

## SMF MOCK EXAMINATION. 18.2.2011

- Q1. (i) State without proof the spectral decomposition for a real symmetric matrix  $A$ . [4]  
(ii) Show how to define the square root and inverse square root of  $A$ . [2, 2]  
(iii) If  $x$  has independent  $N(0, \sigma^2)$  components and  $y := Ox$  with  $O$  an orthogonal matrix, show that  $y$  has the same distribution as  $x$ . [5]  
(iv) If  $A$  is real and symmetric, and  $Q := x^T Ax$  is the quadratic form in  $x$  as in (iii), express  $Q$  as a quadratic form in independent normal variables with diagonal matrix. [4]  
(v) For  $A$  real symmetric, show that  $A$  is idempotent iff all its eigenvalues are 0 or 1. [4]  
(vi) For  $P$  a symmetric projection, show that the rank and trace of  $P$  coincide. [4]

- Q2. (i) Give the definition of the multivariate normal distribution  $N(\mu, \Sigma)$ . [2]  
(ii) Show that if  $x \sim N(\mu, \Sigma)$  and  $y := Ax + b$ , then  $y$  is multivariate normal, and find its mean vector and covariance matrix. [2, 2, 4]  
(iii) Show that any subvector of a multivariate normal vector is multivariate normal. [2]  
(iv) If  $x \sim N(\mu, \Sigma)$  is partitioned as

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix},$$

state without proof the conditional distribution of  $x_1$  given  $x_2$ . [3]

(v) If

$$\Sigma = \begin{pmatrix} 1 & \rho & \rho^2 \\ \rho & 1 & 0 \\ \rho^2 & 0 & 1 \end{pmatrix}$$

and  $x \sim N(\mu, \Sigma)$ , find the conditional distribution of

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} | x_3. \quad [10]$$

- Q3. (i) If  $X = (X_t)$  is  $L_1$ -bounded, i.e.  $\|X\|_1 := \sup_t E[|X_t|] < \infty$ , and  $\psi = (\psi_j) \in \ell_1$  (i.e.  $\|\psi\|_1 = \sum_j |\psi_j| < \infty$ ), show that  $\sum_j \psi_j X_{t-j}$  converges a.s. and in  $\ell_1$ . [5, 5]
- (ii) Show that  $\ell_1 \subset \ell_2$ . [5]
- (iii) If also  $X$  is  $L_2$ -bounded, i.e.  $\|X\|_2 := \sup_t E[|X_t|^2] < \infty$ , show that  $\sum_j \psi_j X_{t-j}$  also converges in  $\ell_2$ , to the same sum. [10]

Q4. Describe briefly, without proofs, the method of principal components analysis. [6]

Discuss the advantages and disadvantages of working with covariances and with correlations. [6]

Give examples, in the financial area, where each might be appropriate. [13]

Q5. (i) Show that

$$\Sigma = \begin{pmatrix} 1 & \rho & \rho \\ \rho & 1 & \rho \\ \rho & \rho & 1 \end{pmatrix},$$

where the parameter  $\rho$  is a correlation, has eigenvalues  $1 + 2\rho$  (simple) and  $(1 - 2\rho)$  (double), with eigenvectors

$$\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \quad \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}, \quad \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}. \quad [4]$$

(ii) Deduce that this matrix can only be a correlation matrix under a restriction on  $\rho$ . Find this restriction, and the further restriction that  $\Sigma$  be non-singular. [4, 4]

(iii) If  $x_i \sim N(\mu, \Sigma)$  and the vector  $y$  has coordinates  $y_1 := x_1 + x_2$ ,  $y_2 := x_2 + x_3$ , find the mean vector and covariance matrix of  $y$ . [4, 9]

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