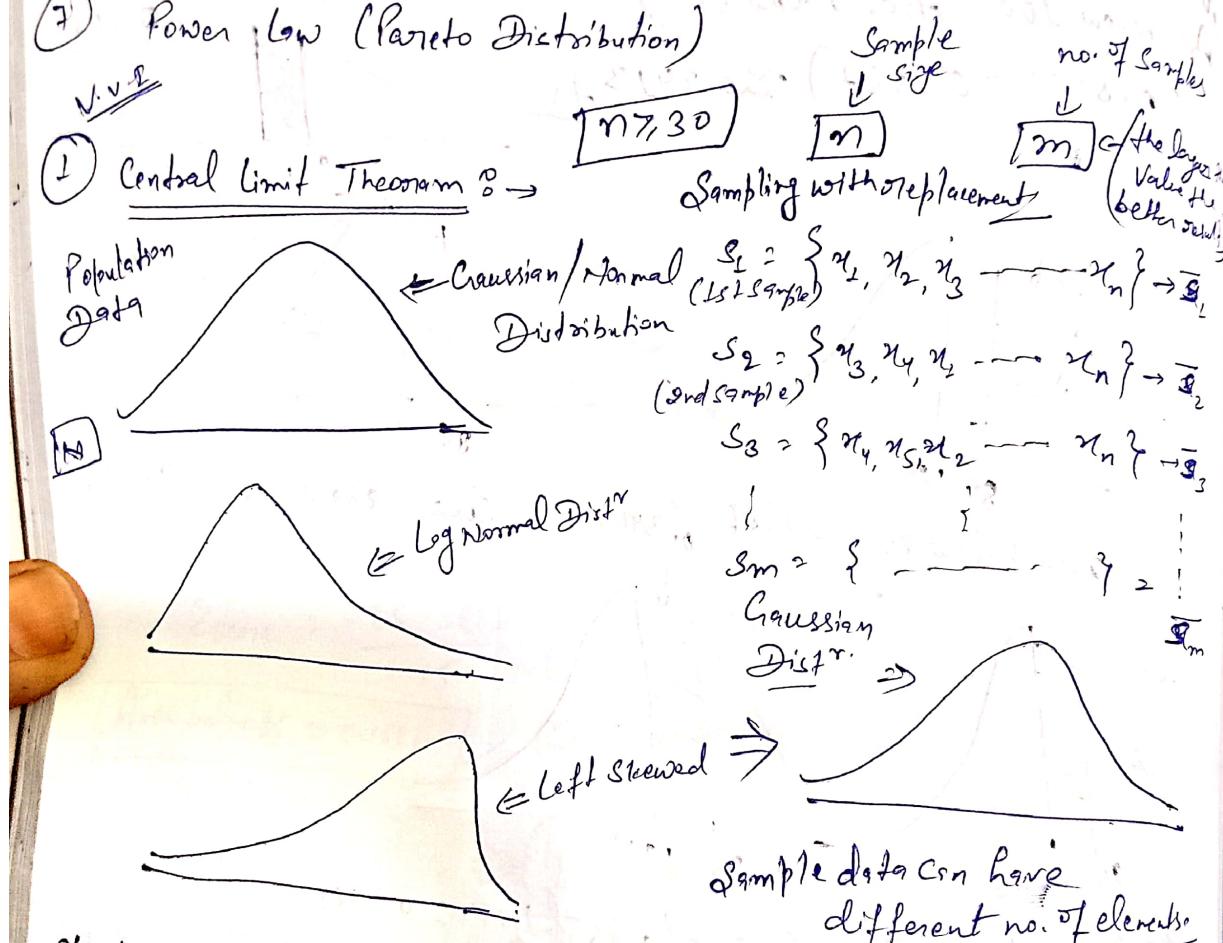


- ① Central Limit Theorem
- ② Probability
- ③ Permutation & Combination
- ④ Covariance, Pearson Correlation, Spearman Rank Correlation.
- ⑤ Bernoulli Distribution
- ⑥ Binomial Distribution.
- ⑦ Power Law (Pareto Distribution)



