

1.  $|\rho(x, y)| = 1$ , precisely when there is equality in Cauchy-Schwarz Inequality.

Thus  $|\rho(x, y)| = 1$

$$\Rightarrow a'(X - EX) = Y - EY \quad \text{for some } a' \in \mathbb{R}.$$

$$\Rightarrow Y = -a'EX + EY + a'X$$

$$\Rightarrow Y = a + bX \quad \text{for some } a, b \in \mathbb{R}.$$

$$2.a) M_{Y,Z}(t_1, t_2) = E[e^{t_1 Y + t_2 Z}] = E[e^{t_1(X_1 + X_2) + t_2(X_1^{\sim} + X_2^{\sim})}]$$

$$= E[e^{t_1 X_1 + t_2 X_1^{\sim}}] E[e^{t_1 X_2 + t_2 X_2^{\sim}}]$$

$$= \left( E[e^{t_1 X + t_2 X^{\sim}}] \right)^{\sim} \quad \text{where } X \sim N(0, 1).$$

$$E[e^{t_1 X + t_2 X^{\sim}}] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{t_1 u + t_2 u^{\sim}} e^{-\frac{u^2}{2}} du \quad \left( \text{Provided } t_2 < \frac{1}{2} \right)$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(1-2t_2)}{2} \left\{ u^{\sim} - 2u \frac{t_1}{1-2t_2} + \frac{t_1^2}{(1-2t_2)^2} - \frac{t_1^2}{(1-2t_2)^2} \right\}} du$$

$$= \frac{1}{(1-2t_2)^{-1/2}} e^{\frac{t_1^2}{2(1-2t_2)}}$$

$$\text{Thus } M_{Y,Z}(t_1, t_2) = \frac{1}{(1-2t_2)^{-1}} e^{\frac{t_1^2}{1-2t_2}} \quad \text{for } t_2 < \frac{1}{2}.$$

$$\begin{aligned}
 b) \quad EY &= \left. \frac{\partial}{\partial t_1} M_{Y,Z}(t_1, t_2) \right|_{t_1=0, t_2=0} \\
 EZ &= \left. \frac{\partial}{\partial t_2} M_{Y,Z}(t_1, t_2) \right|_{t_1=0, t_2=0} \\
 EYZ &= \left. \frac{\partial^2}{\partial t_1 \partial t_2} M_{Y,Z}(t_1, t_2) \right|_{t_1=0, t_2=0} \\
 EY^2 &= \left. \frac{\partial^2}{\partial t_1^2} M_{Y,Z}(t_1, t_2) \right|_{t_1=0, t_2=0} \\
 EZ^2 &= \left. \frac{\partial^2}{\partial t_2^2} M_{Y,Z}(t_1, t_2) \right|_{t_1=0, t_2=0}
 \end{aligned}
 \left. \vphantom{\begin{aligned} EY \\ EZ \\ EYZ \\ EY^2 \\ EZ^2 \end{aligned}} \right\} \text{Do the calculations.}$$

3. The given MGF is the MGF of a  $\text{Bin}(2, \frac{3}{4})$ .

$$\begin{aligned}
 (a) \quad P(X=K) &= \binom{2}{K} \left(\frac{1}{4}\right)^K \left(\frac{3}{4}\right)^{2-K} \text{ for } K=0, 1, 2 \\
 &= 0 \text{ otherwise}
 \end{aligned}$$

$$(b) \quad Y \sim \text{Bin}(6, \frac{1}{4}).$$

$$\begin{aligned}
 \text{Thus } P(Y=K) &= \binom{6}{K} \left(\frac{1}{4}\right)^K \left(\frac{3}{4}\right)^{6-K} \text{ for } K=0, 1, 2, 3, 4, 5, 6 \\
 &= 0 \text{ otherwise.}
 \end{aligned}$$

$$\begin{aligned}
 4. \quad &P(\{X \leq 2\} \mid \{4.999 < Y \leq 5.001\}) \\
 &= \frac{\int_{4.999}^{5.001} \int_0^2 e^{-(2x+3y)} dx dy}{\int_{4.999}^{5.001} \int_0^\infty e^{-(2x+3y)} dx dy}
 \end{aligned}$$

