

**IE 605:Engineering Statistics**

## Tutorial 3 Solution

**Solution 1**

- Say  $Z = \min(X, Y)$ 

$$P(Z > k) = P(\min(X, Y) > k) = P(X > k)P(Y > k) = e^{-\lambda_1 k} e^{-\lambda_2 k}$$

$$= e^{-(\lambda_1 + \lambda_2)k}$$

$$P(Z \leq k) = 1 - e^{-(\lambda_1 + \lambda_2)k}$$

I.e.  $Z \sim \exp(\lambda_1 + \lambda_2)$ .
- Say  $Z = \max(X, Y)$ 

$$P(Z \leq k) = P(\max(X, Y) \leq k) = P(X \leq k)P(Y \leq k)$$

$$= (1 - e^{-\lambda_1 k})(1 - e^{-\lambda_2 k}).$$

**Solution 2**

Given:  $X, Y, Z$  are discrete random variable.

The joint pmf of  $X, Y, Z$  is,  $p_{X,Y,Z}(x, y, z)$  where  $x \in A_1, y \in A_2, z \in A_3$  and  $A_1, A_2, A_3$  are the support of  $x, y, z$  respectively.

- To show:  $\mathbb{E}[Z] = \mathbb{E}[\mathbb{E}[Z|Y]]$ .

$$\begin{aligned} \mathbb{E}[\mathbb{E}[Z|Y]] &= \sum_y \mathbb{E}[Z|Y] \cdot p_Y(y) \quad \text{where } p_Y(y) \text{ is marginal pmf of } Y \in A_2 \\ &= \sum_y p_Y(y) \sum_x \sum_z z \cdot p_{(X,Z)|Y}(x, z|y) \\ &= \sum_x \sum_y \sum_z z \cdot p_Y(y) p_{(X,Z)|Y}(x, z|y) \\ &= \sum_x \sum_y \sum_z z \cdot p_Y(y) \frac{p_{X,Z,Y}(x, y, z)}{p_Y(y)} \\ &= \sum_x \sum_y \sum_z z \cdot p_{X,Z,Y}(x, y, z) \\ &= \mathbb{E}[Z]. \end{aligned}$$

- To show:  $\mathbb{E}[Z] = \mathbb{E}[\mathbb{E}[Z|X, Y]]$ .

$$\begin{aligned}
\mathbb{E}[\mathbb{E}[Z|X, Y]] &= \sum_x \sum_y \mathbb{E}[Z|X, Y] \cdot p_{X,Y}(x, y) \\
&= \sum_x \sum_y p_{X,Y}(x, y) \sum_z z \cdot p_{Z|X,Y}(z|x, y) \\
&= \sum_x \sum_y \sum_z z \cdot p_{X,Y}(x, y) \frac{p_{X,Z,Y}(x, y, z)}{p_{X,Y}(x, y)} \\
&= \sum_x \sum_y \sum_z z \cdot p_{X,Z,Y}(x, y, z) \\
&= \mathbb{E}[Z].
\end{aligned}$$

### Solution 3

$$E[X|Y = 3] = \sum_{x=0}^3 xP(X = x|Y = 3) = \sum_{x=1}^3 xP(X = x|Y = 3)$$

$$P(X = 1|Y = 3) = \frac{P(X = 1 \cap Y = 3)}{P(Y = 3)} = \frac{\binom{6}{3} * \left(\frac{5}{14}\right)^3 * \binom{3}{1} * \left(\frac{3}{14}\right) * \left(\frac{6}{14}\right)^2}{\binom{6}{3} * \left(\frac{5}{14}\right)^3 * \left(\frac{9}{14}\right)^3} = \frac{4}{9}$$

$$P(X = 2|Y = 3) = \frac{P(X = 2 \cap Y = 3)}{P(Y = 3)} = \frac{\binom{6}{3} * \left(\frac{5}{14}\right)^3 * \binom{3}{2} * \left(\frac{3}{14}\right)^2 * \left(\frac{6}{14}\right)}{\binom{6}{3} * \left(\frac{5}{14}\right)^3 * \left(\frac{9}{14}\right)^3} = \frac{2}{9}$$

$$P(X = 3|Y = 3) = \frac{P(X = 3 \cap Y = 3)}{P(Y = 3)} = \frac{\binom{6}{3} * \left(\frac{5}{14}\right)^3 * \binom{3}{3} * \left(\frac{3}{14}\right)^3}{\binom{6}{3} * \left(\frac{5}{14}\right)^3 * \left(\frac{9}{14}\right)^3} = \frac{1}{27}$$

