

CHAP 17

Correlation and Regression

* Types of Data:

① Univariate Data:

- one variable at a time.

② Bivariate Data:

- When data are collected and analysed on two variables, simultaneously, they are known as bivariate.

* The corresponding frequency distribution, derived from it, is known as bivariate frequency distribution.

Marginal Distribution Conditional Distribution

* Correlation:

- In a bivariate data, if change in one variable causes change in another variable either directly or inversely, then the two variables are known to be associated or correlated. $P \uparrow \rightarrow \text{Demand} \downarrow$

Types of Correlation

Positive

$z \geq 0$
 $r = 0$

Negative

Perfect Positive

Positive

Perfect Negative

Negative

$r = 1$

$0 < r < 1$

$r = -1$

$-1 < r < 0$

• Correlation $= r$

• The value of Correlation range from -1 to 1 ,
 $-1 \leq r \leq 1$

Scatter Diagram

Linear

Non-linear / Curve linear

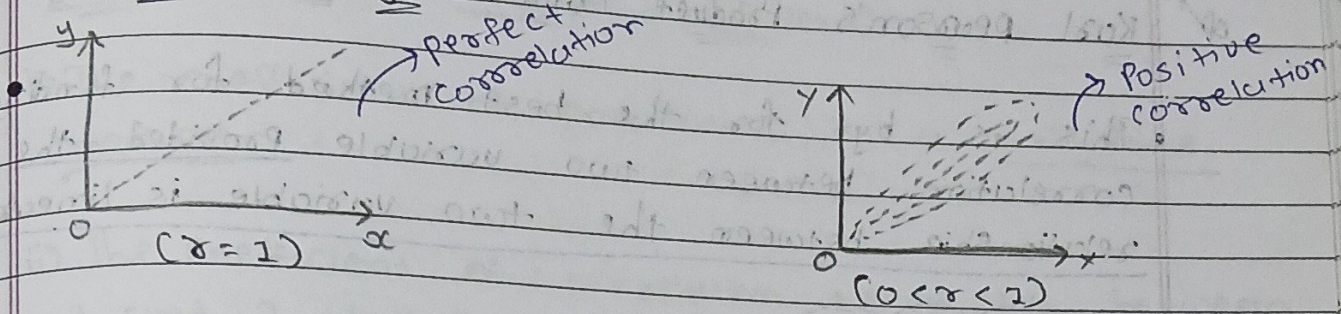
Measures of Correlation

Karl Pearson's Product Moment Correlation coefficient

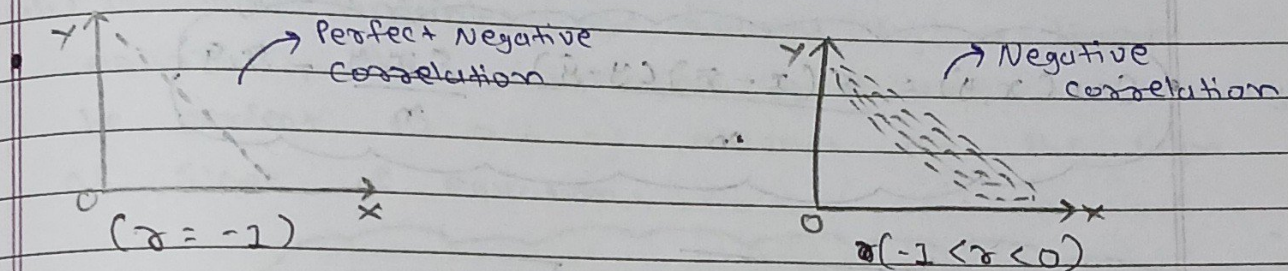
Spearman's Rank correlation coefficient

Coefficient of Concurrent Deviations

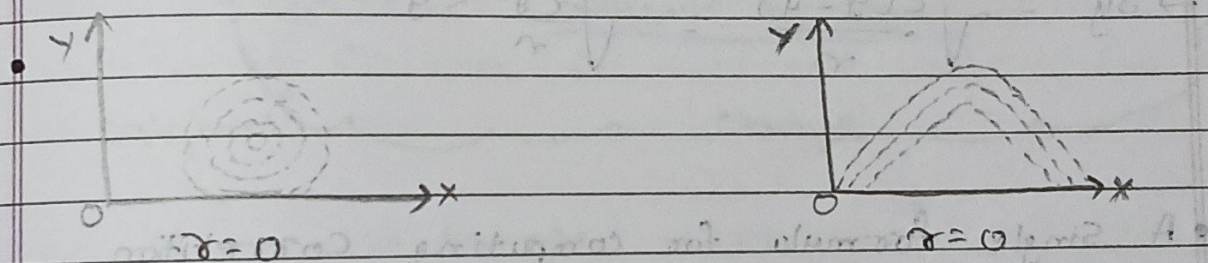
Scatter Diagram:



- The plotted points lie from lower left corner to upper right corner.



- The plotted points concentrate from upper left to lower right.



