

Class Activity Problems

Probability and Statistics 2022
Indian Institute of Management, Udaipur

21st July, 2022

Name:

Name:

- There are a total of 114 points in this Question Paper. Answer as much as you can. If your acquired score is greater than equal to 100, it will be counted as 100%.
- The Exam is scheduled for 3 hours. "Time Left" reminders will be posted in 1.5 hrs, 2:30 hrs, 2:45 hrs from the beginning of the Exam time.
- There are three ★ marked problems that are more involved than the rest. In case you are stuck in one of those, it might be a good idea to consider solving other problems first and then continue with the * marked problems.
- You may take help from the "Exam Assistance Note" containing a few required definitions, lemma and theorem statements.

Let Y be Response variable, X_1, X_2 denote the Explanatory variables, ε be unknown random errors and β_1, \dots, β_3 are unknown parameters of interest. Determine whether the following relationship equation is a linear model. Relationship Equation: $Y = \beta_0 + \beta_1 \sin(X_1) + \beta_2 X_2 + \beta_3 X_1 X_2 + \varepsilon$.

1. (a)

Score:
Total Score: 5

Ans:

☐ Not Linear Model

☐ Linear Model

Identify if the following matrix is a Orthogonal Projection matrix.

$$M = \frac{1}{9} \begin{bmatrix} 3 & 2 & 4 \\ 2 & 1 & 2 \\ 4 & 2 & 1 \end{bmatrix}$$

(b)

Score:
Total Score: 5

Ans:

☐ Orthogonal Projection

☐ Not Orthogonal Projection

Consider the matrices $A = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$. Does the matrix A have a Generalized Inverse ? If yes, Construct a generalized inverse of A .

Score:
Total Score: 1+4

(c) Ans:

Consider the model $\underline{\mathbf{y}} = \mathbf{X}\underline{\boldsymbol{\beta}} + \underline{\boldsymbol{\varepsilon}}$, where $\underline{\boldsymbol{\beta}} = \begin{bmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \end{bmatrix}$, $\mathbf{X} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & -1 \end{bmatrix}$

(d) and $\underline{\boldsymbol{\varepsilon}}$ is a mean zero error vector with variance co-variance matrix $\sigma^2 I_{4 \times 4}$.

Prove that the parameter β_1 is not estimable

Score:
Total Score: 7

Ans:

A spring balance is used to weigh three objects with unknown weights $\beta_1, \beta_2, \beta_3, \beta_4$. The objective is to estimate the $\underline{\beta} = [\beta_1 \ \beta_2 \ \beta_3 \ \beta_4]^T$. Assume that each the measurements in the spring balance are subject to (independent) Normally distributed random errors with mean 0 and unknown variance $\sigma^2 > 0$. Consider the data, when each object are weighted twice to get observations

2.

$$\underline{Y} = [y_{1,1} \ y_{1,2} \ y_{1,3} \ y_{2,1} \ \dots \ y_{4,2} \ y_{4,3}]^T,$$

where $y_{i,j}$ denotes the j^{th} replicated measurement corresponding to the experiment when i^{th} object is placed on the spring balance. Note that $\underline{Y} \in \mathbb{R}^{12}$.

Represent the estimation problem in terms of the matrix notation $\underline{Y} = \underline{X}\underline{\beta} + \underline{\varepsilon}$ by identifying the response vector \underline{Y} , design matrix \underline{X} , regression coefficients $\underline{\beta}$ and the vectors of the random errors $\underline{\varepsilon}$.

(a)

Score:
Total Score: 2+ 5+2+1

Ans:

Prove that all the linear parametric functions of the parameters $\underline{\beta}$ are estimable for the associated design matrix.

(b)

Score:
Total Score: 5

Ans:

Find the **Orthogonal Projection Matrix** for $\mathcal{C}(\mathbf{X})$, the column space of \mathbf{X} . (Justify your steps with the reference to the results/theorem/lemma you are using.)

(c)

Score:
Total Score: 8

Ans:

(d) Consider a linear parametric function $\theta = \beta_1 + \beta_2 + \beta_3$. Identify a vector $\underline{\lambda}$ such that $\theta = \underline{\lambda}^T \underline{\beta}$.

Score:
Total Score: 2

Ans:

- (e) Find the Best Linear Unbiased Estimator for θ (Show your steps and justify your steps with the reference to the results/theorem/lemma you are using.)

Score:
Total Score: 7

Ans:

- (f) A practitioner is interested in estimating all the parameters $\beta_1, \beta_2, \beta_3, \beta_4$ simultaneously. write down the definition of $(1 - \alpha)100\%$ simultaneous confidence intervals for the parameters.

