

EE325 Section 1 HW 2 Due Thursday February 20<sup>th</sup> (23:00 hr.), 2020

Use 4 decimal places for numerical answers

1. In Table 1.a.  $X_i$  is total microeconomics exam point (total points are 100) and  $Y_i$  is GPA of each student.

Table 1.a

Student	$Y_i$	$X_i$
1	2.8	63
2	3.4	72
3	3	78
4	3.5	81
5	3.6	87
6	3.0	75
7	2.7	75
8	3.7	90

$$\bar{Y} = \frac{1}{N} \sum_{i=1}^n Y_i = \frac{1}{8} (25.7) = 3.2125$$

$$\bar{X} = 77.625$$

1.1 Now consider the two-variable  $Y_i = \beta_0 + \beta_1 X_i + u_i$ ,  $u_i \sim NIID(0, \sigma^2)$ . Use OLS to find the estimator of  $\beta_0$  and  $\beta_1$ . (Note: *NIID* = Normally, Identically, and Independently Distributed).

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (Y_i - \bar{Y})(X_i - \bar{X})}{\sum_{i=1}^n (X_i - \bar{X})^2}$$

$$= \frac{17.4375}{511.875} = 0.0341$$

$$\hat{\beta}_0 = \bar{Y} - \beta_1 \bar{X}$$

$$= 3.2125 - 0.0341(77.625)$$

$$= 0.5681$$

1.2 For each observation  $i$ , find  $\hat{Y}_i$  and  $\hat{u}_i$ . Show that  $\sum_{i=1}^N \hat{u}_i = 0$ .

$$\hat{y} = 0.0341X + 0.5681$$

$\hat{u}_i$

1.3 Find  $var(\hat{u}_i)$ ,  $var(\hat{\beta}_0)$ ,  $var(\hat{\beta}_1)$

$$\frac{1}{n-2} \sum_{i=1}^n \hat{u}_i^2 = \frac{SSR}{n-2} = 0$$

$$var(\hat{\beta}_0) = \frac{\sigma^2 n^{-1} \sum_{i=1}^n x_i^2}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{0.3586 \times 8^{-1} \times 48717}{511.875} = 4.2662$$

$$var(\hat{\beta}_1) = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sigma^2}{SST} = \frac{0.3586}{511.875} = 0.0007$$

2. Data is listed in the table

$X_i$	$Y_i$
10	0
12	2
14	5
16	6
18	7
22	10
24	10
26	15
28	16
30	20

$$n = 10$$

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i = 20$$

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i = 9.1$$

2.1 From the simple regression model  $Y_i = \beta_0 + \beta_1 X_i + u_i$ ,  $u_i \sim NIID(0, \sigma^2)$ . Find estimators of  $\beta_0$  and  $\beta_1$  from the OLS method and interpret the meaning.

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2} \quad \left| \quad \begin{aligned} \hat{\beta}_0 &= \bar{y} - \beta_1 \bar{x} \\ &= 9.1 - 0.8955(20) \\ &= -8.8091 \end{aligned}$$

$$\hat{\beta}_1 = 0.8955$$

The slope  $\hat{\beta}_1$  is 0.8955, meaning that if  $x$  increases by 1 unit,  $y$  increases by approximately 0.8955