stmt: Let's suppose take any Population Data (N) either Gaussian or Log Normal or left skewed. Once we select the sample data from any one of them of size $n \geq 30$ and no. of samples is m. Suppose we have taken diff - diff samples i.e. S_1, S_2, \dots, S_m and their mean is $\bar{x}_1, \bar{x}_2, \bar{x}_3, \dots, \bar{x}_m$ then if once we plot the graph of their mean we get Using histogram we get Gaussian Dist. This theorem only holds for $n \geq 30$.

The central limit theorem states that if you have a population with mean μ and standard deviation σ , and take sufficiently large random samples from the population with replacement, then the distribution of the sample means will be approximately normally distributed. Proved

Calculate the size of shark through out the world?

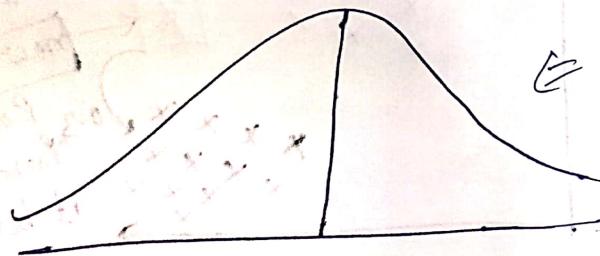
no. of regions = 10 (the larger the value the better the result)
 Sample size = 10
 Lie different regions.

Now find the size of shark from 10 different different regions.

$$R_1 \text{ (Region 1)} = \{x_1, x_2, \dots, x_{n_1}\} = \bar{R}_1$$

$$R_2 \text{ (Region 2)} = \{x_1, x_2, \dots, x_{n_2}\} = \bar{R}_2$$

$$R_{10} \text{ (Region 10)} = \{x_1, x_2, \dots, x_{n_{10}}\} = \bar{R}_{10}$$



Gaussian Dist

Through this we can conclude many things like mean, median, mode etc.
 Based on this lot of assumption can be made.

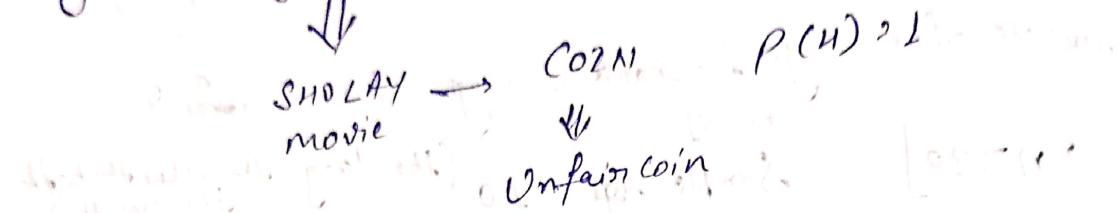
for $n < 30$ central limit theorem fails.

as $n \uparrow \uparrow$ it gets more inclined towards Gaussian Dist.
 (no. of samples increases).

→ Hence with replacement means some of the values get repeated or may not get repeated in sample data.

② Probability : \rightarrow Probability is a measure of the likelihood of an event.