Score:
Total Score: 5

Ans:

- (g) ★ Which type of simultaneous Confidence interval would you prefer for the parameters $\{\beta_1, \beta_2, \beta_3, \beta_4\}$ and why?

Score:
Total Score: 5

Ans:

- (h) Construct a set of 95% simultaneous confidence intervals for the parameters $\{\beta_i\}_{i=1}^4$.

Score:
Total Score: 8

Ans:

3. Consider the standard linear model $\underline{\mathbf{Y}} = \underline{\mathbf{X}}\underline{\beta} + \underline{\varepsilon}$ where $\underline{\varepsilon}$ is Normally distributed ($\underline{\varepsilon} \sim N(\underline{\mathbf{0}}, \sigma^2 V)$), where V is a known positive definite matrix. Assume the design matrix $\underline{\mathbf{X}}$ has full column rank. Also, assume that σ is a known positive constant.

Show that the log likelihood function can be represented as

(a)
$$l_{\underline{\mathbf{Y}}}(\underline{\beta}) = K(V, \sigma^2) - \frac{(\underline{\mathbf{Y}} - \underline{\mathbf{X}}\underline{\beta})^T V^{-1} (\underline{\mathbf{Y}} - \underline{\mathbf{X}}\underline{\beta})}{2\sigma^2}$$

where $K(V, \sigma^2)$ is a constant that does not involve $\underline{\beta}$.

Score:
Total Score: 5

Ans:

- (b) Use part(a) to derive $\hat{\beta}_{MLE}$, the Maximum Likelihood Estimator for the parameter β .

Score:
Total Score:7

Ans:

Consider a linear model given by

$$\mathbf{Y} = \mathbf{X}\tilde{\beta} + \tilde{\varepsilon}$$

4. where $\mathbf{Y} \in \mathbb{R}^n, \tilde{\beta} \in \mathbb{R}^p$ and $\tilde{\varepsilon}$ is a Normally distributed random vector with mean $\mathbf{0}$ and variance $\sigma^2 I_{n \times n}$ and \mathbf{X} is an $n \times p$ matrix with rank $r < p < n$ (i.e. The design matrix \mathbf{X} does not have full column rank). The parameter σ^2 is an unknown positive number. Assume that $\tilde{\varepsilon} \sim N(\mathbf{0}, \sigma^2 I_{n \times n})$. Let $\mathcal{C}(\mathbf{X})$ denotes the column space of \mathbf{X} , the vector space containing all possible linear combination of the column vectors of the matrix \mathbf{X} . It can be shown that $P = \mathbf{X}(\mathbf{X}^T \mathbf{X})^- \mathbf{X}^T$ is the **Orthogonal Projection** matrix for the $\mathcal{C}(\mathbf{X})$, where $(\mathbf{X}^T \mathbf{X})^-$ is any Generalized inverse of $(\mathbf{X}^T \mathbf{X})$.

Show that $E(\mathbf{Y}^T (I_{n \times n} - P) \mathbf{Y}) = (n - r) \sigma^2$.

(a)

Score:
Total Score: 5

Ans:

Prove that the statistics $\mathbf{Y}^T \mathbf{P} \mathbf{Y}$ and $\mathbf{Y}^T (\mathbf{I}_{n \times n} - \mathbf{P}) \mathbf{Y}$ are independent (Mention if you are using any result).

(b)

Score:
Total Score:5

Ans:

Derive the distribution of $\frac{(\mathbf{Y}^T P \mathbf{Y})/r}{(\mathbf{Y}^T (I_{n \times n} - P) \mathbf{Y})/(n-r)}$? (Show your steps)

(c)

Score:
Total Score: 10

Ans:

Let $\hat{\beta}_{LSE}$ be the **Least Square Estimator** for the parameter β . Show that

$$(\mathbf{Y} - \mathbf{X}\hat{\beta}_{LSE})^T P(\mathbf{Y} - \mathbf{X}\hat{\beta}_{LSE}) = (\hat{\beta}_{LSE} - \beta)^T \mathbf{X}^T \mathbf{X} (\hat{\beta}_{LSE} - \beta).$$

(d) ★

Score:
Total Score: 5

Ans:

(e) ★

Derive the distribution of $\frac{\|X(\hat{\beta}_{LSE} - \beta)\|^2}{\sigma^2}$. (You may use the relation in part(d) to get your answer.)

Score:
Total Score: 5

Ans: