

# ECON 709 - PS 3

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1. A random point  $(X, Y)$  is distributed uniformly on the square with vertices  $(1, 1)$ ,  $(1, -1)$ ,  $(-1, 1)$ , and  $(-1, -1)$ . That is, the joint PDF is  $f(x, y) = 1/4$  on the square and  $f(x, y) = 0$  outside the square. Determine the probability of the following events:

(a)  $X^2 + Y^2 < 1$

$$X^2 + Y^2 < 1 \implies -\sqrt{1 - X^2} < Y < \sqrt{1 - X^2}$$

$$\int_{-1}^1 \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} \frac{1}{4} dy dx = \int_{-1}^1 \left[ \frac{1}{4} y \right]_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} dx = \frac{1}{2} \int_{-1}^1 \sqrt{1-x^2} dx$$

Define  $x = \sin \theta \implies dx = \cos \theta d\theta$ .

$$\begin{aligned} \frac{1}{2} \int_{-1}^1 \sqrt{1-x^2} dx &= \frac{1}{2} \int_{-\pi/2}^{\pi/2} \sqrt{1-(\sin \theta)^2} \cos \theta d\theta \\ &= \frac{1}{2} \int_{-\pi/2}^{\pi/2} \cos^2 \theta d\theta \\ &= \frac{1}{4} \int_{-\pi/2}^{\pi/2} 1 + \cos(2\theta) d\theta \\ &= \frac{1}{4} \left[ \theta + \frac{\sin(2\theta)}{2} \right]_{-\pi/2}^{\pi/2} \\ &= \frac{1}{4} \left[ (\pi/2) + \frac{0}{2} - (-\pi/2) - \frac{0}{2} \right] \\ &= \frac{\pi}{4} \end{aligned}$$

(b)  $|X + Y| < 2$

$$|X + Y| < 2 \implies -2 < X + Y < 2 \implies -2 - X < Y < 2 - X. \text{ Since } X \text{ ranges from } -1 \text{ to } 1, \\ -2 - X < Y < 2 - X \implies -1 < Y < 1$$

$$\int_{-1}^1 \int_{-1}^1 \frac{1}{4} dy dx = \frac{1}{4} \int_{-1}^1 [y]_{-1}^1 dx = \frac{1}{2} \int_{-1}^1 dx = \frac{1}{2} [x]_{-1}^1 = 1$$

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2. Let the joint PDF of  $X$  and  $Y$  be given by  $f(x, y) = g(x)h(y) \forall x, y \in \mathbb{R}$  for some functions  $g(x)$  and  $h(y)$ . Let  $a$  denote  $\int_{-\infty}^{\infty} g(x)dx$  and  $b$  denote  $\int_{-\infty}^{\infty} h(x)dx$

(a) What conditions  $a$  and  $b$  should satisfy in order for  $f(x, y)$  to be a bivariate PDF?

For  $f(x, y)$  to be a PDF, it should integrate to one:

$$\begin{aligned} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy &= 1 \\ \implies \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x)h(y) dx dy &= 1 \\ \implies \int_{-\infty}^{\infty} g(x) dx \int_{-\infty}^{\infty} h(y) dy &= 1 \\ &\implies ab = 1 \\ &\implies a = b^{-1} \end{aligned}$$

(b) Find the marginal PDF of  $X$  and  $Y$ .

The marginal PDF of  $X$ :

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_{-\infty}^{\infty} g(x)h(y) dy = g(x) \int_{-\infty}^{\infty} h(y) dy = b \cdot g(x)$$

The marginal PDF of  $Y$ :

$$f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_{-\infty}^{\infty} g(x)h(y) dx = h(y) \int_{-\infty}^{\infty} g(x) dx = a \cdot h(y)$$

(c) Show that  $X$  and  $Y$  are independent.

