

**Advanced Engineering Mathematics**  
**Prof. P. N. Agrawal**  
**Department of Mathematics**  
**Indian Institute of Technology - Roorkee**

**Lecture – 56**  
**Correlation and Regression - II**

Hello friends, welcome to my lecture; second lecture on correlation and regression, the correlation coefficient is independent of the change of origin and scale, we shall prove this result.

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**Theorem 1**

Correlation coefficient is independent of the change of origin and scale.

**Proof**

Let  $U = \frac{X-a}{h}$  and  $V = \frac{Y-b}{k}$  where  $a, b, h, k \in \mathbb{R}$  with  $h > 0$  and  $k > 0$ . We wish to show that

$$\rho(X, Y) = \rho(U, V).$$

We have  $X = a + hU$  and  $Y = b + kV$ .

Hence

$$\mu_X = E(X) = E(a + hU) = a + h\mu_U$$

and

$$\mu_Y = E(Y) = E(b + kV) = b + k\mu_V$$

So, let us say, let us define the random variable  $U$  as  $x - a / h$  and the random variable  $V$  as  $y - b$  over  $k$ , where  $x$  and  $y$  are any  $w$  random variables,  $a, b, h, k$  are real numbers here,  $h$  and  $k$  are positive real numbers, okay, we wish to show that the coefficient of correlation of  $xy$  is same as the coefficient of correlation of  $uv$  that means, it is unaffected by the change of origin, okay and the scale.

So, from  $x =$ ; from  $u = x - a / h$ , okay, we find  $x = a + hu$  and from  $V = y - b$  over  $k$ , we find that, the random variable  $y = b + kv$ , hence the mean of  $x$  that is expectation of  $x =$  expectation of  $a + hu$  and so it is  $a$  times expectation of  $1$ ; expectation of  $1$  is  $1$ , so we have  $a + h$  times expectation of  $U$  that is  $\mu_U$ . Similarly, expectation of  $y$   $\mu_Y = E_y =$  expectation of  $b + kV$  which is  $b$  times expectation of  $1$  which is  $1$ .

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Proof cont...

$$\begin{aligned}
 \text{Cov}(X, Y) &= E[(X - \mu_X)(Y - \mu_Y)] \\
 &= E[(\cancel{a} + hU - \cancel{a} - h\mu_U)(\cancel{b} + kV - \cancel{b} - k\mu_V)] \\
 &= E(hk(U - \mu_U)(V - \mu_V)) \\
 &= hk \text{Cov}(U, V)
 \end{aligned}$$

$$\sigma_X^2 = E[(X - \mu_X)^2] = E[(\cancel{a} + hU - \cancel{a} - h\mu_U)^2] = h^2 E[(U - \mu_U)^2]$$

$\sigma_X = h \sigma_U$

and

$$\sigma_Y^2 = E[(Y - \mu_Y)^2] = E[(\cancel{b} + kV - \cancel{b} - k\mu_V)^2] = k^2 E[(V - \mu_V)^2]$$

$\sigma_Y = k \sigma_V$

Hence

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{hk \text{Cov}(U, V)}{(h\sigma_U)(k\sigma_V)} = \rho(U, V).$$

So,  $b + k$  times expectation of  $b$  that is  $\mu_V$ , okay, so we get  $Y = V + k$  times  $\mu_V$ , now, covariance of  $XY$ , covariance of  $XY$  by definition is expectation of  $X - \mu_X * Y - \mu_Y$ ,  $X = a + hU$ ,  $\mu_X$  we have just now seen is  $a + h \mu_U$ , okay, so we put the value of  $\mu_X$  here and  $Y = b + kV$ ,  $\mu_Y$  is  $b + k * \mu_V$ , so let us put their values of  $\mu_X$  and  $\mu_Y$ , what we get here, this  $a$  will cancel with this  $a$ ,  $b$  cancels with this  $b$  and what we get?

$h$  times  $U - \mu_U$   $k$  times  $V - \mu_V$ , okay, so expectation of  $hk$  times  $U - \mu_U * V - \mu_V$ ,  $hk$  is a constant, so it will come out and we get  $hk$  times covariance of  $U, V$  okay. Now,  $\sigma_X^2$  is expectation of  $X - \mu_X$  whole square,  $X = a + hU$ ,  $\mu_X = a + h * \mu_U$ , so this  $a$  and this  $a$  cancel and we get expectation of  $h$  square times  $U - \mu_U$  whole square, so this is  $h$  square times expectation of  $U - \mu_U$  whole square.

