

Exercises

Multivariate Statistics

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|||| Chapter 1

|||| Exercise 1.1

Consider the matrix

$$\rho = \begin{bmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{bmatrix}$$

- a) For which values of ρ is this a valid dispersion matrix? (Hint: Both dispersion and correlation matrices must be positive semi-definite)

|||| Solution

ρ must be positive semidefinite, i.e. it must have non-negative eigenvalues. We have

$$\det \begin{pmatrix} 1 - \lambda & \rho & \rho \\ \rho & 1 - \lambda & \rho \\ \rho & \rho & 1 - \lambda \end{pmatrix} = (1 - \lambda)^3 + 2\rho^3 - 3\rho^2(1 - \lambda)$$

Substituting $x = (1 - \lambda)$ we get $x^3 - 3\rho^2x + 2\rho^3 = 0$. The roots of which are $\rho, \rho, -2\rho$. This can be trivially solved in e.g. Maple, or we can guess a root ρ and find the remaining by reducing:

$$\frac{x^3 - 3\rho^2x + 2\rho^3}{x - \rho} = x^2 + \rho x - 2\rho^2$$

The eigenvalues then become:

$$1 - \lambda = \begin{cases} \rho \\ \rho \\ -2\rho \end{cases} \implies \lambda = \begin{cases} 1 - \rho \\ 1 - \rho \\ 1 + 2\rho \end{cases}$$

these are all non-negative iff $\underline{\underline{-\frac{1}{2} \leq \rho \leq 1}}$

Assume that ρ is a valid dispersion matrix. Let furthermore the 3-dimensional random variable X be normally distributed

$$X \sim N(0, \rho)$$

- b) Determine the axes in the contour ellipsoid for the probability density function of X

|||| Solution

This is equivalent to finding the eigenvectors. It can easily be solved with Maple - we show how to do it by hand below. We found the eigenvalues in the previous question and will start with $\lambda_1 = 1 + 2\rho$. We subtract it from the diagonal and must then solve:

$$\begin{bmatrix} -2\rho & \rho & \rho \\ \rho & -2\rho & \rho \\ \rho & \rho & -2\rho \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

thus the first eigenvector is of the form

$$s \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

and the normed version is

$$v_1 = \begin{bmatrix} \sqrt{3}/3 \\ \sqrt{3}/3 \\ \sqrt{3}/3 \end{bmatrix}$$

We then consider $\lambda_2 = \lambda_3 = 1 - \rho$ and get

$$\begin{bmatrix} \rho & \rho & \rho \\ \rho & \rho & \rho \\ \rho & \rho & \rho \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

or

$$x_1 + x_2 + x_3 = 0$$

This spans a plane and we may choose

$$v_2 = s \cdot \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} = \begin{bmatrix} \sqrt{2}/2 \\ \sqrt{2}/2 \\ 0 \end{bmatrix}$$

Finally, v_3 must satisfy

$$\begin{aligned} x_1 + x_2 + x_3 &= 0 \\ x_1 - x_2 &= 0 \end{aligned}$$

i.e. $x_3 = -2x_1$ and $x_2 = x_1$, which gives

$$v_3 = s \cdot \begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} \sqrt{6}/6 \\ \sqrt{6}/6 \\ -\sqrt{6}/3 \end{bmatrix}$$

||| Exercise 1.2 EXAM 2013 - Problem 2

We consider a random variable

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

with mean and dispersion matrix respectively equal to

$$\mu = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad \text{and} \quad \Sigma = \begin{bmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{bmatrix}$$

Furthermore we consider the random variables

$$U = X + Y + Z$$

$$V = 2X - Y - Z$$

- a) [org. 2.1] The mean value of the two-dimensional random variable $\begin{bmatrix} U \\ V \end{bmatrix}$ is?

|||| **Solution**

We use theorem 1.3 and 1.2.

