

Recall

Joint distributions

Let X, Y be two random variables. The *joint cumulative distribution function* (joint CDF) of X, Y is

$$F(a, b) := P\{X \leq a, Y \leq b\} \quad , \forall a, b \in \mathbb{R}.$$

Then the *marginal distributions* (marginal CDFs) are

$$F_X(a) = \lim_{b \rightarrow \infty} F(a, b) =: F(a, \infty) \quad , \forall a \in \mathbb{R},$$

$$F_Y(b) = \lim_{a \rightarrow \infty} F(a, b) =: F(\infty, b) \quad , \forall b \in \mathbb{R}.$$

All the joint probability questions about X, Y can be answered in terms of joint CDF. In particular, $P\{X > a, Y > b\} = 1 - F(a, \infty) - F(\infty, b) + F(a, b)$.

- If X, Y are discrete, then the *joint probability mass function* (joint PMF) is

$$p(x, y) := P\{X = x, Y = y\} \quad , \forall x, y \in \mathbb{R}.$$

Moreover, we have the *marginal PMFs* of X, Y

$$p_X(x) = \sum_y p(x, y) \quad \forall x \in \mathbb{R},$$

$$p_Y(y) = \sum_x p(x, y) \quad \forall y \in \mathbb{R}.$$

and the joint CDF becomes $F(a, b) = \sum_{\substack{x \leq a \\ y \leq b}} p(x, y)$ for all $a, b \in \mathbb{R}$.

- Two random variables X, Y are *joint continuous* if there exists a *joint probability density function* (joint PDF) $f: \mathbb{R}^2 \rightarrow [0, \infty)$ such that

$$P\{(X, Y) \in C\} = \iint_C f(x, y) \, dx dy$$

for all ‘measurable’ sets $C \subset \mathbb{R}^2$. Fortunately, the countable intersections or unions of rectangles are ‘measurable’. In particular, the joint CDF becomes

$$F(a, b) = \int_{-\infty}^b \int_{-\infty}^a f(x, y) \, dx dy \quad , \forall a, b \in \mathbb{R}.$$

If f is continuous at (a, b) , then $f(a, b) = \frac{\partial^2}{\partial a \partial b} F(a, b)$.

Moreover, X, Y are continuous random variables with *marginal PDFs* obtained by

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) \, dy \quad , \forall x \in \mathbb{R},$$

$$f_Y(y) = \int_{-\infty}^{\infty} f(x, y) \, dx \quad , \forall y \in \mathbb{R}.$$

Independent random variables

Two random variables X and Y are *independent* if

$$\begin{array}{ccc}
 P\{X \in A, Y \in B\} = P\{X \in A\}P\{Y \in B\}, \forall A, B \subset \mathbb{R} & & \\
 \updownarrow & & \\
 F(a, b) = F_X(a)F_Y(b), \forall a, b \in \mathbb{R} & & \\
 \swarrow \text{X,Y discrete} & & \searrow \text{X,Y joint continuous} \\
 p(x, y) = p_X(x)p_Y(y), \forall x, y \in \mathbb{R} & & f(x, y) = f_X(x)f_Y(y), \forall x, y \in \mathbb{R}.
 \end{array}$$

Examples

Example 1. Let X, Y be random variables with joint PDF

$$f(x, y) = \begin{cases} ce^{-x}e^{-2y} & \text{if } x, y \in (0, +\infty) \\ 0 & \text{otherwise.} \end{cases}$$

Find the value of c , $P\{X > 1, Y < 1\}$, $P\{X < Y\}$ and marginal PDFs f_X, f_Y . Are X and Y independent?

Solution. Since

$$1 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = \int_0^{\infty} \int_0^{\infty} ce^{-x}e^{-2y} dx dy = c \left(-e^{-x} \Big|_0^{\infty} \right) \left(-\frac{1}{2}e^{-2y} \Big|_0^{\infty} \right) = \frac{c}{2},$$

we have $c = 2$.

Then

$$\begin{aligned}
 P\{X > 1, Y < 1\} &= \int_{-\infty}^1 \int_1^{\infty} f(x, y) dx dy = \int_0^1 \int_1^{\infty} 2e^{-x}e^{-2y} dx dy = 2e^{-1} \left(-\frac{1}{2}e^{-2y} + \frac{1}{2} \right) \\
 &= e^{-1} - e^{-3},
 \end{aligned}$$

and

$$P\{X < Y\} = \int_0^{\infty} \int_0^y 2e^{-x}e^{-2y} dx dy = \int_0^{\infty} 2e^{-2y}(1 - e^{-y}) dy = \frac{1}{3}.$$

By formula, if $x \leq 0$, then $f_X(x) = 0$ and if $x > 0$, then

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_0^{\infty} 2e^{-x}e^{-2y} dy = e^{-x}.$$

Similarly, if $y \leq 0$, then $f_Y(y) = 0$ and if $y > 0$, then

$$f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^{\infty} 2e^{-x}e^{-2y} dx = 2e^{-2y}.$$

Hence $f(x, y) = f_X(x)f_Y(y)$ for $x, y \in \mathbb{R}$, thus X and Y are independent. □

Remark. It is optional for us to make a safe check $\int_{-\infty}^{\infty} f_X(x)dx = 1$ to avoid computational mistakes.

Example 2. Let X, Y be random variables with joint PDF

$$f(x, y) = \begin{cases} \frac{1}{x} & \text{if } 0 < y < x < 1 \\ 0 & \text{otherwise.} \end{cases}$$

Find $E[X]$ and $E[Y]$. Are X and Y independent?

Solution. By formula, if $x \notin (0, 1)$, then $f_X(x) = 0$ and if $x \in (0, 1)$, then

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y)dy = \int_0^x \frac{1}{x}dy = \frac{1}{x} \times x = 1.$$

Similarly, if $y \notin (0, 1)$, then $f_Y(y) = 0$ and if $y \in (0, 1)$, then

$$f_Y(y) = \int_{-\infty}^{\infty} f(x, y)dx = \int_y^1 \frac{1}{x}dx = \ln x \Big|_y^1 = -\ln y.$$

This implies $f(x, y) \neq f_X(x)f_Y(y)$. Hence X and Y are not independent.

Then

$$E[X] = \int_{-\infty}^{\infty} x f_X(x)dx = \int_0^1 x dx = \frac{1}{2},$$

and

$$E[Y] = \int_{-\infty}^{\infty} y f_Y(y)dy = \int_0^1 -y \ln y dy = \left(\frac{y^2}{2} \ln y \Big|_0^1 \right) + \int_0^1 \frac{1}{y} \frac{y^2}{2} dy = \mathbf{0} + \int_0^1 \frac{y}{2} dy = \frac{1}{4}$$

where $\mathbf{0}$ follows from $\lim_{y \rightarrow 0} y^2 \ln y = 0$. (“exponential” \geq “polynomial” \geq “logarithmic”.) \square