

**BIRLA INSTITUTE OF TECHNOLOGY, MESRA, RANCHI
(MID-SEMESTER EXAMINATION)**

**CLASS: IMSc
BRANCH: CQEDS**

**SEMESTER: V/ADD
SESSION: MO/2025**

SUBJECT: ED303 - MULTIVARIATE DATA ANALYSIS

TIME: 02 Hours

FULL MARKS: 25

INSTRUCTIONS:

1. The question paper contains 5 questions, each of 5 marks and a total of 25 marks.
2. Attempt all questions.
3. The missing data, if any, may be assumed suitably.
4. Tables/Data handbook/Graph paper, etc., if applicable, will be supplied to the candidates

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|---|-----|----|----|
| Q.1(a) Consider the symmetric matrix
$A = \begin{bmatrix} 2.2 & 0.4 \\ 0.4 & 2.8 \end{bmatrix}$ Find the eigenvalues and the corresponding normalized eigenvectors of A. Also, obtain the spectral decomposition of matrix A.
Let the $p \times p$ standard deviation diagonal matrix be $V^{1/2} = \text{Diag}(\sqrt{\sigma_{ii}})$, | [3] | 1 | 2 |
| Q.1(b) $i = 1, 2, \dots, p$, and $V^{1/2} \rho V^{1/2} = \Sigma$, ρ and Σ are correlation and variance-covariance matrices respectively. Suppose you have given variance and covariance matrix Σ , can you obtain the correlation matrix using given expression?
if yes, write down the expression of ρ matrix. | [2] | 1 | 2 |
| Q.2(a) The random vector $X' = [X_1, X_2, \dots, X_5]$ with mean vector $\mu = [2, 4, -1, 3, 0]$ and variance-covariance matrix | [3] | 2 | 3 |
| $\Sigma = \begin{bmatrix} 4 & -1 & 1/2 & -1/2 & 0 \\ -1 & 3 & 1 & -1 & 0 \\ 1/2 & 1 & 6 & 1 & -1 \\ -1/2 & -1 & 1 & 4 & 0 \\ 0 & 0 & -1 & 0 & 2 \end{bmatrix}$, partition X as $X^{(1)} = (X_1, X_2)$ and $X^{(2)} = (X_3, X_4, X_5)$. Consider | | | |
| $A = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & -2 \end{bmatrix}$ (a) Find the $E(AX^{(1)})$ and $E(BX^{(2)})$.
(b) Find the $\text{Cov}(AX^{(1)})$ and $\text{Cov}(BX^{(2)})$. | | | |
| Q.2(b) Prove that the sample mean vector is an unbiased estimator of the population mean vector. | [2] | 1 | 3 |
| Q.3(a) Let $X = (X_1, X_2, X_3)'$ follow a trivariate normal distribution $N_3(\mu, \Sigma)$ with mean vector $\mu' = (2, -3, 1)$ and covariance matrix | [4] | 2 | 3 |
| $\Sigma = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 3 & 2 \\ 1 & 2 & 2 \end{bmatrix}$ Relabel the variables if necessary, and find a 2×1 vector ' a ' such that $X_2 - a'(X_1, X_3)'$ are independent. | | | |
| Q.3(b) What do you mean by generalised variance? | [1] | 2 | 2 |
| Q.4(a) State and prove the Cramer-Wold theorem. Also, explain the importance of this theorem in real-life scenarios. | [3] | 1 | 3 |
| Q.4(b) If X is distributed as $N_p(\mu, \Sigma)$, what can you say about the distribution of any linear combination of variables, say, $a'X$, where a is a real vector of size $p \times 1$? Support your answer with a proper argument and obtain the distribution of $a'X$.
Let X be a p-dimensional multivariate normal distribution with a mean vector μ and variance-covariance matrix Σ , but $\Sigma = \text{diag}(\sigma_{11}, \sigma_{22}, \dots, \sigma_{pp})$. Derive the maximum likelihood estimators of μ and Σ . | [2] | 1 | 3 |
| Q.5(a) | [4] | 2 | 3 |
| Q.5(b) Explain the Wishart distribution. | [1] | 2 | 1 |