

MATH 451/551

Chapter 6. Joint Distribution

~~6.3 Conditional Expected Values~~

Covariance

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population mean μ .
population variance σ^2
population covariance.



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

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

$$\text{Cov}(X, Y) = E\left\{ \begin{matrix} (X - \mu_X) \\ (Y - \mu_Y) \end{matrix} \right\}$$

- ▶ Symmetric in its arguments: $\text{Cov}(X, Y) = \text{Cov}(Y, X)$
- ▶ Defining formula useful for conceptualizing covariance

$$\text{Cov}(Y, X) = E\left\{ \begin{matrix} (Y - \mu_Y) \\ (X - \mu_X) \end{matrix} \right\}$$

- ▶ 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

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

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 .

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

$$\mathcal{A} = \{(x, y) \mid (x, y) = (0, 2), (1, 1), (2, 0)\}$$

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

$$f(x, y)$$

$$x=0, y=2$$

$$x=1, y=1$$

$$x=2, y=0$$

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

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

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

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

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

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

$$= \sum_{\mathcal{A}} (x-1)(y-1)f(x, y)$$

$$= (0-1)(2-1)\frac{1}{4} + (1-1)(1-1)\frac{1}{2} + (2-1)(0-1)\frac{1}{4}$$

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

$$Y = Z - X$$

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

Special Case

X, Y, Z R.V.
pairwise Cov: $Cov(X, Y), Cov(X, Z), Cov(Y, Z)$



Special Case

$$V(X) = Cov(X, X)$$

Var-Cov Matrix

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

► Bivariate Case:

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

► Trivariate Case:

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

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

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

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

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

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

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 .

$\mathcal{S} = \{(x, y) \mid (0, 0), (0, 1), (0, 2), (1, 0), (1, 1), (2, 0)\}$

$x \backslash y$	0	1	2	$f_X(x)$
0	$\frac{6 \cdot 30}{1326}$	$\frac{4 \cdot 2}{1326}$	$\frac{6 \cdot 6}{1326}$	$\frac{1128}{1326}$
1	$\frac{1 \cdot 44}{1326}$	$\frac{4 \cdot 8}{1326}$	0	$\frac{192}{1326}$
2	$\frac{6}{1326}$	0	0	$\frac{6}{1326}$
$f_Y(y)$	$\frac{780}{1326}$	$\frac{480}{1326}$	$\frac{66}{1326}$	

$\mu_X = \sum x f_X(x) = \frac{2}{13}$
 $\mu_Y = \sum y f_Y(y) = \frac{6}{13}$

$f(0, 0) = \frac{\binom{36}{2}}{\binom{52}{2}}$
 $f(0, 1) = \frac{\binom{12}{1} \binom{36}{1}}{\binom{52}{2}}$
 $f(0, 2) = \frac{\binom{12}{2}}{\binom{52}{2}}$
 $f(1, 0) = \frac{\binom{4}{1} \binom{30}{1}}{\binom{52}{2}}$
 $f(1, 1) = \frac{\binom{4}{1} \binom{12}{1}}{\binom{52}{2}}$
 $f(2, 0) = \frac{\binom{1}{1} \binom{2}{1}}{\binom{52}{2}}$

Thank You



THANK YOU!