

Chennai Mathematical Institute  
Multivariate statistical Analysis

Mid Term Test

Time : 60 Minutes

Max marks: 20

Choose the Best possible answer

1. An advantage of using an experimental multivariate design over separate univariate designs is that using the multivariate analysis is:
- A. Allows you to look at more complex relationships than does univariate strategy  
B. Provides a more powerful test of hypotheses

- a) Both A and B holds ✓  
b) A holds but not B  
c) B holds but not A ✗  
d) Both A and B do not hold ✗

2. If  $X = (X_1, \dots, X_n)$  has a multivariate Normal distribution, then
- A. Each component  $X_i$  has a Normal distribution.  
B. Every subvector of  $X$  has a multivariate Normal distribution.

- a) A is true but not B  
b) B is true but not A ✗  
c) Both A and B are true ✓  
d) Both of them are not true ✗

3. Let  $X$  and  $Y$  be independent standard normal random variables. Define the random variable  $Z$  by

$$Z = X \text{ if } XY > 0 \text{ \& } Z = -X \text{ if } XY < 0$$

- A.  $Z$  is a standard normal random variable  
B. Random Vector  $(Z, Y)$  is not jointly normal

- a) A is true but not B  
b) B is true but not A  
c) Both A and B are true ✓  
d) Both A and B are not true

4. Let  $X \sim N(0, \sigma^2 I)$  and let  $A_1$  and  $A_2$  be symmetric idempotent matrices, then  $X'A_1 X$  and  $X'A_2 X$  are independent, if
- A.  $A_1 A_2 = A_2 A_1 = 0$ .  
B.  $A_1 A_2 = A_2 A_1 = I$

Cochran's theorem?

$$\begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

$$\begin{aligned} a^2 + b^2 &= a \\ b(a+c) &= b \\ b^2 + c^2 &= c \end{aligned}$$

$$\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\begin{aligned} aa' + bb' &= 0 \\ ab' + ba' &= 0 \\ ba' + cb' &= 0 \\ bb' + cc' &= 0 \end{aligned}$$

- a) A is true  
b) B is true  
c) A and B are true  
d) A and B are not true ? (best)

$$E(X'A_1 X - \text{tr}(\sigma^2 A_1))(X'A_2 X - \text{tr}(\sigma^2 A_2)) = 0$$

5. Let  $X \sim N(0, I_n)$ . Then  $X'AX \sim \chi^2(k)$ , and A has rank k, if and only if.

- (C) A. A is symmetric  
B. A is idempotent

- a) B holds but not A  
b) A holds but not B  
c) Both A and B holds ✓  
d) A and B do not hold

6. Examine the following statements

- (D) A. If the components of MVN are not independent, then the vector may not be jointly Normal

B. Uncorrelated jointly normal random variables are independent  $\Rightarrow$  For eg. consider the bivariate density:  
Ind  $\Rightarrow$  uncorrelated

$$f(x, y) = \frac{1}{2\pi\sigma_1\sigma_2(1-\rho)^{1/2}} \exp\left\{-\frac{q}{2}\right\}, -\infty < x < \infty, -\infty < y < \infty$$

$$\text{with } q = \frac{1}{1-\rho^2} \left[ \left( \frac{x-\mu_1}{\sigma_1} \right)^2 - 2\rho \left( \frac{x-\mu_1}{\sigma_1} \right) \left( \frac{y-\mu_2}{\sigma_2} \right) + \left( \frac{y-\mu_2}{\sigma_2} \right)^2 \right]$$

$\Rightarrow$  If  $\rho=0$ , we can write the above density as a product of two normal densities  $\Rightarrow$  RVs are independent...

- a) A holds but not B  
b) B holds but not A  
c) A and B do not hold ✓  
d) A and B holds

7. If Y follows MVN with  $(\mu, \Sigma)$  then the quadratic form  $Y'AY$  follows  $\chi^2(r, \gamma)$  where  $\gamma = \mu'A\mu$  and  $r(A) = r$  if and only if

- (C) A.  $A\Sigma A = \Sigma$   
B.  $A\Sigma$  is idempotent

- a) A holds  
b) B holds  
c) A and B holds ✓  
d) A and B does not hold

8. Examine the following properties on p variate MVN

- (A) A. Invariant under linear transformation  
B. Uncorrelated sub vectors are independent

- a) A and B holds  
b) A holds but not B ✓  
c) B holds but not A  
d) A and B holds under certain other conditions

9. For  $Y \sim N(\mu, \Sigma)$  the following result holds.

- (C) a)  $E(Y'AY) = \text{tr}(A\Sigma)$   
b)  $E(Y'AY) = \mu'A\mu$  ?  
c)  $E(Y'AY) = \text{tr}(A\Sigma) + \mu'A\mu$  (best)  
d)  $E(Y'AY) = \text{tr}(A\Sigma + \mu'A\mu)$

$$E[(Y-\mu)'(Y-\mu)] = \Sigma$$

$$E[Y'Y - Y'\mu - \mu'Y + \mu'\mu] = \Sigma$$

$$E[Y'Y - 2Y'\mu] = \Sigma - 2\mu'\mu$$

$$E[Y'Y] - 2\mu'\mu = \Sigma - 2\mu'\mu$$

$$\therefore E[Y'Y] = \Sigma + \mu'\mu$$

10. The Mahalanobis distance between  $Y$  and  $\mu$  in the metric of  $\Sigma$  is  $D(Y, \mu) = [(Y - \mu)' \Sigma^{-1} (Y - \mu)]^{1/2}$

