

## MODULE - 3

# Statistical Techniques - I

## (I) Measure of Central Tendency :-

(iv) Mean :- for grouped data

$$(\text{Mean}) \bar{x} = \frac{\sum f_i x_i}{N}; N = \sum f_i$$

for ungrouped data.

$$\bar{x} = \frac{\sum x_i}{n}; n = \text{no of frequency}$$

(v) Median :- Median is the central value of the variable when the values are arranged in ascending or descending order of magnitude.

for ungrouped frequency :

(a)  $n$  is odd, the middle value gives the median =  $\left(\frac{n+1}{2}\right)^{\text{th}}$  value.

(b)  $n$  is even, there are two middle values  $\left(\frac{n}{2}\right)^{\text{th}}$  and  $\left(\frac{n+1}{2}\right)^{\text{th}}$

then the arithmetic mean of two middle values gives the median.

$$\text{median} = \frac{\left(\frac{n}{2}\right)^{\text{th}} \text{ value} + \left(\frac{n+1}{2}\right)^{\text{th}} \text{ value}}{2}$$

for grouped frequency:

$$\text{Median} = l + \frac{h}{f} \left( \frac{N - c}{2} \right)$$

$l \rightarrow$  lower limit of median class where median class is the class interval corresponding to cumulative frequency  $C.F$ ) just  $\geq \frac{N}{2}$ .

$h \rightarrow$  width of the median class

$f \rightarrow$  frequency of median class

$N \rightarrow \sum f$

$c \rightarrow$  cumulative frequency ( $C.F$ ) of the preceding class from the median class.

Q. 1 Mode :— It is the value which occurs

most frequently in a set of data observations. Point of maximum frequency.

Frequency	Class Interval ( $C.I$ )	Mid value of $C.I$ ( $x_i$ )	Frequency ( $f_i$ )	$f_i x_i$	Cumulative Frequency ( $C.F$ )
5	6.5-7.5	7	5	35	5
12	7.5-8.5	8	12	96	17
25	8.5-9.5	9	25	225	42
48	9.5-10.5	10	48	480	90
32	10.5-11.5	11	32	352	122
6	11.5-12.5	12	6	72	128
1	12.5-13.5	13	1	13	127
$\sum f_i = 129$		$\sum f_i x_i = 1273$		always equal.	

L14 For a symmetrical distribution mean, mode, median coincide.

Q. 2 Empirical mode formula.

$$\text{Mode} = 3 \text{ Median} - 2 \text{ Mean}$$

where the method of grouping fails then the mode is calculated by Empirical formula

Q. 3 Calculate Mean, median, mode from the following

$$\text{Mean} :- \bar{x} = \frac{\sum f_i x_i}{N} = \frac{1273}{129} = 9.86$$

$$\text{Median} :- \text{Median} = \frac{N}{2} = \frac{129}{2} = 64.5$$

so median class is just  $\geq \left(\frac{N}{2}\right)$

$\left\{ \begin{array}{l} 64.5 \text{ is just close to } 90 \text{ (C.F)} \text{ so} \\ \text{median class is which have closest value of C.F to the median} \end{array} \right.$

$$\text{median class} = 9.5 - 10.5$$

$\text{Lower limit}$        $\text{Upper limit}$

$$l = 9.5$$

$$h = (\text{midvalue of median class})$$

$$f_m = 48$$

$$c = 42 \quad (\text{pre-median class})$$

C.I	Frequency	1-5	6-10	11-15	16-20	21-25
		7	10	16	32	24

$$\text{Upper-Lower} = \frac{6-5}{2} = 0.5$$

-0.5 to lower limit ; +0.5 to upper limit

$$\text{Median} = l + \frac{h}{2} \left( \frac{N - c}{f_m} \right)$$

$$= 9.5 + \frac{1}{2} \left( \frac{129 - 42}{48} \right)$$

$$[ \text{Median} = 9.96 ]$$

Mode :- modal class  $\rightarrow$  C.I of highest frequency

$$l = 9.5$$

$$h = (10.5 - 9.5) = 1$$

$$f_m = 48$$

$$f_1 = 25$$

$$f_2 = 32$$

$$\text{Mode} = l + \frac{f_m - f_1 - f_2}{2f_m - f_1 - f_2} \times h$$

$$[ \text{Mode} = 10.08 ]$$

C.I	Frequency	1-5	6-10	11-15	16-20	21-25
		7	10	16	32	24
		0.5+5.5				
		6-10				
		11-15				
		16-20				
		21-25				

$$C.F$$

$$C.F$$

$$C.F$$

$$C.F$$

$$C.F$$

$$C.F$$

$$1) \bar{x} = \frac{\sum f_i x_i}{\sum f_i} = \frac{1557}{89} = 17.4$$

$$\text{i) median} = \frac{N}{2} = \frac{89}{2} = 44.5$$

$$\text{median class} = 15.5 - 20.5$$

$$l = 15.5$$

$$h = 20.5 - 15.5 = 5$$

$$f = 32$$

$$c = 33$$

$$\text{Median} = l + \frac{h}{f} \left( \frac{N}{2} - c \right)$$

$$= 15.5 + \frac{5}{32} (44.5 - 33)$$

$$\mu_w = \frac{\sum_{i=1}^n f_i (x_i - \bar{x})}{N}$$

where  
 $n = 0, 1, 2, 3, \dots$   
 $\bar{x} = \text{mean}$

$$(a) \text{ If } x_c = 0 \text{ then } \mu_0 = \frac{\sum f_i}{\sum f_i} = 1$$

$$\therefore \boxed{\mu_0 = 1} \text{ always}$$

$$\text{Mode} \Rightarrow \text{modal class} = 15.5 - 20.5$$

$$l = 15.5$$

$$h = 5$$

$$f_m = 32, f_1 = 16, f_2 = 24$$

$$\boxed{\mu_1 = 0} \text{ always}$$

It is also called the variance = square of standard deviation

$$\mu_2 = \frac{\sum f_i (x_i - \bar{x})^2}{\sum f_i} = \frac{(S.D)^2}{\sigma}$$

$$\begin{aligned} \text{Mode} &= l + \frac{f_m - f_1}{2f_m - f_1 - f_2} \times h \\ &= 15.5 + \frac{32 - 16}{2 \times 32 - 16 - 24} \times 5 \end{aligned}$$

$$= 15.5 + \frac{16}{24} \times 5$$

$$= 15.5 + 0.66 \times 5 \\ \therefore 15.5 + 3.33 \end{math>$$

$$\text{Mode} = 18.83$$

## (II) Measure of Dispersion

Moment  $\rightarrow$  deviation of item from some given number. (In case deviation in mid-value)  
' $x$ '  $\rightarrow$  no. of moments

i) Moment about mean ( $\mu_w$ ) / Central Moment

If  $x_1, x_2, x_3, \dots$  are the values of a variable  $x$  with corresponding frequencies  $f_1, f_2, f_3, \dots$ . Then the moment about mean is

$\mu_w = \frac{\sum_{i=1}^n f_i (x_i - \bar{x})}{N}$

$\mu_0 = 1, 2, 3, \dots$

Similarly we can calculate  $\mu_3, \mu_4, \dots$ .

