

**ISLAMIC UNIVERSITY OF TECHNOLOGY (IUT)**  
**ORGANISATION OF ISLAMIC COOPERATION (OIC)**

**Department of Computer Science and Engineering (CSE)**

**MID SEMESTER EXAMINATION**

**DURATION: 1 Hour 30 Minutes**

**SUMMER SEMESTER, 2013-2014**

**FULL MARKS: 75**

**CSE 4835: Pattern Recognition**

**Programmable calculators are not allowed. Do not write anything on the question paper.**

**There are 4 (four) questions. Answer any 3 (three) of them.**

**Figures in the right margin indicate marks.**

- a) A discriminant function is a linear combination of the components of  $x$  and can be written as:  $g(x) = w'x + w_0$ . 10
- Find the weight coefficients of vector  $w$  for a 'minimum distance to class member' classifier.
- b) Write down the differences between Supervised, Unsupervised and Reinforcement Learning. 6
- c) Draw a diagram illustrating the design cycle of a pattern recognition system and explain how feedback information from classifier evaluation may change each component of the design cycle. 4+5
- a) Consider a minimum error rate classification achieved by the use of the following discriminant functions:  $g_i(x) = \ln P(x | \omega_i) + \ln P(\omega_i)$ , where each class has its own covariance matrix. For the multivariate normal case, convert this discriminant function to its general form  $g(x) = w'x + w_0$ . 10
- b) For which special case, the discriminant function will change the classifier in question 2.(a) to a 'minimum distance' classifier? 5
- c) In many pattern classification problems one has the option either to assign the pattern to one of  $c$  classes, or to *reject* it as being unrecognizable. If the cost for rejects is not too high, rejection may be a desirable action. 10
- Let  $\lambda(\alpha_i | \omega_j) = \begin{cases} 0 & i = j \quad i, j = 1, 2, \dots, c \\ \lambda_r & i = c+1 \\ \lambda_s & \text{otherwise} \end{cases}$
- where,  $\lambda_r$  is the loss incurred for choosing the  $(c+1)^{\text{th}}$  action, rejection, and  $\lambda_s$  is the loss incurred for making a substitution error. Show that the minimum risk is obtained if we decide  $\omega_i$  if  $P(\omega_i | x) \geq P(\omega_j | x)$  for all  $j$  and if  $P(\omega_i | x) \geq 1 - \lambda_r / \lambda_s$ , and reject otherwise.
- a) Suppose a set of samples is given for a 2-class problem and they are not linearly separable in the original feature space. How can you design a linear classifier for the same set of samples and find the decision boundary? [Note: Mention the choice of your criterion function.] 10
- b) Briefly discuss why and when Mahalanobis distance is preferred over Euclidean distance. 5
- c) How do nonparametric density estimation techniques estimate the unknown probability density functions? Step by step derive the final estimation function, and list the conditions for convergence. 10



4. a) Let random variable  $x$  have an exponential density:

$$p(x | \theta) = \begin{cases} \theta e^{-\theta x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases} \quad \text{and } \theta > 0.$$

Suppose that  $n$  samples  $x_1, x_2, \dots, x_n$  are drawn independently according to  $p(x | \theta)$ . Show that the maximum likelihood estimate of  $\theta$  is given by

$$\hat{\theta} = \frac{1}{\frac{1}{n} \sum_{k=1}^n x_k}$$

- b) For the univariate Gaussian case, estimate  $p(\mu | D)$  using Bayesian Parameter estimation.