# 40.017 PROBABILITY & STATISTICS

## Homework 2

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Section 2

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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,  $x \neq y$  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  $x \neq y$ , we have  $\mathbb{P}(\max = x \text{ and } \min = x) = \mathbb{P}(\{(x, y)\}, \{(y, x)\}) = 2 \times (1/4)^2$ . Otherwise, it is not possible. The piecewise function for the joint pmf is thus:

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

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

y x	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	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{7}{16}$
$f_Y$	$\frac{7}{16}$	$\frac{5}{16}$	$\frac{3}{16}$	$\frac{1}{16}$	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

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

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:

$$\begin{split} \mathbb{P}(Y=1) &= \mathbb{P}(\mathtt{T}) = \frac{1}{2} \\ \mathbb{P}(Y=2) &= \mathbb{P}(\mathtt{HT}) = \left(\frac{1}{2}\right)^2 \\ \mathbb{P}(Y=3) &= \mathbb{P}(\mathtt{HHT}) = \left(\frac{1}{3}\right)^3 \\ \mathbb{P}(Y=4) &= \mathbb{P}(\mathtt{HHH}) = \left(\frac{1}{4}\right)^4 \end{split}$$

Applying the law of total expectation, we have:

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

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

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:

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

$$\mathbb{E}(X) = \frac{25}{16} + \frac{7}{8}\mathbb{E}(X)$$

$$\mathbb{E}(X) = \boxed{\frac{25}{2}}$$

#### (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  $\lambda$ , 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:

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

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

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

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

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

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

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

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

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

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
$$\approx 0.386$$