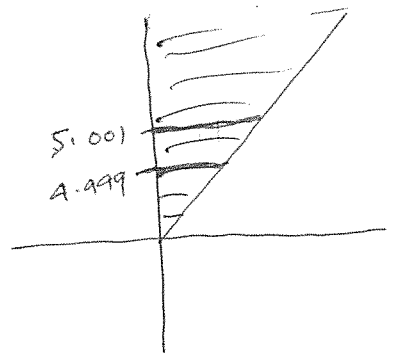
$$4. \quad P(\{X \leq x\} | \{4.999 < Y \leq 5.001\})$$

$$= P(\{X \leq x, 4.999 < Y \leq 5.001\})$$

$$= \frac{P(4.999 < Y \leq 5.001)}{P(4.999 < Y \leq 5.001)}$$

$$P(4.999 < Y \leq 5.001) = 15 \int_{4.999}^{5.001} \int_0^y e^{-(2x+3y)} dx dy$$

$$= \frac{15}{2} \int_{4.999}^{5.001} (e^{-3y} - e^{-5y}) dy$$



$$P(\{X \leq x, 4.999 < Y \leq 5.001\}) = 0 \text{ if } x \leq 0$$

$$= 15 \int_0^x \int_{4.999}^{5.001} e^{-(2x+3y)} dy dx \text{ for } 0 < x \leq 4.999$$

$$= 15 \left[ \int_0^{4.999} \int_{4.999}^{5.001} e^{-2x+3y} dy dx + \int_{4.999}^x \int_x^{5.001} e^{-2x+3y} dy dx \right] \text{ for } 4.999 < x \leq 5.001$$

$$= P(4.999 < Y \leq 5.001) \text{ for } x > 5.001.$$

$$5. \quad P(X=5) = 0.04 + 0.12 + 0.21 + 0.05 \\ = 0.42$$

$$P(Y=1 | X=5) = \frac{0.04}{0.42} = \frac{2}{21}$$

$$P(Y=2 | X=5) = \frac{0.12}{0.42} = \frac{2}{7}$$

$$P(Y=3 | X=5) = \frac{0.21}{0.42} = \frac{1}{2}$$

$$P(Y=4 | X=5) = \frac{0.05}{0.42} = \frac{5}{42}$$

$$6. \quad P(Y=k) = \sum_{j=0}^{\infty} P(Y=k | X=j) P(X=j)$$

$$= \sum_{j=k}^{\infty} P(Y=k | X=j) P(X=j)$$

$$= \sum_{j=k}^{\infty} \binom{j}{k} p^k (1-p)^{j-k} e^{-\lambda} \frac{\lambda^j}{j!}$$

$$= \frac{(\lambda p)^k}{k!} \sum_{j=k}^{\infty} e^{-\lambda} \frac{\{\lambda(1-p)\}^{j-k}}{(j-k)!} = e^{-\lambda} e^{\lambda(1-p)} \frac{(\lambda p)^k}{k!} \\ = e^{-\lambda p} \frac{(\lambda p)^k}{k!}$$

$$P(X=k | Y=y) = \frac{P(X=k, Y=y)}{P(Y=y)} = \frac{P(Y=y | X=k) P(X=k)}{e^{-\lambda p} \frac{(\lambda p)^y}{y!}} \\ \text{for } k=y, y+1, \dots \\ = 0 \text{ otherwise}$$

$$= \frac{\binom{k}{y} p^y (1-p)^{k-y} \times e^{-\lambda} \frac{\lambda^k}{k!}}{e^{-\lambda p} \frac{(\lambda p)^y}{y!}} \quad \text{for } k=y, y+1, \dots$$

$$= 0 \quad \text{otherwise.}$$

$$= \frac{\binom{k}{y} (1-p)^{k-y} \lambda^{k-y} e^{-\lambda(1-p)}}{(k-y)!} \quad \text{for } k=y, y+1, \dots$$

$$= 0 \quad \text{otherwise.}$$

Then conditional distribution of  $X$  given  $Y=y$  is  $y + P((1-\lambda)p)$ .

