

# STAT230: Principles of Probability

## Unit 7: Product moments, covariance and Conditional Expectation

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**1** Product moments and Covariance

**2** Moments of linear combination of random variables

**3** Conditional Expectation

## Definition

Let  $X$  and  $Y$  are two random variables then the  $r$ th and  $s$ th non-central product moment of  $X$  and  $Y$  is defined by  $\mu'_{r,s} = E\{X^r Y^s\}$ .

The  $r$ th and  $s$ th central product moment of  $X$  and  $Y$  is defined by

$$\mu_{r,s} = E\{(X - \mu_X)^r (Y - \mu_Y)^s\}.$$

$\mu_{1,1} = E\{(X - \mu_X)(Y - \mu_Y)\}$  is called the covariance of  $X$  and  $Y$ .

Note that  $\mu_{1,1} = Cov(X, Y) = \mu'_{1,1} - \mu_X \mu_Y = E(XY) - E(X)E(Y)$ .

## Theorem

*If  $X$  and  $Y$  are independent then  $Cov(X, Y) = 0$  i.e.  $E(XY) = E(X)E(Y)$ . The converse is not true.*

Note that  $Var(X) = Cov(X, X)$ .

## Example 1

### Example (1)

Suppose  $X$  and  $Y$  have the following joint distribution

		$x$		
		0	1	2
$y$	0	$1/6$	$1/3$	$1/12$
	1	$2/9$	$1/6$	0
	2	$1/36$	0	0

Find the covariance of  $X$  and  $Y$ .

Are  $X$  and  $Y$  independent?

## Solution for Example 1

		x			$f_Y(y)$
		0	1	2	
y	0	1/6	1/3	1/12	7/12
	1	2/9	1/6	0	7/18
	2	1/36	0	0	1/36
$f_X(x)$		15/36	1/2	1/12	

$$E(XY) = \sum_x \sum_y xyf(x, y) = 0*0*1/6 + 1*0*1/3 + \dots + 2*2*0 = 1/6.$$

$$E(X) = \sum_x xf_X(x) = 0 * 15/36 + 1 * 1/2 + 2 * 1/12 = 2/3$$

$$E(Y) = \sum_y yf_Y(y) = 0 * 7/12 + 1 * 7/18 + 2 * 1/36 = 4/9$$

$$Cov(X, Y) = E(XY) - E(X)E(Y) = 1/6 - 2/3 * 4/9 = -7/54.$$

$X$  and  $Y$  are not independent since  $Cov(X, Y) \neq 0$ .

## Example 2

### Example (2)

Let  $X$  and  $Y$  have joint density

$$f(x, y) = \begin{cases} \frac{x+y+1}{2} & \text{for } 0 < x < 1, 0 < y < 1 \\ 0 & \text{elsewhere.} \end{cases}$$

Find the covariance of  $X$  and  $Y$ .

Are  $X$  and  $Y$  independent?

## Solution for Example 2

$$E(XY) = \int_0^1 \int_0^1 xyf(x,y)dydx = \int_0^1 \int_0^1 xy \frac{x+y+1}{2} dydx = \frac{7}{24}.$$

$$E(X) = \int_0^1 \int_0^1 xf(x,y)dydx = \int_0^1 \int_0^1 x \frac{x+y+1}{2} dydx = \frac{13}{24}.$$

$$E(Y) = \int_0^1 \int_0^1 yf(x,y)dydx = \int_0^1 \int_0^1 y \frac{x+y+1}{2} dydx = \frac{13}{24}.$$

$$\text{Cov}(X,Y) = E(XY) - E(X)E(Y) = \frac{7}{24} - \frac{13}{24} \times \frac{13}{24} = \frac{-1}{576}.$$

$X$  and  $Y$  are dependent since  $\text{Cov}(X,Y) \neq 0$ .



## Example 3

### Example (3)

Let  $X$  and  $Y$  have joint density

$$f(x, y) = \begin{cases} 2 & \text{for } x > 0, y > 0, x + y < 1 \\ 0 & \text{elsewhere.} \end{cases}$$

Find the covariance of  $X$  and  $Y$ .

Are  $X$  and  $Y$  independent?

## Solution for Example 3

$$E(XY) = \int_0^1 \int_0^1 xyf(x,y)dydx = \int_0^1 \int_0^{1-x} 2xydydx = \frac{1}{12}.$$

$$E(X) = \int_0^1 \int_0^1 xf(x,y)dydx = \int_0^1 \int_0^{1-x} 2xdydx = \frac{1}{3}.$$

$$E(Y) = \int_0^1 \int_0^1 yf(x,y)dydx = \int_0^1 \int_0^{1-x} 2ydydx = \frac{1}{3}.$$

$$Cov(X,Y) = E(XY) - E(X)E(Y) = \frac{1}{12} - \frac{1}{3} \times \frac{1}{3} = \frac{-1}{36}.$$

$X$  and  $Y$  are dependent since  $Cov(X,Y) \neq 0$ .

