

## [STAT409] Homework 2

1. In the multiple linear regression model:

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{e}$$

where the definition of vectors are given in the lecture note.

- (a) Show that LSE of  $\boldsymbol{\beta}$ , the minimizer of  $S(\boldsymbol{\beta})$  is given by

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y}. \quad (1)$$

- (b) Prove the following decomposition (a.k.a ANOVA decomposition) in the multiple linear regression.

$$\underbrace{(\mathbf{y} - \bar{\mathbf{y}})^T (\mathbf{y} - \bar{\mathbf{y}})}_{\text{SST}} = \underbrace{(\mathbf{y} - \hat{\mathbf{y}})^T (\mathbf{y} - \hat{\mathbf{y}})}_{\text{SSE}} + \underbrace{(\hat{\mathbf{y}} - \bar{\mathbf{y}})^T (\hat{\mathbf{y}} - \bar{\mathbf{y}})}_{\text{SSR}}$$

2. In the simple linear regression  $y_i = \beta_0 + \beta_1 x_i + \epsilon_i$  where  $\epsilon_i \stackrel{iid}{\sim} N(0, \sigma^2)$ , we have the following OLS slope estimators:

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (x_i - \bar{x}_n)(y_i - \bar{y}_n)}{\sum_{i=1}^n (x_i - \bar{x}_n)^2}, \quad (2)$$

- (a) Compute  $E(\hat{\beta}_1)$  and  $Var(\hat{\beta}_1)$ .  
 (b) It is well known that  $\hat{\beta}_1$  is normally distributed when  $\epsilon_i$ 's are normally distributed. Justify this.
3. For a given  $y_1, \dots, y_n \stackrel{iid}{\sim} F$ , show that the sample median is the minimizer of  $A_n(c)$  where

$$A_n(c) = \frac{1}{n} \sum |y_i - c|$$

(hint: You may use the fact that the derivative of  $|x|$  is simply  $\text{sign}(x)$  and ignore its nondifferentiability at  $x = 0$ .)

4. Suppose  $X_1 \sim \text{Binomial}(n_1, \theta)$  and  $X_2 \sim \text{Binomial}(n_2, \theta)$  are independent random variables.

- (a) Write down the log likelihood of  $\theta$ , denoted by  $\ell(\theta)$ .  
 (b) Compute the maximum likelihood estimator of  $\theta$ , denoted by  $\hat{\theta}$ .  
 (c) Compute the information of  $\theta$ :

$$I(\theta) = E \left[ \left\{ \frac{\partial \ell(\theta)}{\partial \theta} \right\}^2 \right] = -E \left[ \frac{\partial^2 \ell(\theta)}{\partial \theta^2} \right]$$