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## Exercise 7

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**Problem 1.** (*Fisher's Linear Discriminant Function for two classes*) If Fisher's linear discriminant function is used for classification into two classes  $C_1$  and  $C_2$ , prove that an observation  $\mathbf{x}$  is allocated to  $C_1$  if  $\mathbf{a}^T(\mathbf{x} - \frac{1}{2}(\bar{\mathbf{x}}_1 + \bar{\mathbf{x}}_2)) > 0$  with  $\mathbf{a} = \mathbf{W}^{-1}(\bar{\mathbf{x}}_1 - \bar{\mathbf{x}}_2)$ .

**Problem 2.** (*ML Discriminant Rule for two classes*) Suppose that ML discriminant rule is used for classification into two classes  $C_1$  and  $C_2$ . The class distributions are Gaussian and known as  $N_p(\boldsymbol{\mu}_1, \boldsymbol{\Sigma}_1)$  and  $N_p(\boldsymbol{\mu}_2, \boldsymbol{\Sigma}_2)$  with  $\boldsymbol{\Sigma}_1 = \boldsymbol{\Sigma}_2 = \boldsymbol{\Sigma}$ . The densities are:

$$f_l(\mathbf{u}) = \frac{1}{(2\pi)^{p/2} |\boldsymbol{\Sigma}|^{1/2}} \exp \left\{ -\frac{1}{2} (\mathbf{u} - \boldsymbol{\mu}_l)^T \boldsymbol{\Sigma}^{-1} (\mathbf{u} - \boldsymbol{\mu}_l) \right\}, \mathbf{u} \in \mathbb{R}^p, l = 1, 2.$$

Prove that the ML rule allocates  $\mathbf{x}$  to the class  $C_1$  if

$$\boldsymbol{\alpha}^T (\mathbf{x} - \boldsymbol{\mu}) > 0,$$

where  $\boldsymbol{\alpha} = \boldsymbol{\Sigma}^{-1}(\boldsymbol{\mu}_1 - \boldsymbol{\mu}_2)$  and  $\boldsymbol{\mu} = \frac{1}{2}(\boldsymbol{\mu}_1 + \boldsymbol{\mu}_2)$ .

**Problem 3.** (*Eigenvalues in Fisher's Linear Discriminant Analysis*) Let  $\mathbf{X} = [\mathbf{x}_1, \dots, \mathbf{x}_n]^T \in \mathbb{R}^{n \times p}$  be samples and  $\mathbf{W}$  and  $\mathbf{B}$  are matrices corresponding to within-group and between-group sum of squares. Define  $\mathbf{S} = \mathbf{X}^T \mathbf{E}_n \mathbf{X}$ . Suppose that  $\mathbf{W}$  has rank  $p$ . Show that the following three eigenvectors are the same:

- the eigenvector corresponding to the largest eigenvalue of  $\mathbf{W}^{-1}\mathbf{B}$
- the eigenvector corresponding to the largest eigenvalue of  $\mathbf{W}^{-1}\mathbf{S}$
- the eigenvector corresponding to the smallest eigenvalue of  $\mathbf{S}^{-1}\mathbf{W}$