

## Solutions to Assignment #17

1. We have seen in the lecture that if  $X$  has a Poisson distribution with parameter  $\lambda > 0$ , then it has the pmf:

$$p_X(k) = \frac{\lambda^k}{k!} e^{-\lambda} \quad \text{for } k = 0, 1, 2, 3, \dots; \text{ zero elsewhere.}$$

Use the fact that the power series  $\sum_{m=0}^{\infty} \frac{x^m}{m!}$  converges to  $e^x$  for all real values of  $x$  to compute the mgf of  $X$ .

Use the mgf of  $X$  to determine the mean and variance of  $X$ .

**Solution:** Compute

$$\begin{aligned} \psi_X(t) &= E(e^{tX}) \\ &= \sum_{k=0}^{\infty} e^{tk} p_X(k) \\ &= \sum_{k=0}^{\infty} (e^t)^k \frac{\lambda^k}{k!} e^{-\lambda} \\ &= e^{-\lambda} \sum_{k=0}^{\infty} \frac{(\lambda e^t)^k}{k!} \\ &= e^{-\lambda} e^{\lambda e^t} \\ &= e^{\lambda(e^t - 1)}, \end{aligned}$$

for all  $t \in \mathbb{R}$ .

Differentiating the mgf we obtain

$$\psi'_X(t) = \lambda e^t e^{\lambda(e^t - 1)},$$

and

$$\psi''_X(t) = \lambda e^t e^{\lambda(e^t - 1)} + \lambda^2 e^{2t} e^{\lambda(e^t - 1)}.$$

We then get that the expected value of  $X$  is

$$E(X) = \psi'_X(0) = \lambda,$$

and the second moment of  $X$  is

$$E(X^2) = \psi_X''(0) = \lambda + \lambda^2.$$

Consequently, the variance of  $X$  is

$$\text{var}(X) = E(X^2) - \lambda^2 = \lambda.$$

□

2. Let  $X_1, X_2, \dots, X_m$  be independent random variables satisfying  $X_i \sim \text{Poisson}(\lambda)$  for all  $i = 1, 2, \dots, m$  and some  $\lambda > 0$ . Define

$$Y = X_1 + X_2 + \dots + X_m.$$

Determine the distribution of  $Y$ ; that is, compute its pmf.

**Solution:** Use the independence of the random variables  $X_i$ , for  $i = 1, 2, \dots, m$ , to compute the moment generating function of  $Y$ :

$$\begin{aligned} \psi_Y(t) &= \psi_{X_1+X_2+\dots+X_m}(t) \\ &= \psi_{X_1}(t) \cdot \psi_{X_2}(t) \cdots \psi_{X_m}(t) \\ &= e^{\lambda(e^t-1)} \cdot e^{\lambda(e^t-1)} \cdots e^{\lambda(e^t-1)} \\ &= e^{m\lambda(e^t-1)}, \end{aligned}$$

which is the mgf of a  $\text{Poisson}(m\lambda)$  random variable. It then follows that  $Y \sim \text{Poisson}(m\lambda)$  and, therefore, its pmf is given by

$$p_Y(k) = \frac{(m\lambda)^k}{k!} e^{-m\lambda} \quad \text{for } k = 0, 1, 2, 3, \dots; \text{ zero elsewhere.}$$

□

3. [Exercise 2 on page 262 in the text]

Suppose that on a given weekend the number of accidents at a certain intersection has a Poisson distribution with mean 0.7. What is the probability that there will be at least three accidents in the intersection during the weekend?

**Solution:** Let  $X$  denote the number of accidents that occur in the intersection during a weekend. Then, we are assuming that  $X \sim \text{Poisson}(0.7)$  and therefore

$$\Pr(X = k) = \frac{(0.7)^k}{k!} e^{-0.7} \quad \text{for } k = 0, 1, 2, 3, \dots; \text{ zero elsewhere.}$$

Thus, the probability of at least three accidents during that weekend,  $\Pr(X \geq 3)$ , is

$$\begin{aligned} \Pr(X \geq 3) &= 1 - \Pr(X < 3) \\ &= 1 - \Pr(X = 0) - \Pr(X = 1) - \Pr(X = 2) \\ &= 1 - e^{-0.7} - (0.7)e^{-0.7} - \frac{(0.7)^2}{2}e^{-0.7}, \end{aligned}$$

or about 3.4%. □

4. [Exercise 6 on page 262 in the text]

Suppose that a certain type of magnetic tape contains, on average, three defects per 1000 feet. What is the probability that a roll of tape 1200 feet long contains no defects?

