

STA503 (Multivariate Analysis)

Instructions: 1. All questions carry equal marks.

2. Scientific calculator can be used.

Q-1(a) Let $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$ and let \underline{x} , $\underline{\mu}$ and Σ be partition as follows.

$$\underline{x} = \begin{bmatrix} \underline{x}_1 \\ \underline{x}_2 \end{bmatrix} \begin{matrix} r \\ s \end{matrix}, \quad \underline{\mu} = \begin{bmatrix} \underline{\mu}_1 \\ \underline{\mu}_2 \end{bmatrix} \begin{matrix} r \\ s \end{matrix} \quad \text{and} \quad \Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{12} & \Sigma_{22} \end{bmatrix} \begin{matrix} r \\ s \end{matrix}, \quad r + s = p.$$

(i) Show that $\underline{x}_1 - \Sigma_{11}^{-1} \Sigma_{12} \underline{x}_2$ and \underline{x}_2 are independently distributed.

(ii) Obtain the conditional distribution of $(x_1 / X_2 = x_2)$.

OR

(a) Let $\underline{x}_r, r=1,2,\dots,k$, be independently distributed as $N_p(\underline{\mu}_r, \Sigma_r)$. Then for fixed matrices $A_r: m \times p$, obtain the distribution of $\sum_{r=1}^k A_r \underline{x}_r$. If $\underline{\mu}_r = \underline{\mu}$ and $\Sigma_r = \Sigma; r=1,2,\dots,k$,

then obtain the distribution of $\bar{\underline{x}}$.

(b) Define partial Correlation coefficient. In usual notation obtain the expression in terms of elements of $\Sigma^{-1} = (\sigma^{ij})$ for partial correlation coefficient. Hence obtain $r_{12 \cdot}$.

OR

(b) Define canonical correlation coefficients and canonical variates. In usual notation show that the canonical correlation are solution of the determinant equation

$$\begin{vmatrix} -\lambda \Sigma_{11} & \Sigma_{12} \\ \Sigma_{12} & -\lambda \Sigma_{22} \end{vmatrix} = 0.$$

Show that multiple correlation is a special cases of canonical correlation.

Q-2 (a) let $\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n$ be n independent observation from $N_p(\underline{\mu}, \Sigma)$ ($n > p$)

Population. Show that the sample mean $\bar{\underline{x}}$ and $S = \sum_{i=1}^n (\underline{x}_i - \bar{\underline{x}})(\underline{x}_i - \bar{\underline{x}})'$ are

Independently distributed.

OR

(a) Define Wishart matrix. Obtain probability density function of Wishart matrix $V: p \times p$

when $n \geq p$, $\underline{\mu} = 0$ and $\Sigma = I_p$.

E715-2

- (b) Define Hotelling's T^2 statistic. Show that it is used to test the $H_0: \underline{\mu} = \underline{\mu}_0$ against $H_1: \underline{\mu} \neq \underline{\mu}_0$ when $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$. Obtain the distribution of T^2 under H_0 . What is the power of the test?

OR

- (b) Show that Hotelling's T^2 can be used to test $H_0: \rho_{1.23\dots p} = 0$ against $H_1: \rho_{1.23\dots p} \neq 0$, where $\rho_{1.23\dots p}$ is multiple correlation coefficient.

- Q-3(a) Obtain the estimated minimum ECM rule for classifying an object \underline{x}_0 when $\Sigma_1 = \Sigma_2$. Obtain probabilities of errors of misclassification for the classification rule you have obtained.

OR

- (a) Define sample Mahalanobis distance D^2 , obtain the relation between D^2 and Hotelling's T^2 . Hence, obtain the distribution of D^2 .
- (b) Explain orthogonal factor model with K common factors. Give principal component solution of the factor model.

OR

- (b) Obtain null distribution of sample correlation coefficient matrix $R = (r_{ij})$.

- Q-4 (a) Explain the technique of One Way MANOVA for the comparison of several multivariate population means.

OR

- (a) Define principal components. Write its important applications. If $\Sigma = \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}$ where $\rho > 0$, then find the principal components associated with matrix Σ and find the percentage of total variance explained by first principal component.

