

SOLUTIONS—ASSIGNMENT 21

Chapter 7. Problems: p. 385 #80. From Table 7.1 we see that X is Poisson with parameter 2 and Y is binomial with parameters $(10, 3/4)$. Thus from independence:

(a)

$$\begin{aligned} P\{X + Y = 2\} &= P\{X = 0, Y = 2\} + P\{X = 1, Y = 1\} + P\{X = 2, Y = 0\} \\ &= P\{X = 0\}P\{Y = 2\} + P\{X = 1\}P\{Y = 1\} + P\{X = 2\}P\{Y = 0\} \\ &= \frac{e^{-2}}{4^{10}} \left[1 \cdot \binom{10}{2} 3^2 + \frac{2}{1} \cdot \binom{10}{1} 3 + \frac{4}{2} \cdot \binom{10}{0} \right] = \frac{467}{4^{10}} e^{-2}; \end{aligned}$$

(b)

$$\begin{aligned} P\{XY = 0\} &= P(\{X = 0\} \cup \{Y = 0\}) = P\{X = 0\} + P\{Y = 0\} - P\{X = Y = 0\} \\ &= e^{-2} + \frac{1}{4^{10}} - \frac{e^{-2}}{4^{10}}; \end{aligned}$$

(c)

$$E[XY] = E[X]E[Y] = 2 \cdot \left[10 \cdot \frac{3}{4} \right] = 15.$$

Theoretical exercises: p. 389 #46. If X is uniform on $[a, b]$, then

$$M_X(t) = E[e^{tX}] = \frac{1}{b-a} \int_a^b e^{tx} dx = \frac{1}{b-a} \left. \frac{e^{tx}}{t} \right|_a^b = \frac{e^{bt} - e^{at}}{t(b-a)}. \quad (*)$$

Note that $M_X(0)$ is not defined by (*), but that $M_X(0) = E[1] = 1$ directly from the definition of M_X , and also that l'Hospital's rule shows that $\lim_{t \rightarrow 0} M_X(t) = 1$.

Now we want to compute $M'_X(0)$. Because (*) does not make sense at $t = 0$ we cannot differentiate (*) and then set $t = 0$ (in fact, one gets the right answer by differentiating (*) and then taking the $t \rightarrow 0$ limit using l'Hospital's rule, but technically this requires justification). We can, however, compute M'_X from the definition of derivative:

$$E[X] = M'_X(0) = \lim_{t \rightarrow 0} \frac{M_X(t) - M_X(0)}{t} = \lim_{t \rightarrow 0} \frac{e^{bt} - e^{at} - t(b-a)}{t^2(b-a)} = \frac{a+b}{2},$$

where we have used l'Hospital's rule. Of course, this is a familiar answer. We can compute $E[X^2] = M''_X(0) = \lim_{t \rightarrow 0} [M'_X(t) - M'_X(0)]/t = (b^2 + ab + a^2)/3$ similarly.

p. 390 #51.

$$\Psi''(t) = \frac{d^2}{dt^2} \log M(t) = \frac{d}{dt} \frac{M'(t)}{M(t)} = \frac{M''(t)M(t) - M'(t)M'(t)}{M(t)^2}.$$

Setting $t = 0$ and using $M(0) = 1$, $M'(0) = E[X]$, and $M''(0) = E[X^2]$ gives $\Psi''(0) = E[X^2] - E[X]^2 = \text{Var}(X)$.

p. 390 #52. . Let $X = \sum_{i=1}^n X_i$. From Table 7.2, $M_{X_i}(t) = \lambda/(\lambda - t)$ for all $t < \lambda$. Since the X_i are independent,

$$M_X(t) = M_{X_1+\dots+X_n}(t) = M_{X_1}(t) \cdots M_{X_n}(t) = \left(\frac{\lambda}{\lambda - t} \right)^n.$$

According to Table 7.2 this is just the moment generating function of a random variable which has a gamma distribution with parameters (n, λ) ; since the moment generating function determines the distribution, X must have this same distribution.