

# STATISTICS

(SESSION-6)

## COVARIANCE :

- Co means two.
- In generally variance means ,How data varies

$$\text{➤ Variance} = \frac{1}{N} \times \sum_{i=1}^N (x_i - \bar{x})^2$$

- Variance Explained about data variance in column itself.
- For example , we have age data available.

How age is varying is explained by age variance.

AGE	SALARY
20	20000
25	25000
30	30000
35	35000
40	40000
45	45000
50	50000

Here , we observe that how a variable is changing means varying according to the another variable.

- In data we have columns , This columns also called as Features.

Variable = Columns = Features

- For Example , We have age Vs salary

How salary is varying according to the age is , This is given by **Covariance**.

- Variance = single column
- Covariance = two columns

$$\text{Variance (x)} = \frac{1}{N} \times \sum_{i=1}^N (x_i - \bar{x})^2$$

$$\text{CoVariance (x,x)} = \frac{1}{N} \times \sum_{i=1}^N (x_i - \bar{x}) * (x_i - \bar{x})$$

$$\text{CoVariance (x,y)} = \frac{1}{N} \times \sum_{i=1}^N (x_i - \bar{x}) * (y_i - \bar{y})$$

Step – 1 : Calculate the mean of Age ( $\bar{x}$ ) and Income ( $\bar{y}$ )

Step – 2 : ( $x_i - \bar{x}$ ) Subtract each value of Age from mean of Age

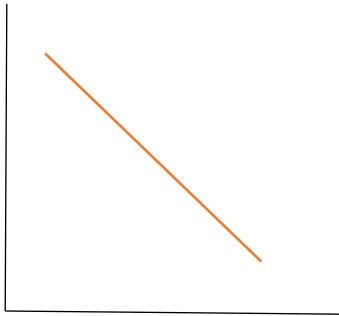
Step – 3 : ( $y_i - \bar{y}$ ) Subtract each value of Income from mean of Income

Step – 4 : Multiply ( $(x_i - \bar{x})(y_i - \bar{y})$ )

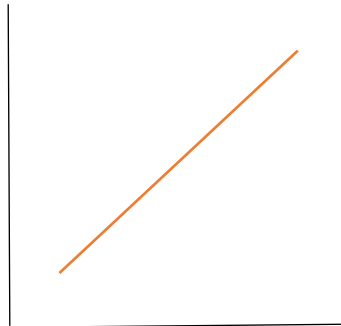
Step – 5 : Addition

Age	Income	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})(y_i - \bar{y})$
20	20000	20-35 = -15	-15000	225000
25	25000	25-35 = -10	-10000	100000
30	30000	30-35 = -5	-5000	25000
35	35000	35-35 = 0	0	0
40	40000	40-35 = 5	5000	25000
45	45000	45-35 = 10	10000	100000
50	50000	50-35 = 15	15000	225000
Age ( $\bar{x}$ ) = 35	Income ( $\bar{y}$ ) = 35000			$\frac{700000}{7}$ = 100000

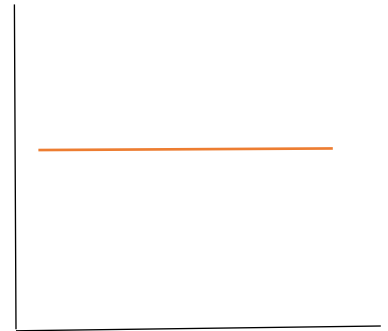
By seeing the covariance values we can say that both are positively related or negative



**Negative correlation**



**positive correlation**



**no relation**

**Plot name :** Scatter plot

➤ Graph between two numerical variables

### **COVARIANCE MATRIX :**

Let us Assume ,

We have Age and Salary

$$\begin{matrix} & \begin{matrix} A & S \end{matrix} \\ \begin{bmatrix} v(A) & cov(A,S) \\ cov(S,A) & v(S) \end{bmatrix} \end{matrix}$$

- There are two columns :

1. Age
2. Salary

- let us know ,

How many combinations possible

0 and 1

00 01 10 11

Age and Salary has 4 Combinations.

- Age with Salary : Covariance
- Age with Age : Variance
- Salary with Age : Covariance
- Salary with Salary : Variance

$$\text{Variance}(x) = \text{covariance}(x,x) = 1$$

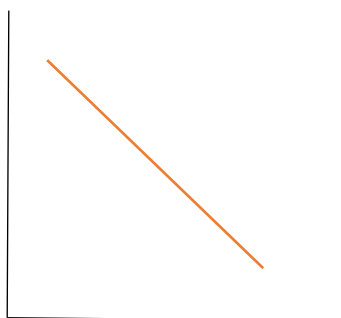
$$\text{Covariance}(x,y) = \text{Covariance}(y,x)$$

- If covariance has positive value , then that means the two variables are positively correlated.
- If covariance has negative value , then that means the two variables are negatively correlated.
- If covariance has zero value , then that means the two variables has no relation.
- Covariance Provides only whether the variables has relation or not.
- It will not provide how much percentage the variables are related with each other.
- Covariance can be anything from information to information.

