

## Exercise Sheet 4

### A: Warm-up questions

1. From the specification of the bivariate normal distribution for  $(X, Y)$ , we can read off:  $\mu_1 = 0$ ,  $\mu_2 = 1$ ,  $\sigma_1 = 1$ ,  $\sigma_2 = 2$ ,  $\rho = \frac{1}{2}$ .

Therefore, the joint density is easy to write down using a formula from the lecture notes:

$$f(x_1, x_2) = \frac{1}{2\pi\sqrt{3}} \exp \left[ -\frac{2}{3} \left\{ x_1^2 - \frac{x_1(x_2 - 1)}{2} + \frac{(x_2 - 1)^2}{4} \right\} \right]$$

The marginal distributions are straightforward, again using results from section 1.5.2 of the lecture notes:  $X_1$  is  $N(0, 1)$ ,  $X_2$  is  $N(1, 4)$ . Finally, the conditional distributions are given as follows:  $X_2|X_1 = x_1$  is  $N(x_1 + 1, 3)$ ,  $X_1|X_2 = x_2$  is  $N(\frac{1}{4}(x_2 - 1), \frac{3}{4})$

2. (a) You can get rid of all those zeros by multiplying the two vectors by 1/1000, so you only have to solve

$$\begin{pmatrix} 5 \\ 1 \\ 1 \end{pmatrix} + \begin{pmatrix} 5 \\ -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 10 \\ 0 \\ 2 \end{pmatrix}$$

I admit that this is a rather simple example and the simplification isn't all that significant but at least here it's easy to see exactly what's going on – this will change when handling multivariate normal distributions.

(b)

$$\begin{aligned} \frac{5}{7} \begin{pmatrix} 5 & 1 & 0 \\ 1 & 3 & 0 \\ 1 & 7 & 7 \end{pmatrix} - \frac{5}{7} \begin{pmatrix} 5 & 1 & 1 \\ 1 & 3 & -7 \\ 0 & 0 & 0 \end{pmatrix}^T &= \frac{5}{7} \begin{pmatrix} 5 & 1 & 0 \\ 1 & 3 & 0 \\ 1 & 7 & 7 \end{pmatrix} - \frac{5}{7} \begin{pmatrix} 5 & 1 & 0 \\ 1 & 3 & 0 \\ 1 & -7 & 0 \end{pmatrix} \\ &= \frac{5}{7} \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 14 & 7 \end{pmatrix} \\ &= \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 10 & 5 \end{pmatrix} \end{aligned}$$

Note that here it was advantageous to carry out the matrix algebra first and exploit all those cancellations before multiplying by 5/7. The resulting matrix is not symmetric, i.e.  $A = A^T$  does not hold, so it cannot be a covariance matrix. (Remember: covariance is symmetric, i.e.  $cov(X, Y) = cov(Y, X)$ .)

### B: Hand-in questions

1. (a) The case of bag 1 is the easiest: since all letters in that bag score one point, the conditional expected value is one,  $\mathbb{E}[T|B = 1] = 1$ .

There are 17 letters in bag 2 altogether, whence

$$\mathbb{E}[T|B = 2] = \dots = \frac{38}{17} \approx 2.235$$

There are 15 letters in bag 3 altogether, whence

$$\mathbb{E}[T|B = 3] = \dots = \frac{81}{15} \approx 5.4$$

The interpretation is that ...

- (b) The variance in bag one is again easy: since all letters have the same number of points, the variance will be zero.

... from which it follows that

$$\text{Var}(T|B = 2) \approx 0.886$$

$$\text{Var}(T|B = 3) \approx 5.04$$

The interpretation is that ...

- (c) The marginal expectation is

$$\mathbb{E}[T] = \dots \approx 3.612$$

and this stands for ...

The marginal variance is

$$\text{Var}(T) = \dots \approx 6.183$$

Again, this is the variance of the number of points ...

2. (a) The sketch for should show a triangle with corners  $(-1, 0)$ ,  $(1, 0)$ ,  $(\alpha, 1)$ .  
 The three lines are  $y = 0$ ,  $x = -1 + (\alpha + 1)y$  and  $x = 1 + y(\alpha - 1)$  ( or equivalent expressions of  $y$  in terms of  $x$ , both are fine).  
 For  $\alpha = -1$ , ... Therefore, we expect negative covariance.  
 For  $\alpha = 0$ , ..., so we expect zero covariance here.  
 For  $\alpha = 1$ ,... we expect positive covariance here.
- (b) Assuming that  $y \in [0, 1]$ , we obtain:

$$f_Y(y) = \dots = 2(1 - y) \text{ on } y \in [0, 1].$$

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy$$

Now, the integrand will only be positive once the point  $(x, y)$  ...

$$f_X(x) = \dots = \frac{x + 1}{\alpha + 1}$$

and a similar result holds for  $\alpha < x \leq 1$ :

$$f_X(x) = \dots = \frac{x-1}{\alpha-1}$$

The expected values are obtained by integration using the given marginal densities:

$$\begin{aligned}\mathbb{E}[Y] &= \dots = \frac{1}{3} \\ \mathbb{E}[X] &= \dots = \frac{\alpha}{3}\end{aligned}$$

(c)

$$\mathbb{E}[XY] = \dots = \frac{\alpha}{6}$$

The covariance is therefore

$$\text{Cov}(X, Y) = \dots = \frac{\alpha}{18}$$

which has the same sign as  $\alpha$  and therefore agrees with the observations made in part (a).

## Section C

later