

## DS203 Assignment 2

### Exercise 1

$$\text{PDF for } X \quad f_X(x) = \begin{cases} \lambda_1 e^{-\lambda_1 x}, & x \geq 0 \\ 0, & x < 0 \end{cases}$$

$$\text{PDF for } Y \quad f_Y(y) = \begin{cases} \lambda_2 e^{-\lambda_2 y}, & y \geq 0 \\ 0, & y < 0 \end{cases}$$

(a) let function  $g = \min(X, Y)$ .

$$\text{CDF for } g, \quad F_g(z) = P(\min(X, Y) \leq z)$$

$$P(\min(X, Y) \leq z) = 0 \quad \forall z < 0 \quad \text{as}$$

X and Y take positive values.

$$P(\min(X, Y) \leq z) = \sum_{j \in (0, \frac{z}{2})} P(X=j, j \leq Y < \infty) + \sum_{j \in (0, \frac{z}{2})} P(j \leq X < \infty, Y=j)$$

if  $z > 0$

$$\begin{aligned}
 &= \int_0^z \int_{-\infty}^{\infty} \lambda_1 \lambda_2 e^{-\lambda_1 j} e^{-\lambda_2 y} dy dj + \int_0^z \int_{-\infty}^{\infty} \lambda_1 \lambda_2 e^{-\lambda_1 x} e^{-\lambda_2 j} dx dj \\
 &= \lambda_1 \int_0^z e^{-(\lambda_1 + \lambda_2)j} dj + \lambda_2 \int_0^z e^{-(\lambda_1 + \lambda_2)j} dj \\
 &= 1 - e^{-(\lambda_1 + \lambda_2)z} \quad \text{if } z \geq 0
 \end{aligned}$$

(b)

$$g = \max(x, y)$$

CDF for  $g$ ,  $F_g(z) = P(\max(x, y) \leq z)$

$$P(\max(x, y) \leq z) = 0 \quad \forall z < 0 \quad \text{as}$$

$X$  and  $Y$  take positive values.

$$\begin{aligned}
 P(\max(x, y) \leq z) &= \sum_{\substack{j \\ \forall j \in (0, z)}} P(x=j, 0 \leq y \leq j) \\
 &\quad + \\
 &\quad \sum_{\substack{j \\ \forall j \in (0, z)}} P(0 \leq x \leq j, y=j)
 \end{aligned}$$

$$\begin{aligned}
 &= \int_0^z \int_{-\infty}^{\infty} \lambda_1 \lambda_2 e^{-\lambda_1 j} e^{-\lambda_2 y} dy dj + \int_0^z \int_{-\infty}^j \lambda_1 \lambda_2 e^{-\lambda_1 x} e^{-\lambda_2 j} dx dj
 \end{aligned}$$

$$\begin{aligned}
 &= \lambda_1 \int_0^z e^{-\lambda_1 j} (1 - e^{-\lambda_2 j}) dj + \lambda_2 \int_0^z e^{-\lambda_2 j} (1 - e^{-\lambda_1 j}) dj \\
 &= (1 - e^{-\lambda_1 z}) + (1 - e^{-\lambda_2 z}) - (1 - e^{-(\lambda_1 + \lambda_2) z}) \\
 &= 1 - e^{-\lambda_1 z} - e^{-\lambda_2 z} + e^{-(\lambda_1 + \lambda_2) z} \quad \text{if } z \geq 0
 \end{aligned}$$

Exercise 2 3 white. 6 red. 5 blue.

Total 6 selections

$X$  = number of white balls selected

$Y$  = number of blue balls selected

$$\begin{aligned}
 E[X | Y=3] &= 1 \cdot P(X=1 | Y=3) + 2 \cdot P(X=2 | Y=3) \\
 &\quad + 3 \cdot P(X=3 | Y=3) \\
 &= \frac{P(X=1 \cap Y=3)}{P(Y=3)} + 2 \cdot \frac{P(X=2 \cap Y=3)}{P(Y=3)} \\
 &\quad + 3 \cdot \frac{P(X=3 \cap Y=3)}{P(Y=3)}
 \end{aligned}$$

