

Module - 3

①

Correlation & Regression :

- * Correlation & Regression
- * Partial & Multiple correlation
- * Multiple Regression

Correlation

- * If the change in one variable affects the change in another variable then the variables are said to be correlated.

+ve Correlation

If the two variables deviate in the same direction then the variables are +vey correlated.

-ve Correlation

If the two variables constantly deviate in the opposite direction then the variables are -vey correlated.

Examples

- * The income & Expenditure is +vey correlated
- * Volume & Pressure of a perfect gas is -vey correlated.

Perfect Correlation:

If the deviation in one variable is followed by a corresponding & proportional deviation in the other.

Note

A measure of intensity (or) degree of linear relationship between two variables.

Karl Pearson's Coefficient of Correlation
(or) Correlation Coefficient:

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

$$\begin{aligned} \text{Where } \text{cov}(x,y) &= E((x - E(x))(y - E(y))) \\ &= \frac{1}{n} \sum (x_i - \bar{x})(y_i - \bar{y}) \\ &= \frac{1}{n} \sum x_i y_i - \bar{x} \bar{y} \end{aligned}$$

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

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

Note

$$\text{cov}(x,y) = \sigma_{xy} = M_{11}$$

which is also known as Product moment correlation coefficient

Properties

- ① $-1 \leq r \leq 1$ (Limits for Correlation Coefficient)
- ② If $r=1$, The correlation is perfect & +ve
If $r=-1$, The Correlation is perfect & -ve
- ③ Correlation Coefficient is independent of change of origin & scale.
- ④ If x & y are Random Variables & a, b, c, d are any nos provided only that $a \neq 0, c \neq 0$
then $r(ax+b, cy+d) = \frac{ac}{|ac|} r(x, y)$
 $\& r(ax, cy) = \frac{ac}{|ac|} r(x, y)$

- ⑤ Two independent Variables are uncorrelated.
i.e., If x & y are uncorrelated then
 $\text{cov}(x, y) = 0$
 $\Rightarrow r(x, y) = 0$
but the converse is not true.

Sums

Calculate the Correlation Coefficient for the following heights (in inches) of fathers (x) & their sons (y):

X	65	66	67	67	68	69	70	72
Y	67	68	65	68	72	72	69	71

Sol

X	Y	$U = X - 68$	$V = Y - 69$	U^2	V^2	UV
65	67	-3	-2	9	4	6
66	68	-2	-1	4	1	2
67	65	-1	-4	1	16	4
67	68	-1	-1	1	1	1
68	72	0	3	0	9	0
69	72	1	3	1	9	3
70	69	2	0	4	0	0
72	71	4	2	16	4	8
<u>Total</u>		$\sum U = 0$	$\sum V = 0$	$\sum U^2 = 36$	$\sum V^2 = 44$	$\sum UV = 24$
		$= 0$	$= 0$	$= 36$	$= 44$	$= 24$

$$\bar{U} = \frac{\sum U}{n} = \frac{0}{8} = 0$$

$$\bar{V} = \frac{\sum V}{n} = \frac{0}{8} = 0$$

$$COV(U, V) = \frac{1}{n} \sum UV - \bar{U}\bar{V} = \frac{1}{8} \times 24 = 3$$

$$\sigma_U^2 = \frac{1}{n} \sum U^2 - (\bar{U})^2 = \frac{1}{8} \times 36 = 4.5$$

$$\sigma_V^2 = \frac{1}{n} \sum V^2 - (\bar{V})^2 = \frac{1}{8} \times 44 = 5.5$$

$$\therefore r = \frac{COV(U, V)}{\sigma_U \sigma_V} = \frac{3}{\sqrt{4.5} \sqrt{5.5}} = 0.603$$

$= r(U, V)$
 $= r(X, Y)$

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② A Computer while calculating Correlation Coefficient between the Variables x & y from 25 pairs of observations obtained ~~from~~ the following results.

$$n=25, \sum x=125, \sum x^2=650, \sum y=100, \sum y^2=460$$

$\sum xy=508$. It was, however, later discovered at the time of checking that he had copied down two pairs as

X	Y
6	14
8	6

 while the correct values were

X	Y
8	12
6	8

.

Obtain the Correct Value of Correlation Coefficient.

Solution

$$\text{Corrected } \sum x = 125 - 6 - 8 + 8 + 6 = 125$$

$$\text{Corrected } \sum y = 100 - 14 - 6 + 12 + 8 = 100$$

$$\text{Corrected } \sum x^2 = 650 - 6^2 - 8^2 + 8^2 + 6^2 = 650$$

$$\text{Corrected } \sum y^2 = 460 - 14^2 - 6^2 + 12^2 + 8^2 = 436$$

$$\begin{aligned} \text{Corrected } \sum xy &= 508 - 6 \times 14 - 8 \times 6 + 8 \times 12 + 6 \times 8 \\ &= 520 \end{aligned}$$

$$\therefore \text{Cov}(x,y) = \frac{1}{n} \sum xy - \bar{x}\bar{y} = \frac{1}{25} (520) - 5 \times 4 = 4/5$$

$$\sigma_x^2 = \frac{1}{n} \sum x^2 - (\bar{x})^2 = \frac{1}{25} \times 650 - (5)^2 = 1$$

$$\sigma_y^2 = \frac{1}{n} \sum y^2 - (\bar{y})^2 = \frac{1}{25} \times 436 - 16 = 36/25$$

$$\text{Corrected } \gamma(x_1y) = \frac{\text{Cov}(x_1y)}{\sigma_x \sigma_y} = \frac{4/5}{1 \times 6/5} = \frac{2}{3} \approx 0.67$$

Note

① If $Z = ax + by$ & γ is the Correlation Coefficient between x & y then $\sigma_z^2 = a^2 \sigma_x^2 + b^2 \sigma_y^2 + 2ab \rho \sigma_x \sigma_y$

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$$\text{Correlation Coefficient when } a=1, b=-1 \text{ & } Z=x-y$$

$$\gamma = \frac{\sigma_x^2 + \sigma_y^2 - \sigma_{x-y}^2}{2\sigma_x \sigma_y}$$

$\sigma_x, \sigma_y, \sigma_{x-y}$ are S.D's of x, y & $x-y$ respectively.

$$③ V(ax+by) = a^2 V(x) + b^2 V(y) + 2ab \text{ Cov}(x,y)$$

$$V(ax-bx) = a^2 V(x) + b^2 V(y) - 2ab \text{ Cov}(x,y)$$

Sum

④ The independent variables x & y are defined as

$$f(x) = \begin{cases} 4an, & 0 \leq x \leq r \\ 0 & \text{otherwise} \end{cases} \quad \text{and} \quad f(y) = \begin{cases} 4by, & 0 \leq y \leq s \\ 0 & \text{otherwise} \end{cases}$$

Find the correlation between U & V where

$$U = x+y \quad \& \quad V = x-y$$

$$\text{Ans: } \gamma(U,V) = \frac{b-a}{b+a} \quad \text{where } a = \frac{1}{2r^2}, b = \frac{1}{2s^2}$$

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Sol Since x & y are independent & total probability is 1