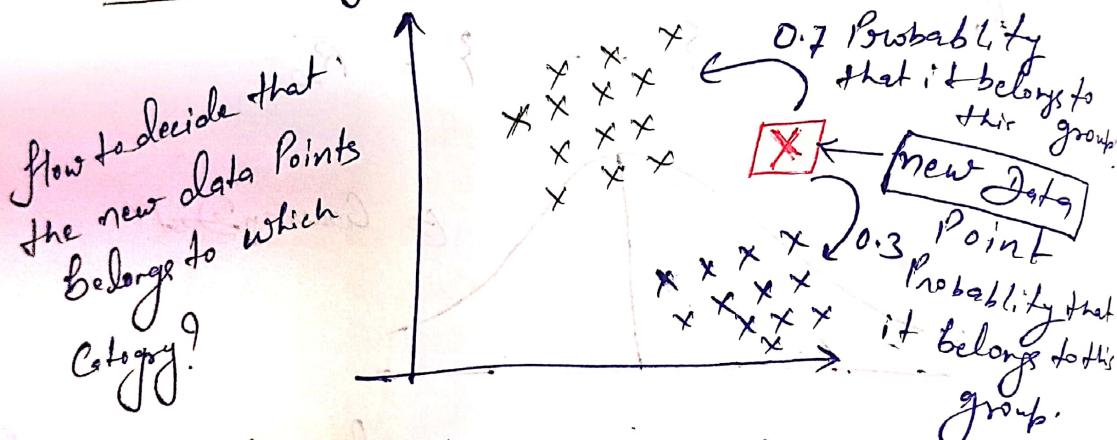
Eg: \rightarrow Tossing a fair coin. $P(H) = 0.5$ $P(T) = 0.5$



Rolling a Dice \rightarrow

$P(1) = \frac{1}{6}$, $P(2) = \frac{1}{6}$, $P(3) = \frac{1}{6}$

\Rightarrow Suppose Solving a Classification Problem.

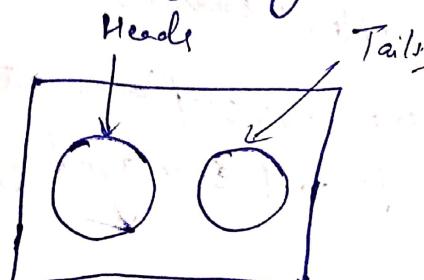


\rightarrow Whenever we try to categorize a new data point that time Probability comes into picture.

③ Mutually Exclusive Event \rightarrow

Two events are mutually exclusive if they don't occur at the same time.

- Eg: \rightarrow
- i) Tossing a coin
 - ii) Winter & Summer
 - iii) Rolling a Dice



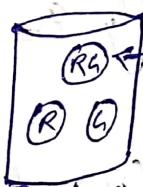
Non-Mutually Exclusive Events:

Q) Non-Mutually Exclusive Events: →
Two events can occur at the same time.

Ex: Q) Picking randomly a card from a deck of card, two events "heart" and "king" can be selected.



Q)



chance of selecting
Red & Green marble
at the same time.

Mutually Exclusive Problem Statement:

Q) What is the probability of coin landing heads or tails.

Sol:

Addition rule of mutually exclusive events

$$P(A \text{ or } B) = P(A) + P(B)$$

$$= \frac{1}{2} + \frac{1}{2} = \underline{\underline{\frac{1}{2} \text{ Ans}}}$$

Q) What is the probability of getting a 1 or 6 or 3 while rolling a dice?

Sol:

$$P(1 \text{ or } 3 \text{ or } 6) = P(1) + P(3) + P(6)$$

$$= \frac{1}{6} + \frac{1}{6} + \frac{1}{6} = \frac{3}{6} = \underline{\underline{\frac{1}{2} \text{ Ans}}}$$

Non Mutually Exclusive Events Problem S.t. 1.1

Q Bag of Marbles: 10 red, 6 green, 3 Red & Green

Q When picking randomly from a bag of marbles, what is the probability of choosing a marble that is red or green?

Sol:

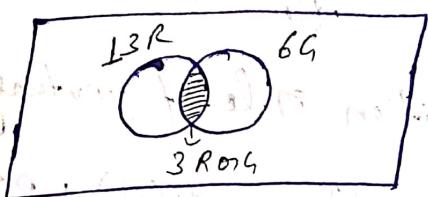
Non Mutually Exclusive Event

Addition Rule for non-mutually Exclusive event

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

$$P(R \text{ or } G) = P(R) + P(G) - P(R \text{ and } G)$$

$$= \frac{10+3}{19} + \frac{6+3}{19} - \frac{3}{19} = \frac{19}{19} \quad \boxed{1}$$



