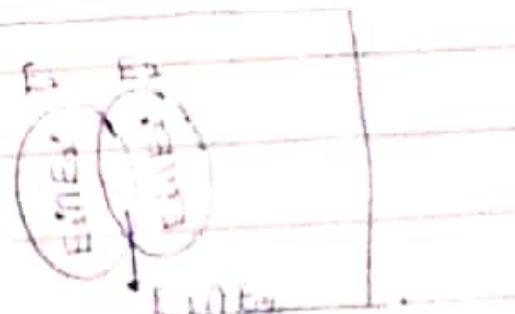
Given:-

$$P(E_1 \cap E_2) = P(E_1) \cdot P(E_2). \quad 1$$

To prove that :-

$$P(E_1 \cap E_2') = P(E_1) \cdot P(E_2'). \quad 2$$

$$\begin{aligned} P(E_1 \cap E_2') &= P(E_2') \cdot P(E_1 \cap E_2) \\ &= P(E_1) \cdot P(E_1) \cdot P(E_2) \\ &= P(E_1) [1 - P(E_2)] \\ &= P(E_1) \cdot P(E_2'). \end{aligned}$$



$$P(E') \cdot [1 - P(E)].$$

Proved !

Ques:- If E_1 & E_2 are independent for given
that E_1 & E_2 are also independent.

Proof :-

Given:-

$$P(E_1 \cap E_2) = P(E_1)P(E_2) \quad 1$$

To prove:-

$$P(E_1' \cap E_2') = P(E_1')P(E_2') \quad 2$$

By DeMorgan's Law:-

$$\begin{aligned} \Rightarrow P(E_1' \cap E_2') &= 1 - P(E_1 \cup E_2) \\ &= 1 - [P(E_1) + P(E_2) - P(E_1 \cap E_2)] \\ &= 1 - P(E_1) - P(E_2) + P(E_1 \cap E_2) \\ &= 1 - P(E_1) - P(E_2) + P(E_1)P(E_2) \\ &= [1 - P(E_1)][1 - P(E_2)] \\ &= P(E_1')P(E_2') \end{aligned}$$

Proved!

* Problem :-

Comp. Sci. department has two O.S.s working independently. The probability that a specific system is available when needed is 0.9.

i) What is Probability that neither is available when needed?

ii) a system is available when needed?

i) $P(O_1) = 0.9$; $P(O_2) = 0.9$.

Soln:-

$$P(O_1') = 1 - P(O_1)$$

$$\begin{aligned} P(O_1' \cap O_2') &= P(O_1') \cdot P(O_2') \\ &= (0.1)(0.1) \end{aligned}$$

$$\Rightarrow P(O_1' \cap O_2') = 0.01 \quad \text{[Ans]}.$$

b) b/c O_1, O_2 are independent:

$$P(O_1 \cup O_2) = P(O_1) + P(O_2) - P(O_1 \cap O_2)$$

$$= (0.9) + (0.9) - P(O_1) \cdot P(O_2)$$

$$= (1.8) - (0.9)(0.9)$$

$$= 1.8 - 0.81$$

$$= 0.99 \quad \text{[Ans]}$$

L # 5 :-

10-Jet-20

Probability :-

* Conditional Probability :-

The probability of any event depending on the outcome of another. The sample space of experiment might often be reduced when some additional information pertaining to the outcome of experiment is received. Probability associated with such a reduced sample space is called Conditional Probability.

Mathematically,

If E_1 and E_2 are define on the sample space S & the occurrence of the event E_2 depend on the occurrence of event E_1 . Then,

$$P(E_2|E_1) = \frac{P(E_1 E_2)}{P(E_1)} = \frac{P(E_1 \cap E_2)}{P(E_1)},$$

$\left\{ \begin{array}{l} \text{read as} \\ \text{probability} \\ E_2 \text{ given } E_1 \end{array} \right\}$ provided $P(E_1) > 0$.

Problem :-

a die shows 3 given that the die shows an odd number?

Solution:-

If a die is thrown

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

$$n(S) = 6 \text{ case.}$$

3 cases are odd.

E_1 : shows 3 , $n(E_1) = 1$.