- (b) Obtain null distribution of the sample correlation coefficient r . Write $E(r)$ and $Var(r)$.

OR

- (b) Obtain MLE of $\underline{\beta}$ and σ^2 in GLM. How do you test $H: \underline{c}'\underline{\beta} = \underline{c}'\underline{\beta}_0$ for a specified real vector \underline{c}' ?

- Q-5 Choose the appropriate answer.

1. If $X = \begin{bmatrix} 2 & 8 & 6 & 8 \\ 12 & 9 & 9 & 10 \end{bmatrix}$ is an observation matrix of order 2×4 then its mean vector is

(A) (6,10)

(B) (24,40)

(C) (14,17,15,18)

(D) (14,17,15,18)/2

2. If $x:2 \times 1$ is distributed as $N_2(\underline{\mu}, \Sigma)$ with $\underline{\mu}' = (1,5)$ and $\Sigma = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}$, then the

distribution of $\underline{c}'x$, where $\underline{C}' = (1,-1)$ is

- (A) $N(-4, 2)$ (B) $N_2(\underline{\mu}, \Sigma)$
 (C) $N(4, -2)$ (D) $N(-4, 3)$

3. Let x_1, x_2 and x_3 be distributed as $N_2(\underline{\mu}, \Sigma)$ with $\underline{\mu}' = (0, 0)$ and $\Sigma = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}$. Then

$E(x_1x_1 + x_2x_2 + x_3x_3)$ is

- (A) 1 (B) 6
 (C) $\Sigma = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}$ (D) $\Sigma = \begin{bmatrix} 3 & 6 \\ 6 & 15 \end{bmatrix}$

4. In usual notations the formula for partial correlation coefficient $r_{12.3}$ is

- (A) $\frac{\sigma^{12}}{\sqrt{\sigma^{11}\sigma^{22}}}$ (B) $\frac{-\sigma^{12}}{\sqrt{\sigma^{11}\sigma^{22}}}$
 (C) $\frac{-\sigma^{12}}{\sigma^{11}\sigma^{22}}$ (D) $\frac{\sigma^{12}}{\sigma^{11}\sigma^{22}}$

5. Let $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$ and consider the partition of \underline{x} , $\underline{\mu}$ and Σ as follows.

$$\underline{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \quad \underline{\mu} = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, \quad \Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}; \quad \text{where } \Sigma_{11} : r \times r \text{ and } \Sigma_{22} : s \times s \text{ are matrices}$$

with $r+s=p$. The conditional distribution of x_1 given $x_2 = \underline{x}_2$ is

- (A) $N_p(\underline{\mu}_1 + \Sigma_{12}\Sigma_{22}^{-1}(x_2 - \underline{\mu}_2), \Sigma_{11} - \Sigma_{12}\Sigma_{22}^{-1}\Sigma_{12})$
 (B) $N_r(\underline{\mu}_1 - \Sigma_{12}\Sigma_{22}^{-1}(x_2 - \underline{\mu}_2), \Sigma_{11} - \Sigma_{12}\Sigma_{22}^{-1}\Sigma_{12})$
 (C) $N_p(\underline{\mu}_2 + \Sigma_{12}\Sigma_{11}^{-1}(x_1 - \underline{\mu}_1), \Sigma_{22} - \Sigma_{12}\Sigma_{11}^{-1}\Sigma_{12})$
 (D) $N_s(\underline{\mu}_2 - \Sigma_{12}\Sigma_{11}^{-1}(x_1 - \underline{\mu}_1), \Sigma_{22} - \Sigma_{12}\Sigma_{11}^{-1}\Sigma_{12})$

6. If \underline{x}_1 and \underline{x}_2 are independent $N_p(\underline{\theta}_i, \Sigma_i)$; $i=1,2$ respectively, then the distribution of $(\underline{x}_1 - \underline{x}_2)$ is

- (A) $N_p(\underline{\theta}_1 - \underline{\theta}_2, \Sigma_1 - \Sigma_2)$ (B) $N_p(\underline{\theta}_1 - \underline{\theta}_2, \Sigma_1 + \Sigma_2)$
 (C) $N_p(\underline{\theta}_1 + \underline{\theta}_2, \Sigma_1 + \Sigma_2)$ (D) $N_p(\underline{\theta}_1 + \underline{\theta}_2, \Sigma_1 - \Sigma_2)$