## ii) Moment about any point ( $\mu_x$ )

The moment about any given point 'A' is

$$\mu_x = \frac{\sum_{i=1}^n f_i (x_i - A)^x}{N}$$

$$(a) \text{ If } x=0, \text{ then } \mu_0 = 1$$

$$(b) \text{ If } x=1, \text{ then } \mu_1 = (\bar{x} - A) \\ = (\text{mean} - A)$$

$$(c) \text{ If } x=2, \text{ then } \mu_2 = \frac{\sum f_i (x_i - A)^2}{N}$$

Similarly, we can calculate  $\mu_3, \mu_4, \dots$ .

## iii) Moment about Origin ( $\mu_x$ )

The  $x^{\text{th}}$  moment about origin is

$$\mu_x = \frac{\sum_{i=1}^n f_i x_i^x}{N}$$

$$(a) \text{ If } x=0, \text{ then } \mu_0 = 1$$

$$(b) \text{ If } x=1, \text{ then } \mu_1 = \frac{\sum f_i x_i}{N} = \bar{x} \text{ (mean)}$$

Similarly we can calculate  $\mu_2, \mu_3, \dots$

$$[ \mu_1 = \bar{x} ]$$

Example question

Relation between  $\mu_x$  and  $\mu'_x$ :

$$\mu_1 = 0$$

$$\mu_2 = \mu'_2 - (\mu')^2$$

$$\mu_3 = \mu'_3 - 3\mu'_2 \mu'_1 + 2(\mu')^3$$

$$\mu_4 = \mu'_4 - 4\mu'_3 \mu'_1 + 6\mu'_2 (\mu')^2 - 3(\mu')^4$$

Relation between  $\mu_x$  and  $\mu_2$ :

$$\mu_1 = \bar{x} \text{ (mean)}$$

$$\star \quad \mu_2 = \frac{(\bar{x})^2}{N} \rightarrow \text{variance}$$

$$\mu_3 = \mu_2 + 3\mu_1 \bar{x} + (\bar{x})^3$$

$$\mu_4 = \mu_2 + 4\mu_1 \bar{x} + 6\mu_2 (\bar{x})^2 + (\bar{x})^4$$

Karl Pearson's Coefficient :

$\beta$ -coefficient :

$$\beta_1 = \frac{\mu_3}{\mu_2^{\frac{3}{2}}} ; \quad \beta_2 = \frac{\mu_4}{\mu_2^2}$$

$\gamma$ -coefficient :  $\gamma_1 = \sqrt{\beta_1} = \frac{\mu_3}{\mu_2^{\frac{3}{2}}}; \quad \gamma_2 = \beta_2 - 3$

for mean :-  $\mu_1 = \bar{x} - A$

$$[\mu_1 = \bar{x}]$$

Numerical on moments.

5 marks  
2014

- Q. In a distribution, the first four moment about a point  $x=4$  are  $-1.5, 17, -30$  and  $308$ . Find the moments about mean and about origin. Also calculate  $\beta_1$  and  $\beta_2$ .

Given A = 4  
 $\mu'_1 = -1.5$ ;  $\mu'_2 = 17$ ;  $\mu'_3 = -30$ ;  $\mu'_4 = 308$

→ Moment about mean (central moment)

$$\boxed{\mu_1 = 0}$$

$$\mu_2 = \mu'_2 - (\mu'_1)^2 = 17 - (-1.5)^2 = \boxed{14.75}$$

$$\begin{aligned} \mu_3 &= \mu'_3 - 3\mu'_2\mu'_1 + 2(\mu'_1)^3 = -30 - 3(17)(-1.5) \\ &\quad + 2(-30)^3 = \boxed{39.75} \end{aligned}$$

$$\mu_4 = \mu'_4 - 4\mu'_3\mu'_1 + 6\mu'_2(\mu'_1)^2 - 3(\mu'_1)^4$$

$$\boxed{\mu_4 = 342.31}$$

→ Moment about origin

$$\nu_1 = \bar{x} \text{ (mean)} = \mu'_1 + A = -1.5 + 4 = 2.5$$

$$\nu_2 = \mu_2 + (\bar{x})^2 = 14.75 + (2.5)^2 = 21$$

$$\nu_3 = \mu_3 + 3\mu_2(\bar{x}) + (\bar{x})^3 = 166$$

$$\nu_4 = \mu_4 + 4\mu_3(\bar{x}) + 6\mu_2(\bar{x})^2 + (\bar{x})^4 = 1332$$

calculation of  $\beta_1$  and  $\beta_2$ :

$$\beta_1 = \frac{\mu_3^2}{\mu_2^3} = \frac{(39.75)^2}{(14.75)^3} = 0.4993$$

$$\beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{342.31}{(14.75)^2} = 1.5733$$

Type - II  
calculate the moment about mean of the given data.

C.I	0-10	10-20	20-30	30-40	40-50
	5	10	40	20	25

$$\text{moment about mean } \mu_x = \frac{\sum_{i=1}^n f_i(x_i - \bar{x})^n}{N}$$

$$N = \sum_{i=1}^n f_i, \bar{x} = \frac{\sum f_i x_i}{N}$$

C.I	$x_i$ (mid)	$f_i$	$x_i f_i$	$(x_i - \bar{x})^1 f_i$	$(x_i - \bar{x})^2 f_i$	$(x_i - \bar{x})^3 f_i$	$(x_i - \bar{x})^4 f_i$
0-10	5	5	25	-25	3125	-78125	1953125
10-20	15	10	150	-15	2250	-33750	506250
20-30	25	40	1000	-5	10000	-50000	250000
30-40	35	20	700	5	500	2500	12500
40-50	45	25	1125	15	5625	84375	1265625
			$\Sigma x_i f_i$		12500	-30000	3762500

$$\bar{x} = \frac{\sum f_i x_i}{N} = \frac{3000}{100} = 30$$

$$\mu_1 = 0$$

$$\mu_2 = \frac{\sum f_i (x_i - \bar{x})^2}{N} = \frac{12500}{100} = 125$$

$$\mu_3 = \frac{\sum f_i (x_i - \bar{x})^3}{N} = \frac{-30000}{100} = -300$$

$$\mu_4 = \frac{\sum f_i (x_i - \bar{x})^4}{N} = \frac{3762500}{100} = 37625$$

#### (iv) Skewness and Kurtosis

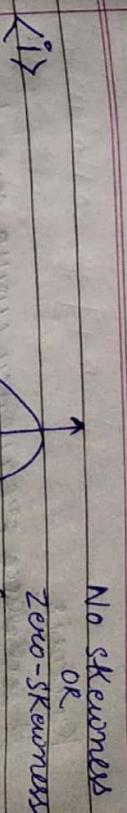
Skewness → Skewness means lack of symmetry or lopsidedness in a frequency distribution.

Skewness indicates whether the curve is turned more to one side than to other or whether the curve has a longer tail on one side. Skewness can be positive as well as negative.

#### Symmetrical distributions

The frequencies are symmetrically distributed about the mean.

Mean, mode and median coincide and median lies half-way between the two quartiles.



*(ii) Positive / right hand Skewness*

Right skewed  
skewness  
(M > Mo > Md)

*(iii) Negative / left hand skewness*

(M < Md < Mo)

Left skewed  
skewness

(M < Md < Mo)

Moment coefficient of Skewness :— It is denoted by ' $\gamma_1$ '.

$$\text{coeff. of skewness } (\gamma_1) = \sqrt{\beta_1} = \sqrt{\frac{\mu_3}{\mu_2^3}}$$

$$\boxed{\text{coeff. of skewness } (\gamma_1) = \frac{\mu_3}{\mu_2^3}}$$

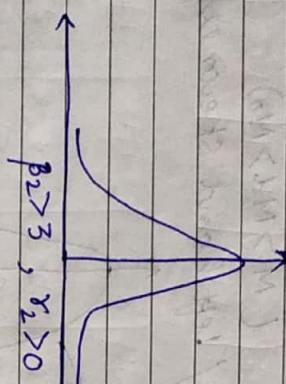
### Kurtosis

Defintion: frequency curve may be symmetrical but it may not be equally flat topped with the normal curve. The relative flatness of the top is called kurtosis and is measured by  $\beta_2$ .