- The plotted points could be equally distributed without depicting any particular pattern.

4/ Karl Pearson's Product Moment Correlation Coefficient

- This is by far the best method for finding correlation between two variable provided the relationship between the two variable is linear.

$$r = \frac{\text{Cov}(x, y)}{S_x \times S_y}$$

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

$$\rightarrow S_x = \sqrt{\frac{\sum (x - \bar{x})^2}{n}} = \sqrt{\frac{\sum x^2}{n} - \bar{x}^2}$$

$$\rightarrow S_y = \sqrt{\frac{\sum (y - \bar{y})^2}{n}} = \sqrt{\frac{\sum y^2}{n} - \bar{y}^2}$$

- A single formula for computing correlation coefficient is given by

$$r = \frac{n \sum xy - \sum x \cdot \sum y}{\sqrt{n \cdot \sum x^2 - (\sum x)^2} \cdot \sqrt{n \cdot \sum y^2 - (\sum y)^2}}$$

Properties of Correlation Coefficient

- i) The coefficient of correlation is a unit-free measure.
- ii) The coefficient of correlation always lies between -1 and 1 , including both the limiting values.
 $-1 \leq r \leq 1$
- iii) If two variables are related by a linear equation, then correlation coefficient will always be perfect $+1$ or -1 depends on the sign of slope of equation.

iv) Change of origin = No Impact

v) Change of scale = No impact of value but affected by sign.

Original

Change

X

→ U

Y

→ V

r_{xy}

→

r_{uv}

- If sign of both change of scale are same.
 $\therefore r_{uv} = r_{xy}$

- If sign of both change of scale are different
 $\therefore r_{uv} = -r_{xy}$

☛ Spearman's Rank Correlation coefficient

- When we need finding correlation between two qualitative characteristics, say, beauty and intelligence, we take recourse to using rank correlation coefficient.
- Rank correlation can also be applied to find the level of agreement (or disagreement) between two ~~variables~~ judges so far as assessing a qualitative characteristic is concerned.
- S R C C is given by

$$r_r = 1 - \frac{6 \sum d_i^2}{n(n^2 - 1)}$$

- Where r_r denotes Rank correlation coefficient and it lies between -1 and 1 inclusive of these two values.
- $d_i = X_i - Y_i$ represents the difference in ranks for the i -th individual and n denotes the number of individuals.

* Coefficient of Concurrent Deviation:

- A very simple & casual method of finding the ~~rough~~ magnitude of the two variables.

$$r_c = \frac{+}{-} \sqrt{\frac{+ (2c - m)}{m}}$$

- Where c is the num. of Concurrent deviation (same direction) m is num. of pairs compared,
 $m = n - 1$

* Regression Analysis:-

- In regression analysis, we are concerned with the estimation of one variable for a given value of another variable on the basis of an average mathematical relationship between the two variables.

* Estimation of Y when X is given $\rightarrow Y$ on X

Y Dependent

X Independent

$$y = a + bx$$

* Estimation of X when y is given $\rightarrow X$ on Y

X Dependent

Y Independent

$$x = a + by$$

Estimation of Y when X is given:-

Regression line of Y on X

$$Y - \bar{Y} = b_{yx} (X - \bar{X})$$

Estimation of X when Y is given:-

Regression line of X on Y

$$X - \bar{X} = b_{xy} (Y - \bar{Y})$$

Method
of
Least
Squares

Regression coefficient :-

$$i) b_{yx} = \frac{\text{cov}(x, y)}{\text{var of } x}$$

$$i) b_{xy} = \frac{\text{cov}(x, y)}{\text{var of } y}$$

$$ii) b_{yx} = r \cdot \frac{SD_y}{SD_x}$$

$$ii) b_{xy} = r \cdot \frac{SD_x}{SD_y}$$

y on x

x on y

Properties of Regression line / coefficient:-

- The two lines of regression intersect at the point $(\bar{X}, \bar{Y})$ mean where x & y are the variable under consideration.

- According to this property, the point of intersection of the regression line of y on x and the regression line x on y is $(\bar{X}, \bar{Y})$ i.e. the solution of the simultaneous equations in x and y .

iii) $r = \sqrt{b_{yx} \times b_{xy}}$

- Product of the regression coefficient must be numerically less than unity.
- This can be applied, unlike correlation, for any type of relationship linear as well as curvilinear.
- The two lines of regression coincide i.e. become identical when $r = -1$ or 1 or in other words, there is a perfect negative or positive correlation between the two variables under discussion.
- If $r = 0$ Regression lines are perpendicular to each other.

Coefficient of Determination / Explained Variance / Accounted Variance:-

- Correlation Coefficient measuring a linear relationship between the two variables indicates the amt. of variation of one variable accounted for by the other variable. r^2

Coefficient of non-determination / unexplained Variance / unaccounted Variance:-

- The coefficient of non-determination is given by $(1 - r^2)$ and can be interpreted as the ratio of unexplained variance to the total variance.

The end