**Solution:** Let  $X$  denote the number of defects in one foot of tape. Then  $X \sim \text{Poisson}(\lambda)$ , where

$$\lambda = \frac{3}{1000} = 0.003.$$

It is reasonable to assume that the number of defects in a foot of the tape is independent of the number of defects in any other foot of the tape. Thus, if we let  $Y$  denote the number of defects in 1200 feet of the tape, it follows from problem 2 in this assignment that

$$Y \sim \text{Poisson}(1200\lambda) \quad \text{or} \quad \text{Poisson}(3.6).$$

Thus,

$$\Pr(Y = k) = \frac{(3.6)^k}{k!} e^{-3.6} \quad \text{for } k = 0, 1, 2, 3, \dots; \text{ zero elsewhere.}$$

Consequently, the probability of no defects in the 1200 feet of tape is

$$\Pr(Y = 0) = e^{-3.6}$$

or about 2.73%. □

5. [Exercise 8 on page 262 in the text]

Suppose that  $X_1$  and  $X_2$  are independent random variables and that  $X_i$  has a Poisson distribution with mean  $\lambda_i$  ( $i = 1, 2$ ). For a fixed value of  $k$  ( $k = 0, 1, 2, 3, \dots$ ), determine the conditional distribution of  $X_1$  given that  $X_1 + X_2 = k$ .

**Solution:** Let  $Y = X_1 + X_2$ . Then, since  $X_1$  and  $X_2$  are independent with  $X_i \sim \text{Poisson}(\lambda_i)$ , for  $i = 1, 2$ ,  $Y \sim \text{Poisson}(\lambda_1 + \lambda_2)$ . Thus, the pmf of  $Y$  is

$$\Pr(Y = k) = \frac{(\lambda_1 + \lambda_2)^k}{k!} e^{-\lambda_1 - \lambda_2} \quad \text{for } k = 0, 1, 2, 3, \dots; \text{ zero elsewhere.}$$

We compute the conditional probability  $\Pr(X_1 = m \mid Y = k)$  for  $m = 0, 1, 2, 3, \dots$

$$\begin{aligned} \Pr(X_1 = m \mid Y = k) &= \frac{\Pr(X_1 = m, Y = k)}{\Pr(Y = k)} \\ &= \frac{\Pr(X_1 = m, X_1 + X_2 = k)}{\Pr(Y = k)} \\ &= \frac{\Pr(X_1 = m, X_2 = k - m)}{\Pr(Y = k)} \\ &= \frac{\Pr(X_1 = m) \cdot \Pr(X_2 = k - m)}{\Pr(Y = k)}, \end{aligned}$$

by the independence of  $X_1$  and  $X_2$ ; so that, for  $m = 0, 1, 2, \dots, k$ ,

$$\begin{aligned} \Pr(X_1 = m \mid Y = k) &= \frac{\frac{\lambda_1^m}{m!} e^{-\lambda_1} \cdot \frac{\lambda_2^{k-m}}{(k-m)!} e^{-\lambda_2}}{\frac{(\lambda_1 + \lambda_2)^k}{k!} e^{-\lambda_1 - \lambda_2}} \\ &= \frac{\frac{\lambda_1^m}{m!} \cdot \frac{\lambda_2^{k-m}}{(k-m)!}}{\frac{(\lambda_1 + \lambda_2)^k}{k!}} \\ &= \frac{k!}{m!(k-m)!} \cdot \frac{\lambda_1^m \lambda_2^{k-m}}{(\lambda_1 + \lambda_2)^k}. \end{aligned}$$

Observing that the first factor in the last expression is the binomial coefficient  $\binom{k}{m}$ , we can write

$$\begin{aligned}\Pr(X_1 = m \mid Y = k) &= \binom{k}{m} \cdot \frac{\lambda_1^m}{(\lambda_1 + \lambda_2)^m} \cdot \frac{\lambda_2^{k-m}}{(\lambda_1 + \lambda_2)^{k-m}} \\ &= \binom{k}{m} \cdot \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^m \cdot \left(\frac{\lambda_2}{\lambda_1 + \lambda_2}\right)^{k-m} \\ &= \binom{k}{m} p^m (1-p)^{k-m},\end{aligned}$$

for  $m = 0, 1, 2, \dots, k$ , where  $p = \frac{\lambda_1}{\lambda_1 + \lambda_2}$ .

Hence, the conditional distribution of  $X_1$  given  $X_1 + X_2 = k$  is Binomial( $p, k$ ) with  $p = \frac{\lambda_1}{\lambda_1 + \lambda_2}$ .  $\square$