7. ~~Problem~~ Problem Set 4, Q. 5 we have seen

that  $X \sim \text{NB}\left(k, 1 - \frac{\theta_1}{1-\theta_2}\right)$  and

$Y \sim \text{NB}\left(k, 1 - \frac{\theta_2}{1-\theta_2}\right)$ .

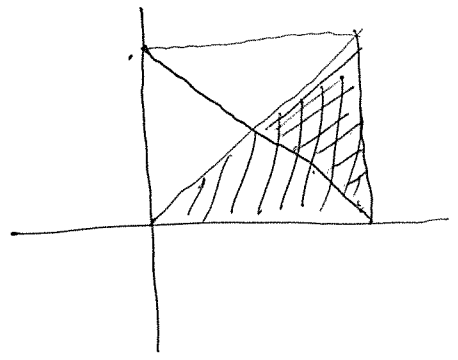
Rest is trivial.

8. Conditional distribution of  $X_2$  given  $X_1 = x$  is  $U(0, x)$ .

$$f_{X_1, X_2}(x_1, x_2) = f_{X_2|X_1}(x_2|x_1) f_{X_1}(x_1)$$

$$= \frac{1}{x_1} \quad \text{for } 0 < x_2 < x_1 < 1.$$

$$P(X_1 + X_2 \geq 1) = \int_{\frac{1}{2}}^1 \int_{1-x_1}^{x_1} \frac{1}{x_1} dx_2 dx_1$$



$$= \int_{\frac{1}{2}}^1 \frac{2x_1 - 1}{x_1} dx_1$$

$$= 2\left(1 - \frac{1}{2}\right) - \log x_1 \Big|_{\frac{1}{2}}^1$$

$$= 1 - \log 2.$$

$$f_{X_2}(x_2) = \int_{x_2}^1 \frac{1}{x_1} dx_1 = -\log x_2.$$

$$E[X_1 | X_2 = x_2] = \frac{\int_{x_2}^1 x_1 \cdot \frac{1}{x_1} dx_1}{-\log x_2} = \frac{1 - x_2}{-\log x_2}$$

$$= \frac{x_2 - 1}{\log x_2}.$$

[Q9] The marginals are found in Problem Set 04,

Question no. 9.

The conditional PDF of  $X$  given  $Y$  is as follows:

For  $y \in (0,1)$ ,

$$f_{X|Y}(x|y) = \begin{cases} \frac{\Gamma(\theta_1 + \theta_3)}{\Gamma(\theta_1) \Gamma(\theta_3)} \frac{x^{\theta_1-1} (1-x-y)^{\theta_3-1}}{(1-y)^{\theta_1+\theta_3-1}} & \text{if } 0 < x < 1-y \\ 0 & \text{o.w.} \end{cases}$$

The conditional PDF of  $Y$  given  $X$  is as follows:

For  $x \in (0,1)$

$$f_{Y|X}(y|x) = \begin{cases} \frac{\Gamma(\theta_2 + \theta_3)}{\Gamma(\theta_2) \Gamma(\theta_3)} \times \frac{y^{\theta_2-1} (1-x-y)^{\theta_3-1}}{(1-x)^{\theta_2+\theta_3-1}} & \text{if } 0 < y < 1-x \\ 0 & \text{o.w.} \end{cases}$$

□

[Q10] Let  $N$  denote the number of accidents in a week. Let  $x_1, x_2, \dots, x_N$  denote the number of injured in accident 1, 2,  $\dots$ ,  $N$ , respectively.

Here  $E(N) = 4$ ,  $E(x_i) = 2$  for all  $i$ .

