40.017 PROBABILITY & STATISTICS

Homework 2

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Section 2

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Y has the following pmf:

$$\mathbb{P}(Y = -1) = 0.4$$
, $\mathbb{P}(Y = 0) = 0.25$, $\mathbb{P}(Y = 1) = 0.35$

In order to simulate Y, we use a basic conditional structure in Python:

We have the individual probabilities $\mathbb{P}(E_1) = \frac{6}{36} = \frac{1}{6}$, $\mathbb{P}(E_2) = \frac{1}{6}$, and $\mathbb{P}(E_3) = \frac{1}{6}$, and we need to show that E_1 and E_2 are not conditionally independent given E_3 . Starting with the definition in Week 3 Class 2, we have the result:

$$\mathbb{P}(B_2|B_1 \cap A) = \mathbb{P}(B_2|A) \tag{1}$$

In the context of the problem, using Equation 1:

 $\mathbb{P}(E_2|E_1 \cap E_3) = \mathbb{P}(\text{'first dice shows 4'}|\text{'sum of two dice is 7' and 'second dice shows 3'})$ = 1

whereas

$$\mathbb{P}(E_2|E_3) = \frac{1}{6}$$

since they are not independent. As we can see,

$$\mathbb{P}(E_2|E_1 \cap E_3) \neq \mathbb{P}(E_2|E_3)$$

and thus we conclude that E_1 and E_2 are not conditionally independent given E_3 .

(a)

Since we have an unloaded dice, the probability of each dice roll $\mathbb{P}(X=i)$, $\forall i \in \{1,2,3,4\}$ is $\frac{1}{4}$. Let x be the result of the first dice roll, and y be the result of the second dice roll. There are three cases for this event, namely when x=y, x>y and otherwise. In the case where x=y, $\mathbb{P}(\max=x \text{ and } \min=x)=\mathbb{P}(\{(x,y)\})=(1/4)^2$. However, if x>y, we have $\mathbb{P}(\max=x \text{ and } \min=y)=\mathbb{P}(\{(x,y)\},\{(y,x)\})=2\times(1/4)^2$. Otherwise, it is not possible. The piecewise function for the joint pmf is thus:

$$f(x,y) = \begin{cases} \frac{1}{16}, & \forall x = y\\ \frac{1}{8}, & \forall x > y\\ 0, & \text{otherwise} \end{cases}$$

We now construct the joint pmf table of X and Y in table 1 below.

у	1	2	3	4	f_X
1	$\frac{1}{16}$	0	0	0	$\frac{1}{16}$
2	$\frac{1}{8}$	$\frac{1}{16}$	0	0	$\frac{3}{16}$
3	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{16}$	0	$ \begin{array}{c c} \hline & \frac{1}{16} \\ \hline & \frac{3}{16} \\ \hline & \frac{5}{16} \\ \hline & \frac{7}{16} $
4	$\frac{1}{8}$	$\frac{\frac{1}{8}}{\frac{1}{8}}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{7}{16}$
f_Y	$\frac{7}{16}$	$\frac{5}{16}$	$\frac{3}{16}$	$\frac{1}{16}$	1

Table 1: Joint pmf of X and Y

(b)

The conditional pmf of X given that Y = 2 is:

$$f_{X|Y}(x|2) = \mathbb{P}((X=x)(Y=2)) = \frac{f(x,2)}{f_Y(2)}$$

where $f_Y(2) = \frac{5}{16}$ from table 1. f(x,2) is given by the 2nd column of table 1, thus we have the conditional pmf table:

X	1	2	3	4
$f_{X Y}(x 2)$	0	$\frac{1}{5}$	$\frac{2}{5}$	$\frac{2}{5}$

Table 2: Conditional pmf of X|Y=2

(a)