$$\therefore \int_0^r f(u) du = 4a \int_0^r u du = 1$$

$$\Rightarrow 4a \left(\frac{u^2}{2} \right)_0^r = 1$$

$$2ar^2 = 1$$

$$\therefore r^2 = 1/2a$$

$$(\text{or}) a = 1/2r^2$$

$$\int_0^s f(y) dy = 1$$

$$\Rightarrow 4b \int_0^s y dy = 1$$

$$2ab \left(\frac{y^2}{2} \right)_0^s = 1$$

$$2b(s^2) = 1$$

$$s^2 = 1/2b$$

$$(\text{or}) b = 1/2s^2$$

$$\therefore f(u) = 4au = 2u/r^2, 0 \leq u \leq r$$

$$\& f(y) = 4by = 2y/s^2, 0 \leq y \leq s$$

Since x & y are independent Variates,

$$\gamma(x,y) = 0 \Rightarrow \text{cov}(x,y) = 0$$

$$\therefore \text{cov}(v,v) = \text{cov}(x+y, x-y)$$

$$= \text{cov}(x,x) - \text{cov}(x,y) + \text{cov}(y,x) - \text{cov}(y,y)$$

$$= \text{cov}(x,x) - \text{cov}(y,y) = \sigma_x^2 - \sigma_y^2$$

$$V(V) = V(x+y) = V(x) + V(y) + 2\text{cov}(x,y)$$

$$= \sigma_x^2 + \sigma_y^2$$

$$V(A) = V(x-y) = \sigma_x^2 - \sigma_y^2$$

$$\therefore r(u,v) = \frac{\text{cov}(u,v)}{\sigma_u \sigma_v} = \frac{\sigma_x^2 - \sigma_y^2}{\sigma_x^2 + \sigma_y^2}$$

$$\therefore E(x) = \int_0^r n f(n) dn = \frac{2}{\pi^2} \int_0^r n^2 dn = \frac{2r^3}{3}$$

$$E(x^2) = \int_0^r n^2 f(n) dn = \frac{2}{\pi^2} \int_0^r n^4 dn = \frac{2}{\pi^2} \left(\frac{n^5}{5} \right)_0^r$$

$$= \frac{1}{2\pi^2} (r^4) = \frac{r^2}{2}$$

$$V(x) = E(x^2) - (E(x))^2 = \frac{r^2}{2} - \frac{4r^2}{9} = \frac{r^2}{18}$$

$$= \frac{1}{36a}$$

IIIrd $E(y) = \frac{2s}{3}$, ~~(cancel)~~

$$\therefore E(y) = \int_0^s y f(y) dy = \frac{4}{2s^2} \int_0^s y^2 dy$$

$$= \frac{2}{s^2} \left(\frac{y^3}{3} \right)_0^s = \frac{2s^3}{3s^2} = \frac{2s}{3}$$

$$E(y^2) = \frac{2}{s^2} \int_0^s y^3 dy = 2s^2 \left(\frac{y^4}{4} \right)_0^s =$$

$$= \frac{1}{2s^2} (s^4) = \frac{s^2}{2}$$
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$$V(y) = E(y^2) - (E(y))^2 = s^2/2 - \frac{4s^2}{9}$$

$$= \frac{9s^2 - 8s^2}{18} = \frac{s^2}{18} = 1/36b$$

$$\therefore r(u,v) = \frac{\frac{1}{36a} - \frac{1}{36b}}{\frac{1}{36a} + \frac{1}{36b}} = \frac{36b - 36a}{36b + 36a}$$

$$= \frac{b-a}{b+a}$$

Rank Correlation (Spearman's Rank Correlation Coefficient)

Rank Correlation Coefficient = $r_s = 1 - \frac{6 \sum_{i=1}^n d_i^2}{n(n^2 - 1)}$

Where $d_i = x_i - y_i$ (difference between ranks of x & y)

Note In Case of Repeated Ranks

In the formula, with $\sum d^2$ we add the correction factor $\frac{m(m^2-1)}{12}$ where m is the

number of times an item is repeated.

* This correction factor is to be added for each repeated value in both the X series & Y series.

Note

limit for the rank Correlation Coefficient

$$-1 \leq r \leq 1.$$

Note

* For Tie Ranks

If two or more observations have the same rank we assign to them the mean rank.

In this case,

$$\text{correlation coefficient } r = 1 - \frac{6 \left(\sum d_i^2 + \frac{\sum m(m-1)}{12} \right)}{n(n^2-1)}$$

where m — number of times a rank is repeated.

* Rank correlation also lies between -1 to 1 .

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Sums

Ten competitors in a beauty contest are ranked by three judges in the following order:

First judge	1	4	6	3	2	9	7	8	10	5
Second judge	2	6	5	4	7	10	9	3	8	1
Third judge	3	7	4	5	10	8	9	2	6	1

Use the method of rank correlation coefficient to determine which pair of judges have the nearest approach to common taste in beauty?

