

## **Data Analytics**

## 109-2 Homework #04 Due at 23h59, March 28, 2021; 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.