- 1. (10 points) Please write down the five major assumptions in Regression Analysis.
- 2. (35 points) Consider a no-intercept model, i.e.

$$y = \beta_1 x + \beta_2 x^2 + \varepsilon,$$

where  $\varepsilon$  is a normal distribution with mean 0 and variance  $\sigma^2$ . Additionally we also assume that the errors are uncorrelated. Given *n* observations,  $(y_i, x_i), i =$ 

1, 2, ..., n, n > 2. Let 
$$Y = (y_1, \dots, y_n)'$$
 and  $X = \begin{pmatrix} x_1 & x_1 \\ \vdots & \vdots \\ x_n & x_n^2 \end{pmatrix}$ .

- **a.** Find the least-square estimator of  $\beta = (\beta_1, \beta_2)', \hat{\beta}$ .
- **b.** Show  $\hat{\beta}$  is an unbiased estimator of  $\beta$  and find the covariance matrix of  $\hat{\beta}$ .
- c. Give a geometrical interpretation of this least-square estimator.
- **d.** Please write down  $SS_{Res}$  in terms of Y and X, and find the unbiased estimator of  $\sigma^2$ .
- e. Find the MLE of  $\beta$  and compare this MLE with the least-square estimator.
- **3.** (15 points) Assume the model is

$$Y = X\beta + \varepsilon$$

where  $Y = (y_1, \dots, y_n)'$  is the vector of the observations;  $X = \begin{pmatrix} x_{11} & x_{12} & \dots & x_{1p} \\ x_{21} & x_{22} & \dots & x_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \dots & x_{np} \end{pmatrix}$ 

is the design matrix;  $\beta = (\beta_1, \dots, \beta_p)'$  is the vector of the unknown parameters, and  $\varepsilon = (\varepsilon_1, \dots, \varepsilon_n)'$  is a vector of errors with mean vector, **0**, and covariance matrix,  $\sigma^2 I_n$ . Let  $\hat{y}_{(i)}$  be the fitted value of the *i*th response based on all observations except the *i*th one, and define

$$e_{(i)} = y_i - \hat{y_{(i)}}$$

to be the ith prediction errors. Show that

$$e_{(i)} = \frac{e_i}{1 - h_{ii}},$$

where  $e_i$  is the original *i*th residual and  $h_{ii}$  is the *i*th diagonal elements of hat matrix.

(Hits: Let  $X_{(i)}$  represent the original X matrix with the *i*th row  $x_i$  withheld.  $h_{ii} = x'_i (X'X)^{-1} x_i$  and  $[X'_{(i)}X_{(i)}]^{-1} = (X'X)^{-1} + \frac{(X'X)^{-1} x_i x'_i (X'X)^{-1}}{1 - h_{ii}}$ .)

- 4. (15 points) Consider the normal probability plot.
  - **a.** How to construct this plot?
  - **b.** What is the purpose of this plot?
  - **c.** If the observations come from a heavy-tailed distribution, then please show the corresponding normal probability plot.
- 5. (15 points) Give 32 observations,  $y_1, \dots, y_{32}$ , and assume the model is  $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \varepsilon$ . Please fill the following ANOVA table up.

Source Variation	Sum of Square	Degree of Freedom	Mean Square	$F_0$
Regression	972.9			
Residual				
Total	1237.54			

6. (10 points) Consider the following two models where  $E(\varepsilon) = \mathbf{0}$  and  $\operatorname{Var}(\varepsilon) = \sigma^2 I$ : Model A:  $y = X_1\beta_1 + \varepsilon$ Model B:  $y = X_1\beta_1 + X_2\beta_2 + \varepsilon$ . Show that  $R_A^2 \leq R_B^2$ .