# 1 Continuous Joint Densities

The joint probability density function of two random variables X and Y is given by f(x,y) = Cxy for  $0 \le x \le 1, 0 \le y \le 2$ , and 0 otherwise (for a constant C).

- (a) Find the constant C that ensures that f(x, y) is indeed a probability density function.
- (b) Find  $f_X(x)$ , the marginal distribution of X.
- (c) Find the conditional distribution of Y given X = x.
- (d) Are *X* and *Y* independent?

### **Solution:**

(a) Since f(x,y) is a probability density function, it must integrate to 1. Then:

$$1 = \int_0^1 \int_0^2 Cxy \, dy \, dx = \int_0^1 2Cx \, dx = C$$

Therefore, C = 1.

(b) To get the marginal distribution of X, we integrate the joint distribution with respect to Y. So:

$$f_X(x) = \int_0^2 f(x, y) dy = \int_0^2 xy dy = 2x$$

This is the marginal distribution for  $0 \le x \le 1$ .

(c) The conditional distribution of Y given by

$$f_{Y|X}(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{xy}{2x} = \frac{y}{2}$$

(d) The conditional distribution of Y given X = x does not depend on x, so they are independent. Alternatively, you could find the marginal distribution of Y and see it is the same as the conditional distribution of Y:

$$f_Y(y) = \int_0^1 f(x, y) dx = \int_0^1 xy dx = \frac{y}{2}$$

Notice that since X and Y are independent,  $f_X(x)f_Y(y) = xy = f_{X,Y}(x,y)$ , i.e. the product of the marginal distributions is the same as the joint distribution.

## 2 Uniform Distribution

You have two spinning wheels, each having a circumference of 10 cm with values in the range [0,10) marked on the circumference. If you spin both (independently) and let X be the position of the first spinning wheel's mark and Y be the position of the second spinning wheel's mark, what is the probability that  $X \ge 5$ , given that  $Y \ge X$ ?

### **Solution:**

First we write down what we want and expand out the conditioning:

$$\mathbb{P}[X \ge 5 \mid Y \ge X] = \frac{\mathbb{P}[Y \ge X \cap X \ge 5]}{\mathbb{P}[Y \ge X]}.$$

 $\mathbb{P}[Y \ge X] = 1/2$  by symmetry. To find  $\mathbb{P}[Y \ge X \cap X \ge 5]$ , it helps a lot to just look at the picture of the probability space and use the continuous uniform law  $\mathbb{P}[A] = (\text{area of } A)/(\text{area of } \Omega)$ . We are interested in the relative area of the region bounded by x < y < 10, 5 < x < 10 to the entire square bounded by 0 < x < 10, 0 < y < 10.

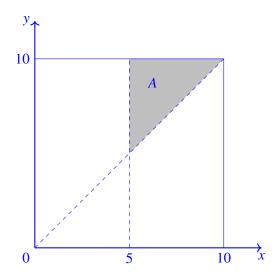


Figure 1: Joint probability density for the spinner.

$$\mathbb{P}[Y \ge X \cap X \ge 5] = \frac{5 \cdot 5/2}{10 \cdot 10} = \frac{1}{8}.$$

So 
$$\mathbb{P}[X \ge 5 \mid Y \ge X] = (1/8)/(1/2) = 1/4$$
.

# 3 Exponential Practice

Let  $X \sim \text{Exponential}(\lambda_X)$  and  $Y \sim \text{Exponential}(\lambda_Y)$  be independent, where  $\lambda_X, \lambda_Y > 0$ . Let  $U = \min\{X,Y\}$ ,  $V = \max\{X,Y\}$ , and W = V - U.

(a) Compute  $\mathbb{P}(U > t, X \leq Y)$ , for  $t \geq 0$ .

- (b) Use the previous part to compute  $\mathbb{P}(X \leq Y)$ . Conclude that the events  $\{U > t\}$  and  $\{X \leq Y\}$  are independent.
- (c) Compute  $\mathbb{P}(W > t \mid X \leq Y)$ .
- (d) Use the previous part to compute  $\mathbb{P}(W > t)$ .
- (e) Calculate  $\mathbb{P}(U > u, W > w)$ , for w > u > 0. Conclude that U and W are independent. [Hint: Think about the approach you used for the previous parts.]

#### **Solution:**

(a) One has

$$\mathbb{P}(U > t, X \le Y) = \mathbb{P}(t < X \le Y) = \int_{t}^{\infty} \int_{x}^{\infty} f_{X,Y}(x,y) \, \mathrm{d}y \, \mathrm{d}x$$

$$= \int_{t}^{\infty} \int_{x}^{\infty} \lambda_{X} \exp(-\lambda_{X}x) \lambda_{Y} \exp(-\lambda_{Y}y) \, \mathrm{d}y \, \mathrm{d}x$$

$$= \lambda_{X} \lambda_{Y} \int_{t}^{\infty} \exp(-\lambda_{X}x) \cdot \frac{\exp(-\lambda_{Y}x)}{\lambda_{Y}} \, \mathrm{d}x = \lambda_{X} \int_{t}^{\infty} \exp(-(\lambda_{X} + \lambda_{Y})x) \, \mathrm{d}x$$

$$= \frac{\lambda_{X}}{\lambda_{X} + \lambda_{Y}} \exp(-(\lambda_{X} + \lambda_{Y})t).$$