7. Let $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$, where Σ is a nonsingular matrix. The characteristic function of the vector $\underline{y} = C\underline{x}$ is given by

- (A) $\phi_{\underline{y}}(\underline{t}) = \exp\left(i\underline{t}'C\underline{\mu} - \frac{1}{2}\underline{t}'C\Sigma C'\underline{t}\right)$ (B) $\phi_{\underline{y}}(\underline{t}) = \exp\left(i\underline{t}'C\underline{\mu} - \frac{1}{2}\underline{t}'C\Sigma C'\underline{t}\right)$
 (C) $\phi_{\underline{y}}(\underline{t}) = \exp\left(i\underline{t}'C\underline{\mu} + \frac{1}{2}\underline{t}'C\Sigma C'\underline{t}\right)$ (D) $\phi_{\underline{y}}(\underline{t}) = \exp\left(i\underline{t}'C\underline{\mu} + \frac{1}{2}\underline{t}'C\Sigma C'\underline{t}\right)$

8. The unbiased estimate of the variance covariance matrix for multivariate normal distribution is

P.T.O

E715-4

$$(A) \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x})$$

$$(B) \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x}) / n$$

$$(C) \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x}) / (n-1)$$

$$(D) \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x}) / n$$

9. If the joint pdf of (x, y) is $\frac{1}{2.4\pi} \exp \left[- \left\{ \frac{(x^2/4) - (1.6xy/2) + y^2}{0.72} \right\} \right]$ then the values of

$\mu_x, \mu_y, \sigma_x, \sigma_y$ and ρ_{xy} are respectively

(A) (0, 0, 2, 1, 0.8)

(B) (0, 0, 2, 1, 0.6)

(C) (1/2, 1, 2, 1, 0.4)

(D) (0, 0, 1, 2, 0.8)

10. The Hotelling's T^2 is a generalization of

(A) Chi-square distribution

(B) t-distribution

(C) Square of t-distribution

(D) F-distribution

11. Let X_1, X_2, \dots, X_{20} be a random sample of size $n=20$ from a $N_6(\underline{\mu}, \Sigma)$ population. If

$B = \begin{bmatrix} 1, 0, 0, 0, 0, 0 \\ 0, 0, 1, 0, 0, 0 \end{bmatrix}$, then the distribution of $B^* = B(19S)B'$ is

(A) Chi-square distribution with degrees of freedom 6.

(B) Non-central Chi-square distribution with degrees of freedom 6

(C) $W_2(B^*, 19, B\Sigma B')$

(D) $W_6(B^*, 19, B\Sigma B')$

12. Let X_1, X_2, \dots, X_{25} be a random sample of size $n = 50$ from a $N_6(\underline{\mu}, \Sigma)$ population. Then the distribution of $(\underline{x}_1 - \underline{\mu})' \Sigma^{-1} (\underline{x}_1 - \underline{\mu})$ is

(A) Chi-square with '50' degrees of freedom

(B) Non central Chi-square with '50' degrees of freedom

(C) Chi-square with '6' degrees of freedom

(D) Non central Chi-square with '6' degrees of freedom

13. If the sample mean vector \bar{x} and variance covariance matrix S of three iid observations

from a bivariate normal distribution are $\begin{bmatrix} 8 \\ 6 \end{bmatrix}$ and $\begin{bmatrix} 4 & -3 \\ -3 & 9 \end{bmatrix}$ respectively then the value of

observed T^2 for testing $H_0 : \mu = \mu_0 = \begin{pmatrix} 9 \\ 5 \end{pmatrix}$ is

(A) 7/9

(B) 9/7

(C) 7/27

(D) 9/27

14. let $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$, $\Sigma > 0$. Then the distribution of $(\underline{x} - \underline{\mu})'(\underline{x} - \underline{\mu})$ is

(A) $\chi_p^2(\underline{\mu}' \Sigma^{-1} \underline{\mu})$

(B) $W_p(1, \Sigma)$

(C) $W_p(n, \Sigma)$

(D) χ_p^2