

## UNIT-IV

### Regression & Functions of R.V

- (a) Fit a SL line  $y = \bar{a}x + b$
- (b) Fit a Parabola  $y = ax^2 + bx + c$
- (c) Regression eqn of  $X$  on  $Y$
- (d) Regression eqn of  $Y$  on  $X$
- (e) Karl Pearson Correlation Coefficient (g1)

### (f) Functions of Random Variables

④

$$Y = T(X)$$

$X \sim R.V$

}  $\begin{cases} \xrightarrow{\text{pdf}} f_X(x) \\ \xrightarrow{\text{cdf}} F_X(x) \end{cases}$

$Y = T(X) \sim R.V$

}  $\begin{cases} \xrightarrow{\text{pdf}} f_Y(y) \\ \xrightarrow{\text{cdf}} F_Y(y) \end{cases}$

Fit a SL line:

$$y = ax + b \rightarrow ①$$

Now it's Normal eqns

$$① \sum y = \omega \sum x + \eta b$$

$$② \sum xy = \alpha \sum \tilde{x} + \beta \sum x$$

Solve above two eqns we get  $\alpha =$   
 $\beta =$

Sub.  $\alpha, \beta$  values in ①

we get required SL line

Fit a Parabola

$$y = \alpha \tilde{x}^2 + \beta x + c \rightarrow ①$$

Now its normal eqns

$$① \sum y = \alpha \sum \tilde{x}^2 + \beta \sum x + nc$$

$$② \sum xy = \alpha \sum \tilde{x}^3 + \beta \sum \tilde{x}^2 + c \sum x$$

$$③ \sum \tilde{x}^2 y = \alpha \sum \tilde{x}^4 + \beta \sum \tilde{x}^3 + c \sum \tilde{x}^2$$

Solve above three eqns we get  $\alpha =$   
 $\beta =$   
 $c =$

Sub.  $\alpha, \beta, c$  values in eqn ①

we get required parabola

① Fit a SL line to the given data

12	13	15	17	20	25
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x	10	12	13	16
y	10	22	24	27

1st line  $y = \omega x + b$  ✓

Now it's normal eqns

$$\left\{ \begin{array}{l} \sum y = \omega \sum x + n b \\ \sum xy = \omega \sum x^2 + b \sum x \end{array} \right\} \quad \begin{array}{l} \text{Hence} \\ n = 7 \end{array}$$

x	y	xy	$x^2$
10	10	100	100
12	22	264	144
13	24	312	169
16	27	432	256
12	29	493	289
20	33	660	400
25	37	925	625
$\sum x$	$\sum y$	$\sum xy$	$\sum x^2$
113	182	3186	1938

These values in above normal eqns

Sub. these values in

$$182 = a(113) + b(7)$$

$$3186 = a(1938) + b(113)$$

Solving above two eqns we get

$$a = 1.56, b = 0.82$$

Hence required SL line

$$y = 1.56x + 0.82$$

② Fit a parabola to the given data

$x$	0	1	2	3	4
$y$	1	1.8	1.3	2.5	6.3

to parabola:

$$y = ax^2 + bx + c$$

Now it's normal eqns

$$\sum y = a \sum x^2 + b \sum x + c n$$

here

$$n = 5$$

$$\sum xy = a \sum x^3 + b \sum x^2 + c \sum x$$

$$\sum xy = \alpha \sum x^2 + \beta \sum x^3 + C \sum x^4$$

$x$	$y$	$x^2$	$x^3$	$x^4$	$xy$	$x^2 y$
0	1	0	0	0	0	0
1	1.8	1	1	1	1.8	1.8
2	5.2	4	8	16	2.6	5.2
3	22.5	9	27	81	7.5	22.5
4	100.8	16	64	256	25.2	100.8
$\sum x$	$\sum y$	$\sum x^2$	$\sum x^3$	$\sum x^4$	$\sum xy$	$\sum x^2 y$
10	12.9	30	100	354	37.1	130.3

Sub. these values in above normal eqns

$$12.9 = \alpha(30) + \beta(10) + C(5) \quad \}$$

$$37.1 = \alpha(100) + \beta(30) + C(10) \quad \}$$

$$130.3 = \alpha(354) + \beta(100) + C(30)$$

From these three eqns we get

$$\alpha = 0.55 \quad \}$$

$$\beta = -1.07 \quad \}$$

$$c = 1.42$$

Hence required parabola

$$y = 0.55x^2 - 1.07x + 1.42$$

Regression of  $X$  on  $Y$

$$x - \bar{x} = g_1 \frac{\sigma_x}{\sigma_y} (y - \bar{y})$$

$\bar{x}$  = mean of  $X$

$\bar{y}$  = mean of  $Y$

Regression coeff of  $X$  on  $Y$

$$g_1 \frac{\sigma_x}{\sigma_y} = b_{XY} = \frac{\sum XY}{\sum Y^2} \rightarrow ①$$

