

Covariance and correlation.

- Covariance. Suppose that X and Y are random variables with $E(X) = \mu_X$ and $E(Y) = \mu_Y$. Then the covariance of X and Y , denoted by $Cov(X, Y)$, is the expectation of $(X - \mu_X)(Y - \mu_Y)$. That is,

$$Cov(X, Y) = E((X - \mu_X)(Y - \mu_Y)) = E(XY) - E(X)E(Y).$$

- Suppose that X and Y are discrete random variables. Then $Cov(X, Y)$ can be determined if $P((X, Y) = (x, y))$ is given for each (x, y) .
- Example 1. Suppose that

$$P((X, Y) = (x, y)) = \begin{cases} 0.1 & \text{if } (x, y) = (1, -2); \\ 0.3 & \text{if } (x, y) = (2, -4); \\ 0.6 & \text{if } (x, y) = (3, -6); \\ 0 & \text{otherwise.} \end{cases}$$

Find $E(X)$, $E(Y)$, $E(XY)$ and $Cov(X, Y)$.

Sol.

$$E(X) = 1 \times 0.1 + 2 \times 0.3 + 3 \times 0.6 = 2.5,$$

$$E(Y) = (-2) \times 0.1 + (-4) \times 0.3 + (-6) \times 0.6 = -5,$$

$$E(XY) = (1)(-2) \times 0.1 + (2)(-4) \times 0.3 + (3)(-6) \times 0.6 = -13.4,$$

and

$$Cov(X, Y) = E(XY) - E(X)E(Y) = -13.4 - (2.5) \times (-5) = -0.9.$$

- If X and Y are independent, then $E(XY) = E(X)E(Y)$ and $Cov(X, Y) = 0$.
- Rules for covariance calculation.
 1. $Cov(X, X) = Var(X)$.
 2. $Cov(X, Y) = Cov(Y, X)$.
 3. Linearity (in one variable when the other variable is held fixed).

$$\begin{aligned} Cov(aX, Y) &= aCov(X, Y) = Cov(X, aY) \\ Cov(X_1 + X_2, Y) &= Cov(X_1, Y) + Cov(X_2, Y) \quad \text{and} \\ Cov(X, Y_1 + Y_2) &= Cov(X, Y_1) + Cov(X, Y_2). \end{aligned}$$

- $Var(aX + bY) = a^2Var(X) + b^2Var(Y) + 2abCov(X, Y)$.

Example 2. Suppose that $Cov(X, Y) = -0.9$, $Var(X) = 0.45$ and $Var(Y) = 1.8$. Find $Var(2X + Y)$.

Sol. $Var(2X + Y) = 4Var(X) + 4Cov(X, Y) + Var(Y) = 4 \times 0.45 + 4 \times (-0.9) + 1.8 = 0$.

- Correlation. Suppose that $Var(X) > 0$ and $Var(Y) > 0$. Then the correlation (or correlation coefficient) between X and Y , denoted by $Corr(X, Y)$, is

$$\frac{Cov(X, Y)}{\sqrt{Var(X)}\sqrt{Var(Y)}}.$$

- Example 3. Consider the (X, Y) in Example 1. Find $Corr(X, Y)$.

Sol. From the solution for Example 1, we have $Cov(X, Y) = -0.9$, $E(X) = 2.5$ and $E(Y) = -5$. Therefore,

$$Var(X) = E(X^2) - (2.5)^2 = 1^2 \times 0.1 + 2^2 \times 0.3 + 3^2 \times 0.6 - (2.5)^2 = 0.45,$$

$$Var(Y) = E(Y^2) - (-5)^2 = ((-2) \times 1)^2 \times 0.1 + ((-2) \times 2)^2 \times 0.3 + ((-2) \times 3)^2 \times 0.6 - (-5)^2 = 1.8,$$

and

$$Corr(X, Y) = \frac{-0.9}{\sqrt{0.45} \times 1.8} = -1.$$

- Properties of correlation. Suppose that $Var(X) > 0$ and $Var(Y) > 0$.
 1. $-1 \leq Corr(X, Y) \leq 1$.
 2. $|Corr(X, Y)| = 1 \Leftrightarrow Y = a + bX$ for some a, b .
 3. Suppose that $Y = a + bX$ for some a, b (with probability one). Then

$$Corr(X, Y) = \begin{cases} 1 & \text{if } b > 0; \\ -1 & \text{if } b < 0. \end{cases}$$

4. If X and Y are independent, then $Corr(X, Y) = 0$.

- Note that

$$Corr(X, Y) = 0 \not\Leftrightarrow X \text{ and } Y \text{ are independent}$$

Example 4. Suppose that $P(X = 0) = P(X = 1) = P(X = -1) = 1/3$ and $Y = X^2$. Then $Cov(X, Y) = 0$ but X and Y are not independent since $P((X, Y) = (0, 0)) \neq P(X = 0)P(Y = 0)$.

Proof of $-1 \leq \text{Corr}(X, Y) \leq 1$ assuming $\text{Var}(X) > 0$ and $\text{Var}(Y) > 0$.

Consider the problem of finding constants a and b so that

$$E(Y - a - bX)^2$$

is minimized. Note that

$$\begin{aligned} E(Y - a - bX)^2 &= \text{Var}(Y - a - bX) + (E(Y - a - bX))^2 \\ &= \text{Var}(Y - bX) + (E(Y) - bE(X) - a)^2 \\ &= b^2 \text{Var}(X) - 2b \text{Cov}(X, Y) + \text{Var}(Y) + (E(Y) - bE(X) - a)^2 \\ &= \text{Var}(X) \left(b - \frac{\text{Cov}(X, Y)}{\text{Var}(X)} \right)^2 + \text{Var}(Y) - \text{Var}(X) \left(\frac{\text{Cov}(X, Y)}{\text{Var}(X)} \right)^2 \\ &\quad + (E(Y) - bE(X) - a)^2, \end{aligned}$$

where

$$\begin{aligned} &\text{Var}(Y) - \text{Var}(X) \left(\frac{\text{Cov}(X, Y)}{\text{Var}(X)} \right)^2 \\ &= \text{Var}(Y) \left(1 - \frac{\text{Cov}(X, Y)^2}{\text{Var}(X) \text{Var}(Y)} \right) \\ &= \text{Var}(Y) (1 - (\text{Corr}(X, Y))^2), \end{aligned}$$

so

$$\begin{aligned} E(Y - a - bX)^2 &= \text{Var}(X) \left(b - \frac{\text{Cov}(X, Y)}{\text{Var}(X)} \right)^2 + (E(Y) - bE(X) - a)^2 \\ &\quad + \text{Var}(Y) (1 - (\text{Corr}(X, Y))^2). \end{aligned} \tag{1}$$

(1) implies that $E(Y - a - bX)^2$ is minimized when

$$b = \frac{\text{Cov}(X, Y)}{\text{Var}(X)} \text{ and } a = E(Y) - bE(X), \tag{2}$$

and the minimum of $E(Y - a - bX)^2$ is

$$\text{Var}(Y) (1 - (\text{Corr}(X, Y))^2),$$

which must be nonnegative. Therefore,

$$1 - \text{Corr}(X, Y)^2 \geq 0$$

and $-1 \leq \text{Corr}(X, Y) \leq 1$.

- Remark. (1) also implies that when $|\text{Corr}(X, Y)| = 1$, $E(Y - a - bX)^2 = 0$ for the a and b in (2).