} just 1 case to 3
occur 3 out of 3
at hi abhi may to
ak hi Jisme aur
skla may.

E_2 : show odd no.

$$n(E_2) = 3.$$

$$\therefore P(E_1/E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)}$$

$$\therefore = \frac{n(E_1 \cap E_2)/n(S)}{n(E_2)/n(S)}$$

$$\therefore = \frac{n(E_1 \cap E_2)}{n(E_2)}.$$

$$\rightarrow n(E_1 \cap E_2) = 1 \quad \left\{ \begin{array}{l} \text{cuz 1 common case} \\ \text{in both.} \end{array} \right\}$$

$$\rightarrow n(E_2) = 3$$

$$\Rightarrow P(E_1/E_2) = \frac{1}{3}$$

ANSWER.

Baye's Theorem:- [Generalization of C.P.J.]

If E_1, E_2, \dots, E_n constitute a partition of sample space S , where $P(E_i) \neq 0$ for $i = 1, 2, 3, \dots, n$ then for any event E define in the sample space S such that $P(E) > 0$. The probability that an event E_n occurs given that event E has occurred is given by:

$$\begin{aligned} P(E_n | E) &= \frac{P(E_n) \cdot P(E | E_n)}{P(E_1)P(E | E_1) + P(E_2)P(E | E_2) + \dots + P(E_n)P(E | E_n)} \\ &= \frac{P(E_n) \cdot P(E | E_n)}{\sum_{i=1}^n P(E_i) \cdot P(E | E_i)} \quad (1) \end{aligned}$$

Proof :-

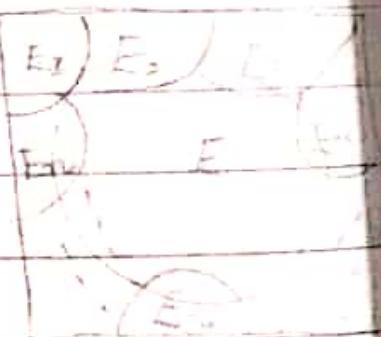
$$E = E \cap S$$

$$E = E \cap (E_1 \cup E_2 \cup \dots \cup E_n)$$

$$= (E \cap E_1) \cup (E \cap E_2) \cup \dots \cup (E \cap E_n) \quad (\text{By Distributive Law})$$

as events are mutually exclusive, so:

$$P(E) = P(E \cap E_1) + P(E \cap E_2) + \dots + P(E \cap E_n) \quad (2)$$



we know that conditional probability is defined as:-

$$P(E/E_1) = \frac{P(E \cap E_1)}{P(E_1)}$$

$$\Rightarrow P(E \cap E_1) = P(E_1) \cdot P(E/E_1). \quad \textcircled{3}$$

so similarly:-

$$P(E) = P(E \cap E_2) = P(E_2) \cdot P(E/E_2)$$

$$P(E \cap E_n) = P(E_n) \cdot P(E/E_n).$$

Substituting eq'n ③ in ② we have:-

$$P(E) = P(E_1) \cdot P(E/E_1) + P(E_2) \cdot P(E/E_2) + \dots + P(E_n) \cdot P(E/E_n)$$

which is called law of total probability.

$$P(E) = \sum_{i=1}^n P(E_i) \cdot P(E/E_i) \quad \textcircled{4}$$

Given that E_k be any other event then:-

$$P(E_k/E) = \frac{P(E_k \cap E)}{P(E)} = \frac{P(E_k) \cdot P(E/E_k)}{P(E_1) \cdot P(E/E_1) + P(E_2) \cdot P(E/E_2) + \dots + P(E_n) \cdot P(E/E_n)}$$

$$= \frac{P(E_k) \cdot P(E/E_k)}{\sum_{i=1}^n P(E_i) \cdot P(E/E_i)} \quad \text{Answer}$$

Question :-

Let 10% of male & 2% of female are color blind and each form 50% of total population.

