x	1	2	3	4
$f_X(x)$	0.19	0.32	0.31	0.18
$f_{X Y}(x 1)$	10/19	5/19	2/19	2/19

y	1	2	3	4
$f_Y(y)$	0.19	0.32	0.31	0.18
$f_{Y X}(y 1)$	10/19	5/19	2/19	2/19

Table 1: Information for question 2.

## Homework 2

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**Question 1** Let X have a pdf of  $f(x) = cx^2$  for  $0 \le x \le 1$  and f(x) = 0 elsewhere.

(a) For this to be a valid pdf, it must integrate to 1 over the support. So

$$\int_0^1 cx^2 \, dx = \frac{cx^3}{3} \Big|_0^1 = \frac{c}{3} \stackrel{\text{set}}{=} 1,$$

which leads to c = 3. So  $f(x) = 3x^2$  for  $0 \le x \le 1$ .

(b) The cdf is given by

$$F(x) = \int_{-\infty}^{x} f(t) dt = \int_{0}^{x} 3t^{2} dt = t^{3} \Big|_{0}^{x} = x^{3}$$

for  $0 \le x \le 1$ . We also have F(x) = 0 when x < 0 and F(x) = 1 when x > 1.

(c) We have

$$\Pr\left(\frac{1}{10} \le X \le \frac{1}{2}\right) = F\left(\frac{1}{2}\right) - F\left(\frac{1}{10}\right) = \frac{1}{2^3} - \frac{1}{10^3} = \frac{31}{250}.$$

**Question 2** Two discrete random variables X and Y are jointly distributed.

- (a) The marginal pmf for X is obtained by summing over every value of Y for each value of X. For example, to find the marginal probability that X = 1, we have  $f_X(x) = 0.10 + 0.05 + 0.02 + 0.02 = 0.19$ . The other values are obtained in the same way. Finding the marginal pmf for Y is done the exact same way, and it turns out that  $f_X(j) = f_Y(j)$  for  $j = \{1, 2, 3, 4\}$ ; they are both found in Table 1.
- (b) X and Y are not independent. For two random variables to be independent, we need  $f(x,y) = f_X(x) \cdot f_Y(y)$  for all possible (x,y) pairs. Here we have f(1,1) = 0.10 and  $f_X(1) \cdot f_Y(1) = 0.19^2 \neq f(1,1)$ , meaning X and Y are dependent.
- (c) To find the conditional pmf of X given that Y = 1, we take each value of f(x, 1) and divide by  $f_Y(1)$ , i.e.  $f_{X|Y}(x|1) = f(x, 1)/f_Y(1)$ . For example, we have  $f_{X|Y}(1|1) = 0.10/0.19 = 10/19$ . Finding the conditional pmf for Y given that X = 1 is done in a similar way, and again they are the same. Both pmfs can be found in be found in Table 1.

Question 3 We are considering points (x, y) uniformly selected within an ellipse given by the equation  $(x/a)^2 + (y/b)^2 = 1$ , where a, b > 0. Therefore, the probability of selecting a point (x, y) from this region is f(x, y) = c for  $-a \le x \le a$ ,  $-b \le y \le b$ , and  $(x/a)^2 + (y/b)^2 \le 1$ , and f(x, y) = 0 elsewhere. To find c, we use the fact that a joint pdf must integrate to 1 over all values of x and y, so

$$1 = \int_{-a}^{a} \int_{-b\sqrt{1 - (x/a)^2}}^{b\sqrt{1 - (x/a)^2}} c \, dy \, dx = c \cdot 2 \int_{-a}^{a} b\sqrt{1 - (x/a)^2} \, dx = c \cdot \pi ab,$$

which implies that  $c = 1/\pi ab$ . The last equality is from the fact that the area of an ellipse is  $\pi ab$ , which is what the integral is computing. To find the marginal density of X(Y), we integrate over all possible values of Y(X):

