

**E 1.3.2** Let  $X$  be the number of defective items.

$$\cdot \Pr(X=1) = \binom{10}{1} (.05)^1 (.95)^9$$

$$\cdot \Pr(X \leq 1) = \binom{10}{0} (.05)^0 (.95)^{10} + \binom{10}{1} (.05)^1 (.95)^9$$

**E 1.3.4**  $\cdot \Pr(X=2) = \frac{\lambda^2}{2!} e^{-\lambda} = \frac{2}{e^2}$

$$\cdot \Pr(X \leq 2) = \frac{\lambda^2}{2!} e^{-\lambda} + \frac{\lambda^1}{1!} e^{-\lambda} + \frac{\lambda^0}{0!} e^{-\lambda} = \frac{2}{e^2} + \frac{2}{e^2} + \frac{1}{e^2} = \frac{5}{e^2}.$$

**P 1.3.3**  $\Pr(X \text{ is odd}) = \sum_{\substack{k=0 \\ k \text{ odd}}}^{\infty} \Pr(X=k) = \sum_{j=0}^{\infty} \Pr(X=2j+1)$

$$= \sum_{j=0}^{\infty} \frac{\lambda^{2j+1}}{(2j+1)!} e^{-\lambda}$$

$$= e^{-\lambda} \left[ \frac{e^{\lambda} - e^{-\lambda}}{2} \right] = \frac{1 - e^{-2\lambda}}{2}.$$

Here I am using the fact that

$$(A) \frac{e^{\lambda} - e^{-\lambda}}{2} = \sum_{\substack{k=0 \\ k \text{ odd}}}^{\infty} \frac{\lambda^k}{k!}, \quad (B) \frac{e^{\lambda} + e^{-\lambda}}{2} = \sum_{\substack{k=0 \\ k \text{ even}}}^{\infty} \frac{\lambda^k}{k!}.$$

These formulas follow from  $e^{\lambda} = \sum_{k=0}^{\infty} \frac{\lambda^k}{k!}$ ,  $e^{-\lambda} = \sum_{k=0}^{\infty} \frac{(-\lambda)^k}{k!}$ .

**P 1.3.4**  $E[V] = \sum_{n=0}^{\infty} \left( \frac{1}{1+n} \right) \binom{\mu^n}{n!} e^{-\mu}$

$$= e^{-\mu} \sum_{n=0}^{\infty} \frac{\mu^n}{(n+1)!} = \frac{e^{-\mu}}{\mu} \sum_{n=0}^{\infty} \frac{\mu^{n+1}}{(n+1)!}$$

$$= \frac{e^{-\mu}}{\mu} [e^{\mu} - 1].$$

**P 1.3.5** For  $Y$ , we need not compute the distribution explicitly.

Since  $X$  is binomial, we have

$$E[X] = Np$$

$$\text{Var}[X] = E[X^2] - (E[X])^2 = Np(1-p),$$

$$\text{so } E[X^2] = Np(1-p) + N^2 p^2.$$

$$\rightarrow E[XY] = E[X(N-X)] = E[NX - X^2]$$

$$= NE[X] - E[X^2]$$

$$= N(Np) - [Np(1-p) + N^2 p^2]$$

$$\rightarrow \text{Cov}[X, Y] = E[XY] - E[X]E[Y]$$

$$= [N^2 p - Np + Np^2 - N^2 p^2] - (Np)(N - Np)$$

$$= Np[p - 1] = 0$$

**P1.3.9**  $p_Z(n)$  for  $n = 0, 1, 2, \dots$ , the mass function for  $Z$ , is given by

$$\begin{aligned} p_Z(n) &= \sum_{k=0}^n p_X(k) p_Y(n-k) \\ &= \sum_{k=0}^n (1-\pi) \pi^k (1-\pi) \pi^{n-k} \\ &= \sum_{k=0}^n (1-\pi)^2 \pi^n \\ &= \left( \sum_{k=0}^n 1 \right) \left[ (1-\pi)^2 \pi^n \right] = (n+1)(1-\pi)^2 \pi^n. \end{aligned}$$

This is indeed a negative binomial.  $Z$  is counting the number of failures before the second success.

**P1.3.13**  $\Pr(\text{sample of 10 gives 2 or more defective parts})$   
 $= 1 - \left[ \binom{10}{0} (.05)^0 (.95)^{10} + \binom{10}{1} (.05)^1 (.95)^9 \right].$

We assume that the sample of 10 is representative and that each part is independent of the others.