Kurtosis refers to the bulgingness of the curve of a frequency distribution.

### Leptokurtic

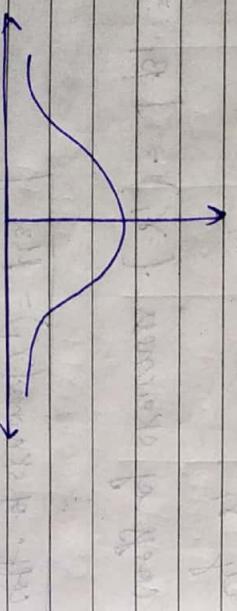
Curve which are more sharply peaked than the normal curve are called leptokurtic.



$$\beta_2 > 3, \gamma_2 > 0 \quad (L)$$

### PlatyKurtic

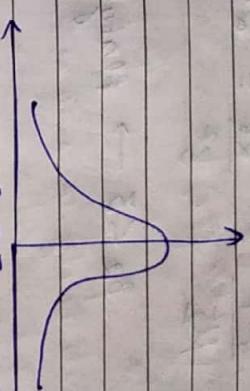
Curves which are flatter than the normal curves are called platykurtic curve.



$$\beta_2 < 3, \gamma_2 < 0 \quad (P)$$

### Mesokurtic

Curves which are neither flat nor sharply peaked are called normal curves are mesokurtic.



### Type - III

Ques. The first four moments about  $x=4$  are 1, 4, 10 and 45. find the coeff. of skewness and kurtosis also comment upon the nature of curve given  $A = 4$ ,  $\mu'_1 = 1$ ,  $\mu'_2 = 4$ ,  $\mu'_3 = 10$ ,  $\mu'_4 = 45$

$$[\mu'_1 = 0]$$

$$\begin{aligned} \mu'_2 &= \mu'_2 - (\mu'_1)^2 = 4 - (1)^2 = [3], \\ \mu'_3 &= \mu'_3 - 3\mu'_2\mu'_1 + 2\mu'_1^3 = [10] \end{aligned}$$

$$\begin{aligned} \mu'_4 &= \mu'_4 - 4\mu'_3\mu'_1 + 6\mu'_2(\mu'_1)^2 - 3\mu'_1^4 \\ &= 45 - 4 \times 10 \times 1 + 6 \times 4 \times 1 - 3 \times 1 \end{aligned}$$

$$\text{Coefficient of Kurtosis } \left[ (\beta_2) = \frac{\mu'_4}{\mu'_2^2} \right]$$

$$\beta_2 = 3, \gamma_2 = 0 \quad (M)$$

$$\gamma_1 = \frac{\mu_3}{\sqrt{\mu_2^3}} = 0$$

$$\mu_2 = \frac{f_i(x_i - \bar{x})^2}{N} = \frac{4260}{80} = 53.25$$

$$\beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{26}{5^2} = 2.88$$

$\beta_2 < 3 \rightarrow$  PlatyKurtic curve.

Type-IV

Q203 calculate first four moments and coeff. of Skewness & kurtosis

C.I	5-10	10-15	15-20	20-25	25-30	30-35	35-40
f <sub>i</sub>	6	8	17	21	15	11	2

C.I	x <sub>i</sub>	f <sub>i</sub>	x <sub>i</sub> f <sub>i</sub>	x <sub>i</sub> - $\bar{x}$	f <sub>i</sub> (x <sub>i</sub> - $\bar{x}$ ) <sup>2</sup>	f <sub>i</sub> (x <sub>i</sub> - $\bar{x}$ ) <sup>3</sup>	f <sub>i</sub> (x <sub>i</sub> - $\bar{x}$ ) <sup>4</sup>
5-10	7.5	6	45	2.3	3174	-73002	1679046
10-15	12.5	8	100	-9.5	-722	-6859	65160.5
15-20	17.5	17	297.5	-4.5	-34425	-1549.12	6971.04
20-25	22.5	21	472.5	6.5	50.25	2.625	1.03125
25-30	27.5	15	412.5	5.5	453.75	2495.625	13725.93
30-35	32.5	11	357.5	10.5	1212.75	12733.875	133705.63
35-40	37.5	2	75	15.5	480.5	7447.75	115490.12

$\sum f_i = 80$	1760	4260	87273.75	2014050.53
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$$\bar{x} = \frac{\sum x_i f_i}{N} = \frac{1760}{80} = 22$$

$$\gamma_1 = \frac{\mu_3}{\sqrt{\mu_2^3}} = \frac{1090.92}{\sqrt{(53.25)^3}} = 25175.63$$

$$\mu_2 = \frac{f_i(x_i - \bar{x})^2}{N} = \frac{1090.92}{388.57} = 2.80$$

$\beta_2 = \frac{\mu_4}{\mu_2^2} = \frac{25175.63}{(2.80)^2} = 8.8$

$\beta_2 > 3 \rightarrow$  Leftskewic curve.

### (III) Moment Generating Function (m.g.f.)

→ for continuous random variable.

If  $x$  is a continuous random variable, then  
 $m.g.f. of (x)$  is

$$M_x(t) = \int_a^b e^{tx} f(x) dx \quad (1)$$

for  $a \leq x \leq b$

$f(x) \rightarrow$  probability density function (p.d.f)  
 $t \rightarrow$  is a parameter

→ for discrete random variable

If  $x$  is a discrete random variable then  
 $m.g.f. of (x)$  is

$$M_x(t) = \sum_{x=a}^b e^{tx} P(x) \quad (2)$$

for  $a \leq x \leq b$

$P(x) \rightarrow$  probability mass function (p.m.f)  
 $t \rightarrow$  is a parameter

Remarks

$$\textcircled{1} \quad \text{Mean } (\mu_1) = \left[ \frac{d}{dt} M_x(t) \right]_{t=0}$$

(i) Variance ( $\mu_2$ ) =  $\mu_2 - (\text{mean})^2$

$$\text{where } \nu_2 = \left[ \frac{d^2}{dt^2} M_x(t) \right]_{t=0}$$

$$\textcircled{2} \quad M_{x+y}(t) = M_{x}(t) \cdot M_y(t)$$

m.g.f. of sum of two independent variable  
 is equal to the product of their respective

m.g.f.

