Probability & Statistics for EECS: Homework #09

Due on Apr 16, 2023 at 23:59

Name: Wang Penghao Student ID: 2021533138

(a) As for the joint PMF of X, Y, N, we have that the PMF is P(X = x, Y = y, N = n), then as we have that N = X + Y, then only when x + y = n, will the PMF be non-zero. So we have that

$$P(X = x, Y = y, N = n) = P(X = x, Y = y) = (1 - p)^{x} * p * (1 - p)^{y} * p = (1 - p)^{x+y}p^{2}$$

, as we have that x + y = n, so we get that

$$P(X = x, Y = y, N = n) = (1 - p)^n p^2$$

(b) As for the joint PMF os X, N, we have that the PMF is P(X = x, N = n), as only when n = x + y will the PMF be non-zero, so we have that

$$P(X = x, N = n) = P(X = x, Y = n - x) = (1 - p)^{x} p(1 - p)^{n - x} p = (1 - p)^{n} p^{2}$$

(c) As for the conditional PMF of X given N = n, we have that the PMF is

$$P(X = x | N = n) = \frac{P(X = x, N = n)}{P(N = n)}.$$

The numerator is the joint PMF of X and N, which is $P(X = x, N = n) = (1 - p)^n p^2$, and the denominator is PMF of N, which is $P(N = n) = \sum_{x=0}^{n} (1 - p)^n p^2 = (n + 1)(1 - p)^n p^2$, so we have that

$$P(X = x | N = n) = \frac{P(X = x, N = n)}{P(N = n)} = \frac{(1 - p)^n p^2}{(n + 1)(1 - p)^n p^2} = \frac{1}{n + 1}.$$

where x = 0, 1, 2, ..., n.

Description: The conditional PMF of X given N=n is a uniform distribution, which is $P(X=x|N=n)=\frac{1}{n+1}$. The event P(X=x) is a Geom distribution, while the event N=n is actually a negative binomial distribution, which denote the fail times before the second success. So the conditional PMF of X given N=n is $\frac{1}{n+1}$, which denote that the first success between the first and the second success is uniformly distributed.

(a) To verify that the conditional distribution of X given X > c is the same as the distribution of c + X, firstly we can find the corresponding CDF of X given X > c, which is $P(X \le x|X > c) = \frac{P(c < X \le x)}{P(X > c)} = \frac{F(x) - F(c)}{1 - F(c)}$. As $X \sim Expo(\lambda)$, so we have $F(x) = 1 - e^{-\lambda x}$. So, the $P(X \le x|X > c) = \frac{e^{-\lambda c} - e^{-\lambda x}}{e^{-\lambda c}} = \frac{1 - e^{-\lambda(x-c)}}{1 - e^{-\lambda(x-c)}}$.

As for the CDF of c + X, we have that $P(c+X \le x) = P(X \le x-c) = 1 - e^{-\lambda(x-c)}$. So we have that $P(X \le x|X > c) = P(c+X \le x)$. So that the conditional CDF of X given X > c is the same as the c + X.

(b) As for the CDF of X given X < c, we have that for x < c, $P(X \le x | X < c) = \frac{P(X \le x, X < c)}{P(X < c)} = \frac{P(X \le x)}{P(X < c)} = \frac{1 - e^{-\lambda x}}{1 - e^{-\lambda c}}$ As for the PDF, we have that $f(x | X < c) = (P(X \le x | X < c))' = (\frac{P(X \le x)}{P(X < c)})' = (\frac{1 - e^{-\lambda x}}{1 - e^{-\lambda c}})' = \frac{\lambda e^{-\lambda x}}{1 - e^{-\lambda c}}$ for x < c, as for $x \ge c$, PDF is zero.

As we have that U_1, U_2, U_3 be i.i.d. Unif(0, 1), and let $L = min(U_1, U_2, U_3)$, $M = max(U_1, U_2, U_3)$

- (a) 1. As for the marginal CDF of M is $F_M(m) = P(M \le m) = P(U_1 \le m, U_2 \le m, U_3 \le m) = P(U_1 \le m)P(U_2 \le m)P(U_3 \le m) = m^3$. For $m \in [0, 1]$
 - 2. As for the marginal PDF of M, we have that $f_M(m) = (F_M(m))' = (m^3)' = 3m^2$. For $m \in [0,1]$
 - 3. As for the joint CDF of M and L, firstly we consider the event $L > l, M \le m$, which is easy to calculate, that is $P(L > l, M \le m) = (m-l)^3$, we have that $P(L \le l, M \le m) = P(M \le m) P(L > l, M \le m) = m^3 (m-l)^3$ for $m \ge l$ and that $m, l \in [0, 1]$.
 - 4. As for the joint PDF of M and L, we have that $f(l,m) = \frac{\partial^2 P(L \leq l, M \leq m)}{\partial l \partial m} = 6(m-l)$ for $m, l \in [0, 1]$ and that $m \geq l$.
- (b) As for the conditional PDF of M given L, we firstly find the CDF, that is