Let $Z \sim \mathcal{N}(0,1)$, and $X = Z^2$. From MU Week 8,

$$F_X(x) = \mathbb{P}(X \le x)$$

$$= \mathbb{P}(Z^2 \le x)$$

$$= \mathbb{P}(-\sqrt{x} \le Z \le \sqrt{x})$$

$$= \Phi(\sqrt{x}) - \Phi(-\sqrt{x})$$

$$f_X(x) = \frac{1}{2\sqrt{x}}\phi(\sqrt{x}) + \frac{1}{\sqrt{x}}\phi(-\sqrt{x}) \quad \text{(differentiating both sides)}$$

$$= \frac{1}{2\sqrt{x}} \left(\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x}{2}\right)\right) + \frac{1}{2\sqrt{x}} \left(\frac{1}{2\sqrt{\pi}} \exp\left(-\frac{x}{2}\right)\right)$$

$$= \frac{1}{\sqrt{2\pi x}} \exp\left(-\frac{x}{2}\right)$$

Thus the pdf of X is given by

$$f_X(x) = \begin{cases} \frac{1}{\sqrt{2\pi x}} \exp\left(-\frac{x}{2}\right), & x > 0\\ 0, & x \le 0 \end{cases}$$
 (2)

which is interestingly also the pdf of a χ^2 distribution with 1 degree of freedom (defined as 0 at x = 0).

The standard Gamma distribution is given by

$$f(x) = \begin{cases} \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha - 1} e^{-\lambda x}, & x \ge 0\\ 0, & x < 0 \end{cases}$$
 (3)

where $X \sim \text{gamma}(\alpha, \lambda)$. To show that X is a gamma random variable, we must transform equation 2 into the form in equation 3. By inspection, it is immediately obvious that $\lambda = \frac{1}{2}$. From equation 2,

$$f_X(x) = \frac{1}{\sqrt{2\pi x}} \exp(-\frac{x}{2})$$
$$= \frac{\left(\frac{1}{2}\right)^{\frac{1}{2}}}{\sqrt{\pi}} x^{\frac{1}{2} - 1} e^{-\frac{1}{2}x}$$

where $\alpha = \frac{1}{2}$ by inspection. To confirm that the value of $\Gamma(\frac{1}{2}) = \sqrt{\pi}$, we can find the integral:

$$\Gamma\left(\frac{1}{2}\right) = \int_0^\infty x^{-\frac{1}{2}} e^{-x} dx$$

$$= \int_0^\infty \left(\frac{t^2}{2}\right)^{-\frac{1}{2}} e^{-\frac{t^2}{2}} \frac{dx}{dt} dt$$

$$= \int_0^\infty \frac{\sqrt{2}}{t} e^{-\frac{t^2}{2}} t dt$$

$$= 2\sqrt{\pi} \int_0^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

$$= \frac{1}{2} (2\sqrt{\pi})$$

$$= \sqrt{\pi}$$

With this, we can conclude that $X \sim \operatorname{gamma}\left(\frac{1}{2}, \frac{1}{2}\right)$.

(b)

If we have two independent standard normal r.v.s Z_1 and Z_2 , the sum of squares is given by $Z_1^2 + Z_2^2 = X_1 + X_2$. In Week 3 Lecture 1, we proved the MGF of the Gamma distribution:

$$M_X(t) = \mathbb{E}(e^{tX})$$

$$= \dots$$

$$= \frac{\lambda^{\alpha}(\lambda - t)^{-\alpha}}{\Gamma(a)} \underbrace{\int_0^{\infty} y^{\alpha - 1} e^{-y} \, dy}_{\Gamma(\alpha)}$$

$$= \lambda^{\alpha}(\lambda - t)^{-\alpha}$$

Using the property for the MGF of the sum of two independent random variables (where $X \sim \text{gamma}(1/2, 1/2)$),

$$\begin{split} M_{X_1+X_2}(t) &= M_{X_1}(t) M_{X_2}(t) \\ &= \left(\frac{1}{2}\right)^{1/2} \left(\frac{1}{2} - t\right)^{-1/2} \left(\frac{1}{2}\right)^{1/2} \left(\frac{1}{2} - t\right)^{-1/2} \\ &= \frac{1/2}{1/2 - t} \end{split}$$

The MGF of an exponential random variable is given by

$$M_X(t) = \frac{\lambda}{\lambda - t}$$

By inspection, we can see that this is actually the MGF of an exponential r.v. with $\lambda = \frac{1}{2}$, thus $X_1 + X_2 \sim \text{exponential}(\frac{1}{2})$.