2011/12/13

$$\begin{aligned} & \text{obtain the m.g.f. of } f(x) = \begin{cases} x, & \text{for } 0 < x < 1 \\ 2-x, & 1 \leq x < 2 \\ 0, & \text{otherwise} \end{cases} \\ & \text{and variance.} \end{aligned}$$

We know that the m.g.f. for continuous random variable 'x' is

$$M_x(t) = \int_a^b e^{tx} f(x) dx$$

$$M_x(t) = \int_0^2 e^{tx} \cdot x dx + \int_2^\infty e^{tx} (2-x) dx + \int_0^2 e^{tx} \cdot 0 dx$$

$$\Rightarrow \int uv dx = uv_1 - u'v_2 + u''v_3 - u'''v_4$$

$$1 = \left[ x \frac{e^{tx}}{t} - \frac{1}{t} \frac{e^{tx}}{t} + 0 \right]_0^2 + \left[ (2-x) \frac{e^{tx}}{t} - (-1) \frac{e^{tx}}{t^2} + 0 \right]_1^2$$

$$= \frac{e^{2t} - e^t}{t} + \frac{1}{t^2} + \frac{e^{2t} - e^t}{t^2} + \frac{-e^t - e^t}{t^2}$$

$$= \frac{1}{t^2} [ e^{2t} + 1 - 2e^t ]$$

$$= \frac{1}{t^2} \left[ (e^t)^2 + 1 - 2e^t \right] = \frac{1}{t^2} [ e^t - 1 ]^2$$

$$= \left[ \frac{e^{xt} - 1}{t} \right]^2 \quad \because e^t = 1 + \frac{t}{1!} + \frac{t^2}{2!} + \frac{t^3}{3!} \dots$$

$$= \left[ \left( 1 + \frac{xt}{1!} + \frac{x^2 t^2}{2!} + \frac{x^3 t^3}{3!} + \dots \right) - 1 \right]^2$$

$$M_x(t) = \left[ \frac{1}{1!} + \frac{xt}{2!} + \frac{x^2 t^2}{3!} + \dots \right]^2$$

m.g.f

$$M_x(t) = \left[ \frac{1}{1!} + \frac{xt}{2!} + \frac{x^2 t^2}{3!} + \dots \right]^2$$

Now, mean( $\bar{Y}_1$ ) =  $\left[ \frac{d}{dt} M_x(t) \right]_{t=0}$

$$\bar{Y}_1 = \left\{ 2 \left[ \frac{1}{1!} + \frac{xt}{2!} + \frac{x^2 t^2}{3!} + \dots \right] \cdot \left[ \frac{1}{1!} + \frac{xt}{2!} + \dots \right] \right\}_{t=0}$$

$$\bar{Y}_1 = 2 \cdot \left( \frac{1}{1!} \right) \left( \frac{1}{2!} \right)$$

$$\boxed{\bar{Y}_1 = 1} \quad \text{mean}$$

~~$$\bar{Y}_2 = \left[ \frac{d^2}{dt^2} M_x(t) \right]_{t=0}$$~~

~~$$\bar{Y}_2 = \left[ 2 \times \left( \frac{1}{2!} + 2t + 3t^2 \right) \left( \frac{1}{1!} + \frac{2t}{3!} + \dots \right) + 2 \left( \frac{1}{1!} + t + \frac{t^2}{2!} \right) \left( \frac{2}{3!} + \frac{6t}{4!} + \dots \right) \right]_{t=0}$$~~

~~$$\bar{Y}_2 = \left[ \frac{1}{2} + \frac{2}{3} \right]$$~~

$$\bar{Y}_2 = \frac{7}{6}$$

~~$$\bar{Y}_2 = \left[ \frac{1}{1! 2!} + \frac{2t^2}{2! 3!} + \frac{12t^2}{3! 4!} \right]_{t=0}$$~~

~~$$1 \cdot \left[ \frac{1}{1! 2!} + \left[ \frac{2t^2}{2! 3!} + \frac{3 \times 4 \times t^3}{3! 4!} \right] \right]_{t=0}$$~~

~~$$\bar{Y}_2 = \frac{1}{2} \left[ \frac{1}{2} + \frac{4t^2}{6} + \frac{3 \times 4 \times t^3}{24} \right]$$~~

~~$$\bar{Y}_2 = \frac{1}{2} + \frac{4t^2}{12} + \frac{4t^3}{24}$$~~

~~$$0.178t$$~~

20/3/11

Q. find the m.g.f of  $P(x) = e^{-\lambda} \frac{\lambda^x}{x!}$ . also find the mean & variance.

m.g.f of discrete random value.

$$M_x(t) = \sum_{x=a}^{\infty} e^{tx} P(x)$$

$$M_x(t) = \sum_{x=0}^{\infty} e^{tx} e^{-\lambda} \frac{\lambda^x}{x!} ; 0 \leq x < \infty$$

$$= e^{-\lambda} \sum_{x=0}^{\infty} e^{tx} \frac{\lambda^x}{x!}$$

$$= e^{-\lambda} \sum_{x=0}^{\infty} \left( e^t \lambda \right)^x \frac{e^{-\lambda}}{x!} ; e^x = 1 + x + \frac{x^2}{2!} + \dots$$

$$= e^{-\lambda} \left[ 1 + \lambda e^t + (\lambda e^t)^2 + (\lambda e^t)^3 + \dots \right]$$

$$= e^{-\lambda} \left[ e^{(\lambda+1)e^t} \right]$$

$$= e^{-\lambda} \cdot e^{\lambda e^t} ; e^x \cdot e^y = e^{x+y}$$

$$= e^{(\lambda e^t - \lambda)}$$

$$\Rightarrow \boxed{e^{\lambda(e^t - 1)} = M_x(t)} \quad \text{m.g.f} \quad \text{①}$$

$$\text{mean } (21) = \left[ \frac{d}{dt} M_x(t) \right]_{t=0}$$

$$21 = \boxed{\lambda(e^{t=0} - 1) + e^{t=0}}$$

$$\text{variance } (22) = \left[ \frac{d^2}{dt^2} M_x(t) \right]_{t=0}$$

$$22 = \left[ \lambda^2 e^{2t} (e^{\lambda(e^t - 1)}) + \lambda e^{\lambda(e^t - 1)} \lambda e^t \right]_{t=0}$$

$$22 = \boxed{\lambda^2 + \lambda}$$

$$\text{Now, variance } (\mu_2) = 22 - (\text{mean})^2$$

$$\mu_2 = \lambda^2 + \lambda - (\lambda^2)^2$$

$$\boxed{\mu_2 = \lambda}$$

## (IV) Method of Least Square [In Curve Fitting]

(a) Fit a straight line

Let  $y = a + bx$  — (1)  
be the straight line to be fitted.

$$\text{or } E = \sum (y_i - a - bx_i)^2$$

$$\text{or } E_i^2 = \sum (y_i - a - bx_i)^2 \quad \text{— (2)} \quad \left[ \text{Let } E_i^2 = U \right]$$

$$U = \sum (y_i - a - bx_i)^2 \quad \text{— (3)}$$

Now differ (3) w.r.t parameter 'a' & 'b' partially  
and equate them equal to zero.

$$\begin{aligned} \frac{\partial U}{\partial a} &= 2 \sum (y_i - a - bx_i) \cdot (-1) = 0 \\ \Rightarrow \sum (y_i - a - bx_i) &= 0 \\ \Rightarrow \sum y_i &= na + b \sum x_i \quad \left\{ \begin{array}{l} n \rightarrow \text{total} \\ \text{no. of data} \end{array} \right\} \\ \Rightarrow \boxed{\sum y_i = na + b \sum x_i} & \quad \text{— (4)} \end{aligned}$$

$$\text{and } \frac{\partial U}{\partial b} = 2 \sum (y_i - a - bx_i) \cdot (-x_i) = 0$$

$$\Rightarrow \sum (x_i y_i) - nx \bar{a} - b \sum x_i^2 = 0$$

$$\Rightarrow \sum x_i y_i - a \sum x_i - b \sum x_i^2 = 0$$

$$\Rightarrow \boxed{\sum xy = a \sum x + b \sum x^2} \quad \text{— (5)}$$

equation (4) and (5) are called Normal Equations

Fit a straight line for the following data :-

x	2	3	4	5	6	7
y	3.0	5.0	5.5	6.0	8.0	9.5

We know that Normal Equations are :-

$$\begin{aligned} \sum y &= na + b \sum x \quad \text{— (1)} \quad \left\{ \begin{array}{l} \text{whereas} \\ n = 6 \end{array} \right. \\ \sum xy &= a \sum x + b \sum x^2 \quad \text{— (2)} \end{aligned}$$

x	y	xy	$x^2$	
2	3	6	4	
3	5	15	9	
4	5.5	22	16	
5	6	30	25	$187.05 = 6a + 27b$
6	8	48	36	
7	9.5	66.5	49	$a = 0.767$
		$\sum y = 37$	$\sum x = 27$	$b = 1.02$
		$\sum xy = 187.5$	$\sum x^2 = 139$	

Equation of straight line

$$y = a + bx$$

$$\text{Ans. } \boxed{y = 0.767 + (1.02)x}$$

(b) Fit a Curve.

