

Probability & Statistics for EECS:

Homework #010

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Problem 1

(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,

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

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

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

Problem 2

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

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

where $t \in (0, 15)$ and CDF is 0 for $t \leq 0$ and 1 for $t \geq 15$.

Problem 3

- (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

$$\begin{aligned}
 P(N = n, X = x, Y = y) &= \sum_{n=0}^{\infty} P(X = x, Y = y | N = n) P(N = n) \\
 &= P(X = x, Y = y | N = x + y) P(N = x + y) \\
 &= P(X = x | N = x + y) P(N = x + y) \\
 &= \binom{x+y}{x} p^x q^y \frac{e^{-\lambda} \lambda^{x+y}}{(x+y)!} \\
 &= \frac{e^{-\lambda p} (\lambda p)^x}{x!} \frac{e^{-\lambda q} (\lambda q)^y}{y!}
 \end{aligned}$$

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 $X \sim \text{Pois}(\lambda p)$, $Y \sim \text{Pois}(\lambda q)$

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

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

where x and y are nonnegative integers

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

$$\text{Cov}(N, X) = \text{Cov}(X + Y, X) = \text{Cov}(X, X) + \text{Cov}(Y, X)$$

As X and Y are independent, then

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

We then have

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

Problem 4

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

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

Then we have that as $X - Y = \sqrt{X}$ where $Z \sim N(0, 1)$, and that $E(|X - Y|) = \sqrt{2}E|Z|$, we then have that By using 1D LOTUS, we have that

$$E|Z| = \int_{-\infty}^{\infty} |z| \frac{1}{\sqrt{2\pi}} e^{-z^2/2} dz = 2 \int_0^{\infty} z \frac{1}{\sqrt{2\pi}} e^{-z^2/2} dz = \sqrt{\frac{2}{\pi}}$$

So we get that

$$E(M) - E(L) = E(M - L) = E(|X - Y|) = \frac{2}{\sqrt{\pi}}$$

So from $E(M) + E(L)$ and that $E(M) - E(L)$, we get that $E(M) = \frac{1}{\sqrt{\pi}}$, and that $E(L) = -\frac{1}{\sqrt{\pi}}$, then we also get that as $ML = XY$ and that $E(X) = E(Y) = 0$, we have that

$$\text{Cov}(M, L) = E(ML) - E(M)E(L) = E(XY) + \frac{1}{\pi} = 0 + \frac{1}{\pi} = \frac{1}{\pi}$$

Then as for $\text{Var}(M)$ and $\text{Var}(L)$, we have that as $E(X) = E(Y) = 0$, then

$$\text{Var}(X - Y) = E[(X - Y)^2] - (E[X - Y])^2 = E[(X - Y)^2] - (E(X) - E(Y))^2 = E[(X - Y)^2] = 2$$

As we also have that $X - Y = M - L$, then we have that

$$\text{Var}(X - Y) = E[(M - L)^2] = E[M^2] + E[L^2] - 2E[ML]$$

Then we have that

$$\text{Var}(X - Y) = E[M^2] + E[L^2] - 2E[x]E[Y] = E[M^2] + E[L^2]$$

As for the property of Variance, we have that $\text{Var}(M) = E(M^2) - EM^2$, the same for L . So we get

$$\text{Var}(X - Y) = \text{Var}(M) + EM^2 + \text{Var}(L) + EL^2 = \text{Var}(M) + \text{Var}(L) = \frac{2}{\pi}$$

Then as for the property of Normal distribution, we have $\text{Var}(M) = \text{Var}(L)$, we get that

$$\text{Var}(M) = \text{Var}(L) = (2 - \frac{2}{\pi})/2 = 1 - \frac{1}{\pi}$$

So we get the Corr is that

$$\text{Corr}(M, L) = \frac{\text{Cov}(M, L)}{\sqrt{\text{Var}(M)\text{Var}(L)}} = \frac{\frac{1}{\pi}}{1 - \frac{1}{\pi}} = \frac{1}{\pi - 1}$$

Problem 5

- (a) As for the relationship between $Cov(X, Y)$ and sample variance, according to the definition of Cov , we have that

$$Cov(X, Y) = E([X - EX][Y - EY])$$

As $EX = \bar{x}$ and $EY = \bar{y}$, then we have that

$$Cov(X, Y) = E([X - \bar{x}][Y - \bar{y}]) = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}).$$

Which is the definition of sample covariance, so we get that

$$Cov(X, Y) = r.$$

- (b) As for the total signed area of the rectangles, we have that the area is that

$$S = \sum_{i < j} (x_i - x_j)(y_i - y_j).$$

Then as for

$$E((X - \tilde{X})(Y - \tilde{Y})) = E(XY) + E(\tilde{X}\tilde{Y}) - E(X\tilde{Y}) - E(Y\tilde{X})$$

As according to the definition of the \tilde{X} and \tilde{Y} , we have that they have the same distribution with X and Y . So, we have that as X and \tilde{Y} , Y and \tilde{X} are independent,

$$E(X\tilde{Y}) = E(Y\tilde{X}) = E(X)E(Y)$$

and that

$$E(\tilde{X}\tilde{Y}) = E(XY)$$

So we have that

$$E((X - \tilde{X})(Y - \tilde{Y})) = 2E(XY) - 2E(X)E(Y) = 2Cov(X, Y)$$

1. As we have that

$$E((X - \tilde{X})(Y - \tilde{Y})) = \frac{0 * n + 2 \sum_{i < j} (x_i - x_j)(y_i - y_j)}{n^2} = \frac{2A}{n^2}$$

2. As we have that

$$E((X - \tilde{X})(Y - \tilde{Y})) = 2Cov(X, Y),$$

we get that

$$Cov(X, Y) = \frac{A}{n^2}.$$

- (c) As for the following property, we have

- (i) As W_1, W_2 are 2 r.v.s, then we as for exchange W_1, W_2 , we consider that as exchange 2 axis in a coordinates, the space stay the same, so we have that

$$Cov(W_1, W_2) = Cov(W_2, W_1).$$

- (ii) By multiply a constant to the axis, we can consider it as stretch the rectangle area on the corresponding axis, so that

$$Cov(a_1 W_1, a_2 W_2) = a_1 a_2 Cov(W_1, W_2)$$

- (iii) By adding a constant on a axis, this operation can be considered as do shift to the rectangle on the corresponding axis, so will not do any effect to the area. So we have that

$$\text{Cov}(W_1 + a_1, W_2, a_2) = \text{Cov}(W_1, W_2).$$

- (iv) As for the $W_2 + W_3$, we can consider that it as the combination of 2 rectangles, so that whether we split it to 2 rectangles or in a single rectangle is the same area. So we have that

$$\text{Cov}(W_1, W_2 + W_3) = \text{Cov}(W_1, W_2) + \text{Cov}(W_1, W_3)$$