

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

$$\hat{\beta}_1 = \frac{\sum x_i y_i - n \bar{x} \bar{y}}{\sum x_i^2 - n \bar{x}^2}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

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).

$$\bar{x} = \frac{63 + 72 + \dots + 90}{8} = \frac{621}{8} = 77.625, \quad \bar{y} = \frac{2.8 + 3.4 + \dots + 3.7}{8} = \frac{25.7}{8} = 3.2125$$

$$\hat{\beta}_1 = \frac{\sum x_i y_i - n \bar{x} \bar{y}}{\sum x_i^2 - n \bar{x}^2} = \frac{2012.4 - (8)(77.625)(3.2125)}{48,979 - (8)(77.625)^2} = \frac{19.4395}{511.875} = 0.034066$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = 3.2125 - (0.034066)(77.625) = 0.5681$$

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

$$\hat{Y}_i = \hat{\beta}_0 + \hat{\beta}_1 X_i \quad \hat{u}_i = y_i - \hat{Y}_i$$

$$\hat{Y}_1 = 0.5681 + (0.034066)(63) = 2.714$$

$$\hat{Y}_2 = 0.5681 + (0.034066)(72) = 3.021$$

$$\hat{Y}_3 = 0.5681 + (0.034066)(78) = 3.225$$

$$\hat{Y}_4 = 0.5681 + (0.034066)(81) = 3.329$$

$$\hat{Y}_5 = 0.5681 + (0.034066)(87) = 3.532$$

$$\hat{Y}_6 = 0.5681 + (0.034066)(75) = 3.123$$

$$\hat{Y}_7 = 0.5681 + (0.034066)(75) = 3.123$$

$$\hat{Y}_8 = 0.5681 + (0.034066)(90) = 3.634$$

$$\hat{u}_1 = y_1 - \hat{Y}_1 = 2.8 - 2.714 = 0.086$$

$$\hat{u}_2 = y_2 - \hat{Y}_2 = 3.4 - 3.021 = 0.379$$

$$\hat{u}_3 = y_3 - \hat{Y}_3 = 3 - 3.225 = -0.225$$

$$\hat{u}_4 = y_4 - \hat{Y}_4 = 3.5 - 3.329 = 0.173$$

$$\hat{u}_5 = y_5 - \hat{Y}_5 = 3.6 - 3.532 = 0.068$$

$$\hat{u}_6 = y_6 - \hat{Y}_6 = 3 - 3.123 = -0.123$$

$$\hat{u}_7 = y_7 - \hat{Y}_7 = 2.7 - 3.123 = -0.432$$

$$\hat{u}_8 = y_8 - \hat{Y}_8 = 3.7 - 3.634 = 0.066$$

$$\begin{aligned} \sum_{i=0}^N \hat{u}_i &= 0.086 + 0.379 - 0.225 + 0.173 + 0.068 \\ &\quad - 0.123 - 0.432 + 0.066 \\ &= 0 \end{aligned}$$

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

$$\text{Var}(\hat{u}_i) = \frac{\sum \hat{u}_i^2}{n-2} = \frac{(0.086)^2 + \dots + (0.066)^2}{6} = \frac{0.4349}{6} = 0.0725 = \sigma^2$$

$$\text{Var}(\hat{\beta}_0) = \frac{\sigma^2}{\sum X_i^2} = \frac{0.0725}{63+72+\dots+90} = 0.0001416$$

$$\text{Var}(\hat{\beta}_1) = \frac{\sum X_i^2}{n \sum X_i^2} \cdot \sigma^2 = \frac{(48717)(0.0725)}{(6)(511.876)} = 0.8625$$

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

2.1 From the simple regression model  $Y_i = \beta_0 + \beta_1 X_i + u_i$ ,  $u_i \sim \text{NIID}(0, \sigma^2)$ . Find estimators of  $\beta_0$  and  $\beta_1$  from the OLS method and interpret the meaning.  $\bar{X} = 20$ ,  $\bar{Y} = 9.1$

$$\hat{\beta}_1 = \frac{\sum X_i Y_i - n \bar{X} \bar{Y}}{\sum X_i^2 - n \bar{X}^2} = \frac{2294 - (10)(20)(9.1)}{4440 - (10)(20)^2} = \frac{394}{440} = 0.895$$

$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X} = 9.1 - (0.895)(20) = -8.809$$