A research working worker studying colorblindness selects a person at random. What is the probability that the so selected is :

a) Male

b) Female

Solution :-

Sex	Color-blind	Not Color-blind	Total
Male	5 ^(10% of 50)	45	50
Female	1 ^(2% of 50)	49	50
Total	6	94	100

$$a) P(M/C) = \frac{P(M \cap C)}{P(C)} = \frac{5/100}{6/100} = \boxed{\frac{5}{6}}$$

$$b) P(F/C) = \frac{P(F \cap C)}{P(C)} = \frac{1/100}{6/100} = \boxed{\frac{1}{6}}$$

AnsweR

17 Feb 2020.

Lecture # 6

* Probability Density Function (PDF).

Everything that can be known about a random variable is contained in its pdf.

Pdf is a rule that allows us to determine the probability that a particular measurement of a random variable might be within some set of values.

* Random Variable:-

If random variable is the concept of making some measurements.

- The value of measurement need not be the same everytime, so we speak of probability distribution of random variable.
- we know everything there is to know about random variable or to we know the probability that the value of random variable might be in any specified set.

e.g.

sex, age, height,

Random Variable:-

A random variable is a function on the sample space. It is basically a device for transferring probabilities from complicated sample spaces to simple sample spaces, where the elements are just natural numbers.

- A random variable x is a function whose domain is the sample space and whose range is the set of real numbers. The random number assign a real value (i.e a number) to every outcomes in the sample space.
- The particular values are called realizations and are denoted by ' x_i '
- If the realizations are countable, i.e., x_1, x_2, \dots, x_n the random variable is said to be discrete. [means in whole no.]
- In contrast, if there are infinite, many, uncountable, the random variable is said to be continuous. [means not in whole no., in decimal.]

• Probability Mass Function: (discrete case).
If x is discrete random variable, the function given by $f(x) = p(x=x)$ for every x within range of x , is called Probability Mass Function or Probability Density Function of x .

A function can serve as a probability distribution of a discrete random variable x if its values $f(x)$ satisfy the conditions as.

1) $f(x) \geq 0$; for all x within range of x .

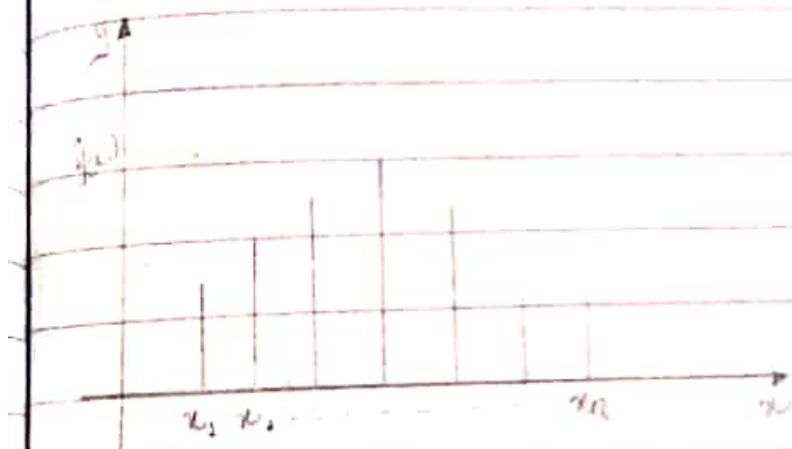
2) $\sum_x f(x) = 1$; where the summation extends over the values within its domain.

Thus a table, formula or graph representing the probabilities which a random variable x takes is called probability distribution.

Table:-

x	x_1	x_2	...	x_n	Total.
$f(x)$	$f(x_1)$	$f(x_2)$...	$f(x_n)$	$\sum f(x_i) = 1$
$P(x)$	$P(x_1)$	$P(x_2)$...	$P(x_n)$	$\sum P(x_i) = 1$

