

TUTORIAL-6

1. The two lines of regression are $3x + 2y = 26$, $6x + y = 31$.

Find (\bar{x}, \bar{y}) , r_{xy} and where $\sigma_x = 1$. find σ_y .

Sol: Since the regression lines always passes through (\bar{x}, \bar{y}) .

$$3\bar{x} + 2\bar{y} = 26$$

$$6\bar{x} + \bar{y} = 31$$

on solving $\bar{x} = 4$, $\bar{y} = 7$

Let us assume $3\bar{x} + 2\bar{y} = 26$ is the regression of y on x .

$$2\bar{y} = -3\bar{x} + 26$$

$$\bar{y} = -\frac{3}{2}\bar{x} + \frac{26}{2}$$

$$\Rightarrow b_{yx} = -\frac{3}{2}$$

Let us assume $6\bar{x} + \bar{y} = 31$ is the regression of x on y

$$6\bar{x} = -\bar{y} + 31$$

$$\bar{x} = -\frac{1}{6}\bar{y} + \frac{31}{6}$$

$$\Rightarrow b_{xy} = -\frac{1}{6}$$

$$r_{xy} = \pm \sqrt{b_{xy} \times b_{yx}}$$

$$= \pm \sqrt{-\frac{1}{6} \times -\frac{3}{2}}$$

$$= \pm \sqrt{\frac{3}{12}} = \pm \sqrt{\frac{1}{4}} = \pm \sqrt{0.25} = \pm 0.5$$

$$r_{xy} = -0.5$$

$$\sigma_x = 1, \sigma_y = ?$$

W.K.T. $b_{xy} = r \frac{\sigma_x}{\sigma_y}$

$$\frac{1}{6} = -0.5 \times \frac{1}{\sigma_y}$$

$$\frac{1}{6} = \frac{0.5}{\sigma_y}$$

$$\sigma_y = 6 \times 0.5$$

$$\sigma_y = 3$$

2. Prove that Correlation coefficient lies between $(-1, 1)$.

i.e. T.P. $-1 \leq r_{xy} \leq 1$

Proof: Let $E[a(x-\bar{x}) + (y-\bar{y})]^2 \geq 0$

$$a^2 E[x-\bar{x}]^2 + E(y-\bar{y})^2 + 2a E(x-\bar{x})(y-\bar{y}) \geq 0$$

$$a^2 \sigma_x^2 + \sigma_y^2 + 2a \text{Cov}(x, y) \geq 0$$

The above eqn ≥ 0 only when $B^2 - 4AC \leq 0$

$$\text{i.e. } [2 \text{Cov}(x, y)]^2 - 4 \sigma_x^2 \sigma_y^2 \leq 0$$

$$4 [\text{Cov}(x, y)]^2 \leq 4 \sigma_x^2 \sigma_y^2$$

$$\text{Cov}^2 xy \leq \sigma_x^2 \sigma_y^2$$

$$\frac{\text{Cov}^2 xy}{\sigma_x^2 \sigma_y^2} \leq 1$$

$$r^2_{xy} \leq 1$$

$$|r_{xy}| \leq 1$$

$$-1 \leq r_{xy} \leq 1$$

Hence proved.

3. Obtain the lines of regression from the following data, also estimate the value of y when $x = 38$ and value of x when $y = 18$.

x : 22 26 29 30 31 31 34 35

y : 20 20 21 29 27 24 27 31

Sol:

$$\bar{x} = \frac{\sum x}{n} = \frac{238}{8} = 29.75 = E(x)$$

$$\bar{y} = \frac{\sum y}{n} = \frac{199}{8} = 24.875 = E(y) \quad E(xy) = 752.375$$

$$\begin{aligned} \sigma_x &= \sqrt{E(x^2) - (E(x))^2} = \sqrt{900.5 - (29.75)^2} \\ &= \sqrt{900.5 - 885.0625} = \sqrt{15.4375} = 3.929 \end{aligned}$$

$$\begin{aligned} \sigma_y &= \sqrt{E(y^2) - (E(y))^2} = \sqrt{634.625 - (24.875)^2} \\ &= \sqrt{634.625 - 618.756} = \sqrt{15.869} = 3.983 \end{aligned}$$