Q. Fit the given data fit a curve  $y = a + bx + cx^2$   
(second degree parabola)

x	1	2	3	4	5
y	1.8	5.1	8.9	14.1	19.8

Given equation  $y = a + bx + cx^2$  — (1)

a, b and c are parameters

$$E = y - a - bx - cx^2 \text{ or } E_i^2 = (y_i - a - bx_i - cx_i^2)^2$$

$$U = \sum (y_i - a - bx_i - cx_i^2)^2$$

Differentiating w.r.t a, b & c partially and equating to zero

$$\frac{\partial U}{\partial a} = 2 \sum (y_i - a - bx_i - cx_i^2)(-1) = 0$$

$$\Rightarrow \sum y_i - na - b \sum x_i - c \sum x_i^2 = 0$$

$$\sum y_i = na + b \sum x_i + c \sum x_i^2$$

$$\boxed{\sum y = na + b \sum x + c \sum x^2} \quad (2)$$

$$\text{and } \frac{\partial U}{\partial b} = 2 \sum (y_i - a - bx_i - cx_i^2)(-x_i) = 0$$

$$\sum x_i y_i - na - b \sum x_i^2 - c \sum x_i^3 = 0$$

$$a = -0.46 \quad c = 0.442$$

$$\sum x_i y_i = \sum a + b \sum x_i^2 + c \sum x_i^3$$

$$\boxed{\sum xy = a \sum x + b \sum x^2 + c \sum x^3} \quad (3)$$

$$\text{and } \frac{\partial U}{\partial c} = 2 \sum (y_i - a - bx_i - cx_i^2)(-x_i^2) = 0$$

$$\Rightarrow \sum y_i x_i^2 - a \sum x_i^2 - b \sum x_i^3 - c \sum x_i^4 = 0$$

$$\Rightarrow \sum x_i^2 y_i = a \sum x^2 + b \sum x^3 + c \sum x^4$$

$$\boxed{\sum x^2 y = a \sum x^2 + b \sum x^3 + c \sum x^4} \quad (4)$$

Equation (3), (4) and (5) are the required Normal equations.

x	y	$x^2$	$x^3$	$x^4$	$xy$	$x^2 y$
1	1.8	1	1	1	1.8	1.8
2	5.1	4	8	16	10.2	20.4
3	8.9	9	27	81	26.7	80.1
4	14.1	16	64	256	56.4	225.6
5	19.8	25	125	625	99.0	495.0
15	49.7	225	979	1941	822.9	

From Normal Equation [n = 5]

$$5a + 15b + 55c = 49.7$$

$$15a + 55b + 225c = 194.1$$

$$55a + 225b + 979c = 822.9$$

curve  $\Rightarrow y = a + bx + cx^2$

Ans. 
$$y = -0.46 + 1.842x + 0.442x^2$$

Q fitting the curve  $y = ax + \frac{b}{x}$  by least sq method

x	1	2	3	4	5	6	7	8
y	5.4	6.2	8.2	10.3	12.0	14.8	17.2	19.5
xy								
$\frac{y}{x}$								
$\frac{1}{x^2}$								

equation :  $y = ax + \frac{b}{x} \quad \text{--- } ①$

$$E = y - ax - \frac{b}{x}$$

$$E_i^2 = (y_i - ax_i - \frac{b}{x_i})^2$$

$$U = \sum (y_i - ax_i - \frac{b}{x_i})^2 \quad \text{--- } ②$$

differentiating eq ② partially w.r.t a & b and equating equal to zero

$$\frac{\partial U}{\partial a} = 2 \sum (y_i - ax_i - \frac{b}{x_i})(-x_i) = 0$$

$$\Rightarrow \sum y_i x_i - a \sum x_i^2 - b \sum n = 0$$

$$\left[ \sum xy = a \sum x^2 + bn \right] \quad \text{--- } ③$$

$$\text{now } \frac{\partial U}{\partial b} = 2 \sum (y_i - ax_i - \frac{b}{x_i}) \left( \frac{-1}{x_i^2} \right) = 0$$

$$\sum y_i - a \sum x_i - \frac{\sum b}{\sum x_i^2} = 0$$

$$\frac{\sum y_i}{\sum x_i} = an + b \frac{\sum 1}{\sum x_i^2}$$

$$\sum \left( \frac{y_i}{x_i} \right) = an + b \sum \left( \frac{1}{x_i^2} \right) \quad \text{--- } ④$$

x	y	xy	$\frac{y}{x}$	$x^2$	$\frac{1}{x^2}$
1	5.4	5.4	5.4	1	1
2	6.2	12.4	3.1	4	0.25
3	8.2	24.6	2.7	9	0.11
4	10.3	41.2	2.5	16	0.0625
5	12.6	63	2.5	25	0.04
6	14.8	88.8	2.4	36	0.0278
7	17.2	120.4	2.4	49	0.0204
8	19.5	156	2.4	64	0.0156
	194.2	511.8	21.9	204	1.52

Normal eqn will be

$$204a + 8b = 511.8$$

$$8a + 1.52b = 21.9$$

$$a = -9.98$$

$$b = 66.97$$

$$\text{Ans. } y = (-9.98)x + \frac{66.97}{x}$$

Q.  $PV^y = (\text{constant}) C$ , find the curve using least sq method :-

$V(\text{cm}^3)$	50	60	70	90	100
$P(\text{kg/cm}^2)$	64.7	51.3	40.5	25.9	18

$PV^y = C$  is the given equation.

