

Regression Analysis Assignment1

November 1, 2021

Assumption. $\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$, where $\mathbf{Y} = (Y_1, Y_2, \dots, Y_n)^t$, $\mathbf{1} = (1, \dots, 1)^t$, $\mathbf{x}_j = (x_{1j}, \dots, x_{nj})^t$, $\boldsymbol{\epsilon} = (\epsilon_1, \dots, \epsilon_n)^t$ are n -dimensional vectors, $\boldsymbol{\beta} = (\beta_0, \dots, \beta_n)^t$ is a $(p+1)$ -dimensional vector, $\mathbf{X} = (\mathbf{1}, \mathbf{x}_1, \dots, \mathbf{x}_p)$ is a $n \times (p+1)$ known matrix of full column rank and $\boldsymbol{\epsilon} \sim \mathbf{N}_n(0, \sigma^2 I_n)$ with $\sigma^2 > 0$.

Problem1. Let \mathbf{A} be a $(p+1) \times q$ matrix of full column rank and \mathbf{c} be q -dimensional vector ($q < p+1$). Let $\hat{\boldsymbol{\beta}}_{\mathbf{r}}$ be the least squared estimator with constraint $\mathbf{A}^t \boldsymbol{\beta} = \mathbf{c}$. It means

$$\hat{\boldsymbol{\beta}}_{\mathbf{r}} = \arg \min_{\mathbf{A}^t \boldsymbol{\beta} = \mathbf{c}} \|\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}\|$$

Find the $\hat{\boldsymbol{\beta}}_{\mathbf{r}}$ using the Lagrangian multiplier method.

Problem2. Let \mathbf{X}_i be i -th row of \mathbf{X} and $\hat{\boldsymbol{\beta}}$ be the least squared estimator based on $(\mathbf{X}_1, Y_1), \dots, (\mathbf{X}_n, Y_n)$. Suppose $Y_i = \mathbf{X}_i^t \hat{\boldsymbol{\beta}}$ for $i = 1, 2, \dots, k$. Then, show that the least squared estimator $\hat{\boldsymbol{\beta}}_{(-)}$ based on $(\mathbf{X}_{k+1}, Y_{k+1}), \dots, (\mathbf{X}_n, Y_n)$ equals to $\hat{\boldsymbol{\beta}}$.

Problem3. Derive $100(1 - \alpha)\%$ confidence interval for mean response and out-of-sample response, referring to slide 16.