$$E[X|Y = 3] = 1 * \frac{4}{9} + 2 * \frac{2}{9} + 3 * \frac{1}{27} = 1$$

### Solution 4

$$X_1 + X_2 = m$$

Since  $X_1$  and  $X_2$  are independent,

$$\mathbb{P}\{X_1 + X_2 = m\} = \binom{n_1+n_2}{m} p^m (1-p)^{n_1+n_2-m}.$$

$$P(X_1 = k|X_1 + X_2 = m) = \frac{P(X_1=k \cap X_1+X_2=m)}{P(X_1+X_2=m)}$$

$$P(X_1 = k|X_1 + X_2 = m) = \frac{P(X_1=k)P(X_2=m-k)}{P(X_1+X_2=m)} \text{ as } X_1, X_2 \text{ are independent}$$

$$P(X_1 = k|X_1 + X_2 = m) = \frac{\binom{n_1}{k} p^k (1-p)^{n_1-k} \binom{n_2}{m-k} p^{m-k} (1-p)^{n_2-m+k}}{\binom{n_1+n_2}{m} p^m (1-p)^{n_1+n_2-m}}$$

$$P(X_1 = k|X_1 + X_2 = m) = \frac{\binom{n_1}{k} \binom{n_2}{m-k}}{\binom{n_1+n_2}{m}}.$$

### Solution 5

Let  $X \sim U([-1, 1])$  and  $Y = X^2$ . Then, it is evident that  $X, Y$  are not independent and  $Cov(X, Y) = E((X - E(X))(Y - E(Y))) = E(X^3) = 0$ .

## Solution 6

Given two random variables  $X, Y$  such that their joint distribution is as follows:

$$f_{X,Y}(x, y) = \begin{cases} c(1 + xy) & \text{if } 2 \leq x \leq 3, 1 \leq y \leq 2 \\ 0 & \text{otherwise.} \end{cases}$$

1. We know that,  $\int_{y=1}^2 \int_{x=2}^3 f(x, y) dx dy = 1$ . Thus, on substitution we get,  $c = 4/19$

2.

$$\begin{aligned} f_X(x) &= \int_{y=1}^2 f_{X,Y}(x, y) dy \\ &= \begin{cases} \frac{4}{19}(1 + 3x/2) & \text{if } 2 \leq x \leq 3 \\ 0 & \text{otherwise.} \end{cases} \end{aligned}$$

Similarly,

$$\begin{aligned} f_Y(y) &= \int_{x=2}^3 f_{X,Y}(x, y) dx \\ &= \begin{cases} \frac{4}{19}(1 + 5y/2) & \text{if } 1 \leq y \leq 2 \\ 0 & \text{otherwise.} \end{cases} \end{aligned}$$

3.

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{f_{X,Y}(x, y)}{f_Y(y)} \\ &= \begin{cases} \frac{2(1+xy)}{2+5y} & \text{if } 2 \leq x \leq 3, 1 \leq y \leq 2 \\ 0 & \text{otherwise.} \end{cases} \end{aligned}$$

## Solution 7

Let  $X$  denote the number of accidents that a randomly chosen policyholder has next year. Letting  $Y$  be the Poisson mean number of accidents for this policyholder, then conditioning on  $Y$  yields

$$\begin{aligned} P\{X = n\} &= \int_0^\infty P\{X = n|Y = \lambda\}g(y)dy \\ &= \int_0^\infty e^{-\lambda} \frac{\lambda^n}{n!} \lambda e^{-\lambda} dy \\ &= \frac{1}{n!} \int_0^\infty e^{-2\lambda} \lambda^{n+1} dy \end{aligned}$$

Since,

$$h(y) = \frac{2e^{-2\lambda}(2\lambda)^{n+1}}{(n+1)!}, \lambda > 0$$

is the density function of a gamma  $(n + 2, 2)$  random variable, its integral is 1. Therefore,

$$1 = \int_0^\infty \frac{2e^{-2\lambda}(2\lambda)^{n+1}}{(n+1)!} d\lambda = \frac{2^{n+2}}{(n+1)!} \int_0^\infty \lambda^{n+1} e^{-2\lambda} d\lambda$$

$$\implies P\{X = n\} = \frac{n+1}{2^{n+2}}$$

## Solution 8

Let  $N_1$  denote the number of women and  $N_2$  the number of men who visit the academy today. Also, let  $N = N_1 + N_2$  be the total number of people who visit. Conditioning on  $N$  gives,

$$\mathbb{P}\{N_1 = n, N_2 = m\} = \sum_{i=0}^{\infty} \mathbb{P}\{N_1 = n, N_2 = m | N = i\} \mathbb{P}\{N = i\}$$

Because  $\mathbb{P}\{N_1 = n, N_2 = m | N = i\} = 0$  when  $i \neq n + m$ , the preceding equation yields

$$\mathbb{P}\{N_1 = n, N_2 = m\} = \mathbb{P}\{N_1 = n, N_2 = m | N = n + m\} e^\lambda \frac{\lambda^{n+m}}{(n+m)!}$$