The  $y$  intercept  $\hat{\beta}_0$  is -8.8091

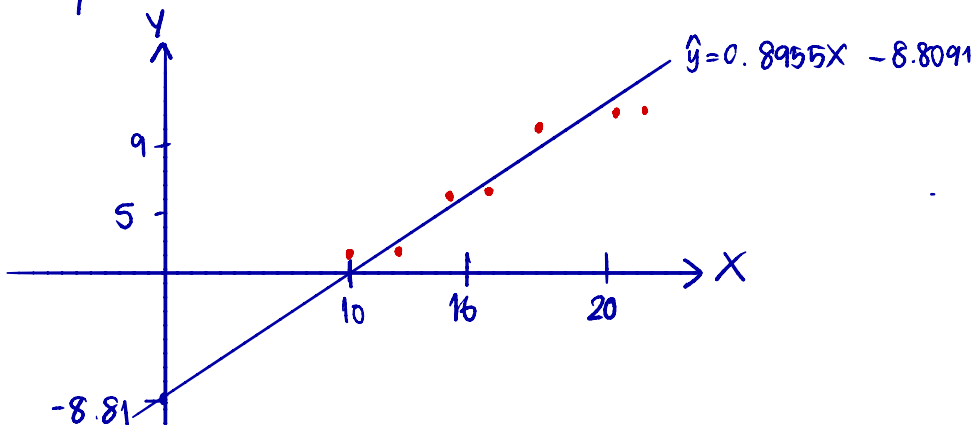
2.2 Find the value of  $\hat{Y}_i$  and  $\hat{u}_i$ . Show that  $\sum_{i=0}^N \hat{u}_i = 0$ .

$$\hat{y} = 0.8955x - 8.8091$$

2.3 Plot graph and draw regression line. Does the line pass  $(\bar{X}, \bar{Y})$ ?

$$\hat{y} = 0.8955x - 8.8091$$

$$\hat{y} = 0.8955(20) - 8.8091 = 9.1, \text{ the regression line passes } (\bar{X}, \bar{Y})$$



2.4 If  $X_i = 16$ , what is the predicted Y?

$$\hat{y} = 0.8955(16) - 8.8091 = 5.5189$$

2.5 Find  $\text{var}(\hat{u}_i)$ ,  $\text{var}(\hat{\beta}_0)$ ,  $\text{var}(\hat{\beta}_1)$

$$\text{var}(\hat{u}_i) = \frac{1}{n-2} \sum_{i=1}^n \hat{u}_i^2 = \frac{\text{SSR}}{n-2} = \frac{0}{10-2} = 0 \neq$$

$$\text{var}(\hat{\beta}_0) = \frac{\sigma^2 n^{-1} \sum_{i=1}^n X_i^2}{\sum_{i=1}^n (X_i - \bar{X})^2}$$

$$\text{var}(\hat{\beta}_1) = \frac{\sigma^2}{\sum_{i=1}^n (X_i - \bar{X})^2} = \frac{\sigma^2}{\text{SST}}$$

$$\sigma^2 = \text{Var}(u|X) = \text{Var}(y|X)$$

$$\text{Var}(y|X) = 6.0572$$

$$\text{var}(\hat{\beta}_0) = \frac{6.0572 \times 10^{-1} \times 4440}{-240} = -11.2058$$

$$\text{var}(\hat{\beta}_1) = \frac{6.0572}{-240} = -0.02524$$

3. Consider the below regression function:

$$Y_i = \beta_1 X_i + u_i,$$

Where  $u_i \sim NIID(0, \sigma^2)$ . Find an OLS estimator of  $\beta_1$ . Then, provide a proof that this is an unbiased estimator. Please state the SLR assumption(s) when used.

$$\hat{\beta} = \frac{\sum_{i=1}^n (X_i - \bar{X}) y_i}{SST_X} = \frac{\sum_{i=1}^n (X_i - \bar{X}) (\beta_0 + \beta_1 X_i + u_i)}{SST}$$

$$\begin{aligned} (X - \bar{X}) &= 0 & \hat{\beta}_1 &= \beta_1 + \frac{\sum_{i=1}^n (X_i - \bar{X}) u_i}{SST} \\ & & &= \beta_1 + \frac{1}{SST} \sum_{i=1}^n d_i u_i \end{aligned}$$

$$\begin{aligned} E(\hat{\beta}_1) &= \beta_1 + E\left(\frac{1}{SST}\right) \sum_{i=1}^n d_i u_i = \beta_1 (1/SST_X) E(d_i u_i) \\ &= \beta_1 + (1/SST) \sum_{i=1}^n d_i E(u_i) \\ &= \beta_1 + (1/SST) \sum_{i=1}^n d_i \times 0 \\ &= \beta_1 \end{aligned}$$