

Probability & Statistics for EECS:

Homework #09

Due on Apr 16, 2023 at 23:59

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Problem 1

Problem 2

- (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 of 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 $x = 0, 1, 2, \dots, n$.

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.

Problem 3

- (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 $F_X(x|X > c) = \frac{P(c < X \leq x)}{P(X > c)} = \frac{F_X(x) - F_X(c)}{1 - F_X(c)}$. Then we can find the corresponding CDF of X given $X > c$, which is $F_X(x|X > c) = \frac{F_X(x) - F_X(c)}{1 - F_X(c)}$, as $X \sim \text{Expo}(\lambda)$, so we have $F_X(x) = 1 - e^{-\lambda x}$. So, the $F_X(x|X > c) = \frac{e^{-\lambda c} - e^{-\lambda x}}{e^{-\lambda c}} = 1 - e^{-\lambda(x-c)}$.

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

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

Problem 4

As we have that U_1, U_2, U_3 be i.i.d. $\text{Unif}(0, 1)$, 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 $F_M(m) = P(M \leq m) = P(U_1 \leq m, U_2 \leq m, U_3 \leq m) = P(U_1 \leq m)P(U_2 \leq m)P(U_3 \leq m) = m^3$.
2. As for the marginal PDF of M, we have that $f_M(m) = (F_M(m))' = (m^3)' = 3m^2$.
3. As for the joint CDF of M and L, firstly we consider the event $L \geq l, M \leq m$, which is easy to calculate, that is $P(L \geq l, M \leq m) = (m-l)^3$ we have that $F_{M,L}(m, l) = P(M \leq m, L \leq l) =$.
4. As for the joint PDF of M and L, we have that $f_{M,L}(m, l) = (F_{M,L}(m, l))'$.
- (b) As for the conditional PDF of M given L, we firstly find the CDF, that is

Problem 5