Given that  $n + m$  people visit it follows, because each of these  $n + m$  is independently a woman with probability  $p$ , that the conditional probability that  $n$  of them are women (and  $m$  are men) is just the binomial probability of  $n$  successes in  $n + m$  trials. Therefore,

$$\begin{aligned} \mathbb{P}\{N_1 = n, N_2 = m\} &= \binom{n+m}{n} p^n (1-p)^m e^\lambda \frac{\lambda^{n+m}}{(n+m)!} \\ &= \frac{(n+m)!}{n!m!} p^n (1-p)^m e^{\lambda p} e^{\lambda(1-p)} \frac{\lambda^n \lambda^m}{(n+m)!} \\ &= e^{\lambda p} \frac{(\lambda p)^n}{n!} e^{\lambda(1-p)} \frac{(\lambda(1-p))^m}{m!} \end{aligned}$$

Because the preceding joint probability mass function factors into two products, one of which depends only on  $n$  and the other only on  $m$ , it follows that  $N_1$  and  $N_2$  are independent. Moreover, because

$$\begin{aligned} \mathbb{P}\{N_1 = n\} &= \sum_{i=0}^{\infty} \mathbb{P}\{N_1 = n, N_2 = m\} \\ &= e^{\lambda p} \frac{(\lambda p)^n}{n!} \sum_{i=0}^{\infty} e^{\lambda(1-p)} \frac{(\lambda(1-p))^m}{m!} \end{aligned}$$

and similarly,

$$\mathbb{P}\{N_2 = m\} = e^{\lambda(1-p)} \frac{(\lambda(1-p))^m}{m!}$$

we can conclude that  $N_1$  and  $N_2$  are independent Poisson random variables with respective means  $\lambda p$  and  $\lambda(1-p)$ . Therefore, this example establishes the important result that when each of a Poisson number of events is independently classified either as being type 1 with probability  $p$  or type 2 with probability  $1-p$ , then the numbers of type 1 and type 2 events are independent Poisson random variables.

## Solution 9

Given: A chicken lays  $n$  eggs. Each egg independently does or doesn't hatch, with probability  $p$  of hatching.

Let  $N$ : number of eggs which hatch out of  $n$  eggs.

$$\implies N \sim \text{Bin}(n, p).$$

For each egg that hatches, the chick does or doesn't survive (independently of the other eggs), with probability  $s$  of survival.

Let  $X$ : the number of chicks which survive out of  $N$  chicks.

$$\implies X|N \sim \text{Bin}(N, s).$$

Let  $Y$ : the number of chicks which hatch but don't survive (so  $X + Y = N$ ).

To find: PMF of  $X$

$$\begin{aligned} p_X(x) = \mathbb{P}\{X = x\} &= \sum_{n'=x}^n p_{X|N}(x|N = n') p_N(n') \\ &= \sum_{n'=x}^n \binom{n'}{x} s^x (1-s)^{n'-x} \binom{n}{n'} p^{n'} (1-p)^{n-n'} \\ &= \sum_{n'=x}^n \frac{n!}{x!(n'-x)!} s^x (1-s)^{n'-x} \cdot \frac{n!}{n!(n-n')!} p^{n'} (1-p)^{n-n'} \end{aligned}$$

Let  $n' - x = r$ . Then,  $n' = x \implies r = 0$  and  $n' = n \implies r = n - x$

$$\begin{aligned} &= \sum_{r=0}^{n-x} \frac{n!}{x!r!(n-r-x)!} s^x (1-s)^r p^{r+x} (1-p)^{n-r-x} \\ &= \frac{n!}{x!(n-x)!} (ps)^x (1-p)^{n-x} \sum_{r=0}^{n-x} \frac{(n-x)!}{r!(n-r-x)!} (1-s)^r p^r (1-p)^{-r} \\ &= \binom{n}{x} (ps)^x (1-p)^{n-x} \sum_{r=0}^{n-x} \binom{n-x}{r} \left( \frac{(1-s)p}{1-p} \right)^r \\ &= \binom{n}{x} (ps)^x (1-p)^{n-x} \left( 1 + \frac{(1-s)p}{1-p} \right)^{n-x} \\ &= \binom{n}{x} (ps)^x (1-p)^{n-x} \left( \frac{1-p + (1-s)p}{1-p} \right)^{n-x} \\ &= \binom{n}{x} (ps)^x (1-p)^{n-x} \left( \frac{1-ps}{1-p} \right)^{n-x} \\ &= \binom{n}{x} (ps)^x (1-ps)^{n-x} \implies X \sim \text{Bin}(n, ps). \end{aligned}$$

