- Due: October 11, Wednesday, 11:59PM
- How to submit: via Blackboard. If you have multiple files, upload a zipped file
- Submission link will disappear after 48 hours
- Homework solution is not required to be typed, but must be legible.

Problem 1 A student fitted a linear regression function for a class assignment. The student plotted the residuals e_i against y_i , and found a positive relation. When the residuals were plotted against the fitted values \hat{y}_i , the student found no relation. How could this difference arise? Which is the more meaningful plot? Explain why.

Problem 2 \hat{y}_0 is the fitted value or predicted value of the response when the predictor is x_0 .

- 1) Derive the variance of \hat{y}_0 , $\sigma^2\{\hat{y}_0\}$, in a Simple Linear Regression setting when there is only one predictor.
- 2) Derive the variance of \hat{y}_0 , $\sigma^2\{\hat{y}_0\}$, in a Multiple Linear Regression setting when there are p predictors.
- 3) Derive the variance of the prediction error, $\sigma^2\{y_0 \hat{y}_0\}$, in a Multiple Linear Regression setting when there are p predictors.

Problem 3 For a multiple linear regression model with 4 variables,

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \beta_4 X_4 + \epsilon,$$

where $\epsilon \sim N(0, \sigma^2)$, what is the test statistic for each of the following tests? and what is the probability distribution for each test statistic? Assume there are n data points.

- 1) testing whether or not $\beta_4 = 0$
- 2) testing whether or not $\beta_2 = \beta_3 = 0$
- 3) testing if all slopes are zero

Problem 4 For multiple linear regression with n data points and p predictors, prove that $\sum_{i=1}^{n} Var(\hat{y}_i) = (p+1)\sigma^2$, where σ^2 is the common variance of the error term in the MLR model.

Problem 5 Let H be the hat matrix for multiple linear regression with p predictors and n data points. The notation h_{ij} represents H[i][j], the element of H at the i-th row and the j-th column.

- 1) Show that $h_{ii} \in [0,1]$. 2) Show that $\sum_{i=1}^{n} h_{ii} = p+1$.
- 3) Plot h_{ii} vs. $|x_i \bar{x}|$ (absolute value) for p = 1, and plot h_{ii} vs $||x_i \bar{x}||_2$ (2-norm) for p > 1. Use the data generated by your own code, and let n = 100, p = 4.