

## 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  $\mu$ .

### 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  $x \geq 0$ .

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