



**Problem 1** Simple matrix calculations

Solve the problems by hand *and* by use of R (when possible).

Let  $A = \begin{pmatrix} 9 & -2 \\ -2 & 6 \end{pmatrix}$ .

- a) Construct  $A$  as a matrix in R. Use the function `matrix`.
- b) Is  $A$  symmetric? Use the R function `t`.
- c) Show that  $A$  is positive definite (not in R – use the definition of positive definiteness).
- d) Find the eigenvalues and the eigenvectors of  $A$ . Are the eigenvectors found by R normalized? Use the function `eigen`.
- e) Find an orthogonal diagonalization of  $A$ . In R, matrix multiplication is performed by `%*%`.
- f) Find  $A^{-1}$ . Use the R function `solve`.
- g) Find the eigenvalues and the eigenvectors of  $A^{-1}$ . Is there a relationship between the eigenvalues and the eigenvectors of  $A$  and  $A^{-1}$ ?
- h) Why can  $A$  be a covariance matrix?
- i) Assume that  $A$  is the covariance matrix of a random vector. Find the correlation matrix, that is, the matrix having the correlation coefficient of the  $i$  and  $j$  entries of the random vector as its  $ij$  entry. The R functions `diag` and `sqrt` may be useful. Check your computations with `cov2cor`.
- j) Let  $\mathbf{X}$  be a random vector, and assume

$$E\mathbf{X} = \begin{pmatrix} 3 \\ 1 \end{pmatrix} \quad \text{and} \quad \text{Cov } \mathbf{X} = A.$$

Find, in R, the expectation and covariance matrices of

$$\begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix} \mathbf{X}, \quad (1 \ 2) \mathbf{X} \quad \text{and} \quad \begin{pmatrix} \mathbf{X} \\ 3\mathbf{X} \end{pmatrix}$$

(the last is a block matrix, in this case the concatenation of the vectors  $\mathbf{X}$  and  $3\mathbf{X}$ ).

**Problem 2 Mean and covariance of linear combinations**

Let  $\mathbf{X}$  be a trivariate (three-dimensional) random vector with mean (expectation)  $(1 \ 1 \ 1)^T$  and covariance matrix  $I$ , a  $3 \times 3$  identity matrix. Find the mean and covariance matrix of

$$\begin{pmatrix} \frac{2}{3} & -\frac{1}{3} & -\frac{1}{3} \\ -\frac{1}{3} & \frac{2}{3} & -\frac{1}{3} \\ -\frac{1}{3} & -\frac{1}{3} & \frac{2}{3} \end{pmatrix} \mathbf{X}.$$

**Problem 3 Covariance formula**

Let  $\mathbf{V}$  and  $\mathbf{W}$  be random vectors of the same dimension. The covariance matrix of  $\mathbf{V}$  and  $\mathbf{W}$  is defined as  $\text{Cov}(\mathbf{V}, \mathbf{W}) = E((\mathbf{V} - E\mathbf{V})(\mathbf{W} - E\mathbf{W})^T)$ . Show that  $\text{Cov}(\mathbf{V}, \mathbf{W}) = E(\mathbf{V}\mathbf{W}^T) - (E\mathbf{V})(E\mathbf{W})^T$ .

Note that this is a generalization of a well-known formula for univariate variables. In the case  $\mathbf{W} = \mathbf{V}$ , we get  $\text{Cov}(\mathbf{V}) = E(\mathbf{V}\mathbf{V}^T) - (E\mathbf{V})(E\mathbf{V})^T$ , which is a generalization of the univariate  $\text{Var } V = EV^2 - (EV)^2$ .

**Problem 4 The square root matrix and the Mahalanobis transform**

Let the expectation (mean) and covariance matrix of a random vector  $\mathbf{X}$  be  $\boldsymbol{\mu} = E\mathbf{X}$  and  $\Sigma = \text{Cov } \mathbf{X}$ . Let  $P$  be an orthogonal matrix having eigenvectors of  $\Sigma$  as columns and  $\Lambda$  a diagonal matrix having the eigenvalue corresponding to the  $i$ th column of  $\Sigma$  as its  $ii$  entry. Then  $\Sigma = P\Lambda P^T$ .

- a) Show that  $\Sigma$  is positive semidefinite. (A symmetric matrix  $A$  is positive semidefinite if  $\mathbf{z}^T A \mathbf{z} \geq 0$  for all vectors  $\mathbf{z}$ .)

Assume that  $\Sigma$  is positive definite. (A symmetric matrix  $A$  is positive definite if  $\mathbf{z}^T A \mathbf{z} > 0$  for all vectors  $\mathbf{z} \neq \mathbf{0}$ .)

- b) Show that all eigenvalues of  $\Sigma$  are positive.

Why does  $\Sigma$  have an inverse? What can you say about the eigenvalues and eigenvectors of  $\Sigma^{-1}$ ? Justify the answer.

- c) Let  $\Lambda^{1/2}$  be the diagonal matrix having as entries the square root of those of  $\Lambda$ , and let  $\Lambda^{-1/2} = (\Lambda^{1/2})^{-1}$ . Define

$$\Sigma^{1/2} = P\Lambda^{1/2}P^T \quad \text{and} \quad \Sigma^{-1/2} = P\Lambda^{-1/2}P^T.$$

Show that both are symmetric, and that

$$\Sigma^{1/2}\Sigma^{1/2} = \Sigma, \quad \Sigma^{-1/2}\Sigma^{-1/2} = \Sigma^{-1} \quad \text{and} \quad \Sigma^{1/2}\Sigma^{-1/2} = I,$$

where  $I$  is an identity matrix.

- d) The transform  $\mathbf{Y} = \Sigma^{-1/2}(\mathbf{X} - \boldsymbol{\mu})$  is called the Mahalanobis transform. Show that  $E\mathbf{Y} = \mathbf{0}$  and  $\text{Cov } \mathbf{Y} = I$ .