

Department of Mathematics
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MATH 451/551

Chapter 6. Joint Distribution

6.3 Covariance & Correlation

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Theorem



Theorem 6.5

If X and Y are random variables with finite population variances and covariance, then

$$V(X + Y) = \underline{V(X)} + \underline{V(Y)} + 2\text{Cov}(X, Y).$$

Proof:
$$\begin{aligned} V(X+Y) &= E\left[\{(X+Y) - \underline{E(X+Y)}\}^2\right] \\ &= E\left[\{(X+Y) - (\mu_X + \mu_Y)\}^2\right] \\ &= E\left[\{(X - \mu_X) + (Y - \mu_Y)\}^2\right] \\ &= E\left\{(X - \mu_X)^2 + (Y - \mu_Y)^2 + 2(X - \mu_X)(Y - \mu_Y)\right\} \\ &= \underline{E\{(X - \mu_X)^2\}} + \underline{E\{(Y - \mu_Y)^2\}} + 2 \underline{E\{(X - \mu_X)(Y - \mu_Y)\}} \\ &= V(X) + V(Y) + 2\text{Cov}(X, Y) \end{aligned}$$

Example 1

The population variance of $X+Y$ must be 0 because the sum of # head and # tails is always 2.



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 variance of $X + Y$.

$$V(X+Y) = \underbrace{V(X)} + \underbrace{V(Y)} + 2\underbrace{\text{Cov}(X,Y)} = \frac{1}{2} + \frac{1}{2} + 2 \cdot (-\frac{1}{2}) = 0$$

$$\textcircled{1}. \mathcal{A} = \{(x,y) | (0,2), (1,1), (2,0)\}$$

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

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

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

$$\begin{aligned} \textcircled{2} V(X) &= E(X^2) - \{E(X)\}^2 = 0^2 \times \frac{1}{4} + 1^2 \times \frac{1}{2} + 2^2 \times \frac{1}{4} = (0 \times \frac{1}{4} + 1 \times \frac{1}{2} + 2 \times \frac{1}{4})^2 \\ &= \frac{3}{2} - 1^2 = \frac{1}{2} \end{aligned}$$

$$V(Y) = \frac{1}{2}$$

$$\text{Cov}(X,Y) = E(XY) - \mu_X \mu_Y = 0 \cdot 2 \cdot \frac{1}{4} + 1 \cdot 1 \cdot \frac{1}{2} + 2 \cdot 0 \cdot \frac{1}{4} = 1^2 = \frac{1}{2} - 1 = -\frac{1}{2}$$

Theorem



Theorem 6.6

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

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

$$X \text{ \& } Y \text{ are independent} \Rightarrow E(XY) = E(X)E(Y) = \mu_X \mu_Y$$

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

Example 2

the converse of THM 6.6 is ~~is~~
NOT true



Example 2

X & Y are indep $\Rightarrow \text{Cov}(X, Y) = 0$

Show that the random variables X and Y that are uniformly distributed over the support $\mathcal{A} = \{(x, y) | 1 < x^2 + y^2 < 4\}$ have a population covariance 0 and are dependent random variables.

$$\pi 2^2 - \pi 1^2 = 3\pi$$

$$f(x, y) = \frac{1}{3\pi}, \quad \boxed{1 < x^2 + y^2 < 4}$$

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

$$\begin{aligned} E(XY) &= \iint_{\mathcal{A}} xy f(x, y) dy dx \\ &= \int_0^{2\pi} \int_1^2 r \cos \theta r \sin \theta \frac{1}{3\pi} dr d\theta \\ &= 0 \end{aligned}$$

$$\Rightarrow \text{Cov}(X, Y) = 0$$

$$\begin{aligned} x &= r \cdot \cos \theta \\ y &= r \cdot \sin \theta \\ dy dx &= r dr d\theta \end{aligned}$$

$$\begin{aligned} \mu_X &= E(X) = 0 \\ \mu_Y &= E(Y) = 0 \end{aligned}$$

Theorem



Theorem 6.7

If X and Y are independent random variables, then

$$V(X + Y) = V(X) + V(Y).$$

In general $V(X + Y) = V(X) + V(Y) + 2\text{Cov}(X, Y)$

X & Y are indep $\Rightarrow \text{Cov}(X, Y) = 0$

$$\Rightarrow \underline{V(X + Y) = V(X) + V(Y)}$$

Correlation

$$Z = \frac{X - \mu_X}{\sigma_X}$$



Correlation

Let X and Y be random variables with finite population means μ_X and μ_Y , and finite population variances $\sigma_X^2 > 0$ and $\sigma_Y^2 > 0$, respectively.

The **population correlation** between X and Y is

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

$$\rho_{XY}, \text{Corr}(X, Y)$$

Theorem



Theorem 6.8

If X and Y are independent random variables, then $\rho = 0$.