In the 6 selections available, if we have to pick  $m$  white and  $n$  blue balls, then the total number of ways are:

Selecting  $n$  blue

from remaining

selections

No	= 1
Date	

$$= {}^6 C_m \left(\frac{3}{14}\right)^m \cdot {}^{(6-m)} C_n \left(\frac{5}{14}\right)^n \cdot {}^{6-m-n} C_{6-m-n} \left(\frac{6}{14}\right)^{6-m-n}$$

- ①

ways of  
having  $m$  white  
selections out of  
6

Probability of  
selecting  $m$   
white balls

Here, I have assumed  $m+n \leq 6$  to  
demonstrate this case.

Using ①,

$$\begin{aligned} E[X | Y=3] &= {}^6 C_1 \left(\frac{3}{14}\right) \cdot {}^5 C_3 \left(\frac{5}{14}\right)^3 \cdot \left(\frac{6}{14}\right)^2 \\ &\quad + 2 \cdot {}^6 C_2 \left(\frac{3}{14}\right)^2 \cdot {}^4 C_3 \left(\frac{5}{14}\right)^3 \cdot \left(\frac{6}{14}\right) \\ &\quad + 3 \cdot {}^6 C_3 \left(\frac{3}{14}\right)^3 \cdot {}^3 C_3 \left(\frac{5}{14}\right)^3 \\ &\quad - \frac{{}^6 C_3 \left(\frac{5}{14}\right)^3 \cdot {}^3 C_3 \left(\frac{9}{14}\right)^3}{-} \end{aligned}$$

$$= \frac{{}^6 C_1 \cdot {}^5 C_3 \cdot 3 \cdot 6^2}{-} + \frac{{}^6 C_2 \cdot {}^4 C_3 \cdot 3^2 \cdot 6 \cdot 2}{-}$$

$$+ \frac{{}^6 C_3 \cdot 3^3 \cdot 3}{-}$$

$$\frac{{}^6 C_3 \cdot 9^3}{-}$$

$$= \frac{6480 + 6480 + 1620}{14580} = \underline{\underline{1}}$$

### Exercise 3

$X_1 \sim \text{Bin}(n_1, p), X_2 \sim \text{Bin}(n_2, p)$

$$\begin{aligned}
 P(X_1 = i \mid X_1 + X_2 = m) &= \frac{P(X_1 = i \cap X_1 + X_2 = m)}{P(X_1 + X_2 = m)} \\
 &= \frac{n_1^i c_i p^i (1-p)^{n_1-i} n_2^{m-i} c_{m-i} p^{m-i} (1-p)^{n_2-m+i}}{\sum_{j=0}^m n_1^j c_j p^j (1-p)^{n_1-j} n_2^{m-j} c_{m-j} p^{m-j} (1-p)^{n_2-m+j}} \\
 &= \frac{n_1^i c_i \cdot n_2^{m-i} p^m \cdot (1-p)^{n_1+n_2-m}}{p^m (1-p)^{n_1+n_2-m} \sum_{j=0}^m n_1^j c_j \cdot n_2^{m-j}} \\
 &= \frac{n_1^i c_i \cdot n_2^{m-i}}{\sum_{j=0}^m n_1^j c_j \cdot n_2^{m-j}}
 \end{aligned}$$

The denominator is the coefficient of  $x^m$  in  
 $(1+x)^{n_1} \cdot (1+x)^{n_2} = (1+x)^{n_1+n_2} = \binom{n_1+n_2}{m}$ .

$$P(X_1 = i \mid X_1 + X_2 = m) = \frac{n_1^i c_i \cdot n_2^{m-i}}{\binom{n_1+n_2}{m}}$$

This formula is however, only valid for  $n_1 + n_2 \geq m$ .

If  $n_1 + n_2 < m$ , the denominator does not make any sense, hence, the PMF of  $X_1$  is identically 0.

If  $n_1 + n_2 = m$ , then the only possible values are taken by  $X_1, X_2$  s.t  $X_1 + X_2 = m$  is  $X_1 = n_1, X_2 = n_2$ .

