



Data Analytics

111-2 Homework #03

Due at 23h59, March 19, 2023; files uploaded to NTU-COOL

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

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 \mathbf{H} . Show that:
- \mathbf{H} is idempotent, i.e., $\mathbf{H}\mathbf{H} = \mathbf{H}$ and $(\mathbf{I} - \mathbf{H})(\mathbf{I} - \mathbf{H}) = \mathbf{I} - \mathbf{H}$
 - $V(\hat{\mathbf{y}}) = \sigma^2 \mathbf{H}$
4. (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.)
5. (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

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

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