

Recap

$X \rightarrow \text{r.v.}$  pdf  $f(x)$

- $E(X) = \sum_x x f(x) \quad \text{or} \quad \int x f(x) dx$
- $E(g(x)) = \sum_x g(x) f(x) \quad \text{or} \quad \int g(x) f(x) dx$
- $\sigma_x^2 = E((X - \mu)^2) = \sum_x (x - \mu)^2 f(x) \quad \text{or} \quad \int (x - \mu)^2 f(x) dx$
- $\sigma_x^2 = E(X^2) - (E(X))^2$

$X$  &  $Y$  are two r.v. with  $f(x, y)$  joint pdf  
 $E[g(x, y)] = \sum_x \sum_y g(x, y) f(x, y)$

$S.D = \sqrt{\text{variance}}$

Today's work

$X$  is a r.v. with pdf  $f(x)$  and  $g(x)$  is any function

$$\sigma_{g(x)}^2 = E[(g(x) - \mu_{g(x)})^2]$$

$$= \sum (g(x) - \mu_{g(x)})^2 f(x)$$

Que

Calculate the variance of  $g(x) = 2x + 3$ , where  $x$  is a r.v. with pdf

$x$	0	1	2	3
$f(x)$	$\frac{3}{4}$	$\frac{1}{8}$	$\frac{1}{2}$	$\frac{1}{8}$

Sol<sup>n</sup>

$$\mu_{g(x)} = \sum_x g(x) f(x) = \sum_x (2x + 3) f(x)$$

$$= (2 \cdot 0 + 3) \frac{3}{4} + (2 \cdot 1 + 3) \frac{1}{8} + (2 \cdot 2 + 3) \frac{1}{2} + (2 \cdot 3 + 3) \frac{1}{8} = \boxed{6}$$

$$\sigma_{g(x)}^2 = E[(g(x) - \overset{\text{mean}}{6})^2]$$

$$= \sum_x (g(x) - 6)^2 f(x) = (2 \cdot 0 + 3 - 6)^2 \cdot \frac{3}{4} + (2 \cdot 1 + 3 - 6)^2 \cdot \frac{1}{8} + (2 \cdot 2 + 3 - 6)^2 \cdot \frac{1}{2} + (2 \cdot 3 + 3 - 6)^2 \cdot \frac{1}{8} = \boxed{4}$$

Que

$X$  is a continuous r.v. with pdf  $f(x) = \begin{cases} \frac{x^2}{3} & -1 < x < 2 \\ 0 & \text{otherwise} \end{cases}$

Find variance of  $g(x) = 4x+3$

Soln

$$E[g(x)] = \int_{-\infty}^{\infty} g(x)f(x)dx = \int_{-1}^2 (4x+3) \frac{x^2}{3} dx = 8$$

$$\sigma_{g(x)}^2 = \int_{-\infty}^{\infty} (g(x) - 8)^2 f(x)dx = \int_{-1}^2 (4x+3-8)^2 \frac{x^2}{3} dx = \frac{51}{5} \text{ Ans.}$$

Next

### Covariance

We have two r.v.  $X$  and  $Y$

Suppose  $Y$  depends on  $X$

We would like to know the kind of dependence.

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will be measured by covariance.

Definition given two r.v.  $X$  &  $Y$  with joint pdf  $f(x,y)$ , their covariance denoted by  $\text{cov}(X,Y)$  is defined as

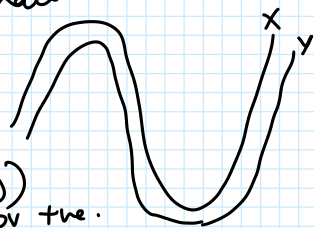
$$E[(X-M_x)(Y-M_y)] \quad \checkmark$$

If  $X$  and  $Y$  are together large or together small

$X$  is large  
 $Y$  is large

$X-M_x \rightarrow \text{positive}$   
 $Y-M_y \rightarrow \text{positive}$

$(X-M_x)(Y-M_y)$   
positive  
 $E[(X-M_x)(Y-M_y)]$   
+ve  $\Rightarrow \text{cov} \rightarrow +ve$



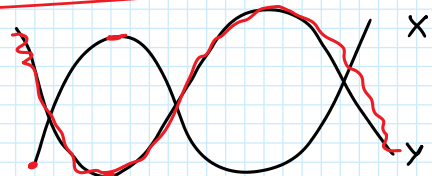
$X$  is small  
 $Y$  is small

$X-M_x \rightarrow -ve$   
 $Y-M_y \rightarrow -ve$

product +ve  
cov  $\rightarrow +ve$

$X$  large  $\rightarrow X-M_x +ve$   
 $Y$  small  $\rightarrow Y-M_y -ve$

product -ve  
 $\therefore \text{cov} \rightarrow -ve$



$X$  small  $\rightarrow X-M_x -ve$   
 $Y$  large  $\rightarrow Y-M_y +ve$

product -ve  
 $\therefore \text{cov} \rightarrow -ve$

~~$E(XY) = E(X)E(Y)$~~

• If  $X$  and  $Y$  are independent  $\Rightarrow \text{cov}(X,Y) = 0$

• Unbiased estimator of covariance is  $\frac{1}{n-1} \sum (x_i - \bar{x})(y_i - \bar{y})$

however if  $\text{cov}(X, Y) = 0 \nRightarrow X$  and  $Y$  are independent.