(b) Take t = 0.

$$\mathbb{P}(X \leq Y) = \frac{\lambda_X}{\lambda_X + \lambda_Y}.$$

Since *X* and *Y* are independent exponentials,  $U = \min\{X, Y\} \sim \text{Exponential}(\lambda_X + \lambda_Y)$ . So,  $\mathbb{P}(U > t) = \exp(-(\lambda_X + \lambda_Y)t)$ , and therefore we have  $\mathbb{P}(U > t, X \leq Y) = \mathbb{P}(X \leq Y)\mathbb{P}(U > t)$ .

(c) One has

$$\mathbb{P}(W > t, X \le Y) = \mathbb{P}(Y - X > t) = \int_0^\infty \int_{x+t}^\infty \lambda_X \exp(-\lambda_X x) \lambda_Y \exp(-\lambda_Y y) \, \mathrm{d}y \, \mathrm{d}x$$

$$= \lambda_X \lambda_Y \int_0^\infty \exp(-\lambda_X x) \cdot \frac{\exp(-\lambda_Y (x+t))}{\lambda_Y} \, \mathrm{d}x$$

$$= \lambda_X \exp(-\lambda_Y t) \int_0^\infty \exp(-(\lambda_X + \lambda_Y) x) \, \mathrm{d}x = \frac{\lambda_X}{\lambda_X + \lambda_Y} \exp(-\lambda_Y t).$$

So, we see that

$$\mathbb{P}(W > t \mid X \leq Y) = \frac{\mathbb{P}(W > t, X \leq Y)}{\mathbb{P}(X < Y)} = \exp(-\lambda_Y t).$$

Alternatively,

$$\mathbb{P}(W > t \mid X \le Y) = \mathbb{P}(Y > X + t \mid X \le Y) = \int_0^\infty \mathbb{P}(Y > X + t \mid Y \ge X) f_X(X) \, \mathrm{d}X$$
$$= \exp(-\lambda_Y t) \int_0^\infty f_X(X) \, \mathrm{d}X = \exp(-\lambda_Y t),$$

where we have used the memoryless property of the exponential distribution. Note that in the first line, we are using conditioning:

$$\mathbb{P}(Y > X + t \mid X \le Y) = \int_0^\infty \mathbb{P}(Y > X + t \mid X \le Y, X = x) f_X(x) dx.$$

The probability inside the integral then becomes  $\mathbb{P}(Y > x + t \mid Y \ge x, X = x)$ , and then one can drop the conditioning on X = x because X and Y are independent.

(d) By switching X and Y in the previous part, we have

$$\mathbb{P}(W > t \mid Y \leq X) = \exp(-\lambda_X t).$$

So, we can use the law of total probability to give

$$\mathbb{P}(W > t) = \mathbb{P}(X \le Y)\mathbb{P}(W > t \mid X \le Y) + \mathbb{P}(Y \le X)\mathbb{P}(W > t \mid Y \le X)$$
$$= \frac{\lambda_X}{\lambda_X + \lambda_Y} \exp(-\lambda_Y t) + \frac{\lambda_Y}{\lambda_X + \lambda_Y} \exp(-\lambda_X t).$$

(e) We calculate

$$\mathbb{P}(U > u, W > w, X \le Y) = \mathbb{P}(u < X \le X + w < Y)$$

$$= \int_{u}^{\infty} \int_{x+w}^{\infty} \lambda_{X} \exp(-\lambda_{X}x) \lambda_{Y} \exp(-\lambda_{Y}y) \, dy \, dx$$

$$= \lambda_{X} \lambda_{Y} \int_{u}^{\infty} \exp(-\lambda_{X}x) \cdot \frac{\exp(-\lambda_{Y}(x+w))}{\lambda_{Y}} \, dx$$

$$= \lambda_{X} \exp(-\lambda_{Y}w) \int_{u}^{\infty} \exp(-(\lambda_{X} + \lambda_{Y})x) \, dx$$

$$= \frac{\lambda_{X}}{\lambda_{Y} + \lambda_{Y}} \exp(-\lambda_{Y}w) \exp(-(\lambda_{X} + \lambda_{Y})u).$$

By switching the roles of X and Y in the above computation, we obtain

$$\mathbb{P}(U > u, W > w, Y \leq X) = \frac{\lambda_Y}{\lambda_X + \lambda_Y} \exp(-\lambda_X w) \exp(-(\lambda_X + \lambda_Y)u).$$

Now, we add the two expressions together to obtain

$$\mathbb{P}(U > u, W > w) = \left(\frac{\lambda_X}{\lambda_X + \lambda_Y} \exp(-\lambda_Y w) + \frac{\lambda_Y}{\lambda_X + \lambda_Y} \exp(-\lambda_X w)\right) \exp(-(\lambda_X + \lambda_Y)u)$$

$$= \mathbb{P}(W > w)\mathbb{P}(U > u).$$

So, U and W are independent!