



**PRINCE OF SONGKLA UNIVERSITY
FACULTY OF ENGINEERING**

Midterm Examination: Semester II
Date: 24 December 2009
Subject: 241-650 Principles of Pattern Recognition

Academic Year: 2009
Time: 13:30-16:30
Room: R300

Instructions:

This exam has 6 problems, 10 pages and 65 points. You may use the back of the pages for scratch work. This exam is closed books and notes. You are allowed to use a calculator. You may consult one A4 sheet of notes (two sides).

Name ID

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1 ภาคการศึกษ โทษสูงสดุไล่ออก**

1. A pattern is represented by a pair of features, $(x_1, x_2)^T$. Suppose there are 2 classes of patterns, each having a mean vector and a covariance matrix shown in the table below:

| Class | Mean | Covariance |
|-------|------------|--|
| 1 | $(1, 2)^T$ | $\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ |
| 2 | $(5, 3)^T$ | $\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ |

Use the technique of minimum distance classifier to determine the equation of the decision boundary, assuming equal prior probability (10 points)

2. Two random variables \mathbf{x} and \mathbf{y} are called statistically independent if $p(\mathbf{x}, \mathbf{y} | \omega) = p(\mathbf{x} | \omega)p(\mathbf{y} | \omega)$. Prove that if $\mathbf{x}_i - \boldsymbol{\mu}_i$ and $\mathbf{x}_j - \boldsymbol{\mu}_j$ are statistically independent (for $i \neq j$) then $\sigma_{ij} = \mathcal{E}[(\mathbf{x}_i - \boldsymbol{\mu}_i)(\mathbf{x}_j - \boldsymbol{\mu}_j)] = 0$
($\boldsymbol{\mu}_i, \boldsymbol{\mu}_j$ are constants) (10 points)

3. Let x has a uniform density $p(x|\theta) = U(0, \theta) = \begin{cases} \frac{1}{\theta} & 0 \leq x \leq \theta \\ 0 & \text{otherwise} \end{cases}$

Suppose that n samples $\mathbf{D} = \{x_1, x_2, \dots, x_n\}$ are drawn independently according to $p(x|\theta)$. Show that the maximum likelihood estimate for θ is $\max[\mathbf{D}]$ —that is, the value of the maximum element in \mathbf{D} . (10 points)

4. Suppose we have 3 classes of patterns whose features are in 2-D feature space. Each has the following underlying distributions:

$$p(\mathbf{x} | \omega_1) \cong N(\mathbf{0}, 2\mathbf{I})$$

$$p(\mathbf{x} | \omega_2) \cong N\left(\begin{bmatrix} 2 \\ 2 \end{bmatrix}, \mathbf{I}\right)$$

$$p(\mathbf{x} | \omega_3) \cong N\left(\begin{bmatrix} 0 \\ 0.5 \end{bmatrix}, \mathbf{I}\right)$$

with $P(\omega_1) = 1/2$ and $P(\omega_2) = P(\omega_3) = 1/4$.

By explicit calculation of posterior probabilities, classify the point $x = \begin{bmatrix} 0.5 \\ 2 \end{bmatrix}$ for minimum probability of error. (15 points)

5. Class 1 and Class 2 has the feature vectors as shown in the following table:

| Class 1 | | Class 2 | |
|---------|----|---------|----|
| x1 | x2 | x1 | x2 |
| -3 | 0 | 6 | 0 |
| -2 | 1 | 5 | 0 |
| -2 | 0 | 5 | 2 |
| -1 | -1 | 4 | -2 |

The features of each class are assumed to have a Gaussian distribution function (normally distributed). Use the maximum likelihood parameter estimation method to determine $p(\mathbf{x}|\boldsymbol{\theta}_1)$ and $p(\mathbf{x}|\boldsymbol{\theta}_2)$ (10 points)

6. Find the eigenvalues and eigenvectors of matrix \mathbf{A} , where $\mathbf{A} = \begin{bmatrix} 1 & 3 \\ 1 & -1 \end{bmatrix}$ (10 points)