

## Homework # 2 Answer Key

*James*

1)

$x$	0.5	1	1.75	3
$P(X=x)$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{2}$

These values can be calculated by measuring the height of the "jumps" in the cdf  $F_X(x)$ .

a) We can use the inverse cdf method as follows:

$$\begin{aligned} F_Y(y) &= P(Y \leq y) = P(e^X \leq y) \\ &= P(X \leq \log(y)) = \int_0^{\log(y)} 1 \, dx \\ &= \log(y), \text{ for } 1 < y < e \end{aligned}$$

It follows that the pdf of  $Y$  is

$$\begin{aligned} f_Y(y) &= \frac{d}{dy} F_Y(y) = \frac{d}{dy} (\log(y)) \\ &= \frac{1}{y}, \quad 1 < y < e \end{aligned}$$

$$\text{So, } f_Y(y) = \frac{1}{y} \mathbb{I}\{1 < y < e\}.$$

3) a) We need to find  $C$  so that the density integrates to 1. So, we solve

$$\int_0^2 C(4x - 2x^2) dx = 1$$

$$\Rightarrow C \left[ 2x^2 \Big|_0^2 - \frac{2}{3} x^3 \Big|_0^2 \right] = 1$$

$$\Rightarrow 8 - \frac{16}{3} = \frac{1}{C} \Rightarrow \frac{8}{3} = \frac{1}{C} \Rightarrow C = \frac{3}{8}$$

$$b) P(X > 1) = \int_1^2 \frac{3}{8} (4x - 2x^2) dx$$

$$= \frac{3}{8} \left[ 2x^2 \Big|_1^2 - \frac{2}{3} x^3 \Big|_1^2 \right]$$

$$= \frac{3}{8} \left[ 8 - 2 - \frac{16}{3} + \frac{2}{3} \right] = \frac{3}{8} \left[ 6 - \frac{14}{3} \right]$$

$$= \frac{3}{8} \left( \frac{4}{3} \right) = \frac{1}{2}$$

4) We need to think about two possibilities for each fixed  $p$ : either  $U < p$  or  $U > p$ . When  $U > p$ , the  $p$  will be on the left stick, which will have ~~densit~~ length  $U$ . However, when  $U < p$ ,  $p$  will be on the right stick, which will have length  $1-U$ . So, let  $L_p(U)$  = length of the stick on which  $p$  lies. We have:

$$L_p(u) = \begin{cases} U, & U > p \\ 1-U, & U < p \end{cases}$$

4) continued...

So, we just need  $E[L_P(U)]$  :

$$\begin{aligned} E[L_P(U)] &= \int_0^1 L_P(u) du \\ &= \int_0^p (1-u) du + \int_p^1 u du \\ &= p - \frac{1}{2} p^2 + \frac{1}{2} - \frac{1}{2} p^2 \\ &= \frac{1}{2} + p - p^2 = \boxed{\frac{1}{2} + p(1-p)} \end{aligned}$$

5) First we calculate the marginal pdfs  $f_X(x)$  &  $f_Y(y)$  :

$$\begin{aligned} f_X(x) &= \int_0^\infty 2e^{-x} e^{-2y} dy = -e^{-x} e^{-2y} \Big|_{y=0}^{y=\infty} \\ &= e^{-x}, \quad x \in (0, \infty) \\ f_Y(y) &= \int_0^\infty 2e^{-x} e^{-2y} dx = -2e^{-x} e^{-2y} \Big|_{x=0}^{x=\infty} \\ &= 2e^{-2y}, \quad y \in (0, \infty) \end{aligned}$$

Then, we use the law of total probability

$$\begin{aligned} P(X < Y) &= \int_0^\infty P(X < y | Y=y) f_Y(y) dy \\ &= 2 \int_0^\infty \left( \int_0^y f_X(x) dx \right) e^{-2y} dy \end{aligned}$$

5) continued...