Q From the Deck of Cards: What is the probability of choosing a \heartsuit or Queen? \Rightarrow Non mutually Exclusive

Sol:

$$P(\heartsuit \text{ or Queen}) = P(\heartsuit) + P(\text{Queen}) - P(\heartsuit \text{ and Queen})$$

$$= \frac{13}{52} + \frac{4}{52} - \frac{1}{52}$$

$$= \boxed{\frac{16}{52}} \quad \underline{\text{Ans}}$$

Multiplication Rule

Q) Dependent Events : → Two events are dependent if they affect one another.

Ex:- Bag of Marble. { 0 0 0 X } → Red Marble
 Choose 1 Red Marble then 1 Black Marble.
 $P(R) = \frac{4}{7}$ → $P(B) = \frac{3}{6}$

Q) What is the probability of rolling a "5" and then a "3" with a normal 6 sided dice? {Independent Events}

so, $P(1) = \frac{1}{6}$, $P(2) = \frac{1}{6}$, $P(3) = \frac{1}{6}$, $P(4) = \frac{1}{6}$

→ with Replacement

Independent Event

Multiplication Rule for Independent Event

$$P(A \text{ and } B) = P(A) * P(B)$$

$$= \frac{1}{6} * \frac{1}{6} = \boxed{\frac{1}{36}}$$

$P(A \text{ or } B)$ → Mutual Exclusive

Non Mutual Exclusive

$$P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$$

Non mutually Exclusive

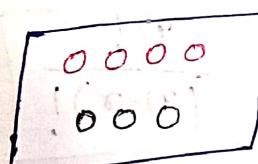
$$P(A \text{ or } B) = P(A) + P(B)$$

Mutually Exclusive

→ without Replacement

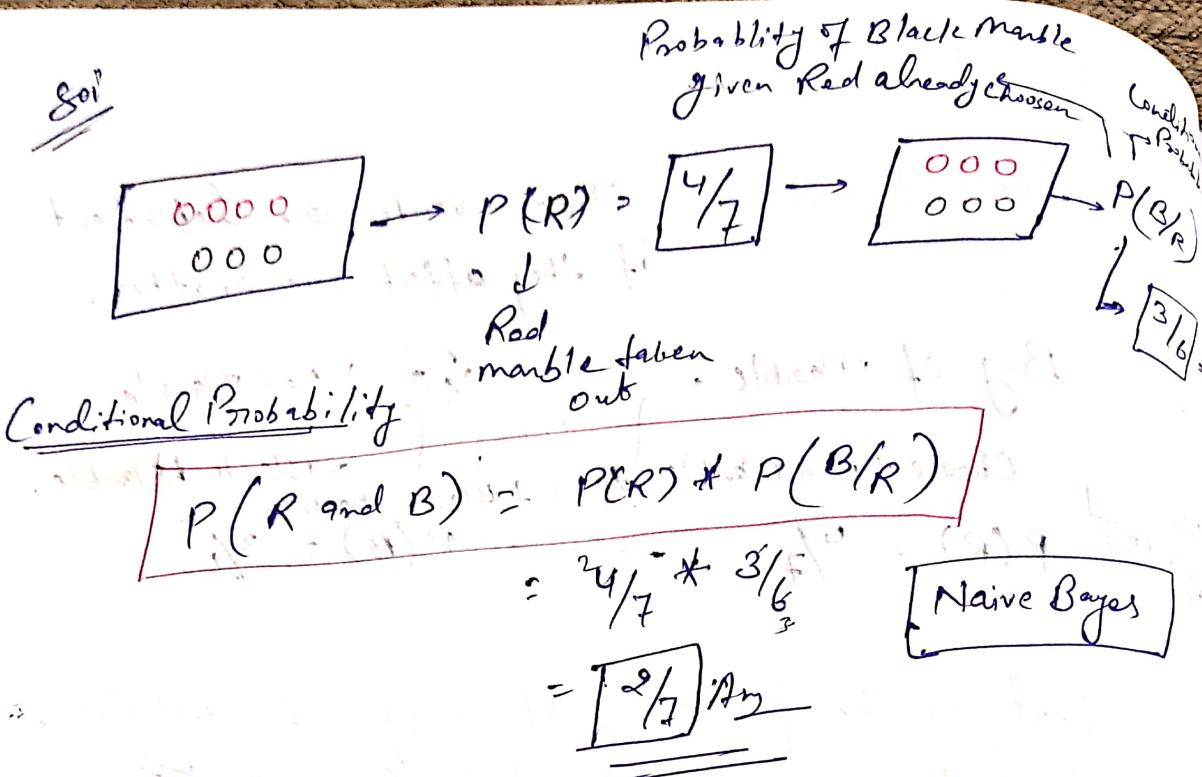
Dependent Event :-

Probability of Drawing a



Dependent Events.

"Red" and then drawing a "Blue" marble from the Bag



Permutation: \rightarrow 60 ways

School of Children