And similarly,  $\sigma_Y^2$  is expectation of  $Y - \mu_Y$  whole square which is  $= k^2 * \text{expectation of } V - \mu_V \text{ whole square}$ , okay, so this  $b$  cancels, so this gives you  $k^2$  times expectation of  $V - \mu_V$  whole square. Now,  $\rho_{XY}$  by definition is covariance of  $xy$  divided by  $\sigma_X \sigma_Y$ , covariance of  $xy$  is  $hk * \text{covariance of } uv$  and  $\sigma_X^2 = h^2 \text{ times expectation of } U - \mu_U \text{ whole square}$ .

So,  $\sigma_X = h$  times  $\sigma_U$ , okay and here  $\sigma_Y^2 = k^2 \text{ times } \sigma_V^2$ , so  $\sigma_Y = k$  times  $\sigma_V$ , okay, so we put their values, so  $\sigma_X = h \sigma_U$ ,  $\sigma_Y = k \sigma_V$ .

sigma v, so this hk cancels with this hk here and we get covariance of Uv divided by sigma u sigma b, so covariance of xy is same as so, rho XY is same as rho UV that is the coefficient of correlation is not affected by the change of origin and scale.

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### Theorem 2

The angle between the two regression lines is

$$\theta = \tan^{-1} \left( \frac{1 - \rho^2}{\rho} \frac{\sigma_X \sigma_Y}{\sigma_X^2 + \sigma_Y^2} \right),$$

where  $\rho$  is the coefficient of correlation between X and Y.

**Proof:** The line of regression of Y on X is given by

$$Y - \mu_Y = \frac{\rho \sigma_Y}{\sigma_X} (X - \mu_X)$$

hence slope  $m_1 = \frac{\rho \sigma_Y}{\sigma_X}$ . Similarly, the slope of the line of regression of X on Y

$$X - \mu_X = \frac{\rho \sigma_X}{\sigma_Y} (Y - \mu_Y) \quad \checkmark \quad \begin{aligned} Y - \mu_Y &= \frac{\sigma_Y}{\rho \sigma_X} (X - \mu_X) \\ m_2 &= \frac{\sigma_Y}{\rho \sigma_X} \end{aligned}$$

Now, let us find the angle between the 2 regression lines, the regression line of y on x and the regression line of x on y, if theta is the acute angle between the regression line of y on x and the regression lines of x on y, then theta is given by  $\tan^{-1} \frac{1 - \rho^2}{\rho} \frac{\sigma_X \sigma_Y}{\sigma_X^2 + \sigma_Y^2}$ , where rho is the coefficient of correlation between x and y. Now, the line of regression of y on x we know is given by  $y - \mu_Y = \rho \frac{\sigma_Y}{\sigma_X} (x - \mu_X)$ .

So, here slope of this regression line of y on x is  $\rho \frac{\sigma_Y}{\sigma_X}$  which we denote by  $m_1$  and the similarly the slope of the regression line of x on y we can find, the regression line of x on y is given by  $x - \mu_X = \rho \frac{\sigma_X}{\sigma_Y} (y - \mu_Y)$  and this can be written as  $y - \mu_Y = \frac{\sigma_Y}{\rho \sigma_X} (x - \mu_X)$ , okay, so here  $m_2$ , the slope of this regression line of x on y =  $\frac{\sigma_Y}{\rho \sigma_X}$ , okay.

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Proof cont...

is given by  $m_2 = \frac{1}{\rho} \frac{\sigma_Y}{\sigma_X}$  ✓

Then, the acute angle  $\theta$  is given by

$$\begin{aligned} \tan \theta &= \frac{m_1 \sim m_2}{1 + m_1 m_2} = \frac{\frac{\rho \sigma_Y}{\sigma_X} \sim \frac{1}{\rho} \frac{\sigma_Y}{\sigma_X}}{1 + \frac{\rho \sigma_Y}{\sigma_X} \frac{1}{\rho} \frac{\sigma_Y}{\sigma_X}} \\ &= \frac{(\rho \sim \frac{1}{\rho}) \sigma_Y \sigma_X}{\sigma_X^2 + \sigma_Y^2} \quad | \rho | \leq 1 \\ &= \frac{(1 - \rho^2) \sigma_Y \sigma_X}{\rho (\sigma_X^2 + \sigma_Y^2)} \quad \frac{(\frac{1}{\rho} - \rho) \sigma_Y \sigma_X}{\sigma_X^2 + \sigma_Y^2} = \frac{(1 - \rho^2) \sigma_Y \sigma_X}{\rho (\sigma_X^2 + \sigma_Y^2)} \end{aligned}$$

or

$$\theta = \tan^{-1} \left( \frac{1 - \rho^2}{\rho} \frac{\sigma_X \sigma_Y}{\sigma_X^2 + \sigma_Y^2} \right) \quad \checkmark$$

So,  $m_1$  is  $\rho \sigma_Y / \sigma_X$ ,  $m_2$  is  $\sigma_Y / \rho \sigma_X$ , okay, so this is  $m_2$ , now  $\theta$  is given by  $\tan^{-1} (m_1 - m_2)$ , now this symbol means we consider the difference  $m_1 - m_2$ , if  $m_1$  is  $> m_2$  and we consider the difference  $m_2 - m_1$ , if  $m_2$  is bigger than  $m_1$ , so that means we will always consider the positive sign here okay, so  $\tan \theta = m_1 - m_2$  or  $m_2 - m_1$  divided by  $1 + m_1 m_2$ , you put the values of  $m_1$  and  $m_2$  here okay, what we get?

