

MATH 451/551

Chapter 6. Joint Distribution

6.3 ~~Conditional Expected Values~~

Covariance

population mean μ .

population variance σ^2

population covariance.

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Covariance

$$V(X) = E\{(X - \mu_X)^2\}$$
$$\text{COV}(X, Y) = E\{(X - \mu_X)(Y - \mu_Y)\}$$



Covariance

Let X and Y be random variables with finite population means μ_X and μ_Y , respectively. The population covariance between X and Y is

$$\text{Cov}(X, Y) = E\{(X - \mu_X)(Y - \mu_Y)\}. \quad \text{Cov}(X, Y) = E\{(X - \mu_X)(Y - \mu_Y)\}$$

- ▶ Symmetric in its arguments: $\text{Cov}(X, Y) = \text{Cov}(Y, X)$
- ▶ Defining formula useful for conceptualizing covariance
 - ▶ If X and Y tend to be on opposite sides of their means together \Rightarrow population covariance negative
 - ▶ If X and Y tend to be on the same sides of their means together \Rightarrow population covariance positive

$$\text{Cov}(X, Y), \text{C}(X, Y), \sigma_{XY}$$

Example 1



Example 1

A fair coin is tossed twice. Let X be the number of heads that appear and Y be the number of tails that appear. Find the population covariance between X and Y .

$$\text{Cov}(X, Y) = E \{ (X - \mu_X)(Y - \mu_Y) \}$$

$$f(x, y)$$

$$g = \{f(x, y) \mid (x, y) = (0, 2), (1, 1), (2, 0)\}$$

$$f(x, y) = \begin{cases} \frac{1}{2} \cdot \frac{1}{2} & x=0, y=2 \\ \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{2} & x=1, y=1 \\ \frac{1}{2} \cdot \frac{1}{2} & x=2, y=0 \end{cases} = \begin{cases} \frac{1}{4} & x=0, y=2 \\ \frac{1}{2} & x=1, y=1 \\ \frac{1}{4} & x=2, y=0 \end{cases}$$

$$f_x(x) = \begin{cases} \frac{1}{4} & x=0 \\ \frac{1}{2} & x=1 \\ \frac{1}{4} & x=2 \end{cases}$$

$$f_y(y) = \begin{cases} \frac{1}{4} & y=0 \\ \frac{1}{2} & y=1 \\ \frac{1}{4} & y=2 \end{cases}$$

x	0	1	2
0	0	0	$\frac{1}{4}$
1	0	$\frac{1}{2}$	0
2	$\frac{1}{4}$	0	0

$$\begin{aligned} \mu_X &= \sum x f_x(x) \\ &= 0 \times \frac{1}{4} + 1 \times \frac{1}{2} + 2 \times \frac{1}{4} \\ &= 1 \end{aligned}$$

$$\mu_Y = \sum y f_y(y) = 1$$

$$\begin{aligned}
 \text{Cov}(X, Y) &= E\{(X - \mu_X)(Y - \mu_Y)\} \\
 &= \sum_{x,y} (x-1)(y-1)f(x, y) \\
 &= (0-1)(2-1)\frac{1}{4} + (1-1)(1-1)\frac{1}{2} + (2-1)(0-1)\frac{1}{4} \\
 &= -\frac{1}{2}
 \end{aligned}$$

$$Y = 2 - X$$

units of Cov: product of the units of X and unit of Y



Special Case

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

Var-Cov Matrix

Variance is a special case of covariance: $V(X) = \text{Cov}(X, X)$.

► Bivariate Case:

$$\Sigma = \begin{pmatrix} \text{Cov}(X, X) & \text{Cov}(X, Y) \\ \text{Cov}(Y, X) & \text{Cov}(Y, Y) \end{pmatrix}$$

► Trivariate Case:

$$\Sigma = \begin{pmatrix} V(X) & \text{Cov}(X, Y) & \text{Cov}(X, Z) \\ \text{Cov}(Y, X) & V(Y) & \text{Cov}(Y, Z) \\ \text{Cov}(Z, X) & \text{Cov}(Z, Y) & V(Z) \end{pmatrix}$$

positive semi-definite



Theorem 6.4

Theorem 6.4

If X and Y are random variables with finite population means μ_X and μ_Y , respectively, then

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

$$\begin{aligned} E\{(X - \mu_X)(Y - \mu_Y)\} &= E\{XY - \mu_X Y - \mu_Y X + \mu_X \mu_Y\} \\ &= E(XY) - \mu_X E(Y) - \mu_Y E(X) + \mu_X \mu_Y \\ &= E(XY) - \mu_X \mu_Y - \cancel{\mu_X \mu_Y} + \cancel{\mu_X \mu_Y} \\ &= E(XY) - \mu_X \mu_Y \end{aligned}$$

Example 2



Example 2

A fair coin is tossed twice. Let X be the number of heads that appear and Y be the number of tails that appear. Find the population covariance between X and Y using the shortcut formula

$$\begin{aligned}\text{Cov}(X, Y) &= \underline{E(XY)} - \mu_X \mu_Y \\ &= \sum \sum xy f(x,y) - 1 * 1 \\ &= 0 * 2 * \frac{1}{4} + 1 * 1 * \frac{1}{2} + 2 * 0 * \frac{1}{4} - 1 \\ &= -\frac{1}{2}\end{aligned}$$

Example 3



Example 3

Deal two cards from a well-shuffled deck. Let X be the number of aces dealt and Y be the number of face cards dealt. Using the shortcut formula, find the population covariance between X and Y .

$x \backslash y$	0	1	2	$f_X(x)$
0	$630/1326$	$432/1326$	$66/1326$	$1128/1326$
1	$144/1326$	$98/1326$	0	$192/1326$
2	$6/1326$	0	0	$6/1326$
$f_Y(y)$	$780/1326$	$480/1326$	$66/1326$	

$$\mu_X = \sum x f_X(x) = \frac{2}{13}$$

$$\mu_Y = \sum y f_Y(y) = \frac{6}{13}$$

$$f_{(2,0)} = \frac{\binom{7}{2}}{\binom{52}{2}}$$

$$f_{(0,0)} = \frac{\binom{36}{2}}{\binom{52}{2}}$$

$$f_{(0,1)} = \frac{\binom{12}{1} \binom{36}{1}}{\binom{52}{2}}$$

$$f_{(0,2)} = \frac{\binom{12}{2}}{\binom{52}{2}}$$

$$f_{(1,0)} = \frac{\binom{4}{1} \binom{31}{1}}{\binom{52}{2}}$$

$$f_{(1,1)} = \frac{\binom{4}{1} \binom{12}{1}}{\binom{52}{2}}$$

Thank You



THANK YOU!