Regression eqn of  $Y$  on  $X$

$$y - \bar{y} = g_1 \frac{\sigma_y}{\sigma_x} (x - \bar{x})$$

$\bar{x}$  = mean of  $X$

$\bar{y}$  = mean of  $Y$

Regression coeff of  $y$  on  $x$

$$g_1 \frac{\sigma_y}{\sigma_x} = b_{yx} = \frac{\Sigma XY}{\Sigma X^2} \rightarrow ②$$

Now  $① \times ②$

$$g_1^2 = b_{xy} \times b_{yx} = \frac{[\Sigma XY]^2}{\Sigma X^2 \Sigma Y^2}$$

$$g_1 = \sqrt{b_{xy} \times b_{yx}} = \frac{\Sigma XY}{\sqrt{\Sigma X^2 \Sigma Y^2}}$$

where  $g_1$  = Karl Pearson Correlation Coeff

$$-1 \leq g_1 \leq 1$$

$g_1 = 1$   $x$  &  $y$  are in same direction

$g_1 = -1$   $x$  &  $y$  are in opposite direction

$g_1 = 0$   $x$  &  $y$  are independent

① Given the bi-variate data

$x$	1	5	3	2	1	1	7	3
$y$	6	1	0	0	1	2	1	5

sum of  $x$  on  $y$

Then find (1) Regression eqn of  $y$  on  $x$   
 (2) Regression eqn of  $x$  on  $y$   
 (3) Karl Pearson's  $c_o.c$  (91)

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

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

$$\textcircled{3} \quad \text{SI} = \frac{\sum xy}{\sqrt{\sum x^2} \sqrt{\sum y^2}}$$

Q Here  $\bar{x} = 2.8$

$$\bar{y} = 2$$

$x$	$y$	$x$ $x - \bar{x}$ $x - 2.8$	$y$ $y - \bar{y}$ $y - 2$	$xy$	$x^2$	$y^2$
1	6	-1.8	4	-7.2	3.24	16
5	1	2.2	-1	-2.2	4.84	1
3	0	0.2	-2	-0.4	0.04	4
2	0	-0.8	-2	1.6	0.64	4
1	1	-1.8	-1	1.8	3.24	1
						3.24

1	2	-10.8	0	-4.2	17.64	1
7	1	4.2	-1	0.6	0.04	9
3	5	0.2	3			
				$\Sigma XY$ = -10	$\Sigma X^2$ = 32.92	$\Sigma Y^2$ = 36

