

**Indian Institute of Technology Guwahati**  
**Statistical Inference (MA682)**  
**Problem Set 07**

1. Consider the simple linear regression model  $y = \beta_0 + \beta_1 x + \varepsilon$ , with usual assumptions on  $\varepsilon$ . Show that

$$\text{Cov}(\widehat{\beta}_0, \widehat{\beta}_1) = -\frac{\bar{x}\sigma^2}{S_{xx}} \quad \text{and} \quad \text{Cov}(\bar{y}, \widehat{\beta}_0) = \frac{\sigma^2}{n}.$$

2. Consider the simple linear regression model  $y = \beta_0 + \beta_1 x + \varepsilon$  with usual assumptions on  $\varepsilon$ . Show that

$$E(MS_R) = \sigma^2 + \beta_1^2 S_{xx} \quad \text{and} \quad E(MS_{Res}) = \sigma^2.$$

3. Suppose that we have fit a simple linear regression model  $\widehat{y} = \widehat{\beta}_0 + \widehat{\beta}_1 x_1$ , but the response is affected by a second variable  $x_2$  such that the true regression is  $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon$ . Is the least square estimator  $\widehat{\beta}_1$  in the simple linear regression model unbiased?
4. Consider the simple linear regression model  $y = \beta_0 + \beta_1 x + \varepsilon$ , where  $\varepsilon$ 's are independent and identically  $N(0, \sigma^2)$  random variables. Find the MLEs of  $\beta_0$ ,  $\beta_1$ , and  $\sigma^2$ . Is the MLE of  $\sigma^2$  UE?
5. Suppose that we are fitting a straight line and wish to make standard error of the slope as small as possible. Suppose that the region of interest for  $x$  is  $-1 \leq x \leq 1$ . Where should the observations  $x_1, x_2, \dots, x_n$  be taken?
6. Consider the simple linear regression model  $y = \beta_0 + \beta_1 x + \varepsilon$ , with usual assumptions on  $\varepsilon$ . Also assume that  $\beta_0$  is known. Find the LSE of  $\beta_1$  and its' variance.
7. Consider the multiple linear regression model

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \varepsilon.$$

Using the procedure for testing a general linear hypothesis, show how to test

- (a)  $H_0 : \beta_1 = \beta_2, \beta_3 = \beta_4$ .  
 (b)  $H_0 : \beta_1 - 2\beta_2 = 4\beta_3, \beta_1 + 2\beta_2 = 0$ .

8. Show that  $\text{Var}(\widehat{\mathbf{y}}) = \sigma^2 H$ .
9. Consider the multiple linear regression model  $\mathbf{y} = X\boldsymbol{\beta} + \boldsymbol{\varepsilon}$ . Show that LSE of  $\boldsymbol{\beta}$  can be written as  $\widehat{\boldsymbol{\beta}} = \boldsymbol{\beta} + R\boldsymbol{\varepsilon}$ , where  $R = (X'X)^{-1} X'$ .
10. Prove that  $R^2$  is the square of the correlation between  $\mathbf{y}$  and  $\widehat{\mathbf{y}}$ .
11. Consider the linear regression  $y = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \epsilon$ , with usual assumptions on  $\epsilon$ . Assuming that  $X$  is a full column rank matrix, show that

$$\sum_{i=1}^n \text{Var}(\widehat{y}_i) = (p+1)\sigma^2.$$