## Week 2

## Multivariate distributions

**4.1.1/5.1.1** (a) The probability of the event  $A = \{X \le 2, Y \le 3\}$  is directly be given by the joint CDF for x = 2 and y = 3:

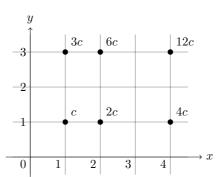
$$P[X \le 2, Y \le 3] = F_{X,Y}(x = 2, y = 3) = (1 - e^{-2})(1 - e^{-3})$$
(2.1)

(b) The marginal cdf can be found by filling in  $y = \infty$ :

$$F_X(x) = F_{X,Y}(x, y = \infty) = \begin{cases} 1 - e^{-x} & x > 0\\ 0 & \text{otherwise.} \end{cases}$$
 (2.2)

(c) This marginal cdf can be found by filling in  $x = \infty$ :

$$F_Y(y) = F_{X,Y}(x = \infty, y) = \begin{cases} 1 - e^{-y} & y > 0\\ 0 & \text{otherwise.} \end{cases}$$
 (2.3)



**4.2.1/5.2.1** First make a picture:

(a) The c should be chosen such, that the PMF adds up to one:

$$c + 2c + 4c + 3c + 6c + 12c = 28c = 1 \rightarrow c = \frac{1}{28}$$
 (2.4)

(b) In the event  $\{Y < X\}$  there are the outcomes (2,1), (4,1) and (4,3), so

$$P[Y < X] = P_{X,Y}(2,1) + P_{X,Y}(4,1) + P_{X,Y}(4,3) = 2c + 4c + 12c = 18/28$$
 (2.5)

(c) In the event  $\{Y > X\}$  there are the outcomes (1,3) and (2,3), so

$$P[Y > X] = P_{X,Y}(1,3) + P_{X,Y}(2,3) = 3c + 6c = 9/28$$
(2.6)

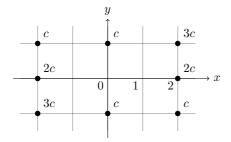
(d) In the event  $\{Y = X\}$  there is only the outcome (1,1) so

$$P[Y = X] = P_{X,Y}(1,1) = c = 1/28$$
(2.7)

(e) In the event  $\{Y=3\}$  there are the outcomes (1,3), (2,3) and (4,3), so

$$P[Y=3] = P_{X,Y}(1,3) + P_{X,Y}(2,3) + P_{X,Y}(4,3) = 3c + 6c + 12c = 21/28$$
(2.8)

## 4.2.2/5.2.2 It is always a good idea to make a picture:



(a) The constant c is found by summing the PMF over all values for X and Y, and equating it to 1.

$$\sum_{x} \sum_{y} P_{X,Y}(x,y) = \sum_{x=-2,0,2} \sum_{y=-1,0,1} c|x+y| = 6c + 2c + 6c = 14c = 1$$
 (2.9)

Therefore c = 1/14.

(b) Simply:

$$P[Y < X] = P_{X,Y}(0,-1) + P_{X,Y}(2,-1) + P_{X,Y}(2,0) + P_{X,Y}(2,1) = 1/2$$
(2.10)

(c) Surprisingly:

$$P[Y > X] = P_{X,Y}(-2,-1) + P_{X,Y}(-2,0) + P_{X,Y}(-2,1) + P_{X,Y}(0,1) = 1/2$$
 (2.11)

- (d) There is no outcome with X = Y so P[X = Y] = 0.
- (e)

$$P[X < 1] = P_{X,Y}(-2,-1) + P_{X,Y}(-2,0) + P_{X,Y}(0,-1) + P_{X,Y}(0,1) + P_{X,Y}(-2,1) = 8/14$$
(2.12)

4.3.2/5.3.2 (a) Use the definition of the marginal probability:

$$P_X(x) = \sum_{y} P_{X,Y}(x,y) = \sum_{y=-1,0,1} P_{X,Y}(x,y) = \begin{cases} 6/14 & x = -2,2\\ 2/14 & x = 0\\ 0 & \text{otherwise} \end{cases}$$
(2.13)

Similarly for  $P_Y(y)$ :

$$P_Y(y) = \sum_{x} P_{X,Y}(x,y) = \sum_{x=-2,0,2} P_{X,Y}(x,y) = \begin{cases} 5/14 & y = -1,1\\ 4/14 & y = 0\\ 0 & \text{otherwise} \end{cases}$$
(2.14)

