MATH477201

This question paper consists of 5 printed pages, each of which is identified by the reference MATH477201.

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Examination for the Module MATH4772 (January 2003)

MULTIVARIATE AND CLUSTER ANALYSIS

Time allowed: 3 hours

Attempt not more than THREE questions of the first FOUR.

All candidates must attempt question FIVE.

All questions carry equal marks.

1. The $p \times 1$ random vector x has a multivariate normal distribution with probability density function

$$f(\mathbf{x}) = |2\pi\Sigma|^{-1/2} \exp\Big\{-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \Sigma^{-1}(\mathbf{x} - \boldsymbol{\mu})\Big\}, \qquad \mathbf{x} \in \mathbb{R}^p,$$

and moment generating function

$$M_{\mathbf{x}}(\mathbf{t}) = \exp\left\{\mathbf{t}^T oldsymbol{\mu} + rac{1}{2} \mathbf{t}^T \Sigma \mathbf{t}
ight\}.$$

The matrix Σ is non-singular. We write $\mathbf{x} \sim N_p(\boldsymbol{\mu}, \Sigma)$.

Let x be partitioned into p_1 and p_2 components, $p_1 + p_2 = p$, with corresponding partitions

$$oldsymbol{\mu} = \left(egin{array}{c} oldsymbol{\mu}_1 \ oldsymbol{\mu}_2 \end{array}
ight) \qquad ext{and} \qquad \Sigma = \left(egin{array}{cc} \Sigma_{11} & \Sigma_{12} \ \Sigma_{21} & \Sigma_{22} \end{array}
ight) \, .$$

(a) Consider $y = x_1 + Mx_2$, where M is a $p_1 \times p_2$ matrix. Show that

$$Cov(\mathbf{y}, \mathbf{x}_2) = \Sigma_{12} + M\Sigma_{22}$$

What value of M results in independence between y and x_2 ?

When y and x_2 are independent the conditional distribution of $y|x_2 = x_2^o$ will be the same as the marginal distribution of y. Use this to show that

$$\mathbf{x}_1|\mathbf{x}_2 = \mathbf{x}_2^{\circ} \sim N_{p_1}(\boldsymbol{\mu}_1 + \Sigma_{12}\Sigma_{22}^{-1}(\mathbf{x}_2^{\circ} - \boldsymbol{\mu}_2), \Sigma_{11.2}),$$

where $\Sigma_{11\;2} = \Sigma_{11} - \Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}$

(b) Suppose

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \sim N_2 \left(\begin{pmatrix} 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 & 0.5 \\ 0.5 & 4 \end{pmatrix} \right)$$

Construct a new random vector $\mathbf{z} = (z_1, z_2)^T$ with $z_1 = 2x_1 + x_2$, $z_2 = x_1 - x_2$. Find the variance matrix and correlation matrix of \mathbf{z} .

2. (a) Suppose x_1, \dots, x_n is a random sample from $N_p(\mu, \Sigma)$ population (where Σ , assumed positive definite, is unknown). Obtain the log likelihood function $l(\mu, \Sigma)$. Show that the maximum likelihood estimator for μ is $\hat{\mu} = \bar{x}$.

Let x_1, \dots, x_n be a random sample from $N_p(\mu_x, \Sigma)$ and y_1, \dots, y_m be a random sample from $N_p(\mu_y, \Sigma)$. The two samples are independent of one another. Show that the union intersection test of the hypothesis $H_0: \mu_x = \mu_y$ vs. $H_1: \mu_x \neq \mu_y$, where Σ is unknown, leads to the test statistic

$$T^2 = \frac{nm}{n+m} (\bar{\mathbf{x}} - \bar{\mathbf{y}})^T S_p^{-1} (\bar{\mathbf{x}} - \bar{\mathbf{y}}),$$

where $\bar{\mathbf{x}}$ and $\bar{\mathbf{y}}$ are sample mean vectors and S_p is the pooled within groups estimate of Σ .

- (b) If H_0 is rejected in the overall test, how can simultaneous confidence intervals be used to give insight into the reasons for rejection?
- (c) In a study to compare male gorilla skulls with female skulls the following variables were measured:

variable 1 = braincase length, variable 2= braincase height.

