

# Covariance and Correlation.

Two RVs  $X, Y$  independent.

$$\mathbb{E}[h(x)g(y)] = \mathbb{E}[h(x)]\mathbb{E}[g(y)].$$

Continuous  $f_{x,y}(x,y) = f_x(x) f_y(y)$ .

$$\begin{aligned} \mathbb{E}[h(x)g(y)] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(x)g(y) f_x(x) f_y(y) dx dy \\ &= \left( \int_{-\infty}^{\infty} h(x) f_x(x) dx \right) \left( \int_{-\infty}^{\infty} g(y) f_y(y) dy \right) \\ &= \mathbb{E}h(x) \mathbb{E}g(y). \end{aligned}$$

Law of iterated expectation.

$$\begin{aligned} \mathbb{E}[h(x)g(y)] &= \mathbb{E}\left\{ \mathbb{E}[h(x)g(y) | Y] \right\} \\ &= \mathbb{E}\left\{ g(y) \mathbb{E}[h(x) | Y] \right\} \\ &= \mathbb{E}\left\{ g(y) \mathbb{E}[h(x)] \right\} \quad \text{by independence.} \\ &= \mathbb{E}[g(y) \mathbb{E}[h(x)]] \\ &= \mathbb{E}[h(x)] \mathbb{E}[g(y)]. \end{aligned}$$

Definition Let  $X, Y$  be two R.V.s.

then the **covariance** is

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

$$\mu_X = \mathbb{E}X, \quad \mu_Y = \mathbb{E}Y.$$

Formula

$$\text{Cov}(X, Y) = \mathbb{E}[XY] - (\mathbb{E}X)(\mathbb{E}Y)$$

Proof:

$$\begin{aligned}\text{Cov}(X, Y) &= \mathbb{E}(X - \mu_X)(Y - \mu_Y) \\ &= \mathbb{E}[XY - \mu_X Y - \mu_Y X + \mu_X \mu_Y] \\ &= \mathbb{E}XY - \mu_X \underbrace{\mathbb{E}Y}_{\mu_Y} - \mu_Y \underbrace{\mathbb{E}X}_{\mu_X} + \mu_X \mu_Y \\ &= \mathbb{E}XY - \mu_X \mu_Y.\end{aligned}$$

# Properties

$$\textcircled{1} \text{Cov}(X, X) = \text{Var}(X)$$

$$\textcircled{2} \text{ if } X \text{ and } Y \text{ are independent } \text{Cov}(X, Y) = 0.$$

$$\textcircled{3} \text{Cov}(X, Y) = \text{Cov}(Y, X) \text{ - symmetric}$$

$$\textcircled{4} \text{Cov}(aX+b, Y) = a\text{Cov}(X, Y) + \cancel{\text{Cov}(b, Y)}$$

$$\textcircled{5} \text{Cov}(X+Y, Z) = \text{Cov}(X, Z) + \text{Cov}(Y, Z)$$

Ex  $X, Y$  jointly continuous

$$f_{X,Y}(x,y) = \begin{cases} 2 & x+y \leq 1, x \geq 0, y \geq 0 \\ 0 & \text{otherwise.} \end{cases}$$

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

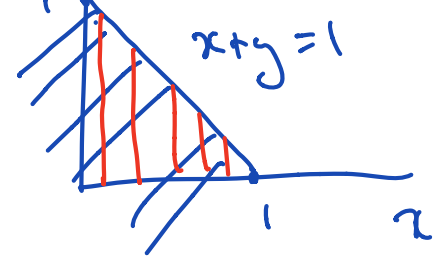
$$0 \leq y \leq 1-x$$

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

$$\mu_X = \mathbb{E}X = \int \int_{\mathbb{R}} x f_{X,Y}(x,y) dy$$

↓

$$= \int_0^1 \int_0^{1-x} x^2 dy dx$$



$$= \int_0^1 x^2(1-x) dx$$

$$= \frac{1}{3}$$

$$EY = \iint_R y f_{xy}(x,y) dA$$

$$= \int_0^1 \int_0^{1-x} 2y dy dx = \int_0^1 (1-x)^2 dx = \frac{1}{3}$$

$$EXY = \iint_R xy f_{xy}(x,y) dA$$

$$= 2 \int_0^1 x \int_0^{1-x} y dy dx = \int_0^1 x(1-x)^2 dx$$

$$= \frac{1}{12}$$

$$\text{Cov}(X, Y) = \frac{1}{12} - \frac{1}{3} \left( \frac{1}{3} \right) = -\frac{1}{36}$$

Warning If  $X, Y$  are independent.

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

The reverse is not true!!

Ex

$X$	$-1, 0, 1$	
$P_X(x)$	$\frac{1}{3}, \frac{1}{3}, \frac{1}{3}$	$Y = X^2$

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

Ex

$$f_{X,Y}(x,y) = \begin{cases} \frac{2}{3} & -1 \leq x \leq 1, 0 \leq y \leq x^2 \\ 0 & \text{otherwise.} \end{cases}$$

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

- You get odd integrals.

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Recall  $X, Y$  are independent. Then

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$$

Sum of two RVs Let  $X, Y$  be two  
RVs

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

More generally

$X_1, X_2, \dots, X_n$  RVs.

$$\text{Var}\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n \text{Var}(X_i) + \sum_{\substack{i=1, j=1 \\ i \neq j}}^n \text{Cov}(X_i, X_j)$$

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Correlation.

Covariance depends on units!

$$\text{Cov}(aX, bY) = ab \text{Cov}(X, Y)$$

Remember units.

Def Correlation of  $X, Y$ .

$$\rho_{xy} = \text{Cor}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_x \sigma_y}$$

↑  
rho.

Property that  $\text{Cor}(aX, bY) = \text{Cor}(X, Y)$ .

Properties.

①  $-1 \leq \rho_{xy} \leq 1$

②  $\rho_{xy} = 1$  iff  $Y = aX + b$   $a \geq 0$

③  $\rho_{xy} = -1$  iff  $X = aY + b$   $a \geq 0$ .

$X, Y$  linearly dependent.

Correlation only measure linear dependence.