(b) Given the marginal pdf's it is easy to compute E[X]:

$$E[X] = \sum_{x} x P_X(x) = \sum_{x=-2,0,2} x P_X(x) = -2 \cdot 6/14 + 2 \cdot 6/14 = 0$$
 (2.15)

$$E[Y] = \sum_{y} y P_Y(y) = \sum_{y=-1,0,1} y P_Y(y) = -1 \cdot 5/14 + 1 \cdot 5/14 = 0$$
 (2.16)

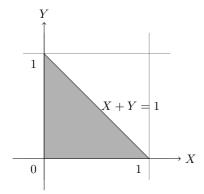
(c) For both variances  $Var(X) = E[X^2] - (E[X])^2 = E[X^2]$ :

$$Var(X) = E[X^2] = \sum_{x} x^2 P_X(x) = (-2)^2 \cdot 6/14 + 2^2 \cdot 6/14 = 24/7$$
 (2.17)

$$Var(Y) = E[Y^2] = \sum_{y}^{\infty} y^2 P_Y(y) = (-1)^2 \cdot 5/14 + 1^2 \cdot 5/14 = 5/7$$
 (2.18)

So  $\sigma_X = \sqrt{24/7}$  and  $\sigma_Y = \sqrt{5/7}$ .

## 4.4.1/5.4.1 Let's first make a picture:



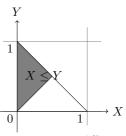
With:  $f_{X,Y} = \begin{cases} c & x+y \le 1, \ x \le 0, \ y \le 0 \\ 0 & \text{otherwise} \end{cases}$ 

(a) The integral over the shaded area should become one:

$$\int f_{X,Y}(x,y)dxdy = \int_0^1 \int_0^{1-x} cdydx = \int_0^1 \left[ cy \right]_0^{1-x} dx = c \int_0^1 (1-x)dx = c \left[ x - \frac{1}{2}x^2 \right]_0^1 = c/2 = 1$$
(2.19)

and therefore c=2.

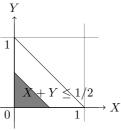
(b) For  $P[X \leq Y]$  we have to integrate over the following area:



That means (first integrating over y, then over x):

$$P[X \le Y] = \int_0^{1/2} \int_x^{1-x} c dy dx = c \int_0^{1/2} [cy]_x^{1-x} dx = c \int_0^{1/2} (1-2x) dx = c \left[x - x^2\right]_0^{1/2} = 1/2$$

(c) For  $P[X+Y \le 1/2]$  we have to integrate over the following area:



That means (first integrating over y, then over x):

$$P[X \le Y] = \int_0^{1/2} \int_0^{1/2-x} c dy dx = c \int_0^{1/2} [cy]_0^{1/2-x} dx = c \int_0^{1/2} (1/2-x) dx = c \left[ x/2 - x^2/2 \right]_0^{1/2} = 1/4$$

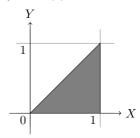
**4.5.2**/5.5.3 (a) This joint pdf is the same as in question 4.4.1. The marginals are computed using the definition:

$$f_X(x) = \int_y f_{X,Y}(x,y)dy = \int_0^{1-x} 2dy = \begin{cases} 2(1-x) & 0 \le x \le 1\\ 0 & \text{otherwise} \end{cases}$$
 (2.20)

(b) And...

$$f_Y(y) = \int_x f_{X,Y}(x,y) dx = \int_0^{1-y} 2dx = \begin{cases} 2(1-y) & 0 \le y \le 1\\ 0 & \text{otherwise} \end{cases}$$
 (2.21)

4.5.6/5.5.9 (a) As always, first make a picture:

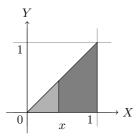


(b) The constant c is determined by the fact that the pdf should integrate to 1. I decide to first integrate over y for a given value of x. For a given value of x variable y runs between 0 and x. The second integral is then over x, between 0 and 1:

$$\int_{0}^{1} \int_{0}^{x} cy dy dx = \int_{0}^{1} \frac{1}{2} cx^{2} dx = \left[ \frac{1}{6} cx^{3} \right]_{0}^{1} = \frac{1}{6} c = 1$$
 (2.22)

And therefore c = 6.

(c) To compute the cdf  $F_X(x) = P[X \le x]$  we have to integrate over the nonzero probability region left of the vertical line at x:



When we choose x < 0 this integral is 0, and if we choose x > 1 then this integral is 1. We now only have to worry for situations where  $0 \le x \le 1$ .

