Tutorial 5

October 22, 2020

Question 1

A discrete random variable N is uniformly distributed on $\{1, 2, 3, \dots, 10\}$.

Let X be the indicator of the event $\{N \leq 5\}$.

Let Y be the indicator of the event $\{N \text{ is even }\}.$

(a) Are X and Y independent?

$$X = \begin{cases} 1 & N \le 5 \\ 0 & \text{otherwise} \end{cases} \qquad Y = \begin{cases} 1 & N \text{ even} \\ 0 & \text{otherwise} \end{cases}$$

We can check for independence of X and Y by checking

$$\mathbf{P}\left(X=1\,\cap\,Y=1\right) \stackrel{?}{=} \mathbf{P}\left(X=1\right) \cdot \mathbf{P}\left(Y=1\right)$$

$$\mathbf{P}(X=1) = \mathbf{P}(N \le 5)$$
$$= \frac{5}{10}$$

$$\mathbf{P}(Y=1) = \mathbf{P}(N \text{ even})$$
$$= \frac{5}{10}$$

$$\mathbf{P}(X = 1 \cap Y = 1) = \mathbf{P}(N \le 5 \cap N \text{ even})$$
$$= \frac{2}{10}$$

Since $\mathbf{P}(X=1 \cap Y=1) \neq \mathbf{P}(X=1) \cdot \mathbf{P}(Y=1)$, X and Y are not independent.

(b) Find **E** $((X + Y)^2)$.

First note that

$$XY = \begin{cases} 1 & N \le 5 \cap N \text{ even} \\ 0 & \text{otherwise} \end{cases}$$

It follows that

$$\mathbf{E}(XY) = 1 \cdot \mathbf{P}(N \le 5 \cap N \text{ even}) + 0 \cdot \mathbf{P}((N \le 5 \cap N \text{ even})^c)$$
$$= \mathbf{P}(N \le 5 \cap N \text{ even})$$

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It should also be noted that the square of an indicator variable is identical to the original indicator variable.

$$\mathbf{E} ((X+Y)^{2}) = \mathbf{E} (X^{2} + 2XY + Y^{2})$$

$$= \mathbf{E} (X^{2}) + 2\mathbf{E} (XY) + \mathbf{E} (Y^{2})$$

$$= \mathbf{E} (X) + 2\mathbf{E} (XY) + \mathbf{E} (Y)$$

$$= \mathbf{P} (N \le 5) + 2\mathbf{P} (N \le 5 \cap N \text{ even}) + \mathbf{P} (N \text{ even})$$

$$= \frac{5}{10} + 2\left(\frac{2}{10}\right) + \frac{5}{10}$$

$$= \frac{14}{10} = 1.4$$

Question 2

13 cards are drawn at random without replacement from an ordinary deck of playing cards. If X is the number of spades in these 13 cards, find the PMF of X. If, in addition, Y is the number of hearts in these 13 cards, find the probability $\mathbf{P}(X=2,Y=5)$. What is the joint PMF of X and Y?

Let X be the number of spades in the 13 cards drawn at random without replacement from an ordinary deck of 52 cards. As there are 13 spades and 39 non-spades in an ordinary deck, the PMF of X is

$$\mathbf{P}(X=x) = \begin{cases} \frac{\binom{13}{x} \binom{39}{13-x}}{\binom{52}{13}} & 0 \le x \le 13\\ 0 & \text{otherwise} \end{cases}$$

Now, let Y be the number of hearts contained in these 13 cards. There are 13 hearts in an ordinary deck, leaving 26 cards that are non-spade and non-heart. The joint probability mass function is

$$\mathbf{P}(X = x, Y = y) = \begin{cases} \frac{\binom{13}{x} \binom{13}{y} \binom{26}{13 - x - y}}{\binom{52}{13}} & 0 \le x \le 13, \ 0 \le y \le 13 - x \\ 0 & \text{otherwise} \end{cases}$$

for $x, y \in \mathbb{Z}$.

