

# Covariance, and a small taste of bivariate probability

[Not in the textbook!]

①

Q.: Suppose that  $X$  and  $Y$  are random variables and we look at  $Z = X + Y$ . Then  $E(Z) = E(X + Y) = E(X) + E(Y)$ , but if  $X$  and  $Y$  are not independent the  $V(Z)$  is harder to calculate

$$[V(X + Y) = V(X) + V(Y) \text{ if } \frac{\text{they}}{1} \text{ are independent.}]$$

If they are not independent we fall on the definition of  $V(Z) = V(X + Y)$

$$\begin{aligned} &= E((X + Y)^2) - [E(X) + E(Y)]^2 \\ &= E(X^2 + 2XY + Y^2) - [E(X)^2 + 2E(X)E(Y) + E(Y)^2] \\ &= E(X^2) + 2E(XY) + E(Y^2) \\ &\quad - [E(X)^2 + 2E(X)E(Y) + E(Y)^2] \\ &= V(X) + V(Y) + 2 \underbrace{[E(XY) - E(X)E(Y)]} \end{aligned}$$

$\text{Cov}(X, Y)$

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

(2)

We will only play with this when  
X and Y are discrete random variables  
with only finitely many possible values each.

It's too complicated at this level to  
handle anything more complicated.

es the continuous case requires

multivariate calculus.

Example:

$Y \setminus X$	-1	1
0	0.1	0.2
1	0.3	0.4

The sum of  
all the  
combined  
probabilities  
must be 1  
 $= 0.1 + 0.2 + 0.3 + 0.4$

so the "joint probability function" of X & Y

$$m(0, -1) = 0.1$$

$$m(0, 1) = 0.2$$

$$m(1, -1) = 0.3$$

$$m(1, 1) = 0.4$$

also  $m_x(-1) = 0.1 + 0.3 = 0.4$

$$m_x(1) = 0.2 + 0.4 = 0.6$$

$$m_y(0) = 0.1 + 0.2 = 0.3$$

$$m_y(1) = 0.3 + 0.4 = 0.7$$

Suppose  $Z = X + Y$ . We want to know  $E(Z)$  and  $V(Z)$ . ③

$$E(X) = (-1) \cdot (0.1 + 0.3) + 1 \cdot (0.2 + 0.4) = -0.4 + 0.6 = 0.2$$

$$E(Y) = 0 \cdot (0.1 + 0.2) + 1 \cdot (0.3 + 0.4) = 0 + 0.7 = 0.7$$

so  $E(Z) = E(X + Y) = E(X) + E(Y) = 0.2 + 0.7 = 0.9$

$$E(X^2) = (-1)^2 \cdot (0.1 + 0.3) + 1^2 \cdot (0.2 + 0.4) = 0.4 + 0.6 = 1$$

$$E(Y^2) = 0^2 \cdot (0.1 + 0.2) + 1^2 \cdot (0.3 + 0.4) = 0 + 0.7 = 0.7$$

$$E(XY) = \overset{x}{(-1)} \cdot \overset{y}{0} \cdot 0.1 + \overset{x}{(-1)} \cdot \overset{y}{1} \cdot 0.3 + 1 \cdot 0 \cdot 0.2 + 1 \cdot 1 \cdot 0.4 = 0 - 0.3 + 0 + 0.4 = 0.1$$

$$V(X) = E(X^2) - [E(X)]^2 = 1 - [0.2]^2 = 1 - 0.04 = 0.96$$

$$V(Y) = E(Y^2) - [E(Y)]^2 = 0.7 - [0.7]^2 = 0.7 - 0.49 = 0.21$$

$$\begin{aligned} \text{Cov}(X, Y) &= E(XY) - E(X) \cdot E(Y) \\ &= 0.1 - 0.2 \cdot 0.7 = 0.1 - 0.14 = -0.04 \end{aligned}$$

$$\begin{aligned} V(X + Y) &= V(X) + V(Y) - 2 \text{Cov}(X, Y) \\ &= 0.96 + 0.21 - 2 \cdot (-0.04) \\ &= 1.17 + 0.08 = 1.25 \end{aligned}$$

If  $X$  &  $Y$  are independent, then

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

so  $\text{Cov}(X, Y) = 0$ . However,

if  $\text{Cov}(X, Y) = 0$ , the  $X$  and  $Y$  might or might not be independent.

Example:

$Y \backslash X$	0	1	2
-1	0.2	0	0.2
0	0	0.2	0
1	0.2	0	0.2

$X$  &  $Y$  are

not independent

since

$$P(X=1|Y=0)$$

$$= 1 \neq P(X=1) = 0.2$$

$$E(X) = 0 \cdot 0.4 + 1 \cdot 0.2 + 2 \cdot 0.4 = 1$$

$$E(X+Y)$$

$$= 1+0$$

$$E(Y) = -1 \cdot 0.4 + 0 \cdot 0.2 + 1 \cdot 0.4 = 0$$

$$= 1$$

$$E(X^2) = 0^2 \cdot 0.4 + 1^2 \cdot 0.2 + 2^2 \cdot 0.4 = 1.8$$

$$V(X) = 1.8 - 1^2 = 0.8$$

$$E(Y^2) = (-1)^2 \cdot 0.4 + 0^2 \cdot 0.2 + 1^2 \cdot 0.4 = 0.8$$

$$V(Y) = 0.8 - 0^2 = 0.8$$

$$E(XY) = 0 \cdot (-1) \cdot 0.2 + 0 \cdot 0 \cdot 0 + 0 \cdot 1 \cdot 0.2$$

$$+ 1 \cdot (-1) \cdot 0 + 1 \cdot 0 \cdot 0.2 + 1 \cdot 1 \cdot 0$$

$$+ 2 \cdot (-1) \cdot 0.2 + 2 \cdot 0 \cdot 0 + 2 \cdot 1 \cdot 0.2 = -0.4 + 0.4 = 0$$

$$\text{so } \text{Cov}(X, Y) = E(XY) - E(X)E(Y) = 0 - 1 \cdot 0 = 0$$

So  $X$  and  $Y$  are "uncorrelated" but not independent

# Basic Properties of Covariance

⑤

(0) If  $X$  and  $Y$  are independent, then  
$$\text{Cov}(X, Y) = 0.$$

$$(1) \quad V(X) = \text{Cov}(X, X)$$

$$(2) \quad \text{Cov}(X+Z, Y) = \text{Cov}(X, Y) + \text{Cov}(Z, Y)$$

$$\text{Cov}(X, Y+Z) = \text{Cov}(X, Y) + \text{Cov}(X, Z)$$

$$\text{Cov}(cX, Y) = c \text{Cov}(X, Y)$$

$$\text{Cov}(X, cY) = c \text{Cov}(X, Y)$$

$$(3) \quad \text{Cov}(X, Y) = \text{Cov}(Y, X)$$

$$(4) \quad \text{V}(X+Y) = V(X) + V(Y) - 2\text{Cov}(X, Y)$$