

(a) For  $x \geq 0$ ,

$$F_X(x) = \int_{-\infty}^x f_X(t) dt = \int_0^x \lambda e^{-\lambda t} dt = [-e^{-\lambda t}]_0^x = 1 - e^{-\lambda x}.$$

For  $x < 0$ , we have  $F_X(x) = \int_{-\infty}^x f_X(t) dt = 0$ . Thus we conclude

$$F_X(x) = \begin{cases} 0, & \text{if } x < 0, \\ 1 - e^{-\lambda x}, & \text{if } x \geq 0. \end{cases}$$

(b) The key step in the following computation uses integration by parts, whereby

$$\int_0^\infty u dv = uv \Big|_0^\infty - \int_0^\infty v du$$

is applied with  $u = x$  and  $v = -e^{-\lambda x}$ :

$$\mathbf{E}[X] = \int_{-\infty}^\infty x f_X(x) dx = \int_0^\infty x \lambda e^{-\lambda x} dx = [-x e^{-\lambda x}]_0^\infty + \int_0^\infty e^{-\lambda x} dx = \frac{1}{\lambda}.$$

(c) Integrating by parts with  $u = x^2$  and  $v = -e^{-\lambda x}$  in the second line below gives

$$\begin{aligned} \mathbf{E}[X^2] &= \int_{-\infty}^\infty x^2 f_X(x) dx = \int_0^\infty x^2 \lambda e^{-\lambda x} dx \\ &= [-x^2 e^{-\lambda x}]_0^\infty + 2 \int_0^\infty x e^{-\lambda x} dx = \frac{2}{\lambda} \mathbf{E}[X] = \frac{2}{\lambda^2}. \end{aligned}$$

Combining with the previous computation, we obtain

$$\text{var}(X) = \mathbf{E}[X^2] - (\mathbf{E}[X])^2 = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2}.$$

(d) The maximum of a set is upper bounded by  $z$  when each element of the set is upper bounded by  $z$ . Thus for any positive  $z$ ,

$$\begin{aligned} \mathbf{P}(Z \leq z) &= \mathbf{P}(\max\{X_1, X_2, X_3\} \leq z) = \mathbf{P}(X_1 \leq z, X_2 \leq z, X_3 \leq z) \\ &= \mathbf{P}(X_1 \leq z) \mathbf{P}(X_2 \leq z) \mathbf{P}(X_3 \leq z) \\ &= (1 - e^{-\lambda z})^3, \end{aligned}$$

where the third equality uses the independence of  $X_1$ ,  $X_2$ , and  $X_3$ . Thus,

$$F_Z(z) = \begin{cases} 0, & \text{if } z < 0, \\ (1 - e^{-\lambda z})^3, & \text{if } z \geq 0. \end{cases}$$

Differentiating the CDF gives the desired PDF:

$$f_Z(z) = \begin{cases} 0, & \text{if } z < 0, \\ 3\lambda e^{-\lambda z}(1 - e^{-\lambda z})^2, & \text{if } z \geq 0. \end{cases}$$

- (e) The minimum of a set is lower bounded by  $w$  when each element of the set is lower bounded by  $w$ . Thus for any positive  $w$ ,

$$\begin{aligned} \mathbf{P}(W \geq w) &= \mathbf{P}(\min\{X_1, X_2\} \geq w) = \mathbf{P}(X_1 \geq w, X_2 \geq w) \\ &= \mathbf{P}(X_1 \geq w) \mathbf{P}(X_2 \geq w) \\ &= (e^{-\lambda w})^2 = e^{-2\lambda w} \end{aligned}$$

where the third equality uses the independence of  $X_1$  and  $X_2$ . Thus,

$$F_W(w) = \begin{cases} 0, & \text{if } w < 0, \\ 1 - e^{-2\lambda w}, & \text{if } w \geq 0. \end{cases}$$

We can recognize this as the CDF of an exponential random variable with parameter  $2\lambda$ . The PDF is

$$f_W(w) = \begin{cases} 0, & \text{if } w < 0, \\ 2\lambda e^{-2\lambda w}, & \text{if } w \geq 0. \end{cases}$$