

## APPENDIX B. DETAILED STEPS OF LINEAR REGRESSION

With the regression model:  $Y_i = \beta_0 + \beta_1 X_{F,i} + \beta_2 X_{C,i} + \beta_3 X_{R,i} + \varepsilon_i$

Our aim is to estimate the coefficients  $\beta_0, \beta_1, \beta_2$  and  $\beta_3$ .

$$\text{Let } \mathbf{Y} = \begin{pmatrix} Y_1 \\ \vdots \\ Y_{10} \end{pmatrix}, \mathbf{X} = \begin{pmatrix} 1 & X_{F,1} & X_{C,1} & X_{R,1} \\ \vdots & \vdots & \vdots & \vdots \\ 1 & X_{F,10} & X_{C,10} & 10 \end{pmatrix}, \boldsymbol{\beta} = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \\ \beta_3 \end{pmatrix}, \boldsymbol{\varepsilon} = \begin{pmatrix} \varepsilon_1 \\ \vdots \\ \varepsilon_{10} \end{pmatrix}$$

$$\therefore \mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

**Note:**  $\varepsilon_i \sim N(0, \sigma^2)$  and  $Y_i \sim N(\mathbf{X}_i^T \boldsymbol{\beta}, \sigma^2)$

The Residual Sum of Squares (RSS) is given by:

$$\text{RSS} = \sum_{i=1}^{10} (Y_i - \mathbf{X}_i^T \boldsymbol{\beta})^2 \text{ and } \mathbf{X}_i = \begin{pmatrix} 1 \\ X_{i1} \\ X_{i2} \\ X_{i3} \end{pmatrix}, i = 1, \dots, 10$$

RSS is equivalent to the sum of the square of the residuals:

$$\sum_{i=1}^{10} \varepsilon_i^2 = (\varepsilon_1 \ \dots \ \varepsilon_{10}) \begin{pmatrix} \varepsilon_1 \\ \vdots \\ \varepsilon_{10} \end{pmatrix} = \boldsymbol{\varepsilon}^T \boldsymbol{\varepsilon} = (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})^T (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}) \text{ where } \boldsymbol{\varepsilon} = (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})$$

$$\therefore \text{RSS} = (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})^T (\mathbf{Y} - \mathbf{X}\boldsymbol{\beta})$$

The RSS is minimised to obtain the regression coefficients, given by:

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

The covariance matrix is

$$\text{Var}(\hat{\boldsymbol{\beta}}) = \sigma^2 (\mathbf{X}^T \mathbf{X})^{-1} \text{ where } \hat{\sigma}^2 = \frac{\text{RSS}}{df} \text{ and } t\text{-statistic for } \beta_j = \frac{\hat{\beta}_j}{\sqrt{\text{Var}(\hat{\beta}_j)}}$$

where the degree of freedom  $df = n - p - 1$  where  $n$  is the total number of data points and  $p$  is the number of predictor variables  $\mathbf{X}_i$  not including the intercept term.

The model selection criterion is based on the values of AIC and BIC that are defined as follows:

$$\text{AIC} = 2(p + 1) - 2 \text{Log}(L)$$

$$\text{BIC} = (p + 1) \text{Log}(n) - 2 \text{Log}(L)$$

where  $\text{Log}(L)$  is the log-likelihood of the regression model.