The number of injured in a week is  $\sum_{i=1}^N x_i$ . We need to find

$$E\left(\sum_{i=1}^N x_i\right) = E\left(E\left(\sum_{i=1}^N x_i \mid N\right)\right) \quad \text{--- (1)}$$

$$\text{Now } E\left(\sum_{i=1}^N x_i \mid N=n\right) = E\left(\sum_{i=1}^n x_i \mid N=n\right) = E\left(\sum_{i=1}^n x_i\right) = 2n$$

$$\text{(1)} \Rightarrow E\left(\sum_{i=1}^N x_i\right) = 2E(N) = 8.$$

□

Q11 Let us define  $X_i = \begin{cases} 1 & \text{if } i^{\text{th}} \text{ Person takes his/her own hat.} \\ 0 & \text{o.w.} \end{cases}$

Then total number of matches  $X = \sum_{i=1}^n X_i$ .

We need to find  $E(X | X_1 = 0)$ .

We know that  $E(X) = 1$ .

Again  $E(X) = E(E(X | X_1))$

$$= E(X | X_1 = 0) P(X_1 = 0) + \underline{E(X | X_1 = 1) P(X_1 = 1)} \quad (*)$$

Now  $P(X_1 = 0) = \frac{n-1}{n}$  and  $P(X_1 = 1) = \frac{1}{n}$ .

$$E(X | X_1 = 1)$$

$$= E\left(\sum_{i=1}^n X_i \mid X_1 = 1\right)$$

$$= E\left(1 + \sum_{i=2}^n X_i \mid X_1 = 1\right)$$

$$= 1 + E\left(\sum_{i=2}^n X_i \mid X_1 = 1\right)$$

$$= 1 + E\left(\sum_{i=2}^n X_i\right),$$

$$= 2$$

As  $X_1 = 1$  means that the 1st person takes his/her own hat and then in the pool of hats, there are ~~only~~ hats of rest of the  $(n-1)$  persons' hats. Hence the conditional expectation ~~is~~  $E\left(\sum_{i=2}^n X_i \mid X_1 = 1\right)$  is same as  $E\left(\sum_{i=2}^n X_i\right)$



Now using ④

$$1 = E(x | x_1=0) \times \frac{n-1}{n} + \frac{2}{n}$$

□.

$$\Rightarrow E(x | x_1=0) = \frac{n-2}{n-1}$$

[Note:  $E(x | x_1=0) \neq E(\sum_{i=2}^n x_i)$ . WHY?]

[Q12] Let us define

$$Y = \begin{cases} 1 & \text{if the 1st door is selected} \\ 2 & \text{if " 2nd " " " "} \\ 3 & \text{if " 3rd " " " "} \end{cases}$$

and  $x$  denote the length of time until the miner reaches safety.

We need to find

$$\begin{aligned} E(x) &= E E(x|Y) \\ &= E(x|Y=1) P(Y=1) + E(x|Y=2) P(Y=2) \\ &\quad + E(x|Y=3) P(Y=3). \end{aligned}$$

$$= \frac{1}{3} [2 + 3 + E(x) + 5 + E(x)]$$

$$\Rightarrow 3 E(x) = 10 + 2 E(x)$$

$$\Rightarrow E(x) = 10.$$

$E(x|Y=2) = 3 + E(x)$ , as the 2nd door leads to a tunnel that returns him to the mine after three hours of travel. ~~and~~ Once he returns to the mine ~~every~~ the problem is as before, and his expected additional time until safety is  $E(x)$ .

□.

Q13 Let  $N_k$  denote the number of trials to get  $k$  consecutive successes, and let  $E(N_k) = M_k$ .

We will find a recursion relation on  $M_k$ 's and then solve for it.