$$\text{Let } P = CV^{-y}$$

$$\log P = \log C + (-y) \log V \quad \text{--- (1)}$$

Comparing with  
 $A + BX$

$$\text{So } \log P = Y, \log C = A, B = -y, \log V = X$$

and so the normal equations will be

$$\sum Y = nA + B\sum X \quad \text{--- (2)}$$

$$\sum XY = A\sum X + B\sum X^2 \quad \text{--- (3)}$$

$$\begin{array}{|c|c|c|c|c|} \hline V & P & Y = \log P & X = \log V & XY \\ \hline 50 & 64.7 & 1.81090 & 1.69897 & 30.7666 \\ \hline 60 & 51.3 & 1.71012 & 1.77815 & 30.4085 \\ \hline 70 & 40.5 & 1.60746 & 1.84510 & 29.6592 \\ \hline 90 & 25.9 & 1.41330 & 1.95429 & 20.76193 \\ \hline 100 & 18 & 1.89209 & 2 & 3.78418 \\ \hline \end{array}$$

$$\begin{array}{|c|c|c|c|c|} \hline & & X^2 & & \\ \hline & & 0.2 & 0.3 & 0.5 & 1 & 2 \\ \hline Y & & 16 & 14 & 11 & 6 & 3 \\ \hline \end{array}$$

$$\Rightarrow PV^y = C$$

$$\text{Now. } \boxed{PV^{0.29091} = 167.7876}$$

Q. fit a curve  $y = Cx + \frac{C_1}{\sqrt{x}}$ , for the data

$$E_i \Rightarrow y = C_0 x + \frac{C_1}{\sqrt{x}} \quad \text{--- (1)}$$

$$E = y - C_0 x + \frac{C_1}{\sqrt{x}}$$

$$E_i = y_i - C_0 x_i + \frac{C_1}{\sqrt{x_i}}$$

$$E_i^2 = \left( y_i - C_0 x_i + \frac{C_1}{\sqrt{x_i}} \right)^2$$

$$\text{Normal eqns :-- } 5A + 9.27646B = 8.43387 \quad \text{--- (2)}$$

$$9.27646A + 17.2717B = 15.6295 \quad \text{--- (3)}$$

$$A = 2.22476 \quad B = -0.29091$$

differentiating eq(2) wrt  $c_0$  &  $c_1$ , partially,

$$\frac{\partial U}{\partial c_0} = 0$$

$$\frac{\partial U}{\partial c_0} = 2 \sum (y_i - c_0 x_i + \frac{c_1}{\sqrt{x_i}}) \times (-x_i) = 0$$

$$\Rightarrow \sum y_i x_i - 2c_0 \sum x_i^2 + \sum c_1 \frac{1}{\sqrt{x_i}} = 0$$

$$\Rightarrow \sum y_i x_i = c_0 \sum x_i^2 - c_1 \sum \frac{1}{\sqrt{x_i}}$$

$$\boxed{\sum x_i y_i = c_0 \sum x_i^2 - c_1 \sum \frac{1}{\sqrt{x_i}}} \quad \text{--- (3)}$$

$$\frac{\partial U}{\partial c_1} = 0$$

$$\frac{\partial U}{\partial c_1} = 2 \sum (y_i - c_0 x_i + \frac{c_1}{\sqrt{x_i}}) \left( \frac{1}{\sqrt{x_i}} \right) = 0$$

$$\Rightarrow \sum \frac{y_i}{x_i} - 2c_0 \sum \frac{1}{\sqrt{x_i}} + \sum \frac{c_1}{x_i} = 0$$

$$\Rightarrow \sum \frac{y_i}{x_i} = c_0 \sum \frac{1}{\sqrt{x_i}} - c_1 \sum \frac{1}{x_i}$$

$$\Rightarrow \boxed{\sum \frac{y_i}{x_i} = c_0 \sum \frac{1}{\sqrt{x_i}} - c_1 \sum \frac{1}{x_i}} \quad \text{--- (4)}$$

eq (3) and (4) are normal equations

$x$	$y$	$x^2$	$y/x$	$x^2$	$\sqrt{x}$	$1/x$
0.2	16	3.2	80	0.04	0.447	5
0.3	14	4.2	46.6	0.09	0.547	3.3
0.5	11	5.5	22	0.25	0.707	2
1	6	6	1	1	1	1
2	3	6	1.5	4	1.414	0.5
4	50	24.9	156.1	5.38	4.1152	11.8

from 3 and 4 we get normal equation

$$c_0 5.38 - c_1 4.1152 = 24.9$$

$$c_0 4.1152 - c_1 11.8 = 156.1$$

$$c_0 = -7.71 \quad c_1 = 16.13$$

$$y = c_0 x + c_1 \left( \frac{1}{\sqrt{x}} \right)$$

$$\text{Ans. } \boxed{y = -7.71 x + 16.13 \left( \frac{1}{\sqrt{x}} \right)}$$

## (V) Correlation

In any distribution when change in one variable affect a change in other variable, then variable are said to be correlated. It is measured by correlation coefficient or coefficient of correlation.

$$\text{for eg: } \frac{x}{y} \begin{array}{|c|c|c|c|c|} \hline & x_1 & x_2 & \dots & x_n \\ \hline & y_1 & y_2 & \dots & y_n \\ \hline \end{array}$$

$x$  and  $y$  are correlated.

(a) Karl - Pearson's coefficient of correlation

or

Moment Product correlation coefficient

for different values of  $x = x_1, x_2, x_3, \dots$ ,  
the value of  $y$  are  $y_1, y_2, y_3, \dots$ . Then  
coefficient of correlation between  $x$  &  $y$   
are defined as

$$\rho = \rho(x, y) = \gamma_{xy} = \frac{\gamma_{xy}}{\sigma_x \sigma_y}$$

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

where  $n$  = no. of data

$$\bar{x} = \text{mean of } x = \frac{\sum x}{n} = \frac{\sum x_i n_i}{N}$$

$$\bar{y} = \text{mean of } y = \frac{\sum y}{n} = \frac{\sum y_i n_i}{N}$$

$$\sigma_x = \text{S.D. (standard deviation) of } x = \sqrt{\frac{\sum (x - \bar{x})^2}{n}}$$

$$\sigma_y = \text{S.D. of } y = \sqrt{\frac{\sum (y - \bar{y})^2}{n}}$$

Remark.

1. Generally  $\gamma$  lies between  $-1$  to  $1$   
i.e.  $-1 \leq \gamma \leq 1$
2. For  $0 < \gamma \leq 1$ , Positive-correlation
3. For  $-1 \leq \gamma < 0$ , Negative-correlation
4. For  $\gamma = 0$ , no correlation

$x$	$y$	$(x - \bar{x})(y - \bar{y})$	$(y - \bar{y})^2$	$(x - \bar{x})^2$
10	18	-10	11	121
14	12	-6	49	49
18	24	-2	9	36
22	6	2	1	4
26	30	6	25	36
30	36	10	81	90
36	10	9	100	100
120	126	286	280	280

$$\bar{x} = \frac{\sum x}{n} = \frac{120}{6} = 20$$

$$\bar{y} = \frac{\sum y}{n} = \frac{126}{6} = 21$$

$$\gamma = \frac{\sum (x-\bar{x})(y-\bar{y})}{n \sigma_x \sigma_y}$$

Remark

$$\sigma_x = \sqrt{\frac{\sum (x-\bar{x})^2}{n}} = \sqrt{\frac{280}{6}} = \sqrt{46.66} = 6.826$$

$$\sigma_y = \sqrt{\frac{\sum (y-\bar{y})^2}{n}} = \sqrt{\frac{286}{6}} = \sqrt{47.66} = 6.904$$

also

$$\gamma = \frac{280}{6 \times 6.826 \times 6.904}$$

$$\gamma = \frac{280}{282.76} = 0.99$$

Note  $\gamma = 0.99$  (much closer to 1)

∴ find the rank correlation coeff. of following data.

$$② \quad \gamma = \frac{\sum (x-\bar{x})(y-\bar{y})}{n \sigma_x \sigma_y} = \frac{\sum (x_i - R_1)(y_i - R_2)}{\sqrt{\sum (x_i - \bar{x})^2} \sqrt{\sum (y_i - \bar{y})^2}}$$

① Since we know that Karl's Pearson coefficient of corr. is :-

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

(b) Rank Correlation Coefficient

$$\gamma = 1 - \frac{6 \sum D_i^2}{n(n^2 - 1)}$$

where  $D_i = R_1 - R_2$

(difference of ranks)