I introduce two integration variables u (over X) and v (over Y), and the integral becomes:

$$F_X(x) = \int_0^x \int_0^u cv \, dv \, du \qquad (2.23)$$

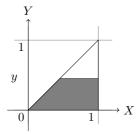
$$= \int_0^x \frac{1}{2} cu^2 du \tag{2.24}$$

$$= \frac{1}{6}cx^3 = x^3 \tag{2.25}$$

So, in total the cdf becomes:

$$F_X(x) = \begin{cases} 0 & x < 0 \\ x^3 & 0 \le x \le 1 \\ 1 & x > 1 \end{cases}$$
 (2.26)

(d) To compute the cdf  $F_Y(y)$ , we have to integrate over the nonzero probability region below the horizontal line at y. When we choose y < 0 this integration becomes 0, and when we choose y > 1, this integration becomes 1. When we have  $0 \le y \le 1$  we have to make the integration over the following region:



Again, I introduce two integration variables u (over X) and v (over Y). I decide to integrate first in the Y direction, then over the X direction:

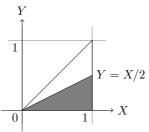
$$F_Y(y) = \int_0^y \int_v^1 cv \, du \, dv$$
 (2.27)

$$= \int_0^y \left[ cvu \right]_v^1 dv \tag{2.28}$$

$$= \int_{0}^{y} cv(1-v)dv \tag{2.29}$$

$$= c \left[ \frac{1}{2}v^2 - \frac{1}{3}v^3 \right]_0^y = c \left( \frac{1}{2}y^2 - \frac{1}{3}y^3 \right)$$
 (2.30)

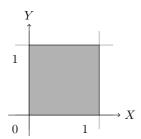
(e) To compute  $P(Y \le X/2)$  we first have to find the event. It is indicated in the figure:



I integrate first over Y, then over X:

$$P[Y \leq X/2] = \int_0^1 \int_0^{x/2} cy \, dy \, dx = \int_0^1 \left[\frac{1}{2}cy^2\right]_0^{x/2} dx = \int_0^1 \frac{1}{8}cx^2 \, dx = \left[\frac{1}{24}cx^3\right]_0^1 = \frac{1}{24}c$$
 (2.31)

4.7.9/5.7.12 Here we go again. First a picture (well, a bit overdone, it is quite simple):



(a) For the computation of  $E[X] = \int x f_X(x) dx$  and Var[X] we need the marginal distribution, so let's do that first:

$$f_X(x) = \int f_{X,Y}(x,y)dy = \int_0^1 4xydy = \left[2xy^2\right]_0^1 = 2x$$
 for  $0 \le x \le 1$ 

The expected value for X is

$$E[X] = \int x f_X(x) dx = \int_0^1 2x^2 dx = \left[\frac{2x^3}{3}\right]_0^1 = \frac{2}{3}$$
 (2.32)

and

$$E[X^{2}] = \int x^{2} f_{X}(x) dx = \int_{0}^{1} 2x^{3} dx = \left[2x^{4}\right]_{0}^{1} = \frac{1}{2}$$
(2.33)

The variance therefore becomes

$$Var[X] = E[X^2] - (E[X])^2 = \frac{1}{2} - \left(\frac{2}{3}\right)^2 = \frac{1}{18}$$
 (2.34)

(b) When we compute the marginal pdf of Y:

$$f_Y(y) = \int f_{X,Y}(x,y)dx = \int_0^1 4xydx = \left[2yx^2\right]_0^1 = 2y$$
 for  $0 \le y \le 1$ 

we see that it is exactly the same as the one for X. So also the expected value and variance is the same! The expected value for Y is

$$E[Y] = \frac{2}{3}, \qquad Var[Y] = \frac{1}{18}$$
 (2.35)

(c) The covariance is defined as Cov[X,Y] = E[XY] - E[X]E[Y], so we need the correlation first:

$$E[XY] = \int xy f_{X,Y}(x,y) dx dy = \int_0^1 \int_0^1 4x^2 y^2 dy dx$$
 (2.36)

$$= \int_0^1 4y^2 \left[ \frac{1}{3} x^3 \right]_0^1 dy \tag{2.37}$$

$$= \frac{4}{3} \left[ \frac{1}{3} y^3 \right]_0^1 = \frac{4}{9} \tag{2.38}$$

so the covariance:

$$Cov[X,Y] = \frac{4}{9} - \frac{2}{3} \cdot \frac{2}{3} = 0$$
 (2.39)

- (d) Well, this is simply E[X + Y] = E[X] + E[Y] = 4/3.
- (e) Use Theorem 4.15,

$$Var[X+Y] = Var[X] + Var[Y] + 2Cov[X,Y] = \frac{1}{18} + \frac{1}{18} + 0 = \frac{1}{9}$$
 (2.40)