$$f_X(x) = \int_{-b\sqrt{1 - (x/a)^2}}^{b\sqrt{1 - (x/a)^2}} \frac{1}{\pi ab} \, \mathrm{d}y = \frac{y}{\pi ab} \Big|_{-b\sqrt{1 - (x/a)^2}}^{b\sqrt{1 - (x/a)^2}} = \frac{2\sqrt{1 - (x/a)^2}}{\pi a} \quad \text{for } -a \le x \le a,$$

$$f_Y(y) = \int_{-a\sqrt{1 - (y/b)^2}}^{a\sqrt{1 - (y/b)^2}} \frac{1}{\pi ab} \, \mathrm{d}x = \frac{x}{\pi ab} \Big|_{-a\sqrt{1 - (y/b)^2}}^{a\sqrt{1 - (y/b)^2}} = \frac{2\sqrt{1 - (y/b)^2}}{\pi b} \quad \text{for } -b \le y \le b.$$

These results make sense intuitively. For  $f_X(x)$ , we can see that the highest probability is obtained when x=0. At x=0, the height of the ellipse is the greatest, so there is more "width" for x=0 to be chosen. And as  $x\to \pm a$ ,  $f_X(x)\to 0$ , meaning that the closer you get to the end of the ellipse, the less likely that point is to be chosen.

- **Question 4** The joint cdf of X and Y is given by  $F(x,y) = (1 e^{-\alpha x})(1 e^{-\beta y})$  for  $x,y \ge 0$  and F(x,y) = 0 elsewhere, where  $\alpha, \beta > 0$  are fixed constants.
  - (a) X and Y are independent. This is because we can express the joing cdf as  $F(x,y) = G(x) \cdot H(y)$ , where  $G(x) = 1 e^{-\alpha x}$  and  $H(y) = 1 e^{-\beta y}$ .
  - (b) The joint pdf is given by

$$f(x,y) = \frac{\partial^2 F}{\partial x \partial y} = \frac{\partial}{\partial x} (1 - e^{-\alpha x}) \cdot \frac{\partial}{\partial y} (1 - e^{-\beta y}) = \alpha e^{-\alpha x} \cdot \beta e^{-\beta y}$$

for  $x, y \ge 0$  and f(x, y) = 0 elsewhere. The marginal densities are given by

$$f_X(x) = \lim_{\phi \to \infty} \int_0^{\phi} \alpha e^{-\alpha x} \cdot \beta e^{-\beta y} dy = \alpha e^{-\alpha x} \cdot \lim_{\phi \to \infty} \left[ -e^{-\beta y} \right]_0^{\phi} = \alpha e^{-\alpha x},$$
  
$$f_Y(y) = \lim_{\phi \to \infty} \int_0^{\phi} \alpha e^{-\alpha x} \cdot \beta e^{-\beta y} dx = \beta e^{-\beta y} \cdot \lim_{\phi \to \infty} \left[ -e^{-\alpha x} \right]_0^{\phi} = \beta e^{-\beta y}.$$

Here we see that  $f_X(x) \cdot f_Y(y) = \alpha e^{-\alpha x} \cdot \beta e^{-\beta y} = f(x, y)$ , another indication that X and Y are independent.

Question 5 Let X and Y have a joint pdf  $f(x,y) = c(x^2 - y^2)e^{-x}$  for  $x \le 0$  and  $-x \le y \le x$ . To help with some of the integrations, this question makes use of the following result:

$$\Omega(\theta, n) := \int_{\theta}^{\infty} x^n e^{-x} dx = n! \cdot e^{-\theta} \sum_{k=0}^{n} \frac{\theta^k}{k!}$$

for all  $\theta \in \mathbb{R}$  and  $n \in \mathbb{Z}_{\geq 0}$ . This specific integral came up so many times in this question that I thought it was worth figuring out a general formula for it. A derivation of this result can be found in Appendix A.

(a) Using the fact that the joint pdf must integrate to 1, we have

$$1 = \int_0^\infty \int_{-x}^x c(x^2 - y^2) e^{-x} dy dx = c \int_0^\infty e^{-x} \left[ x^2 y - \frac{y^3}{3} \right]_{-x}^x dx$$
$$= \frac{4c}{3} \cdot \int_0^\infty x^3 e^{-x} dx = \frac{4c \cdot \Omega(0, 3)}{3} = \frac{4c \cdot 3!}{3} = 8c,$$

and so c = 1/8.

