

**CS4070 – PART 2**  
**EXERCISES RELATED TO LECTURE 1**

- (1) Review some concepts from linear algebra such as: eigenvalue decomposition of a matrix, best approximation theorem, null/column space, spanned subspace of a set of vectors, invertibility of a matrix, trace of a matrix.
- (2) Suppose the square matrix  $A$  has eigenvalue decomposition  $A = V\Lambda V^T$ , where  $\Lambda$  is a diagonal matrix containing the eigenvalues and  $V$  is an orthogonal matrix containing eigenvectors as columns. Show that  $\text{tr}A = \text{tr}\Lambda$  and hence that the trace of  $A$  is the sum of its eigenvalues.
- (3) Show that  $\mathcal{N}(A) = \mathcal{N}(A^T A)$ , where  $\mathcal{N}(A) = \{x : Ax = 0\}$ .
- (4) If

$$Y = \begin{bmatrix} Y_1 \\ \vdots \\ Y_n \end{bmatrix}$$

then the mean-vector and covariance matrix are defined by

$$\mathbb{E}Y = \begin{bmatrix} \mathbb{E}Y_1 \\ \vdots \\ \mathbb{E}Y_n \end{bmatrix}$$

and

$$\text{Cov}Y = \mathbb{E}[(Y - \mathbb{E}Y)(Y - \mathbb{E}Y)^T]$$

respectively.

- (a) Check that element  $[i, j]$  of the matrix  $\text{Cov}(Y)$  is given by  $\text{Cov}(Y_i, Y_j)$ .
- (b) Suppose that  $A$  is a  $k \times n$  matrix. Verify that  $\mathbb{E}[AY] = A\mathbb{E}Y$ . *Hint: note that for  $j \in \{1, \dots, k\}$ , the  $j$ -th element of the vector  $AY$  satisfies  $(AY)[j] = \sum_{i=1}^n A_{ji}Y_i$ .*
- (c) Verify that

$$\text{Cov}(AY) = A(\text{Cov}Y)A^T.$$

- (5) Suppose  $U$  and  $V$  are independent random variables, each with the  $N(0, 1)$ -distribution. for  $\rho \in [-1, 1]$ , define

$$\begin{aligned} X_1 &= \sqrt{1 - \rho^2}U + \rho V \\ X_2 &= V \end{aligned}$$

Verify that the correlation between  $X_1$  and  $X_2$  equals  $\rho$ .

- (6) Consider the linear model

$$\mathbf{y} = X\boldsymbol{\theta} + \boldsymbol{\epsilon}, \quad \boldsymbol{\epsilon} \sim N(0, \Sigma).$$

Assume  $\Sigma$  is nonsingular and known.

- (a) Derive an expression for the maximum-likelihood estimator for  $\boldsymbol{\theta}$ , if it exists. Also explain when the maximum-likelihood estimator is unique.
- (b) Derive an expression for the Hessian matrix of the loglikelihood (this is the matrix containing the second-order derivatives of the loglikelihood).