chocolates available: {Dairy milk, Kitkat, Milky Bar, Snickers}

Q How many ways can we ~~distribute~~ arrange the chocolate at among 3 children? places.

Sol:

$$\underline{5} \times \underline{4} \times \underline{3}$$

$\{ \text{DM, Kitkat, MB} \}, \{ \text{Kitkat, DM, MB} \}, \dots$

All the Possible Arrangements

with Permutation Order matters.

Formula

$${}^n P_r = \frac{n!}{(n-r)!} = \frac{5!}{(5-3)!} = \frac{5 \times 4 \times 3 \times 2!}{2!}$$

Where n : Total no. of Objects

r : no. of Selection

= 60 Ways

Hence we can arrange 5 objects at the 3 places in 60 ways

Combination

~~Repetition~~

Repetition will not occur.

$\{DM, IC, MB\}$

$\times \{MB, IC, DM\}^3$

Unique Combination

Both are same

only the unique combination
will be taken in consideration

$$n_{C_r} = \frac{n!}{r!(n-r)!}$$

$$\frac{5!}{3!(5-3)!} = \frac{5 \times 4 \times 3!}{3! \times 2 \times 1} = 10 \text{ Ans}$$

* Covariance

we use it in Feature Selection

X	Y
Age	Weight
12	40
13	48
15	50
17	60
18	62
$\bar{x} = 15$	

Age ↑ weight ↑
Age ↓ weight ↓



Quantity of the relationship
only using mathematical
question.

$$\text{Cov}(x, y) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

$$\sigma^2 = \frac{\sum (y_i - \bar{y})^2}{n-1}$$

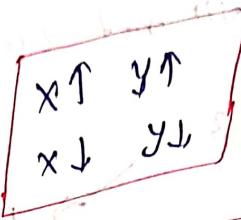
$$\text{Cov}(x, x) \leftarrow$$

$$s^2 = \frac{\sum (x_i - \bar{x})^2}{n-1}$$

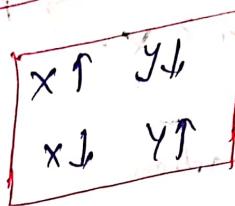
$$\text{Cov}(x, x) = \text{Var}(x)$$

Relationship b/w Covariance
and Variance.