$n \rightarrow$  no. of data

Marks A (R <sub>1</sub> )	Marks B (R <sub>2</sub> )	D <sub>i</sub> = R <sub>1</sub> - R <sub>2</sub>	D <sub>i</sub> <sup>2</sup>
15	50	-35	1225
20	30	-10	100
27	55	-28	784
13	30	-17	289
45	25	20	400
60	10	50	2500
20	30	-10	100
75	70	5	25
		$\sum D_i^2$	5423

Rank - Correlation coefficient

$$n = 8$$

$$\begin{aligned} r &= 1 - \left[ \frac{6 \sum D_i^2}{n(n^2-1)} \right] \\ &= 1 - \left[ \frac{6 \times 5423}{804} \right] = 1 - 64.559 \end{aligned}$$

R <sub>1</sub>	R <sub>2</sub>	R <sub>3</sub>	D <sub>12</sub>	D <sub>23</sub>	D <sub>31</sub>	D <sub>12</sub> <sup>2</sup>	D <sub>23</sub> <sup>2</sup>	D <sub>31</sub> <sup>2</sup>
R <sub>1</sub> -R <sub>2</sub>	R <sub>2</sub> -R <sub>3</sub>	R <sub>1</sub> -R <sub>3</sub>						
1	3	6	-2	-3	-5	4	95	9
6	5	4	1	1	2	1	4	1
5	8	9	-3	-1	-4	9	16	1
10	4	8	6	-4	2	36	4	16
3	7	1	-4	6	2	16	4	36
2	10	2	-8	8	0	64	0	64
4	2	3	2	-1	1	4	1	1
9	1	10	8	-9	-1	64	1	81
7	6	5	1	1	2	1	4	1
8	9	7	1	2	1	1	1	4

n=10

Given that n = 25,  $\sum x = 125$ ,  $\sum x^2 = 650$

$$\sum y = 100, \sum y^2 = 460 \text{ and } \sum xy = 508$$

Find the value of corr coeff  $r$  :-

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

$$r_{12} = 1 - \left[ \frac{6 \sum D_{12}^2}{n(n^2-1)} \right] = 1 - \left[ \frac{6 \times 200}{10(99)} \right] = -0.212$$

$$r_{23} = 1 - \left[ \frac{6 \times 214}{10 \times 99} \right] = -0.297$$

$$r_{13} = 1 - \left[ \frac{6 \times 60}{10 \times 99} \right] = 0.636$$

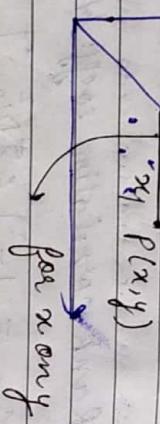
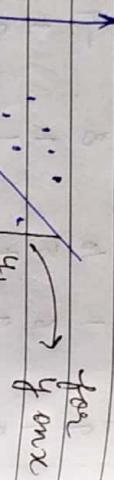
$$r = \frac{12700 - 12500}{\sqrt{16250 - 15625}} / \sqrt{11500 - 10000}$$

$$r = \frac{200}{\sqrt{625} \times \sqrt{1500}} = \frac{200}{25 \times 38.72} = r$$

$$r = 200 / 968.2 = \boxed{0.206} = r$$

## (VII) Regression Analysis

line of Regression :- A straight line which is passing through the data at which maximum numbers of data lies around or on it, then straight line is called line of Regression.



Two types of line of Regression

L1  $y$  on  $x$  : - (line parallel to  $y$ -axis)

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

or

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

where  $b_{yx} = r \frac{\sigma_y}{\sigma_x}$   
(coeff of regression)

(1)  $x$  on  $y$  :- (line parallel to  $x$ -axis)

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

or

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

where  $\bar{x} = \text{mean of } x$ ,  $\bar{y} = \text{mean of } y$

$r = \text{coeff of correlation}$

$$\sigma_x = S.D. \text{ of } x, \quad \sigma_y = S.D. \text{ of } y$$

Remark

- When distance of point to the line is taken along parallel to  $y$ -axis then line of regression is called  $y$  on  $x$  and vice-versa.

$$b_{yx} \times b_{xy} = r \frac{\sigma_y}{\sigma_x} \times r \frac{\sigma_x}{\sigma_y} = r^2$$

$$b_{yx} \cdot b_{xy} = r^2$$

$$r = \sqrt{b_{xy} \cdot b_{yx}}$$

(C.M.)

i.e. coeff. of correlation is Geometric mean between regression coefficient  $b_{xy}$  and  $b_{yx}$ .

## Derivation of Angle between two lines of Regression

OR

Ques:  $\rightarrow$  If  $\theta$  is acute angle between lines of regression then shows that  $\tan \theta = \left( \frac{1-\alpha^2}{\alpha} \right) \left( \frac{\sigma_x \cdot \sigma_y}{\sigma_x^2 + \sigma_y^2} \right)$

where symbols have their usual meaning.  
Also discuss the significance of formula at  $\alpha = 0$  and  $\alpha = \pm 1$ .

Let  $\alpha$  be the acute angle between two lines of regression.

Then slopes of lines of regression will be calculated by the equations of line of regressions.



$$\begin{array}{c} \text{y} \\ \nearrow \\ \text{y on x} \\ \searrow \\ \text{x on y} \end{array}$$

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

$$\begin{aligned} \tan \theta &= \frac{\frac{1-\alpha^2}{\alpha} \cdot \frac{\sigma_x \cdot \sigma_y}{\sigma_x^2 + \sigma_y^2}}{1 + \frac{\sigma_x \cdot \sigma_y}{\sigma_x^2 + \sigma_y^2}} \\ &= \frac{\left( \frac{1-\alpha^2}{\alpha} \right) \frac{\sigma_y}{\sigma_x}}{1 + \frac{(\sigma_x)^2 + (\sigma_y)^2}{(\sigma_x)^2 + (\sigma_y)^2}} \end{aligned}$$

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

$$\boxed{\tan \theta = \left( \frac{1-\alpha^2}{\alpha} \right) \left( \frac{\sigma_x \cdot \sigma_y}{\sigma_x^2 + \sigma_y^2} \right)} \quad \text{Hence proved.}$$

$$\text{y on x} : - (y - \bar{y}) = \alpha \frac{\sigma_y}{\sigma_x} (x - \bar{x}) \quad \text{--- } \textcircled{1}$$

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

$$\text{y on y} : - (x - \bar{x}) = \frac{\sigma_x}{\sigma_y} (y - \bar{y})$$

$$\frac{1}{\sigma_x} \frac{\sigma_y}{\sigma_y} (x - \bar{x}) = (y - \bar{y}) \quad \text{--- } \textcircled{2}$$

Comparing eq  $\textcircled{1}$  &  $\textcircled{2}$  with  $y = mx + c$

i.e. for  $\alpha = \pm 1$  both lines are  $\perp$  to each other.

$$\text{y on y} : - (x - \bar{x}) = \frac{\sigma_x}{\sigma_y} (y - \bar{y})$$

$$\frac{1}{\sigma_x} \frac{\sigma_y}{\sigma_y} (x - \bar{x}) = (y - \bar{y}) \quad \text{--- } \textcircled{2}$$

$$\therefore \text{if } \alpha = \pm 1, \text{ then } \tan \theta = 0$$

$$\text{hence slope of line } \textcircled{1} \quad m_1 = \frac{\sigma_y}{\sigma_x} \quad \text{--- } \textcircled{3}$$

$$\text{and slope of line } \textcircled{2} \quad m_2 = \frac{1}{\sigma_x} \frac{\sigma_y}{\sigma_y} = \frac{\sigma_y}{\sigma_x} \quad \text{--- } \textcircled{4}$$

$$\text{and we know that } \tan \theta = \frac{m_2 - m_1}{1 + m_1 m_2}$$

### Numericals

Q19

For two random variable  $x$  &  $y$  with same mean the two regression equations are :-

$y = ax + b$  and  $x = \alpha y + \beta$ , then show that  
 $\frac{b}{\beta} = \left(\frac{1-\alpha}{1-\alpha}\right)$  also find common mean.