The desired probability is computed as:

$$\mathbf{P}(X=2, Y=5) = \frac{\binom{13}{2} \binom{13}{5} \binom{26}{13-7}}{\binom{52}{13}}$$

$$=\frac{\frac{13!}{2!11!}\frac{13!}{5!8!}\frac{26!}{6!20!}}{\frac{52!}{13!39!}}$$

$$=3.647 \times 10^{-7}$$

Question 3

Consider the multinomial distribution:

- $m \ge 2$ categories
- $n \ge 1$ items chosen at random, with replacement
- $p_k = \mathbf{P}$ (Item of type k chosen), $k = 1, \dots, m$
- X_k = Number of type k chosen, k = 1, ..., m
- (a) Compute $P(X_1 = x_1, X_2 = x_2, ..., X_m = x_m)$.

Computing this probability is equivalent to finding the joint PMF.

$$\mathbf{P}(X_1 = x_1, X_2 = x_2, \dots, X_m = x_m)$$

$$= \binom{n}{x_1} \binom{n - x_1}{x_2} \cdots \binom{n - x_1 - \dots - x_{m-1}}{x_m} p_1^{x_1} p_2^{x_2} \cdots p_m^{x_m}$$

$$= \frac{n!}{x_1! x_2! \cdots x_m!} p_1^{x_1} p_2^{x_2} \cdots p_m^{x_m}$$

provided that $\sum_{i=1}^{m} x_i = n$.

The full form of the joint PMF is:

$$p(x_1, x_2, ..., x_m) = \begin{cases} \frac{n!}{x_1! x_2! \cdots x_m!} p_1^{x_1} p_2^{x_2} \cdots p_m^{x_m} & \sum_{i=1}^m x_i = n \\ 0 & \text{otherwise} \end{cases}$$

(b) Find the marginal distribution of X_k for each k. Are X_i and X_j independent?

Intuitively, we expect that X_k will have a binomial distribution with parameters (n, p_k) . This can be shown formally. Without loss of generality, take k = 1. We know that to find the marginal distribution of X_1 , we will need to sum over everything except X_1 , i.e. X_2 through X_m .

$$p_{X_{1}}(x_{1}) = \sum_{\substack{x_{k} \\ 2 \le k \le m}} \frac{n!}{x_{1}! x_{2}! \cdots x_{m}!} p_{1}^{x_{1}} p_{2}^{x_{2}} \cdots p_{m}^{x_{m}}$$

$$= \sum_{\substack{x_{k} \\ 2 \le k \le m}} \frac{n!}{(n - x_{1})! x_{1}!} \cdot \frac{(n - x_{1})!}{x_{2}! x_{3}! \cdots x_{m}!} p_{1}^{x_{1}} p_{2}^{x_{2}} \cdots p_{m}^{x_{m}}$$

$$= \frac{n!}{(n - x_{1})! x_{1}!} p_{1}^{x_{1}} \sum_{\substack{x_{k} \\ 2 \le k \le m}} \frac{(n - x_{1})!}{x_{2}! x_{3}! \cdots x_{m}!} p_{2}^{x_{2}} p_{3}^{x_{3}} \cdots p_{m}^{x_{m}}$$

$$= \binom{n}{x_{1}} p_{1}^{x_{1}} (p_{2} + p_{3} + \dots + p_{m})^{n - x_{1}} \qquad \text{(Multinomial Theorem)}$$

$$= \binom{n}{x_{1}} p_{1}^{x_{1}} (1 - p_{1})^{n - x_{1}}$$

Thus we have

$$p_{X_1}(x_1) = \begin{cases} \binom{n}{x_1} p_1^{x_1} (1 - p_1)^{n - x_1} & 0 \le x_1 \le n \\ 0 & \text{otherwise} \end{cases}$$

Iterating through the remaining k = 2, ... m, it can be shown that each X_k will have a binomial distribution with parameters (n, p_k) .

While the results of the individual trials still remain independent, X_i will not be independent of X_j since

$$\mathbf{P}(X_i = n, X_j = n) = 0 \neq \mathbf{P}(X_i = n) \cdot \mathbf{P}(X_j = n)$$