Cov only captures the linear relationship b/w  $X$  &  $Y$

$$Y = \underbrace{\alpha + \beta X}_{\text{linear}} + \gamma X^2 + \dots$$

Notation  $\sigma_{xy} = \text{cov}(X, Y)$

Result

$$\sigma_{xy} = E(XY) - E(X)E(Y)$$

$$\sigma_{xy} = E[(X - \mu_x)(Y - \mu_y)]$$

Proof

$$\sigma_{xy} = \sum_x \sum_y (x - \mu_x)(y - \mu_y) f(x, y)$$

$$= \sum_x \sum_y (xy - \mu_y x - \mu_x y + \mu_x \mu_y) f(x, y)$$

$$= \sum_x \sum_y xy f(x, y) - \sum_x \sum_y \mu_y x f(x, y) - \sum_x \sum_y \mu_x y f(x, y) + \sum_x \sum_y \mu_x \mu_y f(x, y)$$

$$= \sum_x \sum_y xy f(x, y) - \mu_y \sum_x \sum_y x f(x, y) - \mu_x \sum_x \sum_y y f(x, y) + \mu_x \mu_y \sum_x \sum_y f(x, y)$$

$$= E(XY) - \mu_y E(X) - \mu_x E(Y) + \mu_x \mu_y$$

$$\sigma_{xy} = E(XY) - \mu_x \mu_y$$

Covariance

$$\sigma_{xy} = E(XY) - E(X)E(Y)$$

~~correlation~~ co

correlation coefficient

$$\rho_{xy} = \frac{\sigma_{xy}}{\sigma_x \sigma_y}$$

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

↓

Not only tells type of relation b/w  $X$  &  $Y$   
it also tells you how strongly they are related.

Que For  $X$  and  $Y$  with joint pdf  $f(x, y)$  given below find  $\sigma_{xy}$  and  $\rho_{xy}$

		$x$			
		0	1	2	$h(y)$
$y$	0	$3/28$	$9/28$	$3/28$	$15/28$
	1	$3/14$	$3/14$	0	$3/7$
	2	0	0	0	$1/28$

$$\begin{aligned} \mu_x &= \sum_x x g(x) \\ &= 0 * \frac{5}{14} + 1 * \frac{15}{28} + 2 * \frac{3}{28} \\ &= \frac{3}{4} \end{aligned}$$

$$U$$

$\begin{pmatrix} 1 \\ 2 \end{pmatrix}$	$\begin{pmatrix} 3/14 \\ 1/28 \end{pmatrix}$	$\begin{pmatrix} 3/14 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 0 \\ 0 \end{pmatrix}$	$\begin{pmatrix} 3/14 \\ 1/28 \end{pmatrix}$
$g(x)$	$\begin{pmatrix} 5/14 \\ 15/28 \end{pmatrix}$	$\begin{pmatrix} 15/28 \end{pmatrix}$	$\begin{pmatrix} 3/28 \end{pmatrix}$	

$$M_y = \sum y h(y) = 0 \cdot \frac{15}{28} + 1 \cdot \frac{3}{7} + 2 \cdot \frac{1}{28} = \frac{3}{4}$$

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

$$E[XY] = \sum_x \sum_y xy f(x,y) = (0 \cdot 0) f(0,0) + (1 \cdot 0) \frac{9}{28} + (2 \cdot 0) \frac{3}{28} + (0 \cdot 1) \frac{3}{14} + (1 \cdot 1) \frac{3}{14} + (0 \cdot 2) \frac{1}{28} = \frac{3}{4}$$

$$\sigma_{xy} = E(XY) - E(X)E(Y) = \frac{3}{4} - \frac{3}{4} \cdot \frac{1}{2} = -\frac{9}{56}$$

$$\sigma_y^2 = E(Y^2) - (E(Y))^2 = \frac{4}{7} - \left(\frac{1}{2}\right)^2 = \frac{9}{28}$$

$$\sigma_{xy} = \frac{\overline{\sigma_{xy}}}{\sigma_x \sigma_y} = \frac{-9/56}{\sqrt{\frac{45}{112}} \sqrt{\frac{9}{28}}} = -\frac{1}{\sqrt{5}}$$

$$\sigma_x^2 = E(X^2) - (E(X))^2$$

$$E(X^2) = \sum x^2 g(x) = 0^2 \cdot \frac{5}{14} + 1^2 \cdot \frac{15}{28} + 2^2 \cdot \frac{3}{28} = \frac{27}{28}$$

$$\sigma_x^2 = \frac{27}{28} - \left(\frac{3}{4}\right)^2 = \frac{45}{112}$$