Proof:  $X$  and  $Y$  are independent if the product of their marginal distributions is their joint distribution:

$$\begin{aligned} f_X(x) \cdot f_Y(y) &= b \cdot g(x) \cdot a \cdot h(y) \\ &= b \cdot g(x) \cdot b^{-1} \cdot h(y) \\ &= g(x) \cdot h(y) \\ &= f(x, y) \end{aligned}$$

□

3. Let the joint PDF of  $X$  and  $Y$  be given by

$$f(x, y) = \begin{cases} cxy & \text{if } x, y \in [0, 1], x + y \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

(a) Find the value of  $c$  such that  $f(x, y)$  is a joint PDF.

$$\begin{aligned} \int_0^1 \int_0^{1-x} f(x, y) dy dx &= 1 \\ \implies \int_0^1 \int_0^{1-x} cxy dy dx &= 1 \\ \implies c \int_0^1 \left[ \frac{xy^2}{2} \right]_{y=0}^{1-x} dx &= 1 \\ \implies \frac{c}{2} \int_0^1 x(1-x)^2 dx &= 1 \\ \implies \frac{c}{2} \left[ \frac{x^2}{2} - \frac{2x^3}{3} + \frac{x^4}{4} \right]_{x=0}^1 &= 1 \\ \implies \frac{c}{2} \left[ \frac{1}{2} - \frac{2}{3} + \frac{1}{4} \right] &= 1 \\ \implies c &= 24 \end{aligned}$$

(b) Find the marginal distributions of  $X$  and  $Y$ .

$$f_X(x) = \int_0^{1-x} f(x, y) dy = \int_0^{1-x} 24xy dy = \left[ 12xy^2 \right]_{y=0}^{1-x} = \begin{cases} 12x(1-x)^2, & x \in [0, 1] \\ 0, & \text{otherwise.} \end{cases}$$

$$f_Y(y) = \int_0^{1-y} f(x, y) dx = \int_0^{1-y} 24xy dx = \left[ 12x^2y \right]_{x=0}^{1-y} = \begin{cases} 12(1-y)^2y, & y \in [0, 1] \\ 0, & \text{otherwise.} \end{cases}$$

(c) Are  $X$  and  $Y$  independent? Compare your answer to Problem 2 and discuss.

$X$  and  $Y$  independent if the product of the marginal distributions equals their joint distribution at all points in the support. If  $x = y = 0.9$ ,  $f(0.9, 0.9) = 0$  because  $(0.9, 0.9)$  is not in the support,  $x + y = 0.9 + 0.9 = 1.8 > 1$ . But each marginal distribution is defined over  $[0, 1]$ , so the product of the marginals is positive at  $(0.9, 0.9)$ :  $f_X(0.9)f_Y(0.9) = [12(0.9)(1 - (0.9))^2][12(1 - 0.9)^2(0.9)] = 0.0117$ .

In (2), the support for the joint distribution is  $\mathbb{R}^2$ , whereas the support for the joint distribution depends on the realization of the random variable.

4. Show that any random variable is uncorrelated with a constant.

Proof: Let  $a \in \mathbb{R}$  and  $X$  be a random variable with distribution  $F_X$ . Define random variable  $Y$  as the degenerate random variable that equals  $a$ . Thus, the distribution  $Y$  is

$$F_Y(y) = \begin{cases} 0, & y < a \\ 1, & y \geq a \end{cases}$$

To show  $X$  is uncorrelated with a constant, I show that  $X$  and  $Y$  are independent and then, by a theorem in the Lecture 3 Notes, we know that  $X$  and  $Y$  are uncorrelated.

To find the joint distribution of  $X$  and  $Y$ , consider two cases:  $y < a$  and  $y \geq a$ . For  $y < a$ ,

$$\begin{aligned} F(x, y) &= P(X \leq x \text{ and } Y \leq y) \\ &= P(X \leq x \text{ and } Y \leq a) \\ &= 0 \end{aligned}$$

For  $y \geq a$ :

$$\begin{aligned} F(x, y) &= P(X \leq x \text{ and } Y \leq y) \\ &= P(X \leq x) \\ &= F_X(x) \end{aligned}$$

Thus, the joint distribution is

$$F(x, y) = \begin{cases} 0, & y < a \\ F_X(x), & y \geq a \end{cases}$$

The joint distribution equals the product of the marginals:

$$F(x, y) = \begin{cases} 0 * F_X(x), & y < a \\ 1 * F_X(x), & y \geq a \end{cases} = \begin{cases} F_Y(y) * F_X(x), & y < a \\ F_Y(y) * F_X(x), & y \geq a \end{cases}.$$

□

5. Let  $X$  and  $Y$  be independent random variables with means  $\mu_X, \mu_Y$  and variances  $\sigma_X^2, \sigma_Y^2$ . Find an expression for the correlation of  $XY$  and  $Y$  in terms of these means and variances.