Hence, the PMF of  $X_1$  given  $X_1 + X_2 = m$  will be :

$$P(X_1 = n_1 \mid X_1 + X_2 = m) = \frac{\binom{n_1}{n_1} \cdot \binom{n_2}{n_2}}{\binom{n_1+n_2}{m}}$$

In all other cases with  $n_1 + n_2 > m$ , the formula derived earlier is valid and gives the PMF of  $X_1$  given  $X_1 + X_2 = m$ .

Exercise 84

For  $X$  and  $Y$  to be uncorrelated,  
it is sufficient that  $\text{Cov}(X, Y) = 0$

$$\begin{aligned}\text{Cov}(X, Y) &= E[XY] - E[X]E[Y] = 0 \\ \Rightarrow E[XY] &= E[X]E[Y]\end{aligned}$$

consider  $X \sim U(-1, 1)$   
 $Y = X^2$

Clearly,  $X$  and  $Y$  are dependent

But,  $E[X] = 0$   
 and since  $XY = X^3 \sim U(-1, 1)$ ,  
 $E[XY] = 0$

Hence,  $\text{Cov}(X, Y) = 0$

So,  $X$  and  $Y$  are uncorrelated but  
not independent

### Exercise 5

$$X \sim \text{Poi}(\lambda)$$

$$\lambda \sim \text{Exp}(1)$$

$$P(X=i, \lambda) = \frac{e^{-\lambda} \lambda^i}{i!}, \quad i \geq 0.$$

PDF of  $\lambda$ ,

$$f_\lambda(x) = \begin{cases} e^{-x}, & x \geq 0 \\ 0, & x < 0. \end{cases}$$

$$P\{X=n\} = \int_{-\infty}^{\infty} P(X=n, \lambda=x) \cdot f_\lambda(x) dx$$

$$= \int_0^{\infty} \frac{e^{-x} x^n}{n!} e^{-x} dx$$

$$= \frac{1}{n!} \int_0^{\infty} x^n e^{-2x} dx$$

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

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

Exercise 6

$$f_{xy}(x, y) = \begin{cases} c(1+xy), & 2 \leq x \leq 3, 1 \leq y \leq 2 \\ 0, & \text{otherwise} \end{cases}$$

$$1. \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{xy}(x, y) = 1$$

$$\Rightarrow \int_2^3 \int_1^3 c(1+xy) dx dy = 1.$$

$$c \left[ 1 \cdot 1 + \frac{3}{2} \cdot \frac{5}{2} \right] = 1.$$

$$c = \underline{\underline{\frac{4}{19}}}$$

$$f_x = \int_{-\infty}^{\infty} f_{xy}(x, y) dy$$

$$= \int_2^3 \frac{4}{19} (1+xy) dy = \frac{4}{19} \left( 1 + \frac{3x}{2} \right)$$

$$f_y = \int_{-\infty}^{\infty} f_{x,y}(x,y) dx$$

$$= \int_2^3 \frac{4}{19} (1+xy) dx = \frac{4}{19} \left( 1 + \frac{5}{2} y \right)$$

### Exercise 7

Let  $X$  = number of accidents random variable.

$$X \sim \text{Poi}(\lambda)$$

where  $\lambda_1$  is a random variable s.t.

P.D.F of  $\lambda_1$  is given by  $f_{\lambda_1}$

$$g(\lambda) = \lambda e^{-\lambda}, \lambda \geq 0$$

$$P(X=n, \lambda_1=p) = \frac{e^{-p} p^n}{n!}, n \geq 0$$

$$P(X=n) = \int_{-\infty}^{\infty} P(X=n, \lambda_1=p) \cdot g(p) dp$$

$$= \int_0^{\infty} \frac{e^{-p} p^n}{n!} p e^{-p} dp$$

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

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

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

### Exercise 8

let  $X$  be number of people visiting for yoga each day.

$$X \sim \text{Poi}(\lambda)$$

$$P(X = n) = \frac{e^{-\lambda} \lambda^n}{n!}, \quad n \geq 0.$$

If  $n$  women &  $m$  men visit today,  
total number visiting =  $(m+n)$ .

