

Data Analytics

110-2 Homework #03 Due at 23h59, March 13, 2022; files uploaded to NTU-COOL

- 1. (10%) Given a simple linear regression model: $y_i = \beta_0 + \beta_1 x_i + \epsilon_i$, i = 1, ..., n, where $\epsilon_i \sim_{iid} N(\mu, \sigma^2)$ Show
 - a. $cov(\hat{\beta}_0, \hat{\beta}_1) = -\bar{x}\sigma^2/S_{xx}$ b. $cov(\bar{y}, \hat{\beta}_1) = 0$
- 2. (10%) Show that the regression sum of squares can be calculated as:

$$SS_R = \left(\sum_{i=1}^n \hat{y}_i^2\right) - n\bar{y}^2$$

- 3. (10%) The matrix, $\mathbf{X}(\mathbf{X}^T\mathbf{X})^{-1}\mathbf{X}^T$, derived in multiple regression is usually defined as **H**. Show that:
 - a. H is idempotent, i.e., HH = H and (I H)(I H) = I H
 - b. $V(\hat{\mathbf{y}}) = \sigma^2 \mathbf{H}$
- (10%) Investigate and explain why R^2 cannot be larger than 1 or smaller than 0. (Do not copy directly from the source you found, but explain in your own words.)
- (15%) In a multiple regression model: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$, it is critical to know if $(\mathbf{X}^T\mathbf{X})^{-1}$ exists. The diagonal elements of $(\mathbf{X}^T\mathbf{X})^{-1}$ in correlation form, i.e., \mathbf{X} is normalized, are often called Variance Inflation Factors (VIFs), and they are important multicollinearity diagnostic. VIF for the j^{th} regression coefficient is expressed as

$$VIF_j = \frac{1}{1 - R_i^2},$$

where R_i^2 is the coefficient of multiple determination obtained from regressing \mathbf{x}_i on the other regressor variables (\mathbf{x}_1 to \mathbf{x}_p , except \mathbf{x}_i). Calculate all the VIFs in the autompg dataset and discuss your observation.