A sample of 11 males and 11 females yielded the following summary statistics:

$$ar{\mathbf{x}} = \left(egin{array}{c} 152 \ 103 \end{array}
ight), \qquad S_x = \left(egin{array}{c} 40 & 10 \ 10 & 25 \end{array}
ight),$$

$$\bar{\mathbf{y}} = \begin{pmatrix} 142 \\ 101 \end{pmatrix}, \qquad S_y = \begin{pmatrix} 32 & 4 \\ 4 & 17 \end{pmatrix}.$$

The pooled covariance matrix for the two samples and its inverse are given by

$$S_p = \left(\begin{array}{cc} 36 & 7 \\ 7 & 21 \end{array} \right), \qquad S_p^{-1} = \left(\begin{array}{cc} 0.03 & -0.01 \\ -0.01 & 0.05 \end{array} \right)$$

Explain how S_p is calculated. Compare the sexes on the basis of the information provided.

[Hints:

- 1. You may use the fact that the Hotelling T^2 and F distribution are related by $T^2(p,\nu) = \{\nu p/(\nu-p+1)\}F(p,\nu-p+1)$.
- 2. Simultaneous 100α percent confidence intervals for this problem can be written in the form

$$(\mathbf{a}^T(\bar{\mathbf{x}}-\bar{\mathbf{y}})-c,\mathbf{a}^T(\bar{\mathbf{x}}-\bar{\mathbf{y}})+c)$$

where $c = \left\{ T_{\alpha}^2(p, \nu) \frac{n+m}{nm} \mathbf{a}^T S_p \mathbf{a} \right\}^{\frac{1}{2}}$ and $T_{\alpha}^2(p, \nu)$ is the 100α percentage point of the $T^2(p, \nu)$ distribution.

- 3. (a) Let x be a p-dimensional random vector with mean vector μ and covariance matrix Σ
 - (i) Define the principal components y of x in terms of the standardized eigenvectors of Σ.
 - (ii) Obtain the variance-covariance matrix of the principal components y
 - (iii) If $\Sigma = \alpha \alpha^T$ for some vector α , find the first principal component. What can you say about the other principal components?
 - (iv) Why do we restrict ourselves to standardized linear combinations when carrying out principal component analysis?
 - (b) Data were collected on 50 irises from the species *Iris setosa*. The variables are: x_1 =sepal length; x_2 =sepal width; x_3 =petal length; x_4 =petal width. The sample mean vector and correlation matrix were

$$\mathbf{x} = \begin{pmatrix} 5.01 \\ 3.43 \\ 1.46 \\ 0.25 \end{pmatrix}, \qquad R = \begin{pmatrix} 1 & 0.74 & 0.27 & 0.28 \\ 0.74 & 1 & 0.18 & 0.23 \\ 0.27 & 0.18 & 1 & 0.33 \\ 0.28 & 0.23 & 0.33 & 1 \end{pmatrix}.$$

Principal component analysis gave eigenvalues 2.06, 1.02, 0.67, 0.25. The corresponding eigenvectors were the columns of

$$\begin{pmatrix} 0.60 & -0.33 & 0.07 & 0.72 \\ 0.58 & -0.44 & 0.00 & -0.69 \\ 0.38 & 0.63 & 0.68 & -0.09 \\ 0.40 & 0.55 & -0.73 & -0.01 \end{pmatrix}$$

Interpret these principal components briefly. Assess their relative contribution to total variation. What methods can be used to decide on the number of components we should retain?

4. (a) Let $f_1(\mathbf{x})$ and $f_2(\mathbf{x})$, $\mathbf{x} \in \mathbb{R}^p$, denote two probability density functions for populations Π_1 and Π_2 respectively, with prior probabilities π_1 and π_2 , where $\pi_1 + \pi_2 = 1$. Consider the allocation rule which assigns \mathbf{x} to Π_1 if

$$\frac{f_1(\mathbf{x})}{f_2(\mathbf{x})} \ge \frac{\pi_2}{\pi_1}$$

and to Π_2 otherwise. Show that this rule is admissible.

(b) If the two populations have $N_p(\mu_1, \Sigma)$ and $N_p(\mu_2, \Sigma)$ distributions and $\pi_1 = 2\pi_2$, show that this rule classifies x as coming from the first population if

$$(\mu_1 - \mu_2)\Sigma^{-1}\mathbf{x} \ge c$$

and to the second population otherwise, for a suitable constant c. What is the value of c?