Sol.

Let x, y, z denotes the ranks by 1st, 2nd & 3rd judges respectively.

x	y	z	Ranking $x \sim y$	Rank of $y \sim z$	Ranking $x \sim z$	d_1	d_2	d_3
1	2	3	-1	-1	-2	1	1	4
4	6	7	-2	-1	-3	4	1	9
6	5	4	1	1	2	1	1	4
3	4	5	-1	-1	-2	1	1	4
2	7	10	-5	-3	-8	25	9	64
9	10	8	-1	2	1	1	4	1
7	9	9	-2	0	-2	4	0	4
8	3	2	5	1	6	25	1	36
10	8	6	2	2	4	4	4	16
5	1	1	4	0	4	16	0	16
Total						82	22	158

$$\therefore n = 10$$

$$P_{xy} = 1 - \frac{6 \sum d_1^2}{n(n^2-1)} = 1 - \frac{6 \times 82}{10 \times 99} = 0.503$$

$$P_{yz} = 1 - \frac{6 \sum d_2^2}{n(n^2-1)} = 1 - \frac{6 \times 22}{10 \times 99} = 0.867$$

$$P_{zx} = 1 - \frac{6 \sum d_3^2}{n(n^2-1)} = 1 - \frac{6 \times 158}{10 \times 99} = 0.04$$

* The Rank Correlation Coefficient between y & z is +ve & highest among all the three.

\therefore Judges y & z have the nearest approach for common taste in beauty.

② Obtain the Rank Correlation Coefficient for the following data:

X	68	64	75	50	64	80	75	40	55	64
Y	62	58	68	45	81	60	68	48	50	70

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Sol

X	Y	Rank of X	Rank of Y	$d = x - y$	d^2
68	62	4	5	-1	1
64	58	6	7	-1	1
75	68	2.5	3.5	-1	1
50	45	9	10	-1	1
64	81	6	1	5	25
80	60	1	6	-5	25
75	68	2.5	3.5	-1	1
40	48	10	9	1	1
55	50	8	8	0	0
64	70	6	2	4	16
<hr/>		Total		$\sum d = 0$	$\sum d^2 = 72$

In X Series,

75 Occurs 2 times

68 Occurs 3 times

In Y Series,

68 occurs twice

$$m_1 = 2$$

$$\therefore m_1 = 2, m_2 = 3$$

Correction factors

$$\begin{aligned}
 CF &= \frac{m_1(m_1^2 - 1)}{12} + \frac{m_2(m_2^2 - 1)}{12} + \frac{m_3(m_3^2 - 1)}{12} \\
 &= \frac{2(4-1)}{12} + \frac{3(9-1)}{12} + \frac{2(4-1)}{12} = \frac{1}{2} + 2 + \frac{1}{2} = 3
 \end{aligned}$$

Rank Correlation Coefficient

$$= \rho = 1 - \frac{6 \left(\sum d^2 + \frac{m_1(m_1^2 - 1)}{12} + \frac{m_2(m_2^2 - 1)}{12} + \frac{m_3(m_3^2 - 1)}{12} \right)}{n(n^2 - 1)}$$

$$\therefore \rho = 1 - \frac{6(72 + 3)}{10 \times 99} = 0.545$$

Multiple & Partial correlation

The value of one variable influenced by many others, is known as multiple correlation.

If x_1 is the dependent Variable & x_2, x_3, \dots, x_n are independent Variables then we denote the multiple correlation by $R_{1,2,3,\dots,n}$.

For Example

The yield of crop may depend upon the rainfall, fertilizer, the average temperature, the period between sowing & harvesting.

Partial correlation

The Correlation between only two Variates eliminating the linear effect of other Variates in them is called as Partial Correlation.

i.e., A Partial Correlation Coefficient measures the relationship between any two variables when the other variables connected with those variables are kept constant.

Example

x. Correlation between the heights & weights of the boys of the same age.