# Moments of linear combinations of R.V.

## Theorem

If  $X_1, X_2, \dots, X_n$  are random variables and  $Y = \sum_{i=1}^n a_i X_i$  where  $a_i$ 's are constants, then  $E(Y) = \sum_{i=1}^n a_i E(X_i)$  and

$$\text{Var}(Y) = \sum_{i=1}^n a_i^2 \text{Var}(X_i) + 2 \sum_{i=1}^{n-1} \sum_{j>i} a_i a_j \text{Cov}(X_i, X_j).$$

If the  $X_i$ 's are independent then

$$\text{Var}(Y) = \sum_{i=1}^n a_i^2 \text{Var}(X_i).$$

Note that

$$\text{Var}(Y) = \sum_{i=1}^n a_i^2 \text{Var}(X_i) + \sum_{i=1}^{n-1} \sum_{j \neq i} a_i a_j \text{Cov}(X_i, X_j).$$

and

$$\text{Var}(Y) = \sum_{i=1}^n \sum_{j=1}^n a_i a_j \text{Cov}(X_i, X_j).$$

## Theorem

If  $X_1, X_2, \dots, X_n$  are random variables and  $Y_1 = \sum_{i=1}^n a_i X_i$  and  $Y_2 = \sum_{i=1}^n b_i X_i$  where  $a_i$ 's and  $b_i$ 's are constants, then

$$\text{Cov}(Y_1, Y_2) = \sum_{i=1}^n a_i b_i \text{Var}(X_i) + \sum_{i=1}^n \sum_{j \neq i}^n a_i b_j \text{Cov}(X_i, X_j).$$

If the  $X_i$ 's are independent then

$$\text{Cov}(Y_1, Y_2) = \sum_{i=1}^n a_i b_i \text{Var}(X_i).$$

## Example 4

### Example (4)

For  $X$  and  $Y$  defined in Examples 1 and 2, let  $Z_1 = 2X + 4Y$  and  $Z_2 = X - 2Y$  find:

$E(Z_1)$ ,  $E(Z_2)$ ,  $Var(Z_1)$ ,  $Var(Z_2)$  and  $Cov(Z_1, Z_2)$ .

## Solution for Example 4

$$Z_1 = 2X + 4Y \text{ and } Z_2 = X - 2Y$$

$$E(Z_1) = 2E(X) + 4E(Y),$$

$$E(Z_2) = E(X) - 2E(Y),$$

$$\text{Var}(Z_1) = 4\text{Var}(X) + 16\text{Var}(Y) + 16\text{Cov}(X, Y),$$

$$\text{Var}(Z_2) = \text{Var}(X) + 4\text{Var}(Y) - 4\text{Cov}(X, Y) \text{ and}$$

$$\text{Cov}(Z_1, Z_2) = 2\text{Var}(X) - 4\text{Cov}(X, Y) + 4\text{Cov}(X, Y) - 8\text{Var}(Y).$$

	$E(X)$	$E(Y)$	$\text{Var}(X)$	$\text{Var}(Y)$	$\text{Cov}(X, Y)$
Example 1	$\frac{2}{3}$	$\frac{4}{9}$	$\frac{7}{18}$	$\frac{49}{162}$	$\frac{-7}{54}$
Example 2	$\frac{13}{24}$	$\frac{13}{24}$	$\frac{47}{576}$	$\frac{47}{576}$	$\frac{-1}{576}$

	$E(Z_1)$	$E(Z_2)$	$\text{Var}(Z_1)$	$\text{Var}(Z_2)$	$\text{Cov}(Z_1, Z_2)$
Example 1	$\frac{28}{9}$	$\frac{-2}{9}$	$\frac{1015}{162}$	$\frac{343}{162}$	$\frac{-133}{81}$
Example 2	$\frac{13}{4}$	$\frac{-13}{24}$	$\frac{77}{48}$	$\frac{239}{576}$	$\frac{-47}{96}$

## Example 5

### Example (5)

Let  $X$  and  $Y$  be two independent random variables with  $\mu_X = 2$ ,  $\sigma_X = 4$ ,  $\mu_Y = 3$  and  $\sigma_Y = 2$ . Let  $Z_1 = X + 2Y + 3$  and  $Z_2 = 3X - Y$ . Find:  $E(Z_1)$ ,  $E(Z_2)$ ,  $Var(Z_1)$ ,  $Var(Z_2)$  and  $Cov(Z_1, Z_2)$ .



## Solution for Example 5

We have  $\mu_X = 2$ ,  $\sigma_X = 4$ ,  $\mu_Y = 3$  and  $\sigma_Y = 2$ ,  $Cov(X, Y) = 0$  and  $Z_1 = X + 2Y + 3$  and  $Z_2 = 3X - Y$ .

$$E(Z_1) = E(X) + 2E(Y) + 3 = 11, \quad E(Z_2) = 3E(X) - E(Y) = 3,$$

$$Var(Z_1) = Var(X) + 4Var(Y) = 16 + 4 \times 4 = 32,$$

$$Var(Z_2) = 9Var(X) + Var(Y) = 9 \times 16 + 4 = 148 \text{ and}$$

$$Cov(Z_1, Z_2) = 3Var(X) - 2Var(Y) = 3 \times 16 - 2 \times 4 = 40.$$

The conditional Expectation of  $u(X)$  given  $Y = y$  is given by

$$E(u(X)|y) = \sum_x u(x)f_{X|Y}(x|y)$$

for discrete random variables and

$$E(u(X)|y) = \int_{-\infty}^{\infty} u(x)f_{X|Y}(x|y)dx$$

for continuous random variables.

## Example 6

### Example (6)

For  $X$  and  $Y$  defined in Example 1 find  $E(X|Y = 1)$ .

For  $X$  and  $Y$  in Example 2 find  $E(Y|X = 1/2)$ .

## Solution for Example 6

For Example 1 we have:

$x$	$f_{X Y=1}(x)$
0	$4/7$
1	$3/7$
2	0

$$E(X|Y = 1) = \sum_x x f_{X|Y=1}(x) = 3/7.$$

For Example 2 we have  $f_{X|Y=1/2}(x) = \frac{f(x, 1/2)}{f_Y(1/2)} = \frac{x}{2} + \frac{3}{4}$  therefore

$$E(Y|X = 1/2) = \int_0^1 x f_{X|Y=1/2}(x) = \int_0^1 x \left( \frac{x}{2} + \frac{3}{4} \right) = \frac{13}{24}.$$

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