Probability & Statistics for EECS: Homework #010

Due on Apr 23, 2023 at 23:59

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(1) As for X discrete, Y discrete, we have that

$$P(X = x) = \sum_{y} P(X = x, Y = y) = \sum_{y} P(X = x | Y = y) P(Y = y)$$

(2) As for X continuous, Y discrete, we have that

$$P(X = x) = \sum_{-\infty}^{\infty} P(X = x | Y = y) f_Y(y) dy$$

Then we have that,

$$\lim_{\varepsilon \to 0} P(X \in (x - \varepsilon, x + \varepsilon)) = \lim_{\varepsilon \to 0} \sum_{y} P(X \in (x - \varepsilon, x + \varepsilon) | Y = y) P(Y = y)$$

So we have that

$$f_X(x) = \sum_{y} f_X(x|Y=y)P(Y=y)$$

(3) As for X discrete, Y continuous, as we have that

$$P(X = x|Y = y) = \frac{f_Y(y|X = x)P(X = x)}{f_Y(y)}$$

Then we have

$$P(X = x|Y = y)f_Y(y) = f_Y(y|X = x)P(X = x)$$

Then we integrate both sides with respect to y, we have that

$$\int_{-\infty}^{\infty} P(X=x|Y=y)f_Y(y)dy = \int_{-\infty}^{\infty} f_Y(y|X=x)P(X=x)dy$$

Then we have that

$$f_X(x) = \int_{-\infty}^{\infty} f_Y(y|X=x)P(X=x)dy = P(X=x)$$

So we get that

$$P(X = x) = \int_{-\infty}^{\infty} P(X = x | Y = y) f_Y(y) dy$$

(4) As for X continuous, Y continuous, we have

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y)dy = \int_{-\infty}^{\infty} f_{Y|X}(y|x)f_X(x)dy = \int_{-\infty}^{\infty} f_{Y|X}(y|x)dy$$

So we get that

$$f_X(x) = \int_{-\infty}^{\infty} f_{Y|X}(y|x)dy$$

1. First, we let U be that the arrival time of the next Blissville company bus, then $U \sim Unif(0,15)$ as the bus comes every 15 minutes, then we let $X \sim Expo(\frac{1}{15})$ is that be the arrival time of the next Blotchville company bus, then we have that

$$P(X < U) = \int_0^{15} P(X < U|U = u) \frac{1}{15} du$$

So we have that

$$P(X < U) = \frac{1}{15} \int_0^{15} (1 - e^{-\frac{u}{15}}) du = \frac{1}{e}$$

2. As for the wait time, that is the wait until the first bus comes. Denote that the wait time is W, then we have W = min(X, U) Then as for the CDF, we firstly calculate the P(W > t), then we have that

$$P(W > t) = P(X > t, U > t) = P(X > t)P(U > t) = e^{-\frac{t}{15}} (1 - \frac{t}{15})$$

So we get that the CDF of the waiting time is that

$$P(W \le t) = 1 - P(W > t) = 1 - e^{-\frac{t}{15}} (1 - \frac{t}{15}),$$

where $t \in (0, 10)$ and CDF is 0 for $t \le 0$ and 1 for $t \ge 10$.

(a) Firstly we denote that p is the probability that an egg hatch, and q = 1 - p. As we have that X is the number which hatch, and that Y is the number which do not hatch, we have that N = X + Y, then we have that N, X, Y are dependent, as N is the sum of two variables. Then we have that

$$P(N = n, X = x, Y = y) = P(X = x, Y = y)$$

$$= \sum_{n=0}^{\infty} P(X = x, Y = y | N = n) P(N = n)$$

$$= P(X = x, Y = y | N = x + y) P(N = x + y)$$

$$= P(X = x | N = x + y) P(N = x + y)$$

$$= {x + y \choose x} p^{x} q^{y} \frac{e^{-\lambda} \lambda^{x+y}}{(x+y)!}$$

$$= \frac{e^{-\lambda p} (\lambda p)^{x}}{x!} \frac{e^{-\lambda p} (\lambda p)^{y}}{y!}$$

Where n, x, y are nonnegative integers and n = x + y. N, X, Y are not independent, but as X and Y are independent, we also get that $X \sim Pois(\lambda p), Y \sim Pois(\lambda q)$

(b) As for the joint PMF of N, X, as from (a) we have that $X \sim Pois(\lambda p)$ and that $Y \sim Pois(\lambda q)$ Then we have

$$P(N = n, X = x) = P(X = x, Y = n - x) = \frac{e^{-\lambda p} \lambda p^x}{x!} \frac{e^{-\lambda q} \lambda q^{n-x}}{(n-x)!},$$

where $n \geq x$ and that X and N are dependent as $N \geq X$

(c) As for joint PMF of X, Y, from (a) we have that

$$P(X = x, Y = y) = \frac{e^{-\lambda p}(\lambda p)^x}{x!} \frac{e^{-\lambda p}(\lambda p)^y}{y!}$$

where x and y are nonnegative integers

(d) As for the relationship of X and N, from (a) we have that $X \sim Pois(\lambda p)$ and $Y \sim Pois(\lambda q)$, then we have that

$$Cov(N, X) = Cov(X + Y, X) = Cov(X, X) + Cov(Y, X)$$

As X and Y are independent, then

$$Cov(N, X) = Var(X) = \lambda p$$

We then have

$$Corr(N, X) = \frac{Cov(X, Y)}{\sqrt{Var(X)Var(Y)}} = \frac{\lambda p}{\sqrt{\lambda \lambda p}} = \sqrt{p}$$

Denote that the two measurements are X,Y, as they are 2 independent standard Normal random variables, so X,Y i.i.d. $\sim N(0,1)$ and denote that $M=\max(X,Y), \ L=\min(X,Y)$ So $\max(x,y)+\min(x,y)=x+y$, and $\max(x,y)-\min(x,y)=|x-y|$, we then have that

$$E(M) + E(L) = E(M + L) = E(X + Y) = E(X) + E(Y) = 0$$