$$r_{xy} = \frac{E(xy) - E(x)E(y)}{\sigma_x \sigma_y}$$

$$= \frac{(752.875) - (29.75)(24.875)}{(3.929)(3.983)}$$

$$= \frac{752.875 - 740.031}{15.808}$$

$$= \frac{12.844}{15.808}$$

$$r_{xy} = 0.8125$$

Regression of X on Y is

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

$$X - 29.75 = (0.8125) \frac{3.929}{3.983} (Y - 24.875)$$

$$X - 29.75 = 0.8015 (Y - 24.875)$$

$$X - 29.75 = 0.8015 Y - 19.937$$

$$X - 0.8015 X = 29.75 - 19.937$$

$$X - 0.8015 X = 9.813 \rightarrow \textcircled{1}$$

Regression of Y on X is

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

$$Y - 24.875 = (0.8125) \frac{3.983}{3.929} (X - 29.75)$$

$$Y - 24.875 = (0.8125)(1.013)(X - 29.75)$$

$$Y - 24.875 = 0.823 (X - 29.75)$$

$$Y - 24.875 = 0.823 X - 24.484$$

$$Y - 0.823 X = 24.875 - 24.484 = 0.391$$

$$Y - 0.823 X = 0.391 \rightarrow \textcircled{2}$$

To find X when Y = 18

put Y = 18 in $\textcircled{1}$

$$X - 0.8015(18) = 9.813$$

$$X - 14.427 = 9.813$$

$$X = 9.813 + 14.427$$

$$X = 24.24$$

To find Y when X = 38

put X = 38 in $\textcircled{2}$

$$Y - 0.823 X = 0.391$$

$$Y - 0.823(38) = 0.391$$

$$Y - 31.274 = 0.391$$

$$Y = 0.391 + 31.274$$

$$Y = 31.665$$

4. Given that x and y are independent random variables which is given by $f(x) = \begin{cases} 4ax, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$, $f(y) = \begin{cases} 4by, & 0 < y < 1 \\ 0, & \text{otherwise} \end{cases}$

Prove that $u = x+y$, $v = x-y$ are uncorrelated.

Sol: u and v are uncorrelated $\Rightarrow \text{Cov}(u, v) = 0$.

$$\begin{aligned} \text{Let } \text{Cov}(u, v) &= E(uv) - E(u)E(v) \\ &= E[(x+y)(x-y)] - E(x+y)E(x-y) \\ &= E(x^2 - y^2) - [E(x) + E(y)][E(x) - E(y)] \\ &= E(x^2 - y^2) - [E(x)^2 - E(y)^2] \end{aligned}$$

To find a

w.t.t $\int_{-\infty}^{\infty} f(x) dx = 1$

$$\int_0^1 4ax dx = 1$$

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

$$2a(1-0) = 1$$

$$2a = 1$$

$$a = \frac{1}{2}$$

To find b .

$$\int_{-\infty}^{\infty} f(y) dy = 1$$

$$\int_0^1 4by dy = 1$$

$$4b \left(\frac{y^2}{2} \right)_0^1 = 1$$

$$2b(1-0) = 1$$

$$2b = 1$$

$$b = \frac{1}{2}$$

$$f(x) = \begin{cases} 4\left(\frac{1}{2}\right)x, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases} \Rightarrow f(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$f(y) = \begin{cases} 4\left(\frac{1}{2}\right)y, & 0 < y < 1 \\ 0, & \text{otherwise} \end{cases} \Rightarrow f(y) = \begin{cases} 2y, & 0 < y < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$\text{Let } E(x) = \int_{-\infty}^{\infty} xf(x) dx = \int_0^1 x(2x) dx = 2 \int_0^1 x^2 dx = 2 \left(\frac{x^3}{3} \right)_0^1$$

$$E(x) = \frac{2}{3}$$

$$E(x^2) = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_0^1 x^2 (2x) dx = 2 \int_0^1 x^3 dx = 2 \left(\frac{x^4}{4} \right)_0^1$$

$$E(x) = \frac{1}{2}$$

$$\text{Let } E(y) = \int_{-\infty}^{\infty} y f(y) dy = \int_0^1 y(2y) dy = \frac{2}{3} \left(\frac{y^3}{3} \right)_0^1 = 2 \left(\frac{y^3}{3} \right)_0^1$$

$$E(y) = \frac{2}{3}$$

$$E(y^2) = \int_{-\infty}^{\infty} y^2 f(y) dy = \int_0^1 y^2(2y) dy = 2 \left(\frac{y^4}{4} \right)_0^1$$

$$E(y^2) = \frac{1}{2}$$

$$\therefore \text{Cov}(u, v) = \left(\frac{1}{2} - \frac{2}{3} \right) - \left[\left(\frac{2}{3} \right)^2 - \left(\frac{2}{3} \right)^2 \right] = 0.$$

$$r_{uv} = \frac{\text{Cov}(u, v)}{\sigma_x \sigma_y} = 0$$

$$r_{uv} = 0$$

Hence u & v are uncorrelated.

5. Find the angle between two lines of regression.

$$\text{w.k.t } \theta = \tan^{-1} \left(\frac{m_2 - m_1}{1 + m_1 m_2} \right)$$

Target equation of the straight line is

$$(y - y_1) = \frac{-1}{m} (x - x_1)$$

Let the regression of y on x is

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

$$\frac{-1}{m} = r \frac{\sigma_y}{\sigma_x}$$

$$\Rightarrow m_1 = -\frac{\sigma_x}{r \sigma_y}$$

Let the regression of x on y is

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

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

$$\Rightarrow \frac{-1}{m} = \frac{\sigma_y}{r\sigma_x}$$

$$\Rightarrow m_2 = \frac{-r\sigma_x}{\sigma_y}$$

$$\theta = \tan^{-1} \left[\frac{-\frac{r\sigma_x}{\sigma_y} + \frac{\sigma_x}{r\sigma_y}}{1 + \frac{\sigma_x}{r\sigma_y} \cdot \frac{r\sigma_x}{\sigma_y}} \right]$$

$$= \tan^{-1} \left[\frac{-r^2\sigma_x + \sigma_x}{r\sigma_y} \cdot \frac{\sigma_y}{\sigma_y^2 + \sigma_x^2} \right]$$

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

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

Yes