$$\Sigma XY = -10$$

$$\bar{x} = 2.8$$

$$\Sigma X^2 = 32.92$$

$$\bar{y} = 2$$

$$\Sigma Y^2 = 36$$

Regression eqn of  $x$  on  $y$

$$x - \bar{x} = g_1 \frac{\Sigma x}{\Sigma y} (y - \bar{y})$$

$$g_1 \frac{\Sigma x}{\Sigma y} = \frac{\Sigma XY}{\Sigma Y^2}$$

$$= \frac{-10}{36}$$

$$= -0.27$$

$$g_1 \frac{\sigma_x}{\sigma_y} = -0.27$$

Hence required Reg. eqn of  $X$  on  $Y$

$$X - 2.8 = -0.27 (Y - 2)$$

$$X = 2.8 - 0.27Y + 0.54$$

$$X = -0.27Y + 3.34$$

Regression eqn of  $Y$  on  $X$

$$Y - \bar{Y} = g_1 \frac{\sigma_y}{\sigma_x} (X - \bar{X})$$

$$g_1 \frac{\sigma_y}{\sigma_x} = \frac{\sum XY}{\sum X^2}$$

$$= \frac{-10}{32.92}$$

$$= -0.303$$

Hence

$$-0.303(X - 2.8)$$

$$Y - 2 = (-0.303)x + 0.8484$$

$$Y = 2 - 0.303x + 0.8484$$

$$Y = -0.303x + 2.8484$$

Now

Karl Pearson Correlation coeff

$$r = \frac{\sum xy}{\sqrt{\sum x^2 \sum y^2}}$$

$$= \frac{-10}{\sqrt{32.92 \times 36}}$$

$$r = -0.29$$

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

Here  $r$  is -ve

$x$  &  $y$  are in opposite direction

① Heights of fathers and sons  
are given in inches

HL of father	65	66	67	67	68	69	71	73
HL of son	67	68	64	68	72	70	69	70

① SL line  $y = \omega x + b$

② Parabola  $y = \omega x^2 + bx + c$

③ Reg. eqn of  $X$  on  $Y$

④ Reg. eqn of  $Y$  on  $X$

⑤  $g_1$

⑥  $\tan \theta = \frac{m_1 - m_2}{1 + m_1 m_2}$

## Functions of Random variable

$X$  is  $\sim R.V$

$f_X(x)$  is p. d. f

$(X, f(x)) \sim P.D.F$

$F(x)$  is C.D.F

$$F(x) = P(X \leq x) = \int_{-\infty}^x f_X(x) dx$$

Define  $\rightarrow$  Functions of  $\mathbb{R}^N$  of  $y$

$$Y = T(X)$$

Here  $Y$  is another  $\mathbb{R}^N$

$$f_Y(y) \text{ is pdf} \quad \int_{-\infty}^{\infty} f_Y(y) dy = 1$$

$F_Y(y)$  is CDF

$$F_Y(y) = P(Y \leq y) = \int_{y=-\infty}^y f_Y(y) dy$$

$$\frac{d}{dy} F_Y(y) = f_Y(y)$$

①  $Y = \omega X + b$

②  $Y = X^2$

③  $\max\{x_1, y\} = T(z)$

④  $\min\{x_1, y\} = T(z)$

⑤  $\dots$

③ sum {x, y}

① Consider the function  $Y = \omega X + b$ , where  
a and b are constants. Then find  
C.D.F of Y and P.D.F of Y

so Given  $Y = \omega X + b$

C.D.F of Y is

$X \sim N$   
 $f_X(x)$   
 $F_X(x)$

$$F_Y(y) = P(Y \leq y)$$

$$= P(\omega X + b \leq y)$$

$$= P(X \leq \frac{y-b}{\omega})$$

$$= F_X\left(\frac{y-b}{\omega}\right)$$

$$F_Y(y) = F_X\left(\frac{y-b}{\omega}\right)$$

P.D.F of Y

$$f_Y(y) = f_X\left(\frac{y-b}{\omega}\right) \cdot \frac{1}{\omega}$$

$$\boxed{\frac{d}{dy} F_Y(y) = f_Y(y)}$$

$$\frac{d}{dy} F_X\left(\frac{y-b}{a}\right) = \frac{1}{a} f_X\left(\frac{y-b}{a}\right)$$

$$\boxed{f_Y(y) = \frac{1}{a} f_X\left(\frac{y-b}{a}\right)} \quad -$$

② Find C.D.F and P.d.f of  
 $Y = 2x + 3$  where  $x$  is  $R.V$

so

Given  $Y = 2x + 3$

$$\begin{aligned} F_Y(y) &= P(Y \leq y) \\ &= P(2x + 3 \leq y) \end{aligned}$$

$$\text{CDF of } Y = P(X \leq \frac{y-3}{2})$$

$$\boxed{F_Y(y) = F_X\left(\frac{y-3}{2}\right)} \quad -$$

$$\text{Now } \frac{d}{dy} F_Y(y) = \frac{d}{dy} F_X\left(\frac{y-3}{2}\right)$$

Part of  $y$  
$$f_Y(y) = \frac{1}{2} f_X\left(\frac{y-3}{2}\right)$$

③ Find the C.D.F and P.D.F of

Part of  
function

$$Y = X^2$$

Q

$$\text{Given } Y = X^2$$

$$F_Y(y) = P(Y \leq y)$$

$$= P(X^2 \leq y)$$

$$= P(X \leq \pm \sqrt{y})$$

$$= P(|X| \leq \sqrt{y})$$

$$= P(-\sqrt{y} \leq X \leq \sqrt{y})$$

$$F_Y(y) = F_X(\sqrt{y}) - F_X(-\sqrt{y})$$

$$\text{Now } \frac{d}{dy} F_Y(y) = \frac{d}{dy} \left[ F_X(\sqrt{y}) - F_X(-\sqrt{y}) \right]$$

$$= \frac{1}{2\sqrt{y}} f_X(\sqrt{y}) + \frac{1}{2\sqrt{y}} f_X(-\sqrt{y})$$

$$f_Y(y) = \frac{f_X(\sqrt{y}) + f_X(-\sqrt{y})}{2\sqrt{y}}$$

④ Maximum of two independent random variables

$$\text{let } \omega = \max\{x, y\}$$

now its CDF

$$\begin{aligned} F_\omega(\omega) &= P(\omega \leq \omega) \\ &= P(\max(x, y) \leq \omega) \\ &= P\{(x \leq \omega) \cap (y \leq \omega)\} \\ &= F_{x,y}(\omega) \end{aligned}$$

