

1 Short Answer

(a) Let X be uniform on the interval $[0, 2]$, and define $Y = 2X + 1$. Find the PDF, CDF, expectation, and variance of Y .

(b) Let X and Y have joint distribution

$$f(x, y) = \begin{cases} cxy + 1/4 & x \in [1, 2] \text{ and } y \in [0, 2] \\ 0 & \text{otherwise.} \end{cases}$$

Find the constant c . Are X and Y independent?

(c) Let $X \sim \text{Exp}(3)$.

(i) Find probability that $X \in [0, 1]$.

(ii) Let $Y = \lfloor X \rfloor$. For each $k \in \mathbb{N}$, what is the probability that $Y = k$? Write the distribution of Y in terms of one of the famous distributions; provide that distribution's name and parameters.

(d) Let $X_i \sim \text{Exp}(\lambda_i)$ for $i = 1, \dots, n$ be mutually independent. It is a (very nice) fact that $\min(X_1, \dots, X_n) \sim \text{Exp}(\mu)$. Find μ .

Solution:

(a) Let's begin with the CDF. It will first be useful to recall that

$$F_X(t) = \mathbb{P}(X \leq t) = \begin{cases} 0 & t \leq 0 \\ \frac{t}{2} & t \in [0, 2] \\ 1 & t \geq 2 \end{cases}.$$

Since Y is defined in terms of X , we can compute that

$$\begin{aligned} F_Y(t) &= \mathbb{P}(Y \leq t) = \mathbb{P}[2X + 1 \leq t] \\ &= \mathbb{P}\left[X \leq \frac{t-1}{2}\right] \\ &= F_X\left(\frac{t-1}{2}\right) \\ &= \begin{cases} 0 & t \leq 1 \\ \frac{t-1}{4} & t \in [1, 5] \\ 1 & t \geq 5 \end{cases} \end{aligned}$$

where in the third line we have used the PDF for X . We know that the PDF can be found by taking the derivative of the CDF, so

$$f_Y(t) = \frac{dF_Y(t)}{dt} = \begin{cases} \frac{1}{4} & t \in [1, 5] \\ 0 & \text{else} \end{cases}.$$

By linearity of expectation $\mathbb{E}[Y] = \mathbb{E}[2X + 1] = 2\mathbb{E}[X] + 1 = 3$, and similarly

$$\text{Var}(Y) = \text{Var}(2X + 1) = 4\text{Var}(X) = 4 \cdot \frac{4}{12} = \frac{4}{3}.$$

(b) To find the correct constant, we use the fact that a PDF must integrate to one. In particular,

$$1 = \int_1^2 \int_0^2 (cxy + 1/4) dy dx = 3c + 1/2,$$

so $c = 1/6$. In order to check independence, we need to first find the marginal distributions of X and Y :

$$f_X(x) = \int_0^2 f(x, y) dy = 1/2 + x/3$$

$$f_Y(y) = \int_1^2 f(x, y) dx = 1/4 + y/4.$$

Since $f_X(x)f_Y(y) = 1/8 + y/8 + x/12 + xy/12 \neq 1/4 + xy/6 = f(x, y)$, the random variables are not independent.

(c) (i) Since $X \sim \text{Exp}(3)$, the CDF of X is $F(x) = 1 - e^{-3x}$. Thus we have

$$\mathbb{P}[X \in [0, 1]] = \int_0^1 f(x) dx = F(1) - F(0) = (1 - e^{-3}) - (1 - e^0) = 1 - e^{-3}.$$

(ii) Similarly, if $Y = \lfloor X \rfloor$, then $Y = k$ exactly when $X \in [k, k+1)$, so

$$\begin{aligned} \mathbb{P}[Y = k] &= \mathbb{P}[X \in [k, k+1)) \\ &= \int_k^{k+1} f(x) dx \\ &= F(k+1) - F(k) \\ &= (1 - e^{-3(k+1)}) - (1 - e^{-3k}) \\ &= e^{-3k} - e^{-3(k+1)} \\ &= e^{-3k} (1 - e^{-3}) = (e^{-3})^k (1 - e^{-3}). \end{aligned}$$

In other words, $Y = W - 1$ for $W \sim \text{Geometric}(1 - e^{-3})$.

(d) Since the X_i are independent,

$$\begin{aligned} \mathbb{P}[\min(X_1, \dots, X_n) \leq t] &= 1 - \mathbb{P}[X_1 > t, X_2 > t, \dots, X_n > t] \\ &= 1 - \mathbb{P}[X_1 > t] \cdot \mathbb{P}[X_2 > t] \cdot \dots \cdot \mathbb{P}[X_n > t] \quad \text{by independence} \\ &= 1 - e^{-\lambda_1 t} e^{-\lambda_2 t} \cdot \dots \cdot e^{-\lambda_n t} \\ &= 1 - e^{-(\lambda_1 + \lambda_2 + \dots + \lambda_n)t}. \end{aligned}$$

This is exactly the CDF of an $\text{Exp}(\lambda_1 + \lambda_2 + \dots + \lambda_n)$ random variable, so $\mu = \lambda_1 + \dots + \lambda_n$.

2 First Exponential to Die

Let X and Y be $\text{Exponential}(\lambda_1)$ and $\text{Exponential}(\lambda_2)$ respectively, independent. What is

$$\mathbb{P}(\min(X, Y) = X),$$

the probability that the first of the two to die is X ?

Solution:

Recall that the CDF of an exponential is $\mathbb{P}[X \leq x] = 1 - \exp(-\lambda x)$ for $x \geq 0$.

$$\begin{aligned}\mathbb{P}(\min(X, Y) = X) &= \mathbb{P}(Y > X) = \int_0^\infty \mathbb{P}(Y > X \mid X = x) f_X(x) \, dx = \int_0^\infty e^{-\lambda_2 x} \cdot \lambda_1 e^{-\lambda_1 x} \, dx \\ &= -\frac{\lambda_1}{\lambda_1 + \lambda_2} e^{-(\lambda_1 + \lambda_2)x} \Big|_{x=0}^\infty = \frac{\lambda_1}{\lambda_1 + \lambda_2}.\end{aligned}$$