X & Y are indep $\Rightarrow \text{Cov}(X, Y) = 0$.

$$\rho = \frac{\text{Cov}(X, Y)}{\underbrace{\sigma_X \sigma_Y}} = 0.$$

Theorem

$$\begin{aligned} \text{Cov}\left(\frac{X}{\sigma_X}, \frac{Y}{\sigma_Y}\right) &= E\left(\frac{XY}{\sigma_X \sigma_Y}\right) - \frac{\mu_X \mu_Y}{\sigma_X \sigma_Y} \\ &= \frac{E(XY) - \mu_X \mu_Y}{\sigma_X \sigma_Y} = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \end{aligned}$$



Theorem 6.9

If X and Y are random variables with population correlation ρ , then $-1 \leq \rho \leq 1$.

$$\begin{aligned} \frac{X}{\sigma_X} + \frac{Y}{\sigma_Y}: \quad & V\left(\frac{X}{\sigma_X} + \frac{Y}{\sigma_Y}\right) = V\left(\frac{X}{\sigma_X}\right) + V\left(\frac{Y}{\sigma_Y}\right) + 2\text{Cov}\left(\frac{X}{\sigma_X}, \frac{Y}{\sigma_Y}\right) \\ &= \frac{V(X)}{\sigma_X^2} + \frac{V(Y)}{\sigma_Y^2} + \frac{2\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \\ &= 1 + 1 + 2\rho = 2(1 + \rho) \geq 0 \Rightarrow \rho \geq -1 \\ \frac{X}{\sigma_X} - \frac{Y}{\sigma_Y}: \quad & V\left(\frac{X}{\sigma_X} - \frac{Y}{\sigma_Y}\right) = V\left(\frac{X}{\sigma_X}\right) + V\left(\frac{Y}{\sigma_Y}\right) - 2\text{Cov}\left(\frac{X}{\sigma_X}, \frac{Y}{\sigma_Y}\right) \\ &= \frac{V(X)}{\sigma_X^2} + \frac{V(Y)}{\sigma_Y^2} - \frac{2\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \\ &= 1 + 1 - 2\rho = 2(1 - \rho) \geq 0 \Rightarrow \rho \leq 1 \end{aligned}$$

Example 3



Example 3

Let the discrete random variables X and Y have joint probability mass function $f(x, y)$ given by the entries in the table. Find the population correlation between X and Y .

	1	2	3	<u>$f_X(x)$</u>
1	0.2	0.1	0.3	0.6
2	0.1	0.1	0.2	0.4
<u>$f_Y(y)$</u>	0.3	0.2	0.5	1

$$\rho = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{0.02}{\sqrt{0.24} \sqrt{0.76}} = 0.0468$$

$$\textcircled{1} \mu_X = E(X) = 1 \times 0.6 + 2 \times 0.4 = 1.4$$

$$\sigma_X^2 = E(X^2) - \{E(X)\}^2 = 1^2 \times 0.6 + 2^2 \times 0.4 - 1.4^2 = 0.24$$

$$\textcircled{2} \mu_Y = E(Y) = 1 \times 0.3 + 2 \times 0.2 + 3 \times 0.5 = 2.2$$

$$\sigma_Y^2 = E(Y^2) - \{E(Y)\}^2 = 1^2 \times 0.3 + 2^2 \times 0.2 + 3^2 \times 0.5 - 2.2^2 = 0.76$$

$$\textcircled{3} \text{Cov}(X, Y) = E(XY) - \mu_X \mu_Y = 1 \cdot 1 \cdot 0.2 + 1 \cdot 2 \cdot 0.1 + 1 \cdot 3 \cdot 0.3 + 2 \cdot 1 \cdot 0.1 + 2 \cdot 2 \cdot 0.1 + 2 \cdot 3 \cdot 0.1 - 1.4 \times 2.2 = 0.02$$



Theorem 6.10

Let X and Y be random variables with population correlation ρ . The population correlation ρ equals -1 iff the support of X and Y lies on a line with negative slope. The population correlation ρ equals 1 iff the support of X and Y lies on a line with positive slope.

- ▶ A population correlation of -1 between the random variables X and Y is often known as a **perfect negative correlation**.
- ▶ A population correlation of 1 between the random variable X and Y is often known as a **perfect positive correlation**.

Example 4



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Example 4

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 correlation between X and Y .

$$\mathcal{A} = \{(x, y) \mid (0, 2), (1, 1), (2, 0)\}$$
$$f_{X,Y}(x, y) = \begin{cases} \frac{1}{4} & x=0, y=2 \\ \frac{1}{2} & x=1, y=1 \\ \frac{1}{4} & x=2, y=0 \end{cases}$$

$$\rho = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{-\frac{1}{2}}{\sqrt{\frac{1}{2}} \sqrt{\frac{1}{2}}} = -1$$

$$Y = 2 - X$$

$$\textcircled{1} \mu_X = E(X) = 1$$

$$\sigma_X^2 = E(X^2) - \{E(X)\}^2 = 0^2 \cdot \frac{1}{4} + 1^2 \cdot \frac{1}{2} + 2^2 \cdot \frac{1}{4} = \frac{1}{2}$$

$$\textcircled{2} \mu_Y = E(Y) = 1$$

$$\sigma_Y^2 = \frac{1}{2}$$

$$\textcircled{3} \text{Cov}(X, Y) = E(XY) - \mu_X \mu_Y = -\frac{1}{2}$$

Thank You



THANK YOU!