It simplifies to this  $\rho$ , now  $\rho$  is mod of  $\rho$  is  $< 1 \leq 1$ , so we shall write it as  $1/\rho - \rho$ , okay, this will be written as  $1/\rho - \rho \sigma_Y \sigma_X$  divided by  $\sigma_X^2 + \sigma_Y^2$  and this gives you  $1 - \rho^2$  divided by  $\rho \sigma_X \sigma_Y$  divided by  $\sigma_X^2 + \sigma_Y^2$ , so  $\theta = \tan^{-1} (1 - \rho^2 \text{ divided by } \rho * \sigma_X \sigma_Y \text{ divided by } \sigma_X^2 + \sigma_Y^2)$ .

Now, from here we can see if the random variable  $x$  and  $y$  are uncorrelated that is  $\rho = 0$ , then  $\theta = \pi/2$ ,  $\tan^{-1} \infty$  is  $\pi/2$ , so that means the regression lines of  $y$  on  $x$  and the regression lines of  $x$  on  $y$  will be perpendicular to each other.

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### Example 1

The tangent of the angle between the lines of regression of  $Y$  on  $X$  and  $X$  on  $Y$  is  $0.6$  and  $\sigma_X = \frac{1}{2}\sigma_Y$ . Find the correlation coefficient.

Ans:  $\rho = \frac{1}{2}$ .

Here  $\tan \theta = 0.6$  So

$$\tan \theta = \frac{(1-\rho^2) \sigma_Y \sigma_Y}{\rho (\sigma_X^2 + \sigma_Y^2)} = \left( \frac{1-\rho^2}{\rho} \right) \left( \frac{\frac{1}{4} \sigma_Y^2}{\frac{1}{4} \sigma_Y^2 + \sigma_Y^2} \right)$$

$$\Rightarrow 0.6 = \left( \frac{1-\rho^2}{\rho} \right) \left( \frac{\frac{1}{4}}{\frac{1}{4} + 1} \right) \Rightarrow \frac{1-\rho^2}{\rho} = \frac{6}{5} \Rightarrow \frac{1-\rho^2}{\rho} = \frac{3}{2}$$

$\sin \theta$   
 $|\rho| \leq 1$   
No  $\rho = -2$  not possible  
Hence  $\rho = \frac{1}{2}$

$$2 - 2\rho^2 = 3\rho$$

$$\Rightarrow 2\rho^2 + 3\rho - 2 = 0 \quad \text{or} \quad 2\rho^2 + 4\rho - \rho - 2 = 0$$

$$2\rho(\rho+2) - 1(\rho+2) = 0 \Rightarrow \rho = \frac{1}{2}, -2$$

So, when  $x$  and  $y$  are uncorrelated,  $\rho = 0$  that is the 2 regression lines are perpendicular to each other, when  $x$  and  $y$  are perfectly correlated that is  $\rho = +1$ , what we get here;  $1 - \rho^2$  becomes 0, so  $\theta = 0$  that is the 2 regression lines coincide. Now, let us do this problem, the tangent of the angles between the lines of regression of  $y$  on  $x$  and  $x$  on  $y$  is  $0.6$ , okay. So, here  $\tan \theta = 0.6$ , okay, so  $\tan \theta = \frac{1 - \rho^2}{\rho} \cdot \frac{\sigma_X^2}{\sigma_X^2 + \sigma_Y^2}$  give you  $1 - \rho^2$  divided by  $\rho \cdot \sigma_X^2 = \frac{1}{2} \sigma_Y^2$ .

So,  $\frac{1}{2} \sigma_Y^2$  and divide by  $\frac{1}{4} \sigma_Y^2 + \sigma_Y^2$  and this is  $=$ ; this gives you  $\tan \theta = 0.6$ , so  $0.6 = \frac{1 - \rho^2}{\rho}$  and here we get  $\frac{1}{2} \cdot \frac{4}{5}$ , okay, so this implies  $1 - \rho^2$  divided by  $\rho = \frac{6}{5}$ , okay, so  $\frac{3}{2}$ , so we get  $2 - 2\rho^2 = 3\rho$ , which implies  $2\rho^2 + 3\rho - 2 = 0$  and this can be written as  $2\rho^2 + 4\rho - \rho - 2 = 0$ , so we get  $2\rho(\rho+2) - 1(\rho+2) = 0$ .