$$F_w(w) = F_x(w) F_y(w)$$

$$f_w(w) = \frac{d}{dw} F_w(w)$$

$$= \frac{d}{dw} [F_x(w) F_y(w)]$$

$$f_w(w) = f_x(w) F_y(w) + F_x(w) f_y(w)$$

Problem

- ⑤ Assume that  $w = \max\{x, y\}$  where  $x \geq y$  are independent r.v. with respective p.d.  $f_x$  and  $f_y$

$$f_x(x) = \lambda e^{-\lambda x} \cdot x \geq 0$$

$$f_y(y) = \mu e^{-\mu y} \cdot y \geq 0$$

where  $\lambda > 0, \mu > 0$  what is p.d. for  $w$

Ans Since

$$F_x(x) = P(X \leq x) = \int_{-\infty}^x f_x(t) dt$$

$$= \int_{-\infty}^x \lambda e^{-\lambda t} dt$$

$$= \lambda \left[ \frac{e^{-\lambda x}}{-\lambda} \right]_0^x$$

$$= -[e^{-\lambda x} - 1]$$

$$\boxed{F_X(x) = 1 - e^{-\lambda x}}$$

$$\text{By } F_Y(y) = P(Y \leq y) = \int_{-\infty}^y f_Y(y) dy$$

$$= \int_0^y \mu e^{-\mu y} dy$$

$$\boxed{F_Y(y) = 1 - e^{-\mu y}}$$

NOW P.d.f of  $w$  is given by

$$f_w(w) = S_X(w) F_Y(w) + F_X(w) f_Y(w)$$

$$= \lambda e^{-\lambda w} (1 - e^{-\mu w}) + \mu e^{-\mu w} (1 - e^{-\lambda w})$$

$$\boxed{f_w(w) = \lambda e^{-\lambda w} + \mu e^{-\mu w} - (\lambda + \mu) e^{-(\lambda + \mu)w}}$$

## ⑥ Minimum of two independent Random Variables

$$\text{let } \omega = \min\{X, Y\}$$

$$\text{now } F_\omega(\omega) = P(\omega \leq \omega)$$

$$= P(\min(X, Y) \leq \omega)$$

$$= P((X \leq \omega, X \leq Y) \cup (Y \leq \omega, Y \leq X))$$

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

$$= F_X(\omega) + F_Y(\omega) - F_{X,Y}(\omega, \omega)$$

$$F_\omega(\omega) = F_X(\omega) + F_Y(\omega) - F_X(\omega) F_Y(\omega)$$

since  $X$  &  $Y$  independent

$$\text{Now } f_\omega(\omega) = \frac{d}{d\omega} F_\omega(\omega)$$

$$= f_X(\omega) + f_Y(\omega) - [f_X(\omega) F_Y(\omega) \\ + F_X(\omega) f_Y(\omega)]$$

$$f_\omega(\omega) = f_X(\omega) [1 - F_Y(\omega)] + f_Y(\omega) [1 - F_X(\omega)]$$

problem

⑦ Assume that  $W = \min\{X, Y\}$  where  $X$  &  $Y$  are independent random variables with respective p.d.f.s

$$f_X(x) = \lambda e^{-\lambda x} \quad x \geq 0$$

$$f_Y(y) = \mu e^{-\mu y} \quad y \geq 0$$

where  $\lambda > 0, \mu > 0$  what is the p.d.f of  $W$

Sol: please solve

$$\text{Same as } W = \max\{X, Y\}$$

sum of two independent random variables

$$\text{let } S = X + Y$$

$$F_S(s) = P(S \leq s)$$

$$= P(X + Y \leq s)$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{s-y} f_{XY}(x, y) dx dy$$

$$y = -\infty, x = -\infty$$

$$x = s - y$$

$$= \int_{y=-\infty}^{\infty} \int_{x=-\infty}^{s-y} f_x(x) f_y(y) dx dy$$

$$= \int_{y=-\infty}^{\infty} \left[ \int_{x=-\infty}^{s-y} f_x(x) dx \right] f_y(y) dy$$

$$F_s(s) = \int_{y=-\infty}^{\infty} F_x(s-y) f_y(y) dy$$

$$\text{Now } \frac{d}{ds} F_s(s) = f_s(s)$$

$$\therefore f_s(s) = \int_{y=-\infty}^{\infty} \frac{d}{ds} F_x(s-y) f_y(y) dy$$

$$f_s(s) = \int_{y=-\infty}^{\infty} f_x(s-y) f_y(y) dy$$

By Convolution Theorem

$$f_s(s) = f_x(s) f_y(s)$$