Note that  $N_k = N_{k-1} + A_{k-1,k}$ ,

where  $A_{k-1,k}$  denotes the number of additional trials needed to go from  $(k-1)$  consecutive successes to having  $k$  in a row. Taking expectation, we have

$$M_k = M_{k-1} + E(A_{k-1,k}).$$

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Let us denote  $X = \begin{cases} 1 & \text{if there is a success on } (N_{k-1}+1)\text{st trial} \\ 0 & \text{if there is a failure on } (N_{k-1}+1)\text{st trial} \end{cases}$

$$\text{Now } E(A_{k-1,k}) = E(A_{k-1,k} | X=1) P(X=1) + E(A_{k-1,k} | X=0) P(X=0)$$

$$E(A_{k-1,k} | X=1) = 1, \text{ as } X=1 \Rightarrow A_{k-1,k} = 1.$$

$E(A_{k-1,k} | X=0) = 1 + M_k$ , as if  $X=0$ , then at that point we are starting all over and the expected additional trials from then on would be  $E(N_k)$ .

$$P(X=1) = p \text{ and } P(X=0) = 1-p$$

$$\Rightarrow E(A_{k-1,k}) = p + (1 + M_k)(1-p) = 1 + (1-p)M_k.$$

$$\Rightarrow M_k = M_{k-1} + 1 + (1-p)M_k$$

$$\Rightarrow M_k = \frac{1}{p} + \frac{M_{k-1}}{p}.$$

Since  $N_1 \sim \text{Geo}(p)$ ,  $M_1 = E(N_1) = \frac{1}{p}$ .

$$M_2 = \frac{1}{p} + \frac{1}{p^2}$$

$$M_3 = \frac{1}{p} + \frac{1}{p^2} + \frac{1}{p^3}$$

$$\vdots$$

$$M_k = \frac{1}{p} + \frac{1}{p^2} + \frac{1}{p^3} + \dots + \frac{1}{p^k} = \frac{1-p^k}{p^k(1-p)} \quad \square.$$

[Q14] Let  $x$  denote the number of accidents that a randomly chosen policy-holder ~~has~~ will have in next year. ~~Let  $x$  denote~~  
~~be a Poisson RV with mean  $\mu$~~

Let  $Y$  denote the mean of ~~accidents~~ number of accidents that a randomly chosen policy-holder will have in next year.

Here  $X|Y=y \sim P(y)$  and  $Y$  has PDF  $g(\cdot)$ .

$$\text{Now } P(X=n) = \int_0^\infty P(X=n|Y=y) g(y) dy$$

$$= \int_0^\infty \frac{e^{-y} y^n}{n!} \times y e^{-y} dy$$

$$= \frac{1}{n!} \int_0^\infty y^{n+1} e^{-2y} dy$$

$$= \frac{1}{n!} \frac{\Gamma(n+2)}{2^{n+2}}$$

$$= \frac{n+1}{2^{n+2}}.$$

$\square.$

**[Q15]** Let  $x$  denote the number of persons visiting the studio today.

Let  $N$  and  $M$  denote the number of female and male, respectively.

$$\begin{aligned}
 & P(N=n, M=m) \\
 &= \sum_{k=0}^{\infty} P(N=n, M=m | X=k) P(X=k) \\
 &= P(N=n, M=m | X=n+m) P(X=n+m) \quad \text{--- (1)}
 \end{aligned}$$

As  $P(N=n, M=m | X=k) = 0$  for  $k \neq n+m$ .

Here  $X \sim P(\lambda)$  and  $N | X=k \sim \text{Bin}(k, p)$ .

Hence (1)  $\Rightarrow$

$$\begin{aligned}
 P(N=n, M=m) &= \binom{m+n}{n} p^n (1-p)^m \times \frac{e^{-\lambda} \lambda^{n+m}}{(n+m)!} \\
 &= \frac{e^{-\lambda} \lambda^{n+m} p^n (1-p)^m}{n! m!} \\
 &= \frac{e^{-(\lambda p)} (\lambda p)^n}{n!} \times \frac{e^{-(\lambda(1-p))} (\lambda(1-p))^m}{m!}
 \end{aligned}$$

Note:  $N$  and  $M$  are indep and  $N \sim P(\lambda p)$  and  $M \sim P(\lambda(1-p))$

□.