(c) Let

$$\Sigma = \left(egin{array}{cc} 2 & 1 \ 1 & 5 \end{array}
ight), \qquad oldsymbol{\mu}_1 = \left(egin{array}{cc} 2 \ 2 \end{array}
ight), \qquad oldsymbol{\mu}_2 = \left(egin{array}{cc} 5 \ 1 \end{array}
ight).$$

Calculate and sketch the boundary between the two classification regions for $\pi_1 = 2\pi_2$ and $\pi_1 = \pi_2$ respectively. Compare these boundaries. How would you classify $\mathbf{x} = (3,2)^T$ under each rule?

- (d) If the population parameters have to be replaced by sample estimates, what problem arises in estimating the misclassification probabilities and how might this be overcome?
- 5. (a) Let a set of centred coordinates of n points in a p dimensional Euclidean space be given by $\mathbf{x}_r = (x_{r1}, \dots, x_{rp})^T$, so that $\sum_{r=1}^n x_{rj} = 0$, $j = 1, \dots, p$. The Euclidean distance between the rth and sth points is given by

$$d_{\pi s}^2 = (\mathbf{x}_r - \mathbf{x}_s)^T (\mathbf{x}_r - \mathbf{x}_s).$$

Let $B = (b_{rs})$ be the positive semi definite inner product matrix

$$B = XX^T$$

where $X = (\mathbf{x}_1, \dots, \mathbf{x}_n)^T$ is the $n \times p$ matrix of coordinates.

Show that the elements of B are determined from the Euclidean distances by

$$b_{rs} = a_{rs} - \bar{a}_r - \bar{a}_s + \bar{a} \tag{1}$$

where $a_{\tau s} = -\frac{1}{2}d_{\tau s}^2$, $\bar{a}_{\tau} = \frac{1}{n}\sum_{s=1}^n a_{\tau s}$, $\bar{a}_{s} = \frac{1}{n}\sum_{r=1}^n a_{rs}$ and $\bar{a}_{r} = \frac{1}{n^2}\sum_{r=1}^n \sum_{s=1}^n a_{rs}$

(b) Assuming only B is known, show how to find X from the spectral decomposition of B. Why is X determined only up to arbitrary location, rotation and reflection? (Note: if the rank of B is p, then B has n-p zero eigenvalues).

(c) Estimated car travel times between the 7 cities Hull, Leeds, Liverpool, London, Manchester, Sheffield and York were calculated by the RAC and are given in the matrix D,

	Hull	0	73	144	273	112	82	61	1
D =	Leeds	73	0	84	209	52	43	45	l
	Liverpool	144	84	0	231	43	102	120	
	London	273	209	231	0	218	183	238	l
	Manchester	112	52	43	218	0	65	86	
	Sheffield	82	43	102	183	65	0	74	l
	York	61	45	120	238	86	74	0	

Consider the estimated car travel time to be a measure of distance between cities. We wish to construct a two dimensional map of the 7 cities from D.

A matrix B was calculated from D using equation (1). Since D is not a distance matrix arising from a two dimensional space, B is not positive semi definite. The eigenvalues of B are 45821, 12401, 1983, 407, 0 -56, -1175. The corresponding standardized eigenvectors are the columns of

$$\begin{pmatrix} 0.38 & -0.45 & 0.39 & -0.15 & 0.38 & -0.01 & -0.58 \\ 0.10 & -0.05 & -0.18 & -0.53 & 0.38 & -0.62 & 0.38 \\ 0.10 & 0.72 & 0.03 & -0.37 & 0.38 & 0.41 & -0.14 \\ -0.88 & -0.14 & -0.04 & -0.04 & 0.38 & 0.00 & -0.24 \\ 0.10 & 0.38 & 0.05 & 0.69 & 0.38 & -0.46 & -0.12 \\ -0.02 & -0.17 & 0.50 & 0.19 & 0.38 & 0.33 & 0.66 \\ 0.22 & -0.28 & -0.75 & 0.21 & 0.38 & 0.35 & 0.05 \end{pmatrix}$$

- (i) Describe how to construct a map from B in p'=2 dimensions.
- (ii) Using an appropriate measure for the proportion of explained variation compare the choice p'=2 with other choices for p'.
- (iii) Obtain the distance between Leeds and London for the map you have described in (i). Compare this value with the original RAC estimate given in D and comment on your findings.

END