So, we get the 2 roots as  $\rho = \frac{1}{2}$  and  $-2$ , okay, now since  $-\rho$  is bounded by 1, okay, mod of  $\rho$  is  $\leq 1$ , so  $\rho = -2$ , not possible and hence  $\rho$  must be  $= \frac{1}{2}$  okay, so the value of  $\rho$  is  $\frac{1}{2}$ , okay.

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$\beta_{yx} = \frac{\rho \sigma_y}{\sigma_x} = 2$       $\beta_{xy} = \frac{\rho \sigma_x}{\sigma_y} = \frac{1}{5}$       $\mu_y = 2 \times \left(\frac{19}{5}\right) - 3 = \frac{20}{5} - 3 = \frac{20-15}{5} = \frac{5}{5} = 1$

**Example 2**

If  $y = 2x - 3$  and  $y = 5x + 7$  are the two regression lines, find

- the mean values of  $X$  and  $Y$ ,
- the correlation coefficient between  $X$  and  $Y$ ,
- find an estimate of  $X$  when  $Y = 1$ .

**Ans:**

(i)  $\mu_x = \frac{10}{3}$ ,  $\mu_y = \frac{29}{3}$

(ii)  $\rho = \sqrt{\frac{2}{5}} = 0.6325$

(iii)  $x = -\frac{6}{5}$

*Handwritten notes:*

Assume that  $y = 2x - 3$  is the regression line of  $y$  on  $x$  and  $y = 5x + 7$  is the regression line of  $x$  on  $y$ .

then  $\beta_{yx} = 2$  and  $\beta_{xy} = \frac{1}{5}$

we know that the two regression lines pass through  $(\mu_x, \mu_y)$

So  $\mu_y = 2\mu_x - 3$  &  $\mu_y = 5\mu_x + 7 \Rightarrow 5\mu_x + 7 = 2\mu_x - 3 \Rightarrow 3\mu_x = -10 \Rightarrow \mu_x = -\frac{10}{3}$

then  $\mu_y = 2 \times \left(-\frac{10}{3}\right) - 3 = -\frac{20}{3} - 3 = -\frac{29}{3}$

then  $\rho = \sqrt{\frac{\beta_{yx} \beta_{xy}}{1}} = \sqrt{\frac{2 \times \frac{1}{5}}{1}} = \sqrt{\frac{2}{5}} = 0.6325$

Now, let us take another example, if  $y = 2x - 3$  and  $y = 5x - 7$  are the 2 regression lines, find the mean value of  $X$  and  $Y$ , the correlation coefficient between  $X$  and  $Y$ , find and estimate of  $X$ , when  $Y = 1$ , okay. Now, sometimes they are given the problem where the; we do not know which one is regression line of  $y$  on  $x$  and which one is the regression line of  $x$  on  $y$ , so we can choose any line as the regression line of  $y$  on  $x$ .

Then the other will be regression line of  $x$  on  $y$  and then we shall find the  $\rho$  from there, okay, if the value; we shall find the  $\beta_{xy}$  and  $\beta_{yx}$ , the product of  $\beta_{xy}$  and  $\beta_{yx}$  should be =  $\rho^2$  and  $\rho$  is  $\leq 1$ , so if  $\rho^2 \leq 1$ , we get; then our choice is correct, if  $\rho^2$  comes out to be  $> 1$  then we will have to our assumption is wrong, so we will have to then take the other line as the regression line of  $y$  on  $x$  and (( )) (11:41).

So, we will change the choice of regression line of  $y$  on  $x$ , so here we are not told which one is the regression line of  $y$  on  $x$  and which one is the regression line of  $x$  on  $y$ , so let us assume that  $y = 2x - 3$  is the regression line of  $y$  on  $x$  and  $y = 5x + 7$  or we can say  $x = \frac{y - 7}{5}$  that is to say  $\frac{1}{5}$  times  $y - 7$ , is the regression line of  $x$  on  $y$ , okay then  $\beta_{xy}$ ;  $\beta_{xy}$  is the regression coefficient of  $y$  on  $x$ , okay, so this will be = 2, okay.

$Y = 2$  times, I can write it as 2 times  $x - \frac{3}{2}$ , okay, so this 2 gives me  $\beta_{xy}$  and similarly  $\beta_{yx}$ , no this is the regression coefficient of  $y$  on  $x$ , so I will write it as  $\beta_{yx}$ , okay  $\beta_{yx}$  is 2

and  $\beta_{xy}$ , okay,  $x - \bar{x}$ ,  $x - \mu_x = \beta_{xy} * y - \mu_y$ , okay, so  $1/5$  is  $\beta_{xy}$ , so now we know that  $\beta_{yx} * \beta_{xy} = \rho^2$ , okay. So, here what do we get;  $2 * 1/5$ , so this is  $2 * 1/5$ , okay, which is  $< 1$ , okay.