~~If the Variable X_1, X_2 are~~

If we want to study the effect of one variable X_2 on another variable X_1 , after eliminating the effects of all other variables the measure of relationship between X_1 & X_2 is called Partial Correlation & it is denoted by

$r_{12.3}$ where X_3 kept as constant &
 X_1 & X_2 are partially correlated.

Formulae

$$* W = \begin{vmatrix} 1 & r_{12} & r_{13} \\ r_{21} & 1 & r_{23} \\ r_{31} & r_{32} & 1 \end{vmatrix} = 1 - r_{12}^2 - r_{13}^2 - r_{23}^2 + 2r_{12}r_{13}r_{23}$$

$$\& W_{11} = \text{The minor of } 1 = \begin{vmatrix} 1 & r_{23} \\ r_{32} & 1 \end{vmatrix} = 1 - r_{23}^2$$

$$W_{12} = \text{The minor of } r_{12} = - \begin{vmatrix} r_{21} & r_{23} \\ r_{21} & 1 \end{vmatrix} = r_{21} + r_{22}r_{31}$$

$$W_{13} = \text{The minor of } r_{13} = \begin{vmatrix} r_{21} & 1 \\ r_{31} & r_{32} \end{vmatrix} = r_{21}r_{32} - r_{31}$$

and so on.

Multiple Correlation

If x_1 is correlated with x_2 & x_3 then the multiple correlation is

$$R_{1,23}^2 = 1 - \frac{W}{W_{11}} = \frac{r_{12}^2 + r_{13}^2 - 2r_{12}r_{13}r_{23}}{1 - r_{23}^2}$$

$$\text{likewise } R_{2,13}^2 = \frac{r_{12}^2 + r_{23}^2 - 2r_{12}r_{13}r_{23}}{1 - r_{12}^2}$$

$$\delta \quad R_{3.12} = \frac{\gamma_{13}^2 + \gamma_{23}^2 - 2\gamma_{12}\gamma_{13}\gamma_{23}}{1 - \gamma_{12}^2}$$

Properties of Multiple Correlation Coefficient

* $0 \leq R_{1.23} \leq 1$

* If $R_{1.23} = 0$ then all the total & Partial correlations involving x_1 are zero.

* $R_{1.23}$ is not less than any total correlation coefficient.

$$R_{1.23} \geq \gamma_{12}, \gamma_{13}, \gamma_{23}.$$

Partial correlation

The correlation coefficients between x_1 & x_2 after the linear effect of x_3 on each of them has been eliminated is called the partial correlation coefficient between x_1 & x_2 & it is denoted by

$$\gamma_{12.3} = \frac{\gamma_{12} - \gamma_{13}\gamma_{23}}{\sqrt{(1-\gamma_{13}^2)(1-\gamma_{23}^2)}}$$

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$$\gamma_{13.2} = \frac{\gamma_{13} - \gamma_{12}\gamma_{23}}{\sqrt{(1-\gamma_{12}^2)(1-\gamma_{23}^2)}}$$

Problems

From the data relating to the yield of dry bark (x_1), height (x_2) & girth (x_3) for 18 cinchona plants, the following correlation coefficient were obtained:

$$r_{12} = 0.77, r_{13} = 0.72 \text{ & } r_{23} = 0.52$$

Find the partial correlation coefficients $r_{12.3}$, $r_{13.2}$, $r_{23.1}$ & multiple correlation coefficient $R_{1.2.3}$.

$$R_{1.2.3}$$

Sol

$$\text{i) } r_{12.3} = \frac{r_{12} - r_{13} r_{23}}{\sqrt{(1-r_{13}^2)(1-r_{23}^2)}} = \frac{0.77 - 0.72 \times 0.52}{\sqrt{(1-(0.72)^2)(1-(0.52)^2)}} \\ = 0.62$$