Graph:-



Eg:-

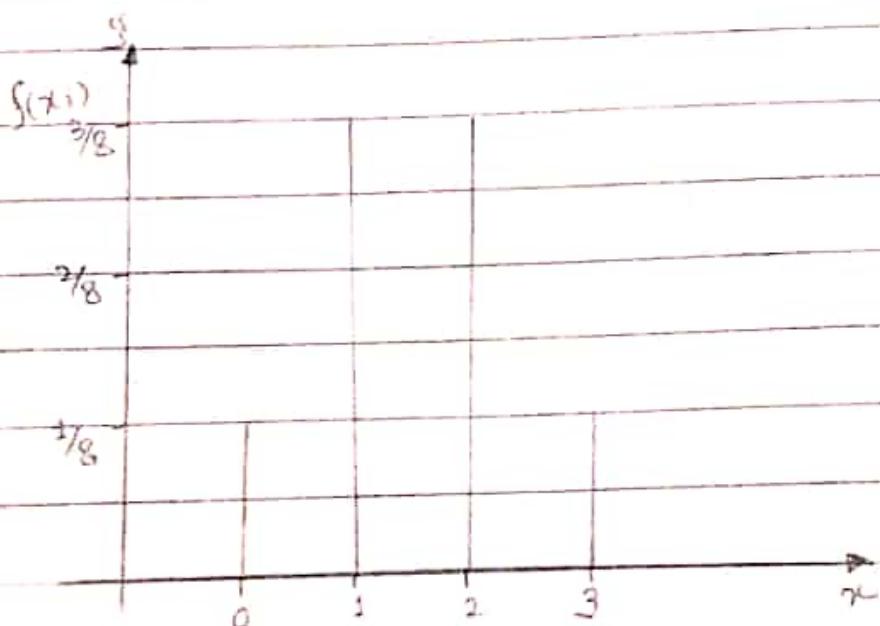
Distribution of boys in a family

having three children. Soln:-

	Children			$S = n(S) = 2^3 = 8$ cases
Cases:	1	2	3	Probabilities for 2 child (each)
1	G	G	G	0
2	G	G	B	1
3	G	B	G	1

$$\Rightarrow \binom{3}{x} = \frac{3!}{(3-x)!x!} \Rightarrow \binom{3}{2} = \frac{3!}{(3-2)!2!} \Rightarrow \binom{3}{2} = \frac{3+2+1}{12+12+3} = \frac{1}{2}$$

					C# of boys.	$P(x)$	P_{ex}	$\binom{3}{x}$
4	G	B	B	2	0	$\frac{1}{8}$		
5	B	G	G	1	1	$\frac{3}{8}$		
6	B	G	B	1	2	$\frac{3}{8}$		
7	B	B	G	2	3	$\frac{1}{8}$		
8	B	B	B	3	Total.	1		
					8 cases.			



Probability Histogram.

* Probability Distribution:

If random variable 'x' is said to be continuous if there exists a non-negative function f , defined for all $x \in (-\infty, \infty)$ having the property that for any set

here
0, 1, 2, 3

set of real numbers

$$P\{x \in A\} = \int_{\pi} f(x) dx.$$

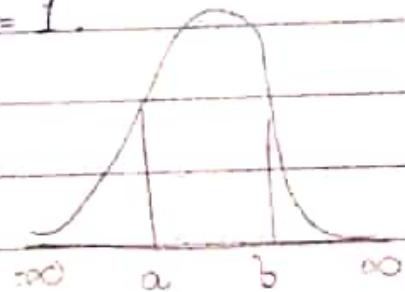
The function f is called the probability density function of random variable x .

Since x must assign some value, f must satisfy:

$$P\{x \in (-\infty, \infty)\} = \int_{-\infty}^{\infty} f(x) dx = 1.$$

if $A = [a, b]$ then,

$$P\{x \in (a, b)\} = \int_a^b f(x) dx.$$



$$\Rightarrow P\{a \leq x \leq b\} =$$

$$\int_a^b f(x) dx = F(b) - F(a)$$

$$\Rightarrow P(x < a) = P(x \leq a)$$

$$= F(a) = \int_{-\infty}^a f(x) dx.$$

* $F(a)$ is called
cumulative density function
or distribution fn.

* Cumulative Distribution Function (cdf) :-

The cdf $F(x)$ of a discrete random function variable x with pmf $P(x)$ or $f(x)$ is defined for every number x by.

