**P 1.** Let  $R_{1,Adj}^2$  denotes the adjusted  $R^2$  when a regressor say  $x_1$  is removed from the model consist of p-1 regressors. Let F be the test statistics for testing the hypothesis  $H_0: \beta_1 = 0$  versus  $\beta_1 \neq 0$ . Show that  $R_{Adj}^2 \geq R_{1,Adj}^2$  if and only if  $F \geq 1$  where  $R_{Adj}^2$  denotes the adjusted  $R^2$  for the full model.

P 2. Consider the following regression model

$$y = X\beta + \epsilon \tag{1}$$

Define

$$X_A = \begin{pmatrix} X \\ \sqrt{k} \mathbf{I}_p \end{pmatrix} \qquad \mathbf{y}_A = \begin{pmatrix} \mathbf{y} \\ \mathbf{0}_p \end{pmatrix}$$

where  $I_p$  is a square matrix of order  $p \times p$  and  $\mathbf{0}_p$  is a  $p \times 1$  vector of zeros. Show that the ridge estimator is the LSE for the following model

$$y_A = X_A \beta_R + \epsilon.$$

- **P 3.** Consider the simple linear regression model  $y_i = \beta_0 + \beta_1 x + \epsilon_i$ , where  $Var(\epsilon_i) = \sigma^2 x_i^2$ .
  - (a) Suppose that we use the transformations y' = y/x and x' = 1/x. Is this a variance–stabilizing transformation?
  - (b) What are the relationships between the parameters in the original and transformed models?
  - (c) Suppose we use the method of weighted least squares with  $w_i = 1/x_i^2$ . Is this equivalent to the transformation introduced in part (a)?
- P 4. Consider the following one-way ANOVA model

$$y_{ij} = \mu + \tau_i + \epsilon_{ij}$$
;  $i = 1, 2, 3$ , and  $j = 1, \dots, n$ .

where  $\mu$  is a parameter common to all 3 treatments (usually called the grand mean),  $\tau_i$  is a parameter that represents the effect of the *i*th treatment, and  $\epsilon_{ij}$  is an NID(0,  $\sigma^2$ ) error component. The associated regression model is

$$y_{ij} = \beta_0 + \beta_1 x_{1j} + \beta_2 x_{2j} + \epsilon_{ij}; \quad i = 1, 2, 3, \text{ and } j = 1, \dots, n.$$

where

$$x_{1j} = \begin{cases} 1 & \text{if observation is from treatment 1} \\ -1 & \text{if observation is from treatment 2} \\ 0, & \text{otherwise} \end{cases}$$

and

$$x_{2j} = \begin{cases} 1 & \text{if observation is from treatment 2} \\ -1 & \text{if observation is from treatment 3} \\ 0, & \text{otherwise} \end{cases}$$

(a) Show that the relationship between the parameters in the regression and analysis - of - variance models is

1

$$\beta_0 = \frac{\mu_1 + \mu_2 + \mu_3}{3} = \bar{\mu}$$

$$\beta_1 = \mu_1 - \bar{\mu}, \ \beta_2 = \mu_2 - \bar{\mu},$$

where  $\mu_i = \mu + \tau_i$ .

(b) Write down the vector  $\boldsymbol{y}$  and matrix X.

- (c) Develop an appropriate sum of squares for testing the hypothesis  $H_0: \tau_1 = \tau_2 = \tau_3 = 0$ . Is this the usual treatment sum of squares in the one- way analysis of variance?
- P 5. Consider the one-way ANOVA model

$$y_{ij} = \mu + \alpha_i + \epsilon_{ij}, \ i = 1, \dots, K, \ j = 1, \dots, n_i$$
 (2)

where  $\epsilon_{ij}$  are i.i.d.  $N(0, \sigma^2)$ .

- (a) Find the residual sum of squares (RSS) for the model (2).
- (b) Find the distribution of  $RSS/\sigma^2$ .
- **P 6.** Consider the ANOVA model (2). The regression sum of squares  $SS_{Reg}$  is defined as

$$SS_{\text{Reg}} = \sum_{i=1}^{K} \sum_{j=1}^{n_i} (\hat{y}_{ij} - \bar{y}),$$
 (3)

where  $\hat{y}_{ij} = \hat{\mu} + \hat{\alpha}_i$ ,  $\hat{\mu}$  and  $\hat{\alpha}_i$  are least squares estimators of  $\mu$  and  $\alpha_i$ ,  $\bar{y} = \sum_{i=1}^K n_i \bar{y}_i / \sum_{i=1}^K n_i$  and  $\bar{y}_i = \sum_{j=1}^{n_i} y_{ij}$  for  $i = 1, \dots, K$ . Find the distribution of  $SS_{\text{Reg}}/\sigma^2$ .

Remark: The hypothesis

$$H_0: \alpha_1 = \cdots = \alpha_K$$
 versus  $H_1: \alpha_i \neq \alpha_j$  for at least one pair of  $(i,j)$ 

can tested using the statistics

$$F = \frac{RSS/df_1}{SS_{\text{Reg}}/df_2} \sim F_{df_1, df_2},$$

where  $df_1$  and  $df_2$  are degrees of freedoms associated to the distributions of RSS and  $SS_{Reg}$  respectively.