**E1.4.1** Let  $X$  denote lightbulb lifetime.

•  $\Pr(X > 1.5 \text{ yr}) = e^{-(2)(1.5)} = e^{-3}$

•  $\Pr(X = 1.5 \text{ yr}) = 0$

**E1.4.2** Let  $\hat{m}$  denote the median. We have  $\int_0^{\hat{m}} \lambda e^{-\lambda t} dt = \frac{1}{2}$ , so  
 $(-e^{-\lambda t}) \Big|_0^{\hat{m}} = \frac{1}{2}$ , or  $(-e^{-\lambda \hat{m}}) + 1 = \frac{1}{2}$ .

Solving for  $\hat{m}$  yields  $\hat{m} = \ln(2)/\lambda$ . The mean is  $\frac{1}{\lambda}$ , so median < mean in this case.

**E1.4.6a** The density function of  $Y$  is computed by change of variable.

Using Eq. (1.15) of Pinsky/Karlin, we have

$$\begin{aligned} f_Y(y) &= \frac{1}{g'(u)} f_U(u) \quad (\text{where } y = g(u) = -\ln(1-u)) \\ &= (1-u) \cdot 1 \\ &= 1 - [1 - e^{-y}] = e^{-y}. \end{aligned}$$

Thus  $Y$  is exponential w/ parameter  $\lambda = 1$ !

**E1.4.7** By convolution,  $R$  has density

$$\begin{aligned} f_R(r) &= \int_0^r \lambda e^{-\lambda(r-s)} \lambda e^{-\lambda s} ds \\ &= \lambda^2 \int_0^r e^{-\lambda r} ds \\ &= \lambda^2 e^{-\lambda r} \int_0^r ds = \lambda^2 e^{-\lambda r} \left( s \Big|_0^r \right) = \lambda^2 r e^{-\lambda r}. \end{aligned}$$

This is a gamma distribution.

|P1.4.3| For an interval  $[a, b]$ , we will write  $\mathbb{1}_{[a, b]}(x) = \begin{cases} 1, & \text{if } a \leq x \leq b; \\ 0, & \text{o.w.} \end{cases}$

for the indicator function of  $[a, b]$ . The density functions for  $X$  and  $(-Y)$  are  $\mathbb{1}_{[0 - \frac{1}{2}, 0 + \frac{1}{2}]}$  and  $\mathbb{1}_{[-0 - \frac{1}{2}, -0 + \frac{1}{2}]}$ , respectively.

The density function for  $W = X - Y$  is given by convolution:

$$\begin{aligned}
 f_W(w) &= \int_{-\infty}^{\infty} \mathbb{1}_{[0 - \frac{1}{2}, 0 + \frac{1}{2}]}(y) \mathbb{1}_{[-0 - \frac{1}{2}, -0 + \frac{1}{2}]}(w - y) dy \\
 &= \int_{0 - \frac{1}{2}}^{0 + \frac{1}{2}} \mathbb{1}_{[-0 - \frac{1}{2}, -0 + \frac{1}{2}]}(w - y) dy \\
 (\text{let } t = w - y) &= \int_{w - [0 + \frac{1}{2}]}^{w - [0 - \frac{1}{2}]} \mathbb{1}_{[-0 - \frac{1}{2}, -0 + \frac{1}{2}]}(t) dt \\
 &= \int_{w - [0 - \frac{1}{2}]}^{w - [0 + \frac{1}{2}]} \mathbb{1}_{[-0 - \frac{1}{2}, -0 + \frac{1}{2}]}(t) dt \\
 &= \text{Length}([w - (0 + \frac{1}{2}), w - (0 - \frac{1}{2})] \cap [-0 - \frac{1}{2}, -0 + \frac{1}{2}]) \\
 &= \text{Length}([w, w + 1] \cap [0, 1]) \\
 &= (1 - |w|) \mathbb{1}_{[-1, 1]}(w) \\
 &= \begin{cases} 1 + w, & \text{for } -1 \leq w < 0; \\ 1 - w, & \text{for } 0 \leq w \leq 1; \\ 0, & \text{for } |w| > 1. \end{cases}
 \end{aligned}$$

|P1.4.5| Since the RVs are independent, the density for  $X, Y$  viewed jointly

is the product  $f(x, y) = 2e^{-2x} \cdot 3e^{-3y} = 6e^{-2x}e^{-3y}$ .

$$\Pr(X < Y) = \int_0^{\infty} \int_x^{\infty} 6e^{-2x}e^{-3y} dy dx$$

$$= 6 \int_0^{\infty} e^{-2x} \int_x^{\infty} e^{-3y} dy dx$$

$$= 6 \int_0^{\infty} e^{-2x} \left( \frac{e^{-3x}}{3} \right) dx$$

$$= 2 \int_0^{\infty} e^{-5x} dx$$

$$= \frac{2}{5}$$