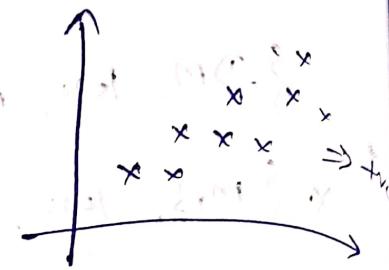
+ve Covariance



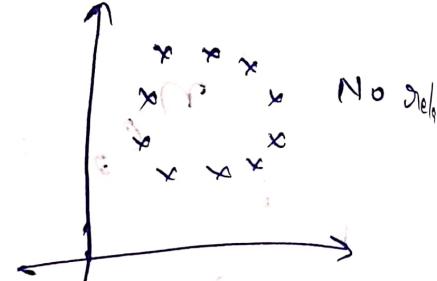
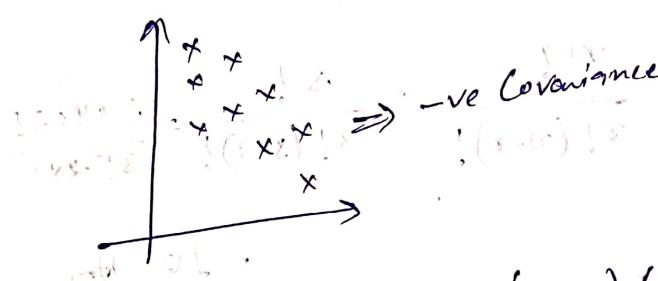
-ve Covariance



~~Covariance~~



Covariance = 0



$$\text{Cov}(x, y) = (12-15)(40-51) + (13-15)(45-51) + \dots$$

$$\bar{x} = \frac{12+13+15+17+18}{5} = \frac{75}{5} = 15$$

$$\bar{y} = \frac{40+46+48+60+62}{5} = \frac{255}{5} = 51$$

$$\begin{aligned} \text{Cov}(x, y) &= (12-15)(40-51) + (13-15)(45-51) + (15-15)(48-51) \\ &\quad + (17-15)(60-51) + (18-15)(62-51) \end{aligned}$$

(5-1)

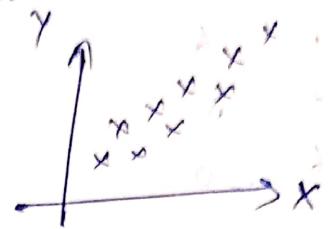
$$= (-3 * -11) + (-2 * -6) + (0 * 3) + (2 * 9) + (3 * 11)$$

$$\begin{aligned} &= \frac{-33 + 12 + 0 + 18 + 33}{4} = \frac{3}{4} = \frac{3}{2} \\ &= \frac{96 - 24}{24} = \frac{24}{24} = 1 \end{aligned}$$

$\text{Cov}(x,y)$ is a + value. Since value of $\text{Cov}(x,y)$ is +ve
so from there we can conclude that

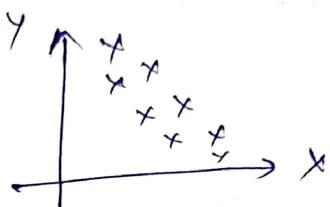
+ve Covariance

$X \uparrow$	$Y \uparrow$
$X \downarrow$	$Y \downarrow$



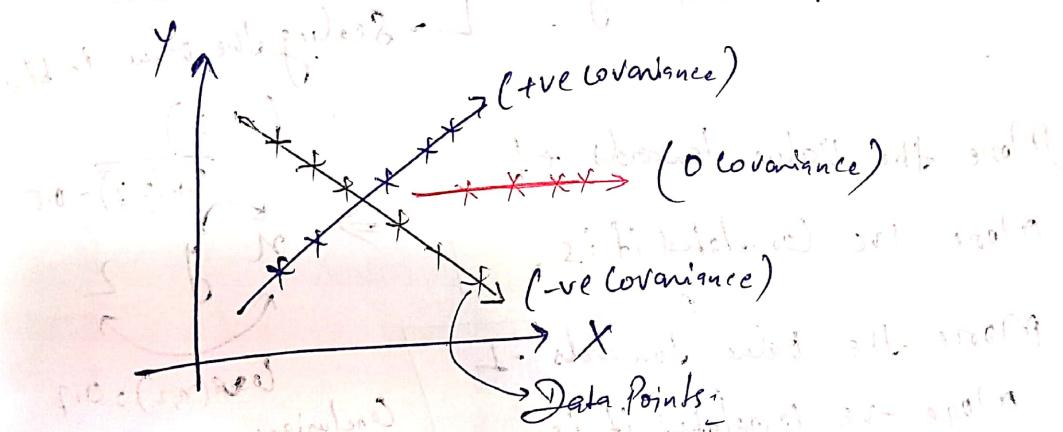
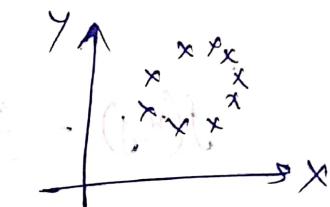
-ve Covariance