(C)

- A. Invariant under appropriate linear transformations
- B. Adjusted Euclidean distance in the metric of reciprocal variance

- a) A holds and not B
- b) B holds and not A
- c) A and B holds together ✓
- d) A and B do not hold

11. Let  $Y$  and  $Q$  be independent random variables where  $Y \sim N_p(\mu, \Sigma)$  and  $Q \sim W_p(n, \Sigma)$ , and  $n > p$ . Then Hotelling's  $T^2$  statistic  $T^2 = nY'Q^{-1}Y$  has a distribution proportional to

(D)

- a) Hotelling Distribution
- b) Multivariate Normal Distribution
- c) Noncentral  $\chi^2$  distribution
- d) Noncentral F distribution ✓

12. Let  $X = [X_1, X_2]'$  be a multivariate normal distribution with mean  $\mu = [1 \ 2]'$  and  $\Sigma =$

(B)

$$\begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix}$$

then the random variable  $Y = X_1 + X_2$  is MVN with mean and variance respectively as

- a) 2 and 5
- b) 3 and 7 ✓
- c) 4 and 3
- d) 5 and 2

13. Examine the following statements

(C)

- A. Independent  $\Rightarrow$  uncorrelated (always holds)
- B. Uncorrelated  $\Rightarrow$  Independent (for MVN)

- a) A is true but not B ✓
- b) B is true but not A
- c) A and B are true
- d) A and B are not true

14. Let  $X_1, X_2, \dots, X_n$  be a random sample from  $N_p[\mu, \Sigma]$  then

(C)

- A. Sample mean has a normal distribution and sample covariance matrix has Wishart distribution
- B. Both the sample distributions are independent

- a) A holds but not B
- b) B holds but not A
- c) A and B holds always
- d) A and B holds conditionally ✓

15. Q-Q plot can be used to picture the Mahalanobis distances for the sample and is useful in

(A)

- a) evaluating multivariate normality and outliers ✓
- b) evaluating chi-square distribution and outliers
- c) evaluating Wishart distribution and outliers
- d) detecting outliers only

16. Examine the following statements in PCA

- (A) A. Construction of principal components do not require that the variables in  $Y$  have a multivariate normal distribution.  
 B. Statistical procedure for transforming a MVN distribution into a set of independent univariate normal distributions. Simply  $\Sigma^{-1/2}(Y-\mu)$  suffices

- a) Both A and B are true  
 b) A alone is true ? ✓  
 c) B alone is true  
 d) A and B are untrue

17. Examine the following statements in PC

- (C) A. PC's are invariant to linear transformation to the original variables. If we standardize the data, PC's change!  
 B. PC's are not sensitive to the presence of outliers

- a) Both A and B are true x  
 b) A alone is true x  
 c) B alone is true (best)  
 d) A and B are untrue

Actually, PCA is sensitive to outliers...  
 Ans (D)

$$Y_{ij} \sim N(\mu_{ij}, \dots)$$

$$Y_1 \dots Y_m$$

$$\begin{bmatrix} y_{11} \\ y_{12} \\ \vdots \\ y_{1p} \end{bmatrix} \dots \begin{bmatrix} y_{m1} \\ y_{m2} \\ \vdots \\ y_{mp} \end{bmatrix}$$

18. Let  $Y_i$  be  $m$  independent MVN random  $p$ -vectors with mean  $\mu$  and covariance Matrix  $\Sigma$ ,  $Y_i \sim \text{IN}_p(\mu, \Sigma)$ . Define  $X_j = \sum_{i=1}^m Y_i^2$  for  $j = 1, 2, \dots, p$ . Then the joint distribution of  $X = [X_1, X_2, \dots, X_p]$  is a

- a) Central chi square distribution with  $m$  df  
 b) Noncentral chi square distribution with  $m$  df  $\rightarrow$  if  $\mu \neq 0$ , we can't choose this option  
 c) Central or Noncentral chi square distribution with  $m$  df (Best) ✓  
 d) Central or Noncentral chi square distribution with  $m-1$  df

19. Examine the following statements on Mahalanobis distance

- (A) A. MD is effectively a multivariate equivalent of Euclidean distance  
 B. MD could be used for classification and Outlier detection  
 C. MD is not possible for highly correlated variables

As  $\rho \rightarrow 1$ , actually,  $\Sigma^{-1} \rightarrow$  singular  
 $(X-Y)^T \Sigma^{-1} (X-Y) \rightarrow$  problematic

- a) A, B and C are true  
 b) A, B is true but not C ✓  
 c) A, C is true but not B  
 d) B, C is true but not A

Actually, MD works well even when two or more variables are highly correlated.  
 Ans (B)

20. MD can be calculated using PCs obtained after PC analysis and need not be corrected for covariance as

- a) number of PCs are smaller  
 b) as PCs are easy to calculate ✓  
 c) as PCs are orthogonal ✓  
 d) as PCs use the maximum variability