So, our choice is correct, okay, so this is regression line of  $y$  on  $x$ , this is regression line of  $x$  on  $y$  and this also gives me the value of  $\rho$ ,  $\rho = \sqrt{2/5}$ , okay, so this is the value of  $\rho$ ,  $\rho = \sqrt{2/5}$ . Now, let us find the mean values of  $X$  and  $Y$ , okay, so mean values of  $x$  and  $y$ ,  $\mu_x$  we have to determine, okay, we know that the 2 regression lines both pass through the point  $\mu_x, \mu_y$ , okay.

So,  $\mu_x, \mu_y$ , so  $\mu_y = 2\mu_x - 3$  and  $\mu_y = 5\mu_x + 7$ , okay, so this will give you what; let us solve these 2 equations, what we will have;  $5\mu_x + 7 = 2\mu_x - 3$ , okay, so this will give you  $3\mu_x = -10$ , okay, this will become  $3\mu_x = -10$ , okay, so this gives me  $\mu_x = -10/3$ , okay and  $\mu_y$  is;  $\mu_y$  is then, what will be  $\mu_y$ ;  $\mu_y = 2\mu_x$ , so  $2 * -10/3 - 3$ , this comes out to be  $-20/3 - 3$ , so  $-20/3 - 9/3$ ,  $-29/3$ .

What is wrong, oh, okay, okay, okay, let me do it again, we have to take  $\rho$  to be; now see we have to take  $\beta_{yx}$ , okay, this is  $\sqrt{2/5}$ , a square root  $2/5$  is  $0.6325$ , okay, this was  $+ - 0.6325$ , okay. Now, our problem is whether we should take positive sign or negative sign, okay, so  $\beta_{yx} =$ ;  $\beta_{yx}$  when written in terms of  $\rho$ , what is the formula for that;  $\beta_{yx} = \frac{\rho \sigma_y}{\sigma_x}$ , yeah,  $\rho \sigma_y$  over  $\sigma_x$ , okay.

And so here, since  $\beta_{yx} = 2$ , okay, so  $\rho$  must be positive okay,  $\beta_{xy}$  also,  $\beta_{xy} = \rho$  times  $\sigma_x$  over  $\sigma_y$ , okay and this is the given to be  $= \beta_{yx}$  is given to be  $2$ , okay,  $\beta_{yx}$  given to be  $1/5$ , okay, so since  $\beta_{xy}$  and  $\beta_{yx}$  are both positive, so  $\rho$  must be positive and therefore, we take the value of  $\rho$  as  $0.6325$ , okay, we choose this sign as positive, because  $\beta_{yx}$  and  $\beta_{xy}$  are positive, okay.

Now, we know that the 2 regression lines are pass through  $\mu_x, \mu_y$ , so this gives you  $\mu_y = 2\mu_x - 3$  and here  $\mu_y = 5\mu_x + 7$ , so  $5\mu_x + 7 = 2\mu_x - 3$ ,  $5\mu_x + 7 = 2\mu_x - 3$ , so  $3\mu_x = -10$ , this is  $-10$  here, so  $\mu_x = -10/3$ , okay, so  $\mu_x = -10/3$  and  $\mu_y = 2 * -10/3 - 3 = -20/3 - 9/3 = -29/3$ .

-  $20/3 - 3$ , so that gives me  $-29/3$ , okay, so this is  $\mu_y$ . Now, let us determined, find and estimate of  $x$ , when  $y = 1$ .

So, we have to find and estimate of  $y$ , estimate of  $x$  when  $y$  is given that means we use the regression line of  $x$  on  $y$ , regression line of  $x$  on  $y$  is this, okay,  $y = 5x + 7$ , so when  $y = 1$ , okay, we have  $y = 1 = 5x + 7$ , so  $x = -6/5$ , okay, so we get  $x = -6/5$ , the value of  $x$  when  $y$  is given is found from the regression line of  $x$  on  $y$ .

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### Example 3

Two random variable  $X$  and  $Y$ , are defined as  $Y = 4X + 9$ . Find the coefficient of correlation between  $X$  and  $Y$ .