From given equations  $y = ax + b$  ————— (1)

$$x = \alpha y + \beta \quad \text{--- (2)}$$

$$\text{we get } byx = a \\ bxy = \alpha$$

Let 'm' be the common mean for  $x$  &  $y$  both

$$\text{i.e. } \bar{x} = \bar{y} = m$$

Now eqns of regression lines

$$(y - \bar{y}) = byx(x - \bar{x})$$

$$y \text{ on } x : \rightarrow (y - m) = a(x - m) \rightarrow \text{given eqn.} \\ \Rightarrow y = ax + m(1-\alpha) \quad \text{--- (3)}$$

$$x \text{ on } y : \rightarrow (x - m) = \alpha(y - m) \\ \Rightarrow x = \alpha y + m(1-\alpha) \quad \text{--- (4)}$$

By comparing eq (3), (4) and (4)

$$b = m(1-\alpha) \quad \text{and} \quad \beta = m(1-\alpha)$$

$$\therefore \frac{b}{\beta} = \frac{m(1-\alpha)}{m(1-\alpha)}$$

$$\frac{b}{\beta} = \left(\frac{1-\alpha}{1-\alpha}\right) \quad \text{Hence proved.}$$

Now for the mean, since 'm' is common mean for both  $x$  &  $y$  then it satisfy the given eqn of regression lines.

$$b = m(1-\alpha)$$

$$b = m - am$$

$$m = b + am$$

$$\text{and } m = \beta + dm$$

By equating them,

$$b + am = \beta + dm$$

$$am - dm = \beta - b$$

$$m(\alpha - \alpha) = \beta - b$$

$$\text{Ans. } m = \frac{\beta - b}{\alpha - \alpha} \quad \text{is the mean}$$

solve

i) Given eqn of regression lines:  $x = -0.4y + 6.4$

and  $y = -0.6x + 4.6$ , Find (i) Mean of  $x$  &  $y$

ii) coeff. of correlation,

$$x = -0.4y + 6.4 \quad \text{--- (1)} \\ y = -0.6x + 4.6 \quad \text{--- (2)}$$

Let  $(\bar{x}, \bar{y})$  be the mean of  $(x, y)$ , then it satisfy the eqns of regression.

$$\text{i.e. } \bar{x} + 0.4\bar{y} - 6.4 = 0$$

$$\frac{0.6\bar{x} + \bar{y} - 4.6}{0.6} = 0$$

$$\text{Ans. } \boxed{\bar{x} = 6 ; \bar{y} = 1}$$

regression eqn of  $y$  on  $x$   $(y - \bar{y}) = b_{yx}(x - \bar{x})$

$$\begin{aligned} x &= -0.4y + 6.4 \quad \rightarrow x \text{ only} \\ (x - \bar{x}) &= b_{yx}(y - \bar{y}) \end{aligned}$$

$\therefore b_{yx} = 0.8$  on comparing

$$\begin{aligned} y &= -0.6x + 4.6 \quad \rightarrow y \text{ on } x \\ (y - \bar{y}) &= b_{yx}(x - \bar{x}) \end{aligned}$$

i. By comparing above equations  
we get  $b_{xy} = -0.4$  &  $b_{yx} = -0.6$

$$\begin{aligned} 40x &= 214 + 18y \\ x &= 18/40y + 214/40 \\ x &= 0.45y + 5.35 \end{aligned}$$

$$\therefore b_{xy} = 0.45$$

$$\begin{aligned} b_{xy} \times b_{yx} &= s_x^2 \\ \sqrt{-0.4 \times -0.6} &= s_x \\ \sqrt{-0.24} &= s_x \end{aligned}$$

$$\begin{aligned} \text{now } s_x^2 &= b_{xy} \times b_{yx} \\ s_x^2 &= 0.45 \times 0.8 \\ s_x &= \sqrt{0.36} \\ s_x &= 0.6 \end{aligned}$$

$\rightarrow$  coeff. of corre.

$$\text{now given } (5x)^2 = 9 \quad \rightarrow 5x = 3$$

2017, 19

Q. In partially destroyed lab records of an analysis of corr. data, variance of  $x = 9$ , regression eqns :  $-3x - 10y + 66 = 0$

$$40x - 18y = 214$$

Find S.D of  $y$  and corr. coeff of  $x$  &  $y$

$$\text{let } y \text{ on } x \Rightarrow 8x - 10y + 66 = 0$$

$$x \text{ on } y \Rightarrow 40x - 18y = 214$$

$$\boxed{\boxed{(x - \bar{x})^2 = 16}} \leftarrow \text{variance of } x$$

The equation of two regression lines, obtained in a correlation analysis of 60 observations are: -  $5x = 6y + 24$  and  $1000y = 768x - 3608$ . Then what is the correlation coefficient? Also show that ratio of coeff. of variability of  $x$  to  $y$  is  $\frac{5}{24}$  & what is the ratio of variance of  $x$  and  $y$ .

$$5x = 6y + 24 \quad (x - \bar{x}) = b_{xy}(y - \bar{y})$$

$$x = 1.2y + 4.8$$

$$\boxed{b_{xy} = 1.2}$$

$$1000y = 768x - 3608$$

$$y = 0.768x - 3.608$$

$$x = \sqrt{b_{xy} \times b_{yx}} = \sqrt{1.2 \times 0.768} = \sqrt{0.9168}$$

$$\boxed{n = 0.9168} \rightarrow \text{coeff of correlation}$$

$$b_{xy} = r \frac{\sigma_x}{\sigma_y} \text{ and } b_{yx} = r \frac{\sigma_y}{\sigma_x}$$

$$\bullet \text{ coeff} = \frac{b_{xy}}{b_{yx}} = \frac{x \frac{\sigma_x}{\sigma_y}}{y \frac{\sigma_y}{\sigma_x}}$$

$$\frac{1.2}{0.768} = \left( \frac{\sigma_x}{\sigma_y} \right)^2$$

$$\boxed{\frac{1.5625}{1.024} = \frac{\sigma_x^2}{\sigma_y^2}}$$

$$\text{coeff of variability of } x = \frac{\sigma_x}{\bar{x}}$$

$$\frac{5\bar{x}}{1000\bar{y}} = \frac{5}{768} \bar{x} - 3608$$

$$\bar{x} = 6 \text{ and } \bar{y} = 1$$

$$\text{coeff of variability of } x = \frac{\sigma_x}{\bar{x}} = \frac{\sigma_x}{6}$$

$$\text{Ratio of variability} = \frac{(r\sigma)}{(\frac{1}{6})}$$

$$= \frac{\sigma_x}{\sigma_y} \times \frac{1}{6}$$

$$\left( \frac{\sigma_x}{\sigma_y} \right)$$

$$\boxed{\text{Ratio of variability} = \frac{5}{24}}$$

Hence proved.