Let X be the number of tosses required to reach our goal of 3 consecutive H's appearing for the first time. We start from the hint, where there are 4 cases for any sequence to start: T, HT, HHT, HHH. Let Y = i for each of the 4 cases accordingly, and $\mathbb{P}(H) = \frac{1}{2}$ (since we have a fair coin), so that:

$$\mathbb{P}(Y=1) = \mathbb{P}(T) = \frac{1}{2}$$

$$\mathbb{P}(Y=2) = \mathbb{P}(HT) = \left(\frac{1}{2}\right)^2$$

$$\mathbb{P}(Y=3) = \mathbb{P}(HHT) = \left(\frac{1}{2}\right)^3$$

$$\mathbb{P}(Y=4) = \mathbb{P}(HHH) = \left(\frac{1}{2}\right)^3$$

Applying the law of total expectation, we have:

$$\mathbb{E}(X) = \sum_{\text{all } y} \mathbb{E}(X|Y=y)\mathbb{P}(Y=y)$$

$$= \left(\frac{1}{2}\right) \mathbb{E}(X|Y=1) + \left(\frac{1}{2}\right)^2 \mathbb{E}(X|Y=2) + \left(\frac{1}{2}\right)^3 \mathbb{E}(X|Y=3) + \left(\frac{1}{2}\right)^3 \mathbb{E}(X|Y=4)$$

To evaluate the conditional expectations, notice that each case involves recursion, where for Y=1, tossing a T at first yields no contribution to our goal of 3 consecutive H's, so we are now back to the beginning with $1+\mathbb{E}(X)$. For $\mathbb{E}(X|Y=2)$ and $\mathbb{E}(X|Y=3)$, a similar reasoning holds and we are back to the beginning after 2 and 3 'useless' tosses. For $\mathbb{E}(X|Y=4)$ however, we have the case for the least number of tosses required to reach our goal, which is 3 tosses. Thus, the expected number of tosses $\mathbb{E}(X|Y=4)$ is trivially 3. We now have:

$$\begin{split} \mathbb{E}(X) &= \left(\frac{1}{2}\right)\left[1 + \mathbb{E}(X)\right] + \left(\frac{1}{2}\right)^2\left[2 + \mathbb{E}(X)\right] + \left(\frac{1}{2}\right)^3\left[3 + \mathbb{E}(X)\right] + \left(\frac{1}{2}\right)^3(3) \\ \mathbb{E}(X) &= \frac{7}{4} + \frac{7}{8}\mathbb{E}(X) \\ \mathbb{E}(X) &= \boxed{14} \end{split}$$

(a)

Using the hint, let X be the amount paid, while Y be the number of claims filed. Since there can only be at most 1 claim filed, Y = 1 or Y = 0, and $\mathbb{P}(Y = 1) = 0.2$, $\mathbb{P}(Y = 0) = 1 - 0.2 = 0.8$. If a claim is filed, then the mean amount paid is \$1000: $\mathbb{E}(X|Y = 1) = 1000$, and if no claim is filed, then nothing is paid: $\mathbb{E}(X|Y = 0) = 0$. Using the alternate formulation of the law of total expectation,

$$\begin{split} \mathbb{E}(X) &= \mathbb{E}(\mathbb{E}(X|Y)) \\ &= \sum_{i=0}^{1} \mathbb{E}(X|Y=i) \mathbb{P}(Y=i) \\ &= \mathbb{E}(X|Y=0) \mathbb{P}(Y=0) + \mathbb{E}(X|Y=1) \mathbb{P}(Y=1) \\ &= 0 \times 0.8 + 1000 \times 0.2 \\ &= \boxed{200} \end{split}$$

(b)

Using the hint, we have the pmf tables for $\mathbb{E}(X|Y)$ and $\mathrm{Var}(X|Y)$ below. Since X is exponentially distributed when Y=1, for an exponential distribution with parameter λ , the mean is given by $\frac{1}{\lambda}$ and the variance is $\left(\frac{1}{\lambda}\right)^2$. Thus, $\mathrm{Var}(X|Y)=\left[\mathbb{E}(X|Y)\right]^2=1000^2$.

