

1. Prove that  $\mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$ .

$$\begin{aligned}\mathbb{E}[(X - \mathbb{E}[X])^2] &= \mathbb{E}[X^2 - 2X\mathbb{E}[X] + \mathbb{E}[X]^2] \\ &= \mathbb{E}[X^2] - 2\mathbb{E}[X]\mathbb{E}[X] + \mathbb{E}[X]^2 \\ &= \mathbb{E}[X^2] - 2\mathbb{E}[X]^2 + \mathbb{E}[X]^2 \\ &= \mathbb{E}[X^2] - \mathbb{E}[X]^2\end{aligned}$$

□

2. Prove that if  $X_1$  and  $X_2$  are independent, then  $\mathbb{E}[X_1X_2] = \mathbb{E}[X_1]\mathbb{E}[X_2]$ .

### Continuous

$$\begin{aligned}\mathbb{E}[X_1X_2] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1x_2p_{X_1X_2}(x_1, x_2)dx_1dx_2 \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1x_2p_{X_1}(x_1)p_{X_2}(x_2)dx_1dx_2 & X_1 \perp\!\!\!\perp X_2 \implies p_{X_1X_2}(x_1, x_2) = p_{X_1}(x_1)p_{X_2}(x_2) \\ &= \int_{-\infty}^{\infty} x_1p_{X_1}(x_1)dx_1 \int_{-\infty}^{\infty} x_2p_{X_2}(x_2)dx_2 \\ &= \mathbb{E}[X_1]\mathbb{E}[X_2]\end{aligned}$$

### Discrete

$$\begin{aligned}\mathbb{E}[X_1X_2] &= \sum_{x_1} \sum_{x_2} x_1x_2p_{X_1, X_2}(x_1, x_2) \\ &= \sum_{x_1} \sum_{x_2} x_1x_2p_{X_1}(x_1)p_{X_2}(x_2) & X_1 \perp\!\!\!\perp X_2 \implies p_{X_1X_2}(x_1, x_2) = p_{X_1}(x_1)p_{X_2}(x_2) \\ &= \left( \sum_{x_1} x_1p_{X_1}(x_1) \right) \left( \sum_{x_2} x_2p_{X_2}(x_2) \right)\end{aligned}$$

□

3. Prove that  $\text{Var}(X_1 + X_2) = \text{Var}(X_1) + \text{Var}(X_2) + 2\text{Cov}(X_1, X_2)$ , where  $\text{Cov}(X_1, X_2) = \mathbb{E}[(X_1 - \mathbb{E}[X_1])(X_2 - \mathbb{E}[X_2])] = \mathbb{E}[X_1 X_2] - \mathbb{E}[X_1] \mathbb{E}[X_2]$ .

$$\begin{aligned}
 \text{Var}(X_1 + X_2) &= \mathbb{E}[(X_1 + X_2)^2] - \mathbb{E}[X_1 + X_2]^2 \\
 &= \mathbb{E}[X_1^2 + 2X_1 X_2 + X_2^2] - (\mathbb{E}[X_1] + \mathbb{E}[X_2])^2 \\
 &= \mathbb{E}[X_1^2] + 2\mathbb{E}[X_1 X_2] + \mathbb{E}[X_2^2] - (\mathbb{E}[X_1]^2 + 2\mathbb{E}[X_1] \mathbb{E}[X_2] + \mathbb{E}[X_2]^2) \\
 &= \mathbb{E}[X_1^2] + 2\mathbb{E}[X_1 X_2] + \mathbb{E}[X_2^2] - \mathbb{E}[X_1]^2 - 2\mathbb{E}[X_1] \mathbb{E}[X_2] - \mathbb{E}[X_2]^2 \\
 &= \underbrace{\mathbb{E}[X_1^2] - \mathbb{E}[X_1]^2}_{\text{Var}(X_1)} + \underbrace{\mathbb{E}[X_2^2] - \mathbb{E}[X_2]^2}_{\text{Var}(X_2)} + 2 \cdot \underbrace{(\mathbb{E}[X_1 X_2] - \mathbb{E}[X_1] \mathbb{E}[X_2])}_{\text{Cov}(X_1, X_2)} \\
 &= \text{Var}(X_1) + \text{Var}(X_2) + 2\text{Cov}(X_1, X_2) \quad \square
 \end{aligned}$$

4. Prove or disprove via a counterexample: If  $\text{Corr}(X_1, X_2) = 0$  then  $X_1$  and  $X_2$  are independent

$$\text{Corr}(X_1, X_2) = 0 \implies \text{Cov}(X_1, X_2) = 0 \implies \mathbb{E}[X_1 X_2] = \mathbb{E}[X_1] \mathbb{E}[X_2]$$

Need to find an example where  $\mathbb{E}[X_1 X_2] = \mathbb{E}[X_1] \mathbb{E}[X_2]$  but  $X_1 \not\perp X_2$ .