Ans:  $\rho = 1$ .

$$\begin{aligned}
 Y &= 4X + 9 \\
 E(Y) &= \mu_y = E(4X + 9) = 4\mu_x + 9 \checkmark \\
 \text{Var}(Y) &= E((Y - \mu_y)^2) = E((4X + 9 - 4\mu_x - 9)^2) \\
 &= 16 E((X - \mu_x)^2) = 16 \text{Var}(X) \Rightarrow \sigma_y = 4\sigma_x \\
 E(XY) &= E(X(4X + 9)) = E(4X^2 + 9X) = 4E(X^2) + 9\mu_x \\
 \text{Now, } \rho &= \frac{E(XY) - E(X)E(Y)}{\sigma_x \sigma_y} = \frac{4E(X^2) + 9\mu_x - \mu_x(4\mu_x + 9)}{\sigma_x \cdot (4\sigma_x)} = \frac{4E(X^2) - 4\mu_x^2}{4\sigma_x^2} = 1
 \end{aligned}$$

Now, let me go to another question, 2 random variables  $X$  and  $Y$  are defined as  $Y = 4X + 9$ , find the coefficient of correlation between  $X$  and  $Y$ , okay, so coefficient of correlation  $Y = 4X + 9$ , okay, so let us first find the expected value of  $Y$ , okay, expected value of  $Y$  is  $\mu_y$ , so this = expected value of  $4X + 9$ , okay, so this is = 4 times expected value of  $X$  that is  $\mu_x$  + expected value of 9, which is 9, okay.

So, expected value of  $y$  is  $4 \mu_x + 9$ , okay, variance of  $Y$ , okay, let me write directly the variance of  $Y$ , so variance of  $Y$ , okay, variance of  $Y$  is expectation of  $Y - \mu_y$  whole square, okay, so expectation of  $Y = 4X + 9$ , okay -  $\mu_y$ ,  $\mu_y = 4 \mu_x + 9$ , so we get this, okay, so we cancel, this will cancel with this and we get 4 square, 4 square means 16, so 16 times variance of expectation of  $X - \mu_x$  whole square.



So, 16 times variance of X, thus we get  $\text{Mu } Y = 4 \text{ times Mu } X + 9$  are variance of Y = this, so this implies  $\sigma_Y = 4 \text{ times } \sigma_X$ , okay. Now, expectation of we need to find covariance, so expectation of X Y, so expectation of X Y is expectation of  $X * 4X + 9$ , so this is 4 times expectation of X square, so expectation of  $4X^2 + 9X$ , so this is 4 times expectation of X square + 9 times Mu X, okay. Now,  $\rho = \frac{EXY - EX EY}{\sigma_X \sigma_Y}$  divided by  $\rho = \frac{EXY - EX EY}{\sigma_X \sigma_Y}$ , okay.

So, EXY is  $4EX^2 + 9 \text{ Mu } X - \text{expectation of } EX$ ; expectation of  $X * \text{expectation of } Y = 4 \text{ times Mu } X + 9$ , 4 times; expectation of X we can write as Mu X, so this is Mu X, okay and then I get here 4 times Mu X + 9, okay divided by  $\sigma_X$  and  $\sigma_Y$  is 4  $\sigma_X$ , okay, now what is this; so, 9 Mu X will cancel with 9 Mu X, what we get here, this = 4 times  $EX^2 - \text{Mu } X^2$  whole square.

So, we get variance of X that is 4 times  $\sigma_X^2$ , okay, 9 Mu X, 9 Mu X cancel and we get here  $4/\sigma_X^2$ , so this cancels with this and this cancels with this and we get 1, okay, so  $\rho = 1$ .

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$\sigma_X = \sqrt{\frac{\sum x^2}{n} - \left(\frac{\sum x}{n}\right)^2} = \sqrt{\frac{150}{10} - \left(\frac{10}{10}\right)^2} = \sqrt{14}$      $\sigma_Y = \sqrt{\frac{\sum y^2}{n} - \left(\frac{\sum y}{n}\right)^2}$

**Example 4**

From the given data find,

(i) the two regression equations     $\mu_X = A + \frac{\sum x}{n} = 31 + \frac{10}{10} = 32 = \sqrt{408 - \left(\frac{10}{10}\right)^2}$

(ii) the coefficient of correlation between the marks in Economics and Statistics     $\mu_Y = B + \frac{\sum y}{n} = 41 - \frac{10}{10} = 38 = \sqrt{488 - 9} = \sqrt{39.8}$

(iii) the most likely marks in Statistics when the marks in Economics are 30.     $\text{Cov}(X, Y) = \frac{\sum xy}{n} - \left(\frac{\sum x}{n}\right)\left(\frac{\sum y}{n}\right) = \frac{123}{10} - \left(\frac{10}{10}\right)\left(\frac{38}{10}\right) = 12.3 - 3.8 = 8.5$

Marks in Economics (X)	25	28	35	32	31	36	29	38	34	32
Marks in Statistics (Y)	43	46	49	41	36	32	31	30	33	39

**Ans:**

(i)  $y = -0.6643x + 59.2570$  and  $x = -0.2337y + 40.8806$

(ii)  $\rho = -0.3940$

(iii) 39.3286.