$$\text{ii) } r_{13.2} = \frac{r_{13} - r_{12} r_{23}}{\sqrt{(1-r_{12}^2)(1-r_{23}^2)}} = \frac{0.72 - (0.77)(0.52)}{\sqrt{(1-(0.77)^2)(1-(0.52)^2)}} \\ = \frac{0.72 - 0.4004}{\sqrt{0.4071 \times 0.7296}} = \frac{0.3196}{0.4655} \\ = 0.6866$$

$$\text{iii) } \gamma_{23.1} = \frac{\gamma_{23} - \gamma_{12} \gamma_{13}}{\sqrt{(1-\gamma_{12}^2)(1-\gamma_{13}^2)}} = \frac{0.52 - 0.77 \times 0.72}{\sqrt{(1-(0.77)^2)(1-(0.72)^2)}} \\ = \frac{-0.0344}{0.3073} = -0.1119$$

$$\text{iv) } R_{1.23}^2 = \frac{\gamma_{12}^2 + \gamma_{13}^2 - 2\gamma_{12}\gamma_{13}\gamma_{23}}{1-\gamma_{23}^2} \\ = \frac{(0.77)^2 + (0.72)^2 - 2(0.77)(0.72)(0.52)}{1-(0.52)^2} \\ = 0.7334$$

$$R_{1.23}^2 = 0.7334$$

$$R_{1.23} = 0.8564 \quad (\text{Since multiple correlation coefficient is non-ve})$$

Formulae

$$\text{or } b_{12.3} = \gamma_{12.3} \frac{\sigma_{1.3}}{\sigma_{2.3}} = \text{regression coefficient of } X_{1.3} \text{ on } X_{2.3}$$

where $\gamma_{12.3} = \frac{\gamma_{12} - \gamma_{13}\gamma_{23}}{\sqrt{(1-\gamma_{13}^2)(1-\gamma_{23}^2)}}$

$$\sigma_{1.3} = \sigma_1 \sqrt{(1 - r_{13}^2)}$$

$$\sigma_{2.3} = \sigma_2 \sqrt{(1 - r_{23}^2)}$$

III^{by} $b_{13.2} = r_{13.2} \frac{\sigma_{1.2}}{\sigma_{3.2}}$ = Coefficient of regression of $x_{2.3}$ on $x_{1.3}$

Where

$r_{13.2}$ = Partial correlation coefficient of x_1 & x_3 keeping x_2 as constant

$$= \frac{r_{13} - r_{12} r_{32}}{\sqrt{(1 - r_{12}^2)(1 - r_{32}^2)}}$$

$$\sigma_{1.2} = \sigma_1 \sqrt{(1 - r_{12}^2)}$$

$$\sigma_{3.2} = \sigma_3 \sqrt{(1 - r_{32}^2)}$$

* $\sigma_{1.23} = \sigma_1 \sqrt{\frac{w}{w_{11}}}$

Solve

- ① In a trivariate distribution $\sigma_1 = 2, \sigma_2 = \sigma_3 = 3, r_{12} = 0.7, r_{23} = r_{31} = 0.5$ then find i) $r_{23.1}$ ii) $R_{1.23}$, iii) $b_{12.3}, b_{13.2}$ & iv) $\sigma_{1.23}$

$$\text{i) } \gamma_{23,1} = \frac{\gamma_{23} - \gamma_{12}\gamma_{13}}{\sqrt{(1-\gamma_{12}^2)(1-\gamma_{13}^2)}} = \frac{0.5 - (0.5)(0.5)}{\sqrt{(1-0.49)(1-0.25)}}$$

$$= 0.2425$$

$$\text{ii) } R_{1,23}^2 = \frac{\gamma_{11}^2 + \gamma_{13}^2 - 2\gamma_{12}\gamma_{13}\gamma_{23}}{1 - \gamma_{23}^2}$$

$$= \frac{0.49 + 0.25 - 2(0.7)(0.5)(0.5)}{1 - 0.25}$$

$$= 0.52$$

$$R_{1,23} = 0.7211$$

$$\text{iii) } b_{12,3} = \gamma_{12,3} \frac{\sigma_{1,3}}{\sigma_{2,3}} \neq b_{13,2} = \gamma_{13,2} \frac{\sigma_{1,2}}{\sigma_{2,2}}$$