Question 4

Let $(X_1, X_2, X_3) \sim \text{Multi}(n, p_1, p_2, p_3)$. Find the conditional distribution of X_1 given that $X_3 = x_3$. Intuitively, we expect that

$$X_1 | X_3 = x_3 \sim \text{Binomial}\left(n - x_3, \frac{p_1}{p_1 + p_2}\right)$$

$$\mathbf{P}(X_1 | X_3 = x_3) = \frac{\mathbf{P}(X_1 = x_1, X_3 = x_3)}{\mathbf{P}(X_3 = x_3)}$$

$$= \frac{\mathbf{P}(X_1 = x_1, X_2 = n - x_1 - x_3, X_3 = x_3)}{\mathbf{P}(X_3 = x_3)}$$

$$= \frac{\frac{n!}{x_1! (n - x_1 - x_3)! x_3!} p_1^{x_1} p_2^{n - x_1 - x_3} p_3^{x_3}}{\frac{n!}{(n - x_3)! x_3!} p_3^{x_3} (1 - p_3)^{n - x_3}}$$

$$= \frac{(n-x_3)!}{(n-x_1-x_3)!} \frac{p_1^{x_1} p_2^{x_2}}{(p_1+p_2)^{x_1+x_2}}$$

$$= \binom{n-x_3}{x_1} \left(\frac{p_1}{p_1+p_2}\right)^{x_1} \left(\frac{p_2}{p_1+p_2}\right)^{x_2}$$

$$= \binom{n-x_3}{x_1} \left(\frac{p_1}{p_1+p_2}\right)^{x_1} \left(\frac{p_2}{p_1+p_2}\right)^{n-x_3-x_1}$$

for $0 \le x_1 \le n - x_3$, and zero elsewhere.

As expected,

$$X_1 | X_3 = x_3 \sim \text{Binomial}\left(n - x_3, \frac{p_1}{p_1 + p_2}\right)$$

Question 5

Suppose $X \sim \text{Bin}(N, p)$, where the number of trials, N, is also a random variable (but independent of the trials themselves). Then conditioned on the fact that N = n, the number of successes, X, would have distribution Bin(n, p). What can be said about the unconditional distribution of X, in particular the case when N is a Poisson random variable?

This is actually the exact same question as Tutorial 3 Question 1, but worded slightly differently... We saw in Tutorial 3 Question 1 that if

$$N \sim \text{Poisson}(\lambda)$$

and the conditional distribution of $X \mid N = n$ was

$$X \mid N = n \sim \text{Binomial}(n, p)$$

then applying the law of total probability, the unconditional distribution of X was

$$X \sim \text{Poisson}(\lambda p)$$

Question 6

Following the setup of the previous question, let Y = N - X represent the number of failures. It is implied that Y has (unconditional) distribution $Poisson(\lambda \cdot (1-p))$. Show that X and Y are independent. [Note that this is strongly due to the Poisson distribution of N, and does not happen otherwise (i.e. with deterministic N).]

We can show that X and Y are independent by finding their joint PMF. Before beginning, we recall that

$$X \mid N = n \sim \text{Binomial}(n, p)$$

$$Y \mid N = n \sim \text{Binomial}(n, (1-p))$$

$$\begin{aligned} \mathbf{P}\left(X=x,\,Y=y\right) &= \mathbf{P}\left(X=x,\,Y=y\,|\,N=n\right) \cdot \mathbf{P}\left(N=n\right) \\ &= \mathbf{P}\left(X=x,\,Y=y\,|\,N=x+y\right) \cdot \mathbf{P}\left(N=x+y\right) \\ &= \frac{(x+y)!}{x!\,y!}\,p^x(1-p)^y \cdot e^{-\lambda}\frac{\lambda^{x+y}}{(x+y)!} \\ &= e^{-\lambda} \cdot \frac{\lambda^x\,p^x}{x!} \cdot \frac{\lambda^y\,(1-p)^y}{y!} \\ &= e^{-\lambda+\lambda p-\lambda p} \cdot \frac{(\lambda p)^x}{x!} \cdot \frac{(\lambda(1-p))^y}{y!} \\ &= e^{-\lambda p-\lambda(1-p)} \cdot \frac{(\lambda p)^x}{x!} \cdot \frac{(\lambda(1-p))^y}{y!} \\ &= e^{-\lambda p}\frac{(\lambda p)^x}{x!} \cdot e^{-\lambda(1-p)}\frac{(\lambda(1-p))^y}{y!} \end{aligned}$$

for $x \ge 0$, $y \ge 0$, and zero otherwise.

Since the joint PMF of X and Y is a product of their (unconditional) marginal PMFs, X and Y are independent.