$X \uparrow$	$Y \downarrow$
$X \downarrow$	$Y \downarrow$



Covariance = 0

No relation b/w X & Y	
------------------------------	--



Example for -ve Covariance →

$$\begin{array}{ll} \textcircled{1} & \begin{array}{l} \text{X} \quad \text{Y} \\ 10 \quad 6 \\ 8 \quad 7 \\ 7 \quad 8 \\ 6 \quad 10 \end{array} \end{array}$$

Cov(x,y) = $\frac{(10-7.75)(4-7.75) + (8-7.75)(6-7.75) + (7-7.75)(8-7.75) + (6-7.75)(10-7.75)}{(4-1)}$
 $= \frac{(2.25 * -3.75) + (0.25 * -1.75) + (-0.75 * 0.25) + (-1.75 * 2.25)}{3}$

$$\bar{x} = 7.75 \quad \bar{y} = 7.75$$

$$\begin{array}{l} \bar{x} = \frac{10+8+7+6}{4} = \frac{31}{4} = 7.75 \\ \bar{y} = 7.75 \end{array}$$

$= -2.25 + (-0.4375)$
 $+ (-0.1875) + (-3.9375)$
 $= -6.8125$
 (-ve covariance)

Q) $X = \begin{matrix} 10 & 8 & 7 \\ 4 & 6 & 8 \\ 7 & 8 & 6 \end{matrix}$ $\text{Cov}(X, Y) = \frac{(10-7.75)(4-7) + (8-7.75)(6-7) + (7-7.75)(8-7) + (6-7.75)(10-7)}{3}$

$$\frac{6+10}{X=7.75} \quad \frac{9-3.25}{Y=7} \quad \boxed{-3.25} \rightarrow \text{neg Covariance}$$

Pearson Correlation Coefficient \rightarrow They try to restrict

$$f(x, y) = \frac{\text{Cov}(x, y)}{\sigma_x \cdot \sigma_y}$$

Scaling the value in b/w (-1, 1).

More the value towards +1

More +ve correlated it is.

More the value towards -1

More -ve correlated it is.

(-1, 1)

$$\frac{\text{Cov}(x, y)}{\sigma_x \cdot \sigma_y} = 0.5$$

x y z

$$\text{Cov}(x, z) = 0.7$$

Conclusion:-

x is more correlated towards z.

Difference b/w Covariance & Correlation.

In case of Covariance there is no range. We can't say how much +ve or -ve it will be.

But in case of Correlation the value will be in the

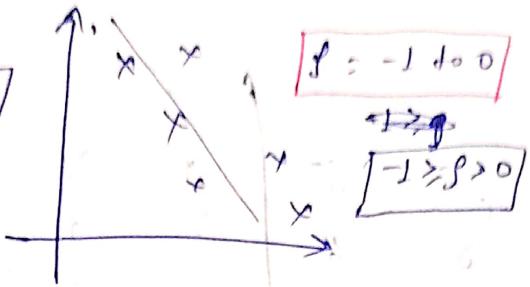
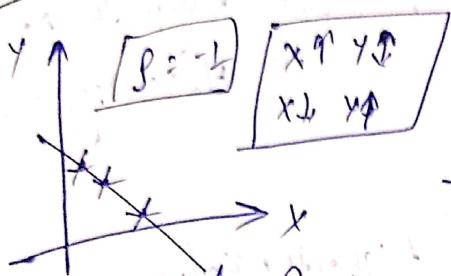
range of (-1, 1) so that we can say how much +ve or -ve it is.