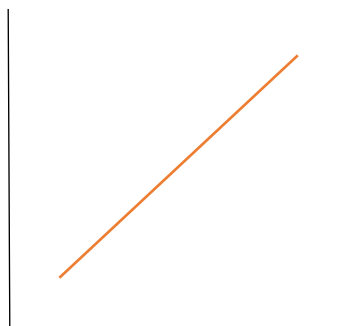
$$\begin{bmatrix} 25 & 5 \\ 5 & 30 \end{bmatrix} \rightarrow \text{covariance value is Positive So, Positive relation.}$$

$$\begin{bmatrix} 25 & -5 \\ -5 & 30 \end{bmatrix} \rightarrow \text{covariance value is Negative So, Negative relation.}$$

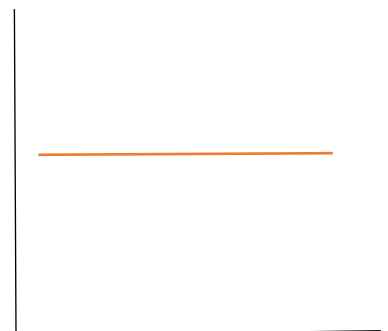
$$\begin{bmatrix} 25 & 0 \\ 0 & 30 \end{bmatrix} \rightarrow \text{covariance value is Zero So , No relation.}$$



**Negative correlation**



**Positive correlation**



**No correlation**

**Plot name :** Scatter plot

- Graph between two numerical variables.

### **CORRELATION COEFFICIENT (r) :**

We already know that , Covariance will provide that there is a relation or not , But it will not provide the amount of relation.

- Correlation will provide the Amount of relation.
- It will Explain how two variables related to each other.

In order to get the percentage of relationship we will use :

#### **Pearson correlation coefficient**

- It is denoted by 'r'
- r varies from -1 to 1
- r = -1 to 0 means Negative relation
- r = 0 means no relation
- r = 1 means Positive relation

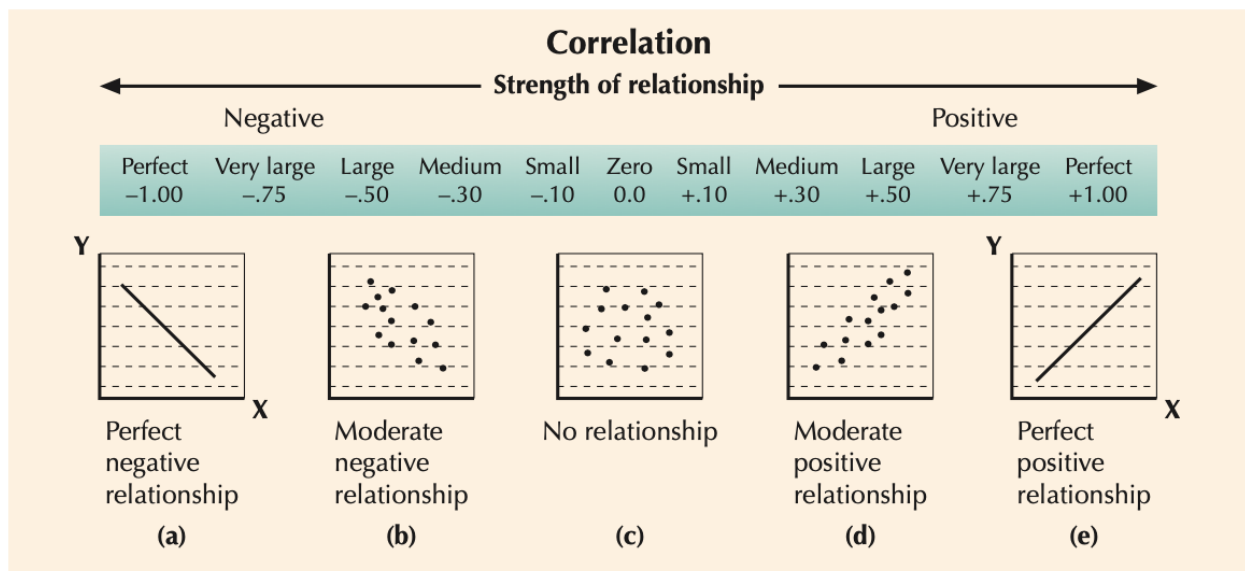
For Example :

- If r = 0.7
  - There is a 70% Positive relation between Age and Salary
- If r = -0.7
  - There is a 70% Negative relation between Age and Salary
- If r = 0
  - There is a no relation between Age and Salary

$$r = \frac{cov(x,y)}{\sigma_x * \sigma_y}$$

$$r = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum (x_i - \bar{x})^2 \sum (y_i - \bar{y})^2}}$$

- Features or Variables are having No relation.
- No relation also called as Independent each other.
- Independence also called as Perpendicular each other.
- Perpendicular is also called as Orthogonal each other.
- Orthogonal is also called as 90 Degrees phase shift.



$$\begin{bmatrix} 5 & 20 \\ 20 & 10 \end{bmatrix} \quad \begin{bmatrix} 5 & -20 \\ -20 & 10 \end{bmatrix} \quad \begin{bmatrix} 5 & 0 \\ 0 & 10 \end{bmatrix}$$

(1)

(2)

(3)

Positive

negative

no relation