$$E(U) = E(X + Y + Z) = E(X) + E(Y) + E(Z) = 1 + 1 + 1 = 3$$

$$E(V) = E(2X - Y - Z) = E(2X) - E(Y) - E(Z) = 2 - 1 - 1 = 0$$

Collecting the terms: $E \left(\begin{bmatrix} U \\ V \end{bmatrix} \right) = \begin{bmatrix} 3 \\ 0 \end{bmatrix}$

b) [org. 2.2] The variance of U is?

|||| **Solution**

We use theorem 1.8.

$$\begin{aligned} D(U) &= D(X + Y + Z) = C(X + Y + Z, X + Y + Z) \\ &= C(X, X) + C(X, Y) + C(X, Z) + \\ &\quad C(Y, X) + C(Y, Y) + C(Y, Z) + \\ &\quad C(Z, X) + C(Z, Y) + C(Z, Z) \\ &= 1 + \rho + \rho + \\ &\quad \rho + 1 + \rho + \\ &\quad \rho + \rho + 1 \\ &= 3 + 6\rho \end{aligned}$$

Note that $C(X, X) = D(X)$

c) [org. 2.3] The variance of V is?

|||| Solution

We use theorem 1.8 and 1.6

$$\begin{aligned}
 D(V) &= D(2X - Y - Z) = C(2X - Y - Z, 2X - Y - Z) \\
 &= C(2X, 2X) + C(2X, -Y) + C(2X, -Z) + \\
 &\quad C(-Y, 2X) + C(-Y, -Y) + C(-Y, -Z) + \\
 &\quad C(-Z, 2X) + C(-Z, -Y) + C(-Z, -Z) \\
 &= 4 - 2\rho - 2\rho \\
 &\quad - 2\rho + 1 + \rho \\
 &\quad - 2\rho + \rho + 1 \\
 &= 6 - 6\rho
 \end{aligned}$$

Note that $C(-X, -X) = -1 \cdot C(X, X) \cdot (-1) = D(X)$ as the minuses cancels each other.

d) [org. 2.4] The covariance between U and V is?

|||| Solution

We use theorem 1.8 and define our original variables as $W = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$, then

$$C(U, V) = C(AW, BW)$$

where

$$A = [1 \ 1 \ 1]$$

and

$$B = [2 \ -1 \ -1]$$

thus

$$\begin{aligned}
 C(U, V) &= C(AW, BW) = A C(W, W) B^T \\
 &= [1 \ 1 \ 1] \begin{bmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \\ -1 \end{bmatrix} \\
 &= 0
 \end{aligned}$$

Note that $C(W, W) = D(W)$ which was given in the exercise.

e) [org. 2.5] The conditional mean $E\left(\begin{bmatrix} X \\ Y \end{bmatrix} \middle| Z = z\right)$ is equal to?

|||| Solution

We use theorem 1.27

$$\begin{aligned} E\left(\begin{bmatrix} X \\ Y \end{bmatrix} \middle| Z = z\right) &= \mu_1 + \Sigma_{12}\Sigma_{22}^{-1}(x_2 - \mu_2) \\ &= \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} \rho \\ \rho \end{bmatrix} [1]^{-1}(z - 1) \\ &= \begin{bmatrix} 1 + \rho(z - 1) \\ 1 + \rho(z - 1) \end{bmatrix} \end{aligned}$$

|||| Exercise 1.3

The goal of this exercise, is to further familiarise you with calculations of statistical moments on multivariate data, as well as the interpretation of partial correlations.

We consider a three-dimensional random variable

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

with dispersion matrix

$$\Sigma = \begin{bmatrix} 1 & \rho & \rho^2 \\ \rho & 1 & \rho \\ \rho^2 & \rho & 1 \end{bmatrix}$$

We now consider the linear combination $Y + aZ$. We want to estimate a such that the quantity

$$\text{Corr}(X, Y + aZ)^2$$

is maximised.