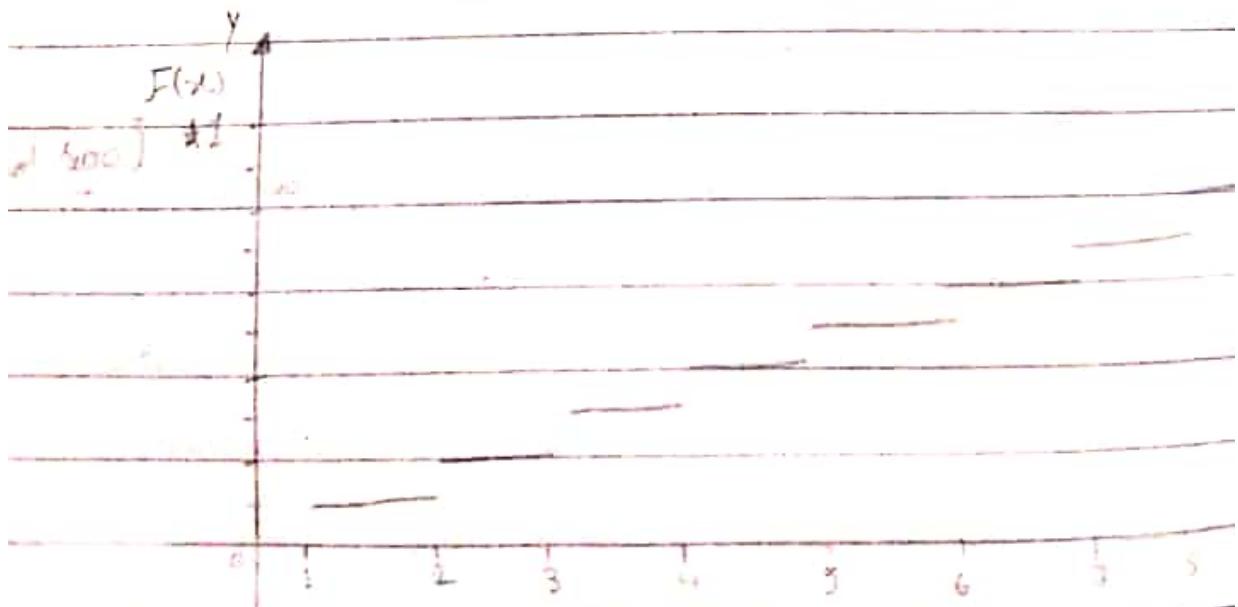
$$F(x) = P(x \leq x) = \sum_{x=0}^{\infty} f(x) = \sum_{x=0}^{\infty} P(x).$$

For any no. x , $F(x)$ is the probability that the observed value of x will be at most x .

| For a Continuous random variable:

$$F(a) = P(x < a) = P(x \leq a) = \int_{-\infty}^a f(x) dx$$

For x is a discrete random variable, The graph will have a jump at every possible value of x and will be flat b/w any two possible values of x such a graph is called Step Function.



Q # 1.

Let x be the random variable with pdf:

$$f(x) = \frac{1}{k} \left(x - \frac{1}{2} \right); \quad k < x < 2.$$

Find the value of k so that $f(x)$ be a pdf?

Solution :-

For a pdf:-

$$\therefore \int f(x) dx = 1.$$

$$\Rightarrow \int_k^2 \left(x - \frac{1}{2} \right) dx = k. \quad \text{or } \frac{1}{k} \int_k^2 \left(x - \frac{1}{2} \right) dx = 1$$

$$\Rightarrow \frac{1}{k} \left[\frac{x^2}{2} - \frac{1}{2}x \right]_k^2 = 1 \quad \Rightarrow \left\{ \frac{2^2 - (2)}{2} - \frac{k^2 - k}{2} \right\} = 1$$

$$\Rightarrow \frac{4 - k^2 + k}{2} = 1$$

$$\Rightarrow k^2 - k - 2 = 0$$

$$\Rightarrow k^2 + 2k - k - 2 = 0$$

$$\Rightarrow k(k+2) - 1(k+2) = 0$$

$$\Rightarrow (k+2)(k-1) = 0$$

$$\therefore k = 1, k = -2$$

Next Class :-

Mathematical expectation (Random variable)