*Handwritten calculations:*

$\sum x = 150$      $\sum y = 380$      $\sum x^2 = 150$      $\sum y^2 = 1488$      $\sum xy = 1230$

$\bar{x} = 31$      $\bar{y} = 38$      $\sum (x - \bar{x})^2 = 140$      $\sum (y - \bar{y})^2 = 100$      $\sum (x - \bar{x})(y - \bar{y}) = -123$

$\sigma_X = \sqrt{14}$      $\sigma_Y = \sqrt{10}$      $\rho = \frac{-123}{\sqrt{140} \sqrt{100}} = -0.3940$

Now, let us take the last problem from the given data, find the 2 regression equations; regression lines of y on x and regression line of x on y, the coefficient of correlation between the marks in economics and statistics is are given in this table, the most likely marks and statistics, when the

marks in economics are 30, okay, so let us say we are given x and y, okay, x is 25, okay, 28, 35, we have 32, okay, 32, 36, then we have 29, 38, okay, 34 and 32, okay.

And y values are 33, 46, 49, 41, 36, 32, 31, 30, 33, 39, okay, let us since the coefficient of correlation is independent, it does not change with the change in origin and scale, let us shift the origin, so let me write here,  $X_i = \text{say } x - 31$ , okay and  $y = \text{eta} = y - 41$ , okay, so if we do that then  $X_i =$ ; so we are subtracting 31 here, so -6, okay, this will be -31 means -3, 31 we subtract from 35, we get 4, 31 we subtract here, we get 1, here we get 0, here we get 31 subtracted gives 5, here we get -2, here we get 7, here we get 3, here we get 1, okay.

And eta; eta will be when we are subtracting 41, so 2 and then here 5, here we get 8, here we get 0, okay, here we get -5, here we get -9, here we get -10, here we get -11, here we get -8 and here we get -2, okay, so what is  $\sum X_i$ , let us find. So, -6, -3, -9, -9 + 4; -5, -5 + 1; -4, -4 + 5; +1, +1 -2 is -1, -1 + 7 is +6, 6 and 3, 9 and 1, 10, so  $\sum X_i$  is 10,  $\sum \text{eta} = 2 + 5; 7, 7 + 8; 15, 15 - 5$  is 10, 10 - 9 is 1, 1 - 10 is -9, -9 - 11; -20, -20 - 8; -28 - 2; -30, so we get -30, okay.

Now, then we need to get  $X_i \text{ eta}$ , okay, so -6 \* 2 is -12, then -3 \* 5 is -15, then we get 4 \* 8, 32, okay, we get 1 \* 0; 0, 0 \* -5; 0, 5 \* -5; -25, -2 \* 10; -10, + 20, 7 \* -11; -77 and then we get 3 \* -8; -24, then 1 \* -2 is -2, okay, we also need to find  $X_i^2$  square, okay, so  $X_i^2$  square is 36, okay, then we have -3 square is 9, 4 square is 16, 1 square is 1, 0 square 0, then we have 25, then we have 4, then we have 49 and then we have here 9 and then we have 1, okay.

And we need to find  $\text{eta}^2$  square, so 4, then we have 25, then we have 64, then we have 0, we have 25, we have 31, okay, -9 square, -10 square is 100 and then we get 121 and then we get here 64 and here we get 4, okay. Now, what is  $\sum X_i^2$  square, okay,  $\sum X_i^2$  square is 36 + 9; 45, 45 + 16 is 61, okay, so we get 61 then 1, 62, 62 + 25; 62 + 25 means 87, okay, 87 + 4; 91, 91 + 49; so we get here 140, then 10, 150.

So, we get 150 here and  $\sum \text{eta}^2$  square we can find, okay, so this is 4 + 25; 29, 29 + 64, so we get here 93, 93 + 25, we get here 118, 118 + 81, we get here 199, okay, then 100, so we get 299, okay, 299 and then we need to add 121, 64, 4; 121 + 64 + 4; 68 we add here, so this gives me

189, okay, so 189, so 189 we add to 299, okay, so 299 we add, okay, this comes out 488, okay, so we have the values here, now  $\sum X_i$  also we have to find.

So, we have here -12, -15, -27, -27 + 32; + 5, + 5 - 45; - 40, -40 + 20; -20, -20 -77; -97, -97 -24 and also we have -2, so 123, okay, so - 123, okay now let us find; first we find  $\mu_X$ , okay,  $\mu_X = \text{assumed mean}$ , okay,  $A + \frac{\sum X_i}{n}$ , A is the assumed mean, assumed mean = 31, okay, so 31 and  $\sum X_i = 10$ ; 10/10 because there are 10 values, okay, so this is 32 and  $\mu_Y$ ;  $\mu_Y = \text{assumed mean}$ , assumed mean is 41 here, so  $B + \frac{\sum Y_i}{n}$ , okay.