$$\begin{aligned} &= 2 \int_0^{\infty} \left( \int_0^y e^{-x} dx \right) e^{-2y} dy \\ &= 2 \int_0^{\infty} \left( -e^{-x} \Big|_{x=0}^{x=y} \right) e^{-2y} dy \\ &= 2 \int_0^{\infty} (1 - e^{-y}) e^{-2y} dy \\ &= 2 \int_0^{\infty} e^{-2y} - e^{-3y} dy \\ &= 2 \left[ -\frac{1}{2} e^{-2y} \Big|_0^{\infty} + \frac{1}{3} e^{-3y} \Big|_0^{\infty} \right] \\ &= 2 \left[ \frac{1}{2} - \frac{1}{3} \right] = 2 \left( \frac{1}{6} \right) = \frac{1}{3} \end{aligned}$$

$$\text{So, } P(X < Y) = \frac{1}{3}$$

6) we can use the convolution formula for each of these calculations.

First, note that  $f_X(x) = 1$ ,  $x \in (0,1)$

$$F_X(x) = x$$

$$f_Y(y) = e^{-y}, \quad y > 0$$

$$F_Y(y) = 1 - e^{-y}$$

$$\begin{aligned} \text{Thus, } F_Z(a) &= F_{X+Y}(a) = \int_y f_X(a-y) f_Y(y) dy \\ &= \int_y (a-y) e^{-y} dy \quad (*) \end{aligned}$$

We have to be smart about the limits of integration here. Note that



6) continued...

$$x + y = a \Rightarrow y = a - x$$

Since  $x \in (0, 1)$ , it follows that  $y$  must be between  $a-1$  &  $a$ . Moreover, since  $y > 0$  it follows that  $y \in (\max\{0, a-1\}, a)$

So, we can calculate (\*) as :

$$= \int_{\max\{0, a-1\}}^a (a-y) e^{-y} dy = a e^{-\max\{0, a-1\}} - a e^{-a} - \int_{\max\{0, a-1\}}^a y e^{-y} dy.$$

Then, the density can be calculated by taking derivatives. (Not all work is shown here...)

$$\begin{aligned} 7) \quad P(X=x | X+Y=n) &= \frac{P(X=x, X+Y=n)}{P(X+Y=n)} \\ &= \frac{P(X=x, Y=n-x)}{P(X+Y=n)} \\ &= \frac{P(X=x) P(Y=n-x)}{P(X+Y=n)} \quad \text{by independence} \end{aligned}$$

Now, it is easy to show that if  $X \sim \text{Po}(\lambda_1)$  and  $Y \sim \text{Po}(\lambda_2)$  and they are independent,

7) continued...

then  $X + Y \sim \text{Po}(\lambda_1 + \lambda_2)$ .

It follows that

$$P(X=x | X+Y=n) = \frac{\lambda_1^x e^{-\lambda_1} \lambda_2^{n-x} e^{-\lambda_2}}{(\lambda_1 + \lambda_2)^n e^{-(\lambda_1 + \lambda_2)}} \frac{n!}{x!(n-x)!}$$

$$= \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^x \left(\frac{\lambda_2}{\lambda_1 + \lambda_2}\right)^{n-x} \binom{n}{x}$$
$$= \left(\frac{\lambda_2}{\lambda_1 + \lambda_2}\right)^{n-x} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^x \binom{n}{x}$$

which is the form of a  $\text{Bin}(n, \frac{\lambda_1}{\lambda_1 + \lambda_2})$

random variable! Hence,  $X | X+Y=n \sim \text{Bin}(n, \frac{\lambda_1}{\lambda_1 + \lambda_2})$

8) Note: One can calculate these probabilities by investigating each of 36 total outcomes. I don't do this here, but the calculation (though long) is straightforward.

Notes on computation:

①  $X+Y$  is just a  $\text{Gamma}(1, 2)$  random variable. This can be found using the convolution formula.

But the point of this exercise is to show that simulation can be used to obtain the distribution of a convolution and can be done pretty quickly.

② Strategy 2 turns out to be more efficient.