$$\gamma_{12,3} = \frac{\gamma_{12} - \gamma_{13}\gamma_{23}}{\sqrt{(1-\gamma_{12}^2)(1-\gamma_{13}^2)}} = 0.6$$

$$\gamma_{13,2} = \frac{\gamma_{13} - \gamma_{12}\gamma_{23}}{\sqrt{(1-\gamma_{12}^2)(1-\gamma_{23}^2)}} = \frac{0.5 - (0.7)(0.5)}{\sqrt{(1-(0.7)^2)(1-(0.5)^2)}}$$

$$= 0.2425$$

$$\sigma_{1,3} = \sigma_1 \sqrt{1 - \gamma_{13}^2} = 2 \sqrt{1 - 0.25} = 0.7320$$

$$\sigma_{2,3} = \sigma_2 \sqrt{(1 - \gamma_{23}^2)} = 3 \sqrt{(1 - 0.25)} = 2.5980$$

$$\sigma_{1,2} = \sigma_1 \sqrt{(1 - \gamma_{12}^2)} = 2 \sqrt{(1 - 0.49)} = 1.4282$$

$$\sigma_{3,2} = \sigma_3 \sqrt{(1 - \gamma_{32}^2)} = 3 \sqrt{(1 - 0.25)} = 2.5980$$

$$b_{12,3} = 0.6 \times \frac{1.7320}{2.5980} = 0.4$$

$$b_{13,2} = 0.2425 \times \frac{1.4282}{2.5980} = 0.1333$$

iv) $\sigma_{1,23} = \sigma_1 \sqrt{\omega/\omega_{11}}$

$$\text{where } \omega = \begin{vmatrix} 1 & \gamma_{12} & \gamma_{13} \\ \gamma_{21} & 1 & \gamma_{23} \\ \gamma_{31} & \gamma_{32} & 1 \end{vmatrix} = 1 - \gamma_{12}^2 - \gamma_{13}^2 - \gamma_{23}^2 + 2\gamma_{12}\gamma_{23}\gamma_{13}$$

$$= 1 - 0.49 - 0.25 - 0.25 + 2(0.7)(0.5)(0.5)$$

$$= 0.36$$

$$\delta \omega_{11} = \text{Minor of } 1 = \begin{vmatrix} 1 & \gamma_{23} \\ \gamma_{23} & 1 \end{vmatrix} = 1 - \gamma_{23}^2$$

$$= 1 - 0.25 = 0.75$$

$$\therefore \sigma_{1,23} = 2 \times \sqrt{\frac{0.36}{0.75}} = 2 \times \sqrt{0.48}$$

$$= 2(0.6928) = 1.3856$$

Relation between partial & Multiple Correlation

Coefficient

$$\star 1 - R_{1.23}^2 = (1 - r_{12}^2)(1 - r_{13.2}^2)$$

* If $R_{1.23} = 0$ then X_1 is uncorrelated with X_2 & X_3

$$\star R_{1.23} \geq r_{12}$$

Linear Regression

* If the Variable in a bivariate distributions are related we will find the point in the Scatter diagram will clustered round some curve called the curve of regression.

* If the curve is a straight line then it is called the line of regression

Formulae

Equation of the line of regression of Y on X

$$Y - \bar{Y} = r \frac{\sigma_y}{\sigma_x} (X - \bar{X})$$

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

Where $b_{yx} = \text{regression Coefficient of } Y \text{ on } X$

$$= r \frac{\sigma_y}{\sigma_x}$$

$r = \text{Correlation coefficient}$

$\sigma_y = \text{S.D of } Y, \sigma_x = \text{S.D of } X.$

Evaluation of the line of Regression of X on Y :

$$(x - \bar{x}) = r \cdot \frac{\sigma_x}{\sigma_y} (y - \bar{y})$$

(or)

$$(x - \bar{x}) = b_{xy} (y - \bar{y})$$

Where $b_{xy} = \text{Regression coefficient of } X \text{ on } Y$

$$= r \cdot \frac{\sigma_x}{\sigma_y}$$

Properties of Regression Coefficients:

i) Correlation coefficient is the geometric mean between the regression coefficients.

ii) i.e., $r = \pm \sqrt{b_{xy} b_{yx}}$

Note:

- ① If both b_{xy} & b_{yx} are +ve then r is +ve.
- ② If both b_{xy} & b_{yx} are -ve then r is -ve.