- a) What is the variance of X , the variance of $Y + aZ$ and the covariance $\text{Cov}(X, Y + aZ)$?

|||| Solution

We use theorem 1.8 and remark 1.9.

$$\begin{aligned}
 V(X) &= 1 \\
 V(Y + aZ) &= V(Y) + V(aZ) + 2 \operatorname{Cov}(Y, aZ) \\
 &= V(Y) + a^2 V(Z) + 2a \operatorname{Cov}(Y, Z) \\
 &= 1 + a^2 + 2a\rho \\
 \operatorname{Cov}(X, Y + aZ) &= \operatorname{Cov}(X, Y) + \operatorname{Cov}(X, aZ) \\
 &= \operatorname{Cov}(X, Y) + a \operatorname{Cov}(X, Z) = \\
 &= \rho + a\rho^2
 \end{aligned}$$

- b) Using the results above, write up the expression for $\operatorname{Corr}(X, Y + aZ)^2$.

|||| Solution

We use the formula for correlation found in section 1.1.3, and insert the values found above in a).

$$\begin{aligned}
 \operatorname{Corr}(X, Y + aZ)^2 &= \left(\frac{\operatorname{Cov}(X, Y + aZ)}{\sqrt{V(X)V(Y + aZ)}} \right)^2 \\
 &= \frac{\operatorname{Cov}(X, Y + aZ)^2}{V(X)V(Y + aZ)} \\
 &= \frac{(\rho + a\rho^2)^2}{1 + a^2 + 2a\rho}
 \end{aligned}$$

- c) Determine a such that $\operatorname{Corr}(X, Y + aZ)^2$ is maximised (Hint: You may consider using a symbolic solver, but it is possible by hand).

|||| Solution

We had that

$$\text{Corr}(X, Y + aZ)^2 = \frac{(\rho + a\rho^2)^2}{1 + a^2 + 2a\rho}$$

which we can expand to

$$\rho^2 \frac{1 + a^2\rho^2 + 2a\rho}{1 + a^2 + 2a\rho}$$

To find the maximum as a function of a we take the derivative of the numerator with respect to a

$$\rho^2 [(2a\rho^2 2\rho)(1 + a^2 + 2a\rho) - (1 + a^2\rho^2 + 2a\rho)(2a + 2\rho)] = 2\rho^2(\rho^2 - 1)a(a\rho + 1)$$

The zeros for this expression is $a = 0$ for the maximum and $a = -\frac{1}{\rho}$ for the minimum.

The answer is thus $a = 0$!

- d) Using the value found for a , determine the maximum of the squared correlation.

|||| Solution

We insert in the equation found in b)

$$\text{Corr}(X, Y + aZ)^2 = \frac{(\rho + a\rho^2)^2}{1 + a^2 + 2a\rho} = \frac{(\rho + 0\rho^2)^2}{1 + 0^2 + 2 \cdot 0\rho} = \rho^2$$

- e) Compare the result in d) with the squared multiple correlation between X and $(Y, Z)^T$

|||| Solution

We use theorem 1.42

$$\rho_{X|YZ}^2 = 1 - \frac{\det \Sigma}{\det \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}} = 1 - \frac{1 + \rho^4 - 2\rho^2}{1 - \rho^2} = \frac{-\rho^4 + \rho^2}{1 - \rho^2} = \rho^2$$

The same result as in d)!

- f) Find the partial correlation between X and Z given Y , and between X and Y given Z .

|||| Solution

We use the equation on the bottom of page 34.

$$\rho_{XZ|Y} = \frac{\rho^2 - \rho\rho}{\sqrt{(1-\rho^2)(1-\rho^2)}} = 0$$

$$\rho_{XY|Z} = \frac{\rho - \rho^2\rho}{\sqrt{(1-\rho^4)(1-\rho^2)}} = \frac{\rho}{\sqrt{1+\rho^2}}$$

- g) Comment on the results. Can you relate the value you found for a , with the partial correlation values?

|||| Solution

The result in c) tells us that when we want to predict X based on Y and Z we should only use Y . Z will not contribute any further. This is in accordance with the result in f) namely that $\rho_{XZ|Y} = 0$, i.e. given Y then X and Z are uncorrelated, i.e. Z contains no additional information on X if we know Y .