Note that  $Var(X) = E(X^2) - E(X)^2 \implies E(X^2) = Var(X) + E(X)^2$ .

$$\begin{aligned}
Corr(XY, Y) &= \frac{Cov(XY, Y)}{\sqrt{Var(XY)Var(Y)}} \\
&= \frac{E(XY^2) - E(XY)E(Y)}{\sigma_Y \sqrt{E((XY)^2) - E(XY)^2}} \\
&= \frac{E(X)E(Y^2) - E(X)E(Y)E(Y)}{\sigma_Y \sqrt{E(X^2)E(Y^2) - (E(X)E(Y))^2}} \\
&= \frac{\mu_X(Var(Y) + E(Y)^2) - \mu_X\mu_Y^2}{\sigma_Y \sqrt{(Var(X) + E(X)^2)(Var(Y) + E(Y)^2) - (\mu_X\mu_Y)^2}} \\
&= \frac{\mu_X\sigma_Y^2 + \mu_X\mu_Y^2 - \mu_X\mu_Y^2}{\sigma_Y \sqrt{\sigma_Y^2\sigma_X^2 + \mu_X^2\sigma_Y^2 + \mu_Y^2\sigma_X^2 + \mu_Y^2\mu_X^2 - \mu_X^2\mu_Y^2}} \\
&= \frac{\mu_X\sigma_Y}{\sqrt{\sigma_Y^2\sigma_X^2 + \mu_X^2\sigma_Y^2 + \mu_Y^2\sigma_X^2}}
\end{aligned}$$

6. Prove the following: For any random vector  $(X_1, X_2, \dots, X_n)$ ,

$$\text{Var}\left(\sum_{i=1}^n X_i\right) = \sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{1 \leq i < j \leq n} \text{Cov}(X_i, X_j).$$

Proof by induction.

7. Suppose that  $X$  and  $Y$  are joint normal, i.e. they have the joint PDF:

$$f(x, y) = \frac{1}{2\pi\sigma_X\sigma_Y\sqrt{1-\rho^2}} \exp(-(2(1-\rho^2))^{-1}(x^2/\sigma_X^2 - 2xy/\sigma_X\sigma_Y + y^2/\sigma_Y^2))$$

(a) Derive the marginal distributions of  $X$  and  $Y$ , and observe that both normal distributions.

Gaussian integrals

(b) Derive the conditional distribution of  $Y$  given  $X = x$ . Observe that it is also a normal distribution.

(c) Derive the joint distribution of  $(X, Z)$  where  $Z = (Y/\sigma_Y) - (\rho X/\sigma_X)$ , and then show that  $X$  and  $Z$  are independent.

Equation 23 in lecture 3.

8. Consider a function  $g : \mathbb{R} \rightarrow \mathbb{R}$ . Recall that the inverse image of a set  $A$ , denoted  $g^{-1}(A)$  is  $g^{-1}(A) = \{x \in \mathbb{R} : g(x) \in A\}$ . Let there be functions  $g_1 : \mathbb{R} \rightarrow \mathbb{R}$  and  $g_2 : \mathbb{R} \rightarrow \mathbb{R}$ . Let  $X$  and  $Y$  be two random variables that are independent. Suppose that  $g_1$  and  $g_2$  are both Borel-measurable, which means that  $g_1^{-1}(A)$  and  $g_2^{-1}(A)$  are both in the Borel  $\sigma$ -field whenever  $A$  is in the Borel  $\sigma$ -field. Show that the two random variables  $Z := g_1(X)$  and  $W := g_2(Y)$  are independent. (Hint: use the 1st or the 2nd definition of independence.)