

# MTH 464 HW 4

Brandyn Tucknott

12 February 2025

1. Let  $X$  be a random variable with  $\mu = \mathbb{E}(X)$  and  $\sigma^2 = \text{Var}(X)$ . Determine  $a, b$  such that  $Y = a + bX$  satisfies  $\mathbb{E}(Y) = 0$ ,  $\text{Var}(Y) = 4\sigma^2$ , and  $\text{Corr}(X, Y) = -1$ .

$$\mathbb{E}(Y) = \mathbb{E}(a + bX) = \mathbb{E}(a) + \mathbb{E}(bX) = a + b\mathbb{E}(X) = a + b\mu = 0 \quad (1)$$

$$\text{Var}(Y) = \text{Var}(a + bX) = \text{Var}(a) + \text{Var}(bX) = b^2\text{Var}(X) = b^2\sigma^2 = 4\sigma^2 \quad (2)$$

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_x \sigma_y} = \frac{\mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y)}{\sigma_x \sigma_y} = \frac{\mathbb{E}(X(a + bX))}{\sigma_x \sigma_y} = -1 \quad (3)$$

Solving for equation (1) gives  $a = -b\mu$ . Using this, we can further reduce equation (3) into

$$\begin{aligned} \text{Corr}(X, Y) &= \frac{\mathbb{E}(aX + bX^2)}{\sigma_x \sigma_y} = \frac{a\mathbb{E}(X) + b\mathbb{E}(X^2)}{\sigma_x \sigma_y} = \frac{a\mu + b(\sigma^2 + \mu^2)}{\sigma \cdot 2\sigma} = \frac{-b\mu^2 + b\sigma^2 + b\mu^2}{2\sigma^2} = \\ &= \frac{b}{2} = -1 \longrightarrow b = -2 \end{aligned}$$

Since  $a = -b\mu$ , we can sub-in  $b = -2$  and conclude that  $a = 2\mu, b = -2$ .

2. Assume that the joint density of  $X, Y$  is given by

$$f_{X,Y}(x, y) = \frac{1}{y} e^{-(y + \frac{x}{y})} \mathbb{1}_{[0, \infty) \times [0, \infty)}(x, y)$$

(a) Check that  $f$  is a pdf.

Solution.

To verify that  $f$  is a pdf, we need to check that it integrates to 1 over its domain, and also that  $0 \leq f_{X,Y}(x, y) \leq 1$  for all  $x, y \in \mathbb{R}$ .

Examining  $f$  reveals it is negative only when  $y < 0$ , and since  $y \in [0, \infty)$ , we conclude that  $f \geq 0$ .

To check that it integrates to 1 over its domain, we compute

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x, y) dx dy = \int_0^{\infty} \int_0^{\infty} \frac{1}{y} e^{-(y + \frac{x}{y})} dx dy = \int_0^{\infty} e^{-y} dy = 1$$

(b) Find  $\mathbb{E}(X)$  and  $\mathbb{E}(Y)$ .

Solution.

$$\mathbb{E}(X) = \int_0^{\infty} \int_0^{\infty} x f_{X,Y}(x, y) dx dy = \int_0^{\infty} \int_0^{\infty} \frac{x}{y} e^{-y - \frac{x}{y}} dx dy = \int_0^{\infty} y e^{-y} dy = 1$$

$$\mathbb{E}(Y) = \int_0^{\infty} \int_0^{\infty} y f_{X,Y}(x, y) dx dy = \int_0^{\infty} \int_0^{\infty} e^{-y - \frac{x}{y}} dx dy = \int_0^{\infty} y e^{-y} dy = 1$$

(c) Show that the  $\text{Cov}(X, Y) = 1$ .

*Proof.* Note that

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

Since we calculated  $\mathbb{E}(X), \mathbb{E}(Y)$  in Part (b), all that remains is to calculate  $\mathbb{E}(XY)$  and put them together.

$$\mathbb{E}(XY) = \int_0^{\infty} \int_0^{\infty} x e^{-y - \frac{x}{y}} dx dy = \int_0^{\infty} y^2 e^{-y} dy = 2$$

We compute  $\text{Cov}(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X) \mathbb{E}(Y) = 2 - 1 \cdot 1 = 1$ , and we are done.  $\square$

3. Let  $Z \sim N(0, 1)$  be a standard normal random variable. Let  $I$  be independent of  $Z$  such that  $P(I = -1) = P(I = 1) = \frac{1}{2}$ . Define  $Y = Z \times I$ .

(a) Show that  $Y$  is a standard normal random variable.

*Proof.* Recall that for a standard normal distribution,  $\Phi(a) = 1 - \Phi(-a)$ . Then

$$\begin{aligned} P(Y \leq a) &= P(ZI \leq a) = \frac{1}{2}P(Z \leq a) + \frac{1}{2}P(Z \geq -a) = \\ &= \frac{1}{2}\Phi(a) + \frac{1}{2}(1 - \Phi(-a)) = \frac{1}{2}\Phi(a) + \frac{1}{2}\Phi(a) = \Phi(a) \end{aligned}$$

Since  $Y$  has the same cdf as a standard normal, we conclude that  $Y$  is standard normal.  $\square$

(b) Show that  $\text{Cov}(Y, Z) = 0$ .

*Proof.*

$$\text{Cov}(Y, Z) = \mathbb{E}(YZ) - \mathbb{E}(Y)\mathbb{E}(Z) = \mathbb{E}(ZI \cdot Z) - 0 = \mathbb{E}(Z \cdot Z) = \mathbb{E}(Z^2) = 0$$

$\square$

(c) Show that  $Z, Y$  are not independent. This provides an example of uncorrelated variables which are not independent.

*Proof.* Consider the case where  $z = y = a$ . If  $Z, Y$  were independent, given both  $Z, Y \sim N(0, 1)$  we would expect

$$P(Z \leq a \wedge Y \leq a) = P(Z \leq a)P(Y \leq a) = (\Phi(a))^2$$

We now directly calculate the joint probability to be

$$\begin{aligned} P(Z \leq a \wedge Y \leq a) &- P(Z \leq a \wedge ZI \leq a) = \\ &= P(I = 1)P(Z \leq a \wedge ZI \leq a|I = 1) + P(I = -1)P(Z \leq a \wedge ZI \leq a|I = -1) = \\ &= \frac{1}{2}P(Z \leq a \wedge Z \leq a) + \frac{1}{2}P(Z \leq a \wedge Z \geq -a) = \frac{1}{2}\Phi(a) + \frac{1}{2}(\Phi(a) - \Phi(-a)) = \\ &= \Phi(a) - \frac{\Phi(-a)}{2} \neq (\Phi(a))^2 \end{aligned}$$

We conclude that since  $P(Z \leq a \wedge Y \leq a) \neq P(Z \leq a)P(Y \leq a)$ ,  $Z, Y$  are not independent.  $\square$