Y	0	1
$\mathbb{E}(X Y=i)$	0	1000
$\mathbb{P}(Y=i)$	0.8	0.2

Table 3: pmf table of $\mathbb{E}(X|Y)$

Y	0	1
Var(X Y=i)	0	1000000
P(Y=i)	0.8	0.2

Table 4: pmf table of Var(X|Y)

Now to find Var(X), we use the law of total variance:

$$Var(X) = \mathbb{E}(Var(X|Y)) + Var(\mathbb{E}(X|Y))$$

$$= \mathbb{E}(Var(X|Y)) + \mathbb{E}(\mathbb{E}(X|Y)^2) - (\mathbb{E}(X))^2$$

$$= \mathbb{E}(Var(X|Y)) + \mathbb{E}(Var(X|Y)) - (\mathbb{E}(X))^2$$

$$= 2 \cdot \mathbb{E}(Var(X|Y)) - (\mathbb{E}(X))^2$$

$$= 2 (0 \times 0.8 + 1000000 \times 0.2) - 200^2$$

$$= \boxed{360000}$$

(a)

We first compute the \mathbb{E} and Var of 2X + 3Y (linear combination of X and Y):

$$\mathbb{E}(2X + 3Y) = 2\mathbb{E}(X) + 3\mathbb{E}(Y)$$

$$= 2\mu_X + 3\mu_Y$$

$$= 2(2) + 3(-1)$$

$$= 1$$

$$\operatorname{Var}(2X + 3Y) = \operatorname{Var}(2X) + \operatorname{Var}(3Y) + 2 \cdot \operatorname{Cov}(2X, 3Y)$$

$$= 4 \cdot \operatorname{Var}(X) + 9 \cdot \operatorname{Var}(Y) + 2 \cdot (6) \cdot \operatorname{Cov}(X, Y)$$

$$= 4 + 36 + 12 \left(\rho \sqrt{\sigma_X^2 \sigma_Y^2}\right)$$

$$= 40 + 12(-\frac{1}{2}) \left(\sqrt{1(4)}\right)$$

$$= 28$$

Thus, we have the normally distributed 2X + 3Y: $2X + 3Y \sim \mathcal{N}(1, 28)$. To obtain $\mathbb{P}(2X + 3Y \geq 6)$, we use a standard normal Z:

$$\mathbb{P}(2X + 3Y \ge 6) = \mathbb{P}\left(Z \ge \frac{6 - 1}{\sqrt{28}}\right)$$
$$= 0.17235$$
$$\approx 0.172$$

(b)

We first find the conditional pdf of X|Y=2 using the joint pdf of the bivariate normal:

$$f_{X|Y}(X|Y=2) = \frac{f(X,Y=2)}{f_Y(Y=2)}$$

$$= c_1 f(X,2)$$

$$= c_2 \exp\left[-\frac{1}{2(1-(\frac{1}{2})^2)} \left(\left(\frac{x-2}{1}\right)^2 - 2\left(-\frac{1}{2}\right) \frac{x-2}{1} \frac{2-(-1)}{2} + \left(\frac{2-(-1)}{2}\right)^2\right)\right]$$

$$= c_2 \exp\left[-\frac{2}{3} \left((x-2)^2 + \frac{3}{2}(x-2) + \frac{9}{4}\right)\right]$$

$$= c_2 \exp\left[-\frac{2}{3} \left(x^2 - 4x + 4 + \frac{3}{2}x - 3 + \frac{9}{4}\right)\right]$$

$$= c_2 \exp\left[-\frac{2}{3} \left(x^2 - \frac{5}{2}x + \frac{13}{4}\right)\right]$$

$$= c_2 \exp\left[-\frac{2}{3} \left(\left(x - \frac{5}{4}\right)^2 + \frac{13}{4} - \left(\frac{5}{4}\right)^2\right)\right] \qquad \text{(Completing the square)}$$

$$= c_3 \exp\left[-\frac{2}{3}(x - \frac{5}{4})^2\right] \qquad \text{(Constant absorbed to } c_3)$$

$$= c_3 \exp\left[-\frac{(x - \frac{5}{4})^2}{2(\frac{3}{4})}\right]$$

Comparing the result with the pdf of a univariate normal, we can see that

$$\mathbb{E}(X|Y=2) = \frac{5}{4}, \quad \text{Var}(X|Y=2) = \frac{3}{4}$$

where $c_3 \in \mathbb{R}$. Thus, to compute $\mathbb{P}(X < 1|Y = 2)$, we use a standard normal Z:

$$\mathbb{P}(X < 1 | Y = 2) = \mathbb{P}\left(Z < \frac{1 - 5/4}{\sqrt{3/4}}\right)$$
= 0.386414
$$\approx 0.386$$