(ii) If one of the regression coefficients is greater than unity then the other must be less than unity.

i.e., If $b_{xy} > 1$ then $b_{yx} < 1$ vice versa

$$(iii) \left| b_2(b_{xy} + b_{yx}) \right| > |r|$$

iv) Regression coefficients are independent of change of origin but not scale.

Angle between two lines of regression:

$$\theta = \tan^{-1} \left\{ \frac{1 - r^2}{|r|} \left(\frac{\sigma_x \sigma_y}{\sigma_x^2 + \sigma_y^2} \right)^{\frac{1}{2}} \right\}$$

Note

① If $r=0$ then $\theta = \pi/2$

i.e., The lines of regression become \perp^r to each other

② If $r=\pm 1$ then $\theta = 0$ (or) 0°

i.e., The two lines are \parallel^l or coincide.

But both lines of regression are passing through ~~(x̄, ȳ)~~ (\bar{x}, \bar{y}) . They cannot be \parallel^l .

\therefore They coincide.

Sums

- ① In a partially destroyed laboratory, result of an analysis of correlation data, the following results are legible:

Variance of $x = 9$, Regression equations :

$$8\bar{x} - 10\bar{y} + 66 = 0 \quad ; \quad 40\bar{x} - 18\bar{y} = 214$$

- What are i) the mean values of x & y
 ii) The correlation coefficient between x & y
 & iii) The standard deviation of y ?

Sol

Since both lines of regression passes through (\bar{x}, \bar{y})

$$\therefore 8\bar{x} - 10\bar{y} + 66 = 0 \quad \text{--- (1)}$$

$$40\bar{x} - 18\bar{y} = 214 \quad \text{--- (2)}$$

$$\textcircled{1} \quad x \Rightarrow \quad 40\bar{x} - 50\bar{y} = -330$$

$$\textcircled{2} \quad \Rightarrow \quad \underbrace{\begin{array}{rcl} 40\bar{x} - 18\bar{y} & = & 214 \\ (+) & & (-) \end{array}}_{\cancel{40\bar{x}}} \quad \cancel{- 32\bar{y}} = 214$$

$$+ 32\bar{y} = 214$$

$$\bar{y} = \frac{544}{32} = 17$$

Put $\bar{y} = 17$ in (1)

$$8\bar{x} = -66 + 170$$

$$\bar{x} = \frac{104}{8} = 13$$

$$\therefore (\bar{x}, \bar{y}) = (13, 17)$$

ii) Let

Regression of y on x

$$8x - 10y + 66 = 0$$

$$y = 8/10 x + 66/10$$

Regression of x on y

$$40x - 18y = 214$$

$$x = 18/40 y + 214/40$$

$$\begin{aligned} \therefore b_{yx} &= \text{Regression coefficient of } y \text{ on } x \\ &= 8/10 = 4/5 \end{aligned}$$

$$\begin{aligned} b_{xy} &= \text{Regression coefficient of } x \text{ on } y \\ &= 18/40 = 9/20 \end{aligned}$$

$$\begin{aligned} \therefore r^2 &= b_{xy} \cdot b_{yx} = 4/5 \cdot 9/20 \\ &= 9/25 \end{aligned}$$

$$r = \pm 3/5 = \pm 0.6$$

Since b_{xy} & b_{yx} are +ve

$$\therefore r = 0.6$$

iii) we know that

$$b_{yx} = r \cdot \frac{\sigma_y}{\sigma_x}$$

$$\gamma_{xs} = \beta_{xs} \times \frac{\sigma_y}{\beta}$$

$$\sigma_y = 4$$