Indian Statistical Institute, Bangalore B. Math (III) First Semester 2013-2014 End-Semester Examination : Statistics (III) Maximum Score 60

Duration: 3 Hours

- 1. (a) For an $m \times n$ matrix A, define minimum norm g-inverse.
 - (b) Prove that the following two statements are equivalent:
 - i) G is a minimum norm *g*-inverse of A.
 - ii) For any $y \in C(A)$, the column space of A, x = Gy is a minimum norm solution to the system of equations Ax = y.

[2+8=10]

2. Consider the following generalized linear model (GLM)

$$\begin{aligned} \mathbf{Y} &= \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon} \\ E\left(\boldsymbol{\varepsilon}\right) &= \mathbf{0} \text{ and } Var\left(\boldsymbol{\varepsilon}\right) = \sigma^{2} \mathbf{I}_{n}. \end{aligned}$$

- (a) Write down Γ , the class of all solutions to the normal equations $\mathbf{X}'\mathbf{X}\boldsymbol{\beta} = \mathbf{X}'\mathbf{Y}$.
- (b) Define estimability of a parametric function $l'\beta$.
- (c) Show that $l'\beta$ is estimable iff $l'\gamma$ is linear in **Y** and is same for all $\gamma \in \Gamma$.
- (d) Show that $l'\beta$ is estimable iff l'(I K) = 0 where $K = (X'X)^{-}X'X$.

[3+2+8+7=20]

3. Consider the following model

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$$\begin{aligned} Y_1 &= \theta_2 - \theta_1 + \varepsilon_1 \\ Y_2 &= 2\theta_2 + \varepsilon_2 \\ Y_3 &= \theta_1 + \theta_2 + \varepsilon_3, \end{aligned}$$

where the errors $\varepsilon_1, \varepsilon_2, \varepsilon_3$ are *iid* $\mathcal{N}(0, \sigma^2)$.

- (a) Is $\theta_2 \theta_1$ estimable?
- (b) Derive a test for

$$H_0: \theta_2 - \theta_1 = 0 \ Versus \ H_1: \theta_2 - \theta_1 \neq 0. \tag{1}$$

(c) For y' = (2, 1, 3) carry out a suitable test for the testing problem in (1).

[2+8+8=18]

[PTO]

4. Consider the following linear regression model

$$egin{array}{rcl} m{Y} &=& m{X}m{eta}+m{arepsilon} \ m{arepsilon} & & \ m{arepsilon} & & \ m{arepsilon} & \ m{arepsilon} & \ m{\mathcal{N}}\left(m{0},\sigma^2m{I}_n
ight) \end{array}$$

Let us partition the full column rank matrix X as (X_1, X_2) , and β' as (β'_1, β'_2) conformally. Derive a test for

$$H_0: \boldsymbol{\beta}_2 = \mathbf{0} \ Versus \ H_1: \boldsymbol{\beta}_2 \neq \mathbf{0}$$

using the notion of extra sum of squares.

[10]

- 5. Let us consider $(1+k) \times 1$ random vector $\begin{pmatrix} Y \\ \mathbf{X} \end{pmatrix}$ with a *positive definite* variance covariance matrix $\mathbf{V} = \begin{pmatrix} \sigma_y^2 & \mathbf{\delta}' \\ \mathbf{\delta} & \mathbf{\Sigma} \end{pmatrix}$. Here $Var(Y) = \sigma_y^2$, and \mathbf{X} is a $k \times 1$ random vector with variance covariance matrix $\mathbf{\Sigma}$. $\mathbf{X}' = (X_1, X_2, \cdots, X_k)$.
 - (a) Obtain $\rho^2_{Y(X_1, X_2, \dots, X_k)}$, $\rho_{Y(X_1, X_2, \dots, X_k)}$ being the multiple correlation coefficient between Y and (X_1, X_2, \dots, X_k) .
 - (b) Let w_{11} be the (1,1) element of V^{-1} then express $\rho^2_{Y(X_1,X_2,\cdots,X_k)}$ in terms of w_{11} .

[6+6=12]