Let  $X_1$  be a random variable with possible values of  $-1, 1$  where

$$\mathbb{P}(X_1 = -1) = \mathbb{P}(X_1 = 1) = 0.5$$

Let  $X_2$  be a random variable with possible values of  $-1, 0, 1$

$$X_2 = \begin{cases} 0 & \text{if } X_1 = -1 \\ -1 \text{ or } 1 & \text{if } X_1 = 1 \text{ (both with probability 0.5)} \end{cases}$$

Note that  $X_2$  is dependent on  $X_1$

$$\begin{aligned}
 \mathbb{E}[X_1] &= \sum_{x_1} x_1 p_{X_1}(x_1) \\
 &= -1 \cdot p_{X_1}(-1) + 1 \cdot p_{X_1}(1) \\
 &= -1 \cdot 0.5 + 1 \cdot 0.5 \\
 &= 0 \\
 \mathbb{E}[X_2] &= \sum_{x_2} x_2 p_{X_2}(x_2) \\
 &= 0 \cdot \underbrace{p_{X_2}(0)}_{p_{X_1}(-1)} + -1 \cdot \underbrace{p_{X_2}(-1)}_{0.5 p_{X_1}(1)} + 1 \cdot \underbrace{p_{X_2}(1)}_{0.5 p_{X_1}(1)} \\
 &= 0 \cdot 0.5 - 1 \cdot 0.25 + 1 \cdot 0.25 \\
 &= 0
 \end{aligned}$$

$$\begin{aligned}
 \mathbb{E}[X_1 X_2] &= \sum_{x_1} \sum_{x_2} x_1 x_2 p_{X_1 X_2}(x_1, x_2) \\
 &= -1 \cdot 0 \cdot \mathbb{P}(X_1 = -1) + 1 \cdot -1 \cdot \mathbb{P}(X_1 = 1, X_2 = -1) + 1 \cdot 1 \cdot \mathbb{P}(X_1 = 1, X_2 = 1) \\
 &= 0 + \mathbb{P}(X_1 = 1, X_2 = -1) - \mathbb{P}(X_1 = 1, X_2 = 1) \\
 &\quad \text{Note: } \mathbb{P}(X_1 = 1, X_2 = -1) = \mathbb{P}(X_1 = 1, X_2 = 1) \\
 &= 0
 \end{aligned}$$

$$\mathbb{E}[X_1] \mathbb{E}[X_2] = \mathbb{E}[X_1 X_2] \implies \text{Corr}(X_1, X_2) = 0 \text{ but } X_1 \not\perp X_2. \quad \square$$

5. Let  $f$  and  $g$  be non-decreasing on  $\mathbb{R}$ . Assume that  $\mathbb{E}[g^2(X)] < \infty$  and  $\mathbb{E}[f^2(X)] < \infty$ . Show  $\text{Cov}(f(X), g(X)) \geq 0$ . Hint: Let  $Y$  be independent of  $X$ , with the same distribution as  $X$ . Consider  $\mathbb{E}[\{g(X) - g(Y)\} \{f(X) - f(Y)\}]$ .

$X \perp\!\!\!\perp Y$   $X, Y$  are from the same distribution

Look at  $\{g(X) - g(Y)\} \{f(X) - f(Y)\}$

Option 1:  $X < Y$

$$\left. \begin{array}{l} g(X) - g(Y) \leq 0 \\ f(X) - f(Y) \leq 0 \end{array} \right\} \implies \{g(X) - g(Y)\} \{f(X) - f(Y)\} \geq 0$$

Option 2:  $X = Y$

$$\left. \begin{array}{l} g(X) - g(Y) = 0 \\ f(X) - f(Y) = 0 \end{array} \right\} \implies \{g(X) - g(Y)\} \{f(X) - f(Y)\} = 0$$

Option 3:  $X > Y$

$$\left. \begin{array}{l} g(X) - g(Y) \geq 0 \\ f(X) - f(Y) \geq 0 \end{array} \right\} \implies \{g(X) - g(Y)\} \{f(X) - f(Y)\} \geq 0$$

$$\implies \{g(X) - g(Y)\} \{f(X) - f(Y)\} \geq 0$$

$$\implies \mathbb{E}[\{g(X) - g(Y)\} \{f(X) - f(Y)\}] \geq 0$$

Let  $g(Y) = \mathbb{E}[g(X)]$  and  $f(Y) = \mathbb{E}[f(X)]$

$$\mathbb{E}[\{g(X) - \mathbb{E}[g(X)]\} \{f(X) - \mathbb{E}[f(X)]\}] \geq 0$$

$$\text{Cov}(f(X), g(X)) \geq 0$$

□