- (b) X and Y are not independent since the joint pdf does not have rectangular support.
- (c) When considering all possible  $x \ge 0$  values, we see that we must have  $-x \le y \le x$ . So the marginal density of X is given by

$$f_X(x) = \int_{-x}^{x} \frac{(x^2 - y^2)e^{-x}}{8} dx = \frac{e^{-x}}{8} \left[ x^2 y - \frac{y^3}{3} \right]_{-x}^{x} = \frac{e^{-x}}{8} \cdot \frac{4x^3}{3} = \frac{x^3 e^{-x}}{6}$$

for  $x \ge 0$ . When considering all possible  $-\infty < y < \infty$ , we must split the support into two regions: where  $y \ge 0$  and where y < 0. When  $y \ge 0$ , we have  $0 \le y \le x < \infty$ , so we have  $y \le x < \infty$ . In this region, the marginal density is given by

$$f_Y(y) = \int_y^\infty \int_{-x}^x \frac{(x^2 - y^2)e^{-x}}{8} dx = \frac{1}{8} \left( \int_y^\infty x^2 e^{-x} dx - y^2 \int_y^\infty e^{-x} dx \right)$$
$$= \frac{1}{8} \left( \Omega(y, 2) - y^2 \cdot \Omega(y, 0) \right) = \frac{1}{8} \left( e^{-y} (2 + 2y + y^2) - y^2 e^{-y} \right) = \frac{(1 + y)e^{-y}}{8}.$$

In the second region, where y < 0, we have  $0 > y \ge -x > -\infty$ , and so we can have  $-y < x < \infty$ . The marginal density is this region is given by

$$f_Y(y) = \int_{-y}^{\infty} \frac{(x^2 - y^2)e^{-x}}{8} dx = \frac{1}{8} \left( \int_{-y}^{\infty} x^2 e^{-x} dx - y^2 \int_{y}^{\infty} e^{-x} dx \right)$$
$$= \frac{1}{8} \left( \Omega(-y, 2) - y^2 \cdot \Omega(-y, 0) \right) = \frac{1}{8} \left( e^y (2 - 2y + y^2) - y^2 e^y \right) = \frac{(1 - y)e^y}{8}.$$

As a whole, the joint density is given by

$$f_Y(y) = \begin{cases} (1+y)e^{-y}/8 & \text{when } y \ge 0, \\ (1-y)e^{y}/8 & \text{when } y < 0. \end{cases}$$

(d) The conditional distribution  $f_{X|Y}(x|y)$  is found by dividing f(x,y) by  $f_Y(y)$ :

$$f_{X|Y}(x|y) = \begin{cases} \frac{(x^2 - y^2)e^{y-x}}{2(1+y)}, & y \ge 0, \ y \le x < \infty \\ \frac{(x^2 - y^2)e^{-(y+x)}}{2(1-y)}, & y < 0, \ -y < x < \infty \end{cases}$$

Similarly,

$$f_{Y|X}(y|x) = \frac{3(x^2 - y^2)}{4x^2}, -x \le y \le x.$$

Question 6 Suppose the rv X has some density  $f_X(x)$  and let Y = aX + b for some  $a \neq 0$ , meaning Y is a linear combination of X. We see that g(X) = aX + b is monotonic with inverse  $h(Y) = \frac{Y - b}{a}$  and h'(Y) = 1/a. Then the density of Y is given by

$$f_Y(y) = f_X(h(y)) \cdot |h'(y)| = f_X\left(\frac{y-a}{b}\right) \cdot \frac{1}{|a|} = \frac{y}{|a|}.$$

## Question 7

**Question 8** Let X be the number of cars in the left lane at a randomly chosen red light, where  $x = \{0, 1, ..., 7\}$ . The engineer also believes  $f(x) \propto (x+1)(8-x)$ .