$B = 41$  and  $\sum Y_i = -30$ , so -30 divided by 10, so this is -3 that means -38, okay then we need to find variance  $\sigma_X$ ,  $\sigma_Y$ , so  $\sigma_X$ , okay,  $\sigma_X$  is given by square root  $\frac{\sum X_i^2}{n} - \left(\frac{\sum X_i}{n}\right)^2$ , okay, so this is square root  $\frac{150}{10} - \left(\frac{10}{10}\right)^2$ , 15 - 1 that is 14, okay.

And  $\sigma_Y$  similarly = square root  $\frac{\sum Y_i^2}{n} - \left(\frac{\sum Y_i}{n}\right)^2$ , okay, so this  $\sigma_Y = \text{then } \frac{488}{10} - 9$ , so 48.8 - 9, okay, so this is square root 39.8, okay, so this is the value of  $\sigma_Y$ , okay. Now, we have to find the 2 regression equations.

We have the value of  $\mu_X$ ,  $\mu_Y$ , okay so, we need to find the value of coefficient of correlation also we need to find and the 2 regression lines are given by  $\beta_{YX}$  and  $\beta_{XY}$ , okay, so we need to find, okay, so, okay,  $E_{XY}$ , okay, the covariance also we can find, covariance of  $X_i Y_i$ , okay, so covariance of  $XY$ , this is also =  $\frac{\sum X_i Y_i}{n} - \left(\frac{\sum X_i}{n}\right) \left(\frac{\sum Y_i}{n}\right)$ , okay.

So, this comes out to be covariance of  $XY$   $\frac{\sum X_i Y_i}{n}$ ,  $\sum X_i Y_i$  is -123 divided by 10, okay -  $\frac{\sum X_i}{n} \frac{\sum Y_i}{n}$ ,  $\sum X_i$  is 10/10 that is 1, okay and  $\sum Y_i$  is -30/10, so what we get, this cancels, and this cancels and we get +3, so here -123, -12.3 and + 3, so how much we get; -9.3, okay, -9.3 that is covariance; covariance of  $XY$  divided by  $\sigma_X \sigma_Y$ , okay.

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$$\rho = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{-9.3}{\sqrt{14} \sqrt{39.8}} = -0.3940$$

Regression line of Y on X is given by

$$Y - \mu_Y = \frac{\rho \sigma_Y}{\sigma_X} (X - \mu_X)$$

$$Y - 38 = \frac{(-0.3940) \sqrt{39.8}}{\sqrt{14}} (X - 32) \quad \checkmark$$

Regression line of X on Y is given by

$$X - \mu_X = \frac{\rho \sigma_X}{\sigma_Y} (Y - \mu_Y)$$

$$X - 32 = \frac{-0.3940 \sqrt{14}}{\sqrt{39.8}} (Y - 38) \quad \checkmark$$

So, we need to find covariance, so sigma rho = covariance of XY divided by sigma X sigma Y, so -9.3 divided by sigma X sigma Y; sigma X = root 14, sigma Y = root 39.8, so root 14 and root 39.8, okay, it comes out to be = rho = -0.3940, so -0.3940, so this rho, okay. Now, the regression line of y on x is given by Y - Mu Y rho sigma Y over sigma X \* X - Mu X, okay, so Y - Mu Y we have found.

Mu Y = 378, okay and Mu X is 32, okay, so 38, rho we have found, rho = -0.3940, okay and what is sigma Y and sigma X, sigma Y = this one, root 39.8, okay and sigma X is root 14, so root 39.8 divided by root 14 and X - Mu X is 32, okay, this Mu X 32, so this is the regression line of y on x and regression line of x on y is given by X - Mu X = sigma rho sigma X over sigma Y Y - Mu Y, okay.

So, we have X - Mu X is 32, okay, this should be small x = rho is -0.3940 \* sigma X is square root 14 divided by square root 39.8, okay \* Y - Mu Y which is 38, so these 2 are the regression lines of y on x and x on y, okay, the most likely marks in statistics when the marks in economics are 30, okay. So, marks in economics we have denoted by the random variable X and marks in statistics we have denoted by the random variable Y.

We want to find the most likely marks in statistics that means we want to find the likely marks in Y, when the marks in X are given, so they can be found from the regression line of y on x, okay because the value of Y is to be found, when the value of X is given, so regression line of y on x is this one, okay, this is the regression line of y on x, in this you put  $x = 30$ , okay, so you get the value of the likely marks in statistics.

When you put S 30, Y will come out to be  $-0.6643 * 30 + 59.2570$  which will be  $= 39.3286$ , so that is how we solve this problem, with that I would like to end my lecture, thank you very much for your attention.