$$P(X = (m+n)) = \frac{e^{-\lambda} \lambda^{m+n}}{(m+n)!}$$

$$\begin{aligned}
 P(n \text{ women, } m \text{ men}) &= P(X = (m+n)) \cdot \binom{m+n}{n} p^n (1-p)^m \\
 &= \frac{e^{-\lambda} \lambda^{m+n}}{(m+n)!} \frac{(m+n)!}{m! n!} p^n (1-p)^m \\
 &= \frac{e^{-\lambda} \lambda^{m+n}}{m! n!} p^n (1-p)^m
 \end{aligned}$$

### Exercise 9

$$\begin{aligned}
 (a) \text{ cov}(ax_1 + b, cx_2 + b) &= E[(ax_1 + b)(cx_2 + b)] - E[ax_1 + b] E[cx_2 + b] \\
 &= E[acx_1x_2 + abx_1 + bcx_2 + b^2] - (aE[x_1] + bE[1])(cE[x_2] + bE[1]) \\
 &= acE[x_1x_2] + abE[x_1] + bcE[x_2] + b^2 \\
 &\quad - acE[x_1]E[x_2] - abE[x_1] - bcE[x_2] - b^2 \\
 &\quad (\because E[1] = 1) \\
 &= ac(E[x_1x_2] - E[x_1]E[x_2]) \\
 &= ac \text{ cov}(x_1, x_2)
 \end{aligned}$$

$$(b) \text{Cov}(x_1 + x_2, x_3)$$

$$\begin{aligned}
 &= E[(x_1 + x_2)x_3] - E[x_1 + x_2]E[x_3] \\
 &= E[x_1x_3] + E[x_2x_3] - (E[x_1] + E[x_2])E[x_3] \\
 &= (E[x_1x_3] - E[x_1]E[x_3]) + \\
 &\quad (E[x_2x_3] - E[x_2]E[x_3]) \\
 &= \text{Cov}(x_1, x_3) + \text{Cov}(x_2, x_3).
 \end{aligned}$$

### Exercise 10

(a) number of samples  $n = 100$ .

Estimated mean  $\hat{p} = 0.45$

let the true mean after taking large number of samples be  $p$ .

Let  $|p - \hat{p}| > \varepsilon$  for some  $\varepsilon > 0$ .

Then, we know

$$P(|p - \hat{p}| > \varepsilon) \leq 2e^{-n\varepsilon^2}$$

Thus,  $\delta = 2e^{-n\varepsilon^2}$  gives the maximum probability with which  $p$  lies outside  $\hat{p} - \varepsilon$  and  $\hat{p} + \varepsilon$ .

Since we want  $p$  to lies between  $\hat{p} - \varepsilon$  and  $\hat{p} + \varepsilon$  with probability at least 0.95.

$$\Rightarrow \delta = 1 - 0.95 = 0.05$$

$$\varepsilon = \sqrt{\frac{-1}{n} \ln\left(\frac{\delta}{2}\right)}$$

$$= \sqrt{\frac{-1}{100} \ln\left(\frac{0.05}{2}\right)}$$

$$= 0.192$$

So, the confidence interval is  $(\hat{p} - \varepsilon, \hat{p} + \varepsilon)$

$$= (0.45 - 0.192, 0.45 + 0.192)$$

$$= (0.258, 0.642)$$

(b) For confidence interval of  $(\hat{p} - \varepsilon, \hat{p} + \varepsilon)$  to shrink by half  $(\hat{p} - \frac{\varepsilon}{2}, \hat{p} + \frac{\varepsilon}{2})$ ,  $\varepsilon$  must shrink by half.

$$\text{let } \varepsilon = \sqrt{\frac{1}{n_1} \ln\left(\frac{\delta}{2}\right)}, \quad \frac{\varepsilon}{2} = \sqrt{\frac{1}{n_2} \ln\left(\frac{\delta}{2}\right)}$$

$$\Rightarrow n_2 = 4n_1$$

since  $n_1 = 2n$

$$\Rightarrow n_2 = 4n.$$

So, we need  $(n_2 - n_1) = 3n$   
more samples.