(a) The engineer's assumption implies that f(x) = c(x+1)(8-x) for some c. So

$$\begin{split} 1 &= \sum_{x=0}^{7} c(x+1)(x-8) = c \sum_{i=0}^{7} \left(8 + 7x - x^2\right) = c \left(8 \sum_{i=0}^{7} 1 + 7 \sum_{i=0}^{7} x - \sum_{i=0}^{7} x^2\right) \\ &= c \left(8 \cdot 8 + \frac{7 \cdot 7 \cdot 8}{2} - \frac{7 \cdot 8 \cdot 15}{6}\right) = 120c, \end{split}$$

which means c = 120 and f(x) = (x+1)(x-8)/120.

(b) We have

$$\Pr(X \ge 5) = \sum_{x=5}^{7} \frac{(x+1)(8-x)}{120} = \frac{6 \cdot 3 + 7 \cdot 2 + 8 \cdot 1}{120} = \frac{40}{120} = \frac{1}{3}.$$

**Question 9** Let X have a pdf given by f(x) = x/8 for  $0 \le x \le 4$ . The cdf is then given by

$$F(x) = \int_0^x \frac{t}{8} dt = \frac{t^2}{16} \Big|_0^x = \frac{x^2}{16}$$

for  $0 \le x \le 4$ , F(x) = 0 if x < 0, and F(x) = 1 if x > 4.

- (a) We have  $\Pr(X \le t) = F(t) = t^2/16 \stackrel{\text{set}}{=} 1/4$ , and solving for t gives us t = 2.
- (b) We have  $\Pr(X \ge t) = 1 F(t) = (16 t^2)/16 \stackrel{\text{set}}{=} 1/2$ , and solving for t gives us  $t = 2\sqrt{2}$ .

**Appendix A** Here we will prove the result used throughout question 5. By using integration by parts with  $u = x^n$  and  $dv = e^{-x}$ , we have

$$\int x^n e^{-x} dx = -x^n e^{-x} + n \int x^{n-1} e^{-x} dx.$$

The next integral must also be solved via integration by parts, with  $u = x^{n-1}$  and  $dv = e^{-x}$ :

$$\int x^n e^{-x} dx = -x^n e^{-x} + n \left( -x^{n-1} e^{-x} + (n-1) \int x^{n-2} e^{-x} dx \right)$$
$$= -x^n e^{-x} - nx^{n-1} e^{-x} - n(n-1) \int x^{n-2} e^{-x} dx.$$

By now a pattern starts to appear. We have to repeat the integration by parts for the right most integral n times in total. Completing this pattern and factoring out an n! from each term, we get

$$\int x^n e^{-x} dx = -x^n e^{-x} - nx^{n-1} e^{-x} - n(n-1)x^{n-2} e^{-x} - \dots - n! + C$$

$$= -e^{-x} \left( \frac{n!}{n!} x^n + \frac{n!}{(n-1)!} x^{n-1} + \frac{n!}{(n-2)!} x^{n-2} + \dots + \frac{n!}{0!} x^0 \right) + C$$

$$= -n! \cdot e^{-x} \left( \frac{x^n}{n!} + \frac{x^{n-1}}{(n-1)!} + \frac{x^{n-2}}{(n-2)!} + \dots + \frac{x^0}{0!} \right) + C = -n! \cdot e^{-x} \sum_{k=0}^{n} \frac{x^k}{k!} + C,$$

where C is the integration constant. Evaluating this integral from  $\theta$  to  $\infty$  gives us

$$\int_{\theta}^{\infty} x^n e^{-x} d = \left[ -n! \cdot e^{-x} \sum_{k=0}^{n} \frac{x^k}{k!} \right]_{\theta}^{\infty} = n! \cdot e^{-\theta} \sum_{k=0}^{n} \frac{\theta^k}{k!} - n! \lim_{\phi \to \infty} e^{-\phi} \sum_{k=0}^{n} \frac{\phi^k}{k!},$$

and using L'Hopital's rule n times on the right limit shows that

$$\lim_{\phi \to \infty} e^{-\phi} \sum_{k=0}^{n} \frac{\phi^{k}}{k!} = \lim_{\phi \to \infty} (-1)^{n} e^{-\theta} = 0.$$

which gives us our desired result.