

2.

Temos que :  $X \sim \exp(\lambda)$

$Y \sim \exp(\theta)$

$X$  e  $Y$  independentes

$\delta = I(X \leq Y)$        $Z = \min(X, Y)$

2.a

Sem perda de generalidade, no lugar de  $P(X \in [t, t + \Delta t])$  com  $\Delta t \rightarrow 0$ , vamos escrever simplesmente  $P(X = t)$ .

Assim:

$$P(\delta = 1) = \int_0^{\infty} P(X = t, Y \geq t) dt \stackrel{\text{ind}}{=} \int_0^{\infty} P(X = t) P(Y \geq t) dt$$

$$P(X = t) = \lambda e^{-\lambda t}$$

$$P(Y \geq t) = e^{-\theta t}$$

$$P(\delta = 1) = \int_0^{\infty} \lambda e^{-(\lambda + \theta)t} dt$$

$$P(\delta = 1) = \lambda \cdot \left[ -\frac{e^{-(\lambda + \theta)t}}{(\lambda + \theta)} \right]_{t=0}^{t \rightarrow \infty}$$

Lembremos que:  $\lim_{t \rightarrow \infty} \frac{e^{-at}}{a} = 0$ , se  $a > 0$

Assim:

$$P(\delta = 1) = \frac{\lambda}{\lambda + \theta}$$

2. b

$$P(Z > z) = P(X > z ; Y > z) \stackrel{\text{ind}}{=} P(X > z) P(Y > z)$$

$$P(Z > z) = e^{-(\lambda + \theta)z}, \quad z > 0$$

$$f_z(z) = \frac{d}{dz} F_z(z) = \frac{d}{dz} \left[ 1 - e^{-(\lambda + \theta)z} \right]$$

Assim:

$$F_z(z) = 1 - e^{-(\lambda + \theta)z}$$

$$f_z(z) = (\lambda + \theta) e^{-(\lambda + \theta)z}, \quad z > 0$$

$$\therefore Z \sim \exp(\lambda + \theta)$$

2. c

Se forem independentes:

$$P(\delta = k | Z = z) = P(\delta = k) \quad \forall k \in \{0, 1\} \\ z > 0$$

Para  $k = 1$ :

$$P(\delta = 1 | Z = z) = \frac{P(X \leq Y \cap \min(X, Y) = z)}{P(Z = z)}$$

$$P(\delta = 1 | Z = z) = \frac{P(X = z) P(Y \geq z)}{P(Z = z)}$$

$$= \frac{\lambda e^{-\lambda z} e^{-\theta z}}{(\lambda + \theta) e^{-(\lambda + \theta)z}}$$

$$\therefore P(\delta = 1 | Z = z) = \frac{\lambda}{\lambda + \theta} = P(\delta = 1) \quad \forall z > 0$$

Analogamente:

$$P(\delta = 0 | Z = z) = \frac{\theta}{\lambda + \theta} = P(\delta = 0) \quad \forall z > 0$$

Portanto

$\delta$  e  $Z$  são independentes //

2. d

Lembremos que:

$$P(\delta=1) = \frac{\lambda}{\lambda+\theta}$$

$$P(\delta=0) = \frac{\theta}{\lambda+\theta} = 1 - \frac{\lambda}{\lambda+\theta}$$

Assim, escrevendo  $p = \frac{\lambda}{\lambda+\theta}$ , temos que:

$$P(\delta=1) = p$$

$$P(\delta=0) = 1-p$$

$$f_{\delta}(x) = P(\delta=x) = \left[ \frac{\lambda}{\lambda+\theta} \right]^x \left[ \frac{\theta}{\lambda+\theta} \right]^{1-x} \mathbb{1}_{\{0,1\}}^{(x)}$$

Logo, torna-se evidente que

$$\delta \sim \text{Bernoulli}\left(\frac{\lambda}{\lambda+\theta}\right)$$

De forma então que a distribuição da soma de  $n$  Bernoullis iid é:

$$D \sim \text{Binomial}\left(n, \frac{\lambda}{\lambda+\theta}\right)$$

2. e

$$E(\hat{\lambda}) = E\left[\frac{\sum \delta_i}{\sum Z_i}\right] = E\left[\frac{\delta_1}{\sum Z_i} + \frac{\delta_2}{\sum Z_i} + \dots + \frac{\delta_n}{\sum Z_i}\right]$$

Onde  $\sum Z_i \sim \text{Gamma}(n, \lambda+\theta)$ , pois  $Z_i \sim \exp(\lambda+\theta)$

Além disso:

$$E\left[\frac{\delta_i}{\sum z_i}\right] = E\left[E\left(\frac{\delta_i}{\sum z_i} \mid \delta_i\right)\right]$$

$$= P(\delta_i=0) \cdot E\left[\frac{0}{\sum z_i}\right] + P(\delta_i=1) E\left[\frac{1}{\sum z_i}\right]$$

$$= 0 + \frac{\lambda}{\lambda+\theta} \cdot \int_0^{\infty} \frac{1}{x} \cdot \frac{[\lambda+\theta]^n}{\Gamma(n)} \cdot e^{-(\lambda+\theta)x} \cdot x^{n-1} dx$$

$$= \lambda \cdot \int_0^{\infty} \frac{1}{\Gamma(n)} \cdot [\lambda+\theta]^{n-1} \cdot e^{-(\lambda+\theta)x} \cdot x^{n-2} dx$$

Sabemos que, sendo  $n$  um inteiro positivo:

$$\Gamma(n) = (n+1)! = (n+1) \cdot n! = (n+1) \Gamma(n-1)$$

Assim:

$$E\left[\frac{\delta_i}{\sum z_i}\right] = \frac{\lambda}{(n+1)} \cdot \underbrace{\int_0^{\infty} \frac{[\lambda+\theta]^{n-1}}{\Gamma(n-1)} \cdot e^{-(\lambda+\theta)x} \cdot x^{n-2} dx}_{= \int_0^{\infty} \text{Gamma}(n-1, \lambda+\theta) dx = 1}$$

Assim:

$$E\left[\frac{\delta_i}{\sum z_i}\right] = \frac{\lambda}{n+1} \quad \therefore E[\hat{\lambda}] = \frac{n}{n+1} \lambda$$