

**Theorem** If  $X_i \sim \text{exponential}(\lambda_i)$  for  $i = 1, 2, \dots, n$  are mutually independent random variables with survivor functions

$$S_i(x) = e^{-\lambda_i x} \quad x > 0,$$

for  $i = 1, 2, \dots, n$  and  $\lambda_i \neq \lambda_j$  for  $i \neq j$ , then  $X_1 + X_2 + \dots + X_n$  has the hypoexponential probability density function

$$f_{X_1+X_2+\dots+X_n}(x) = \sum_{i=1}^n \lambda_i e^{-\lambda_i x} \prod_{j=1, j \neq i}^n \left( \frac{\lambda_j}{\lambda_j - \lambda_i} \right) \quad x > 0.$$

**Proof** Given in Ross (2007) *Introduction to Probability Models*, 8th ed., Academic Press by using induction by first showing the case of  $n = 2$ , then that showing that case  $n$  implies case  $n + 1$ .

**Illustration.** We show the case of  $n = 2$  here.

*Option 1.* Use the cumulative distribution function technique.

$$\begin{aligned} F_{X_1+X_2}(x) &= P(X_1 + X_2 \leq x) \\ &= \int_0^x \int_0^{x-x_1} \lambda_1 e^{-\lambda_1 x_1} \lambda_2 e^{-\lambda_2 x_2} dx_2 dx_1 \\ &= \dots \\ &= 1 - \left( \frac{\lambda_1}{\lambda_1 - \lambda_2} \right) e^{-\lambda_2 x} - \left( \frac{\lambda_2}{\lambda_2 - \lambda_1} \right) e^{-\lambda_1 x} \quad x > 0. \end{aligned}$$

So the probability density function of  $X_1 + X_2$  is

$$f_{X_1+X_2}(x) = \left( \frac{\lambda_2}{\lambda_2 - \lambda_1} \right) \lambda_1 e^{-\lambda_1 x} + \left( \frac{\lambda_1}{\lambda_1 - \lambda_2} \right) \lambda_2 e^{-\lambda_2 x} \quad x > 0.$$

*Option 2.* Use the convolution formula. Since  $X_1$  and  $X_2$  are independent,

$$\begin{aligned} f_{X_1+X_2}(x) &= \int_0^x f_{X_1}(t) f_{X_2}(x-t) dt \\ &= \int_0^x \lambda_1 e^{-\lambda_1 t} \lambda_2 e^{-\lambda_2(x-t)} dt \\ &= \dots \\ &= \left( \frac{\lambda_2}{\lambda_2 - \lambda_1} \right) \lambda_1 e^{-\lambda_1 x} + \left( \frac{\lambda_1}{\lambda_1 - \lambda_2} \right) \lambda_2 e^{-\lambda_2 x} \quad x > 0. \end{aligned}$$

*Option 3.* Use the transformation technique. (Tedious due to the dummy transformation.)

**APPL illustration:** For  $n = 3$ , the statement

`Convolution(ExponentialRV(lam1), ExponentialRV(lam2), ExponentialRV(lam3))`

gives

$$f_X(x) = \left( \frac{\lambda_2}{\lambda_2 - \lambda_1} \right) \left( \frac{\lambda_3}{\lambda_3 - \lambda_1} \right) \lambda_1 e^{-\lambda_1 x} + \left( \frac{\lambda_1}{\lambda_1 - \lambda_2} \right) \left( \frac{\lambda_3}{\lambda_3 - \lambda_2} \right) \lambda_2 e^{-\lambda_2 x} + \left( \frac{\lambda_1}{\lambda_1 - \lambda_3} \right) \left( \frac{\lambda_2}{\lambda_2 - \lambda_3} \right) \lambda_3 e^{-\lambda_3 x} \quad x > 0.$$