By using Pearson Correlation

b/w (-1, 1) by dividing it ...

... by Pearson Correlation Coefficient we are restricting the

examples



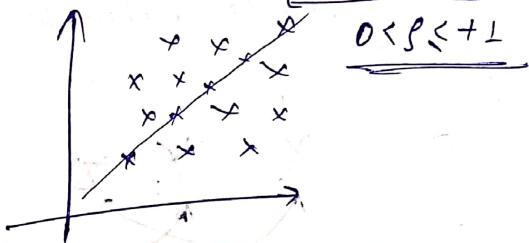
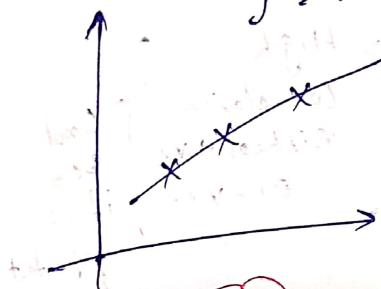
Since all the data points are on the same line & it is in-ve sign hence the value of Pearson Correlation

$$\text{Coefficient } r = -1$$

Since all the data points can't fall on some line,

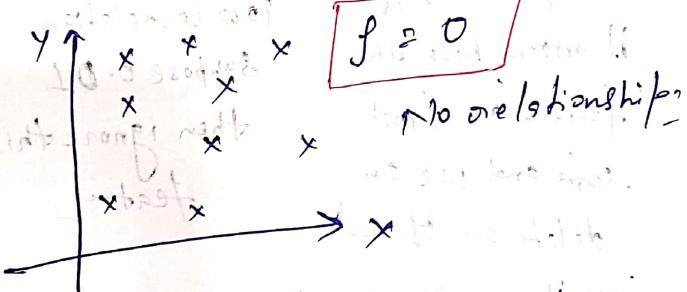
$$-1 < r < 1$$

$$r = 1$$



$$-1 < r < 1$$

Pearson Correlation holds good only for linear data.



$$r = 0$$

No relationship

Spearman Rank Correlation : We use Spearman

Rank Correlation mainly for non-linear data.

Basically here we calculate the Rank

Formula

$$r_s = \frac{\text{Cov}(R(x), R(y))}{\sigma(R(x)) \cdot \sigma(R(y))}$$

for linear data
for non-linear data

Pearson Correlation = 0.89
Spearman Correlation = 1



How to calculate Rank?

X	Y	R(X)	R(Y)
10	4	3	1
8	6	2	3
7	8	1	4
6	10	4	2

We Assign the Rank in Ascending order.

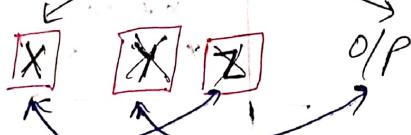
lower the value
lower the Rank

Higher the Value
Higher the Rank

Rank Table
Use this for our calculation

Why This Correlation will be Used?

Ex:



High Correlation either +ve or -ve

Good because if it is important feature to predict the o/p.

$$\text{Cor}(x, z) = 0.95$$

it means both the features are almost same and we can delete one of them.

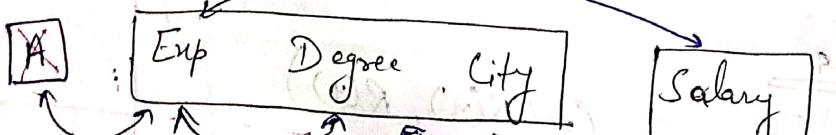
low Correlation,

suppose 0.01
then ignore this feature

Since they are 95% correlated

Using one of them will be enough to get the 'O/P'

Practical Example →



Since A and Exp, no correlation is given. So we can drop A.

we are not losing data by dropping A.

+ve or -ve Correlation