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i \quad \hat{u}_i = y_i - \hat{y}_i$$

$$\hat{u}_i = y_i - \hat{y}_i$$

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

$$\hat{y}_1 = -8.809 + 0.895(10) = 0.141 \quad \hat{y}_6 = -8.809 + 0.895(22) = 10.881$$

$$\hat{y}_2 = -8.809 + 0.895(12) = 1.931 \quad \hat{y}_7 = -8.809 + 0.895(24) = 12.671$$

$$\hat{y}_3 = -8.809 + 0.895(14) = 3.721 \quad \hat{y}_8 = -8.809 + 0.895(26) = 14.461$$

$$\hat{y}_4 = -8.809 + 0.895(16) = 5.511 \quad \hat{y}_9 = -8.809 + 0.895(28) = 16.251$$

$$\hat{y}_5 = -8.809 + 0.895(18) = 7.301 \quad \hat{y}_{10} = -8.809 + 0.895(30) = 18.041$$

$$\hat{u}_1 = 0 - 0.141 = -0.141 \quad \hat{u}_6 = -0.881$$

$$\hat{u}_2 = 2 - 1.931 = 0.069 \quad \hat{u}_7 = -2.671$$

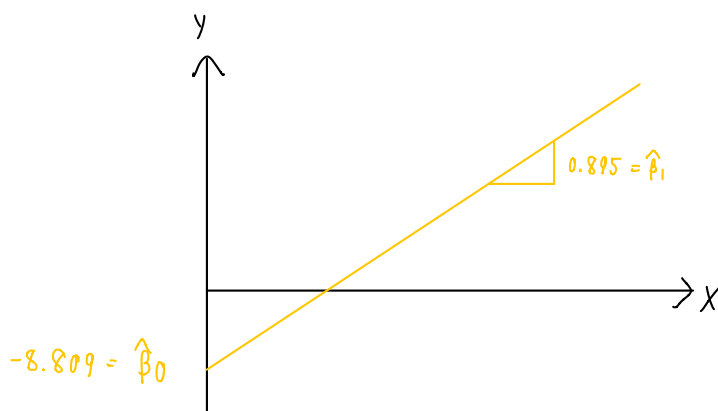
$$\hat{u}_3 = 5 - 3.721 = 1.279 \quad \hat{u}_8 = 0.539$$

$$\hat{u}_4 = 6 - 5.511 = 0.489 \quad \hat{u}_9 = -0.251$$

$$\hat{u}_5 = 7 - 7.301 = -0.301 \quad \hat{u}_{10} = 1.959$$

$$\sum_{i=0}^N \hat{u}_i = -0.141 + 0.069 + \dots + 1.959 = 0$$

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



$$\downarrow \downarrow$$

$$20 \quad 9.1$$

$$y = \beta_0 + \beta_1 X$$

$$y = -8.809 + 0.895(20)$$

$$y = -8.809 + 17.9$$

$$y = 9.091 \approx 9.1$$

$\therefore$  The line passes  $(\bar{X}, \bar{Y})$

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

$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 X_i$$

$$= -8.809 + 0.895(16) = 5.511$$

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

$$\text{var}(\hat{u}_i) = \frac{\sum \hat{u}_i^2}{n-2} = \frac{14.09181}{8} = 1.7615$$

$$\text{var}(\hat{\beta}_0) = \frac{\sigma^2}{\sum X_i^2} = \frac{1.7615}{4440} = 0.0003967$$

$$\text{var}(\hat{\beta}_1) = \frac{\sigma^2}{\sum (X_i - \bar{X})^2} = \frac{1.7615}{440} = 0.0040034$$

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.

$$x_i = X_i - \bar{X}$$

$$k_i = \frac{x_i}{\sum_i x_i^2}$$

$$\begin{aligned} \hat{\beta}_1 &= \sum_{i=1}^n (Y_i - \bar{Y}) k_i \\ &= \sum_{i=1}^n (\beta_0 + \beta_1 X_i + u_i - \beta_0 - \beta_1 \bar{X}) k_i \\ &= \sum_{i=1}^n \beta_1 (X_i - \bar{X}) k_i + \sum_{i=1}^n u_i k_i \\ &= \beta_1 \sum_{i=1}^n x_i k_i + \sum_{i=1}^n u_i k_i \\ &= \beta_1 \sum_{i=1}^n x_i \frac{x_i}{\sum_{i=1}^n x_i^2} + \sum_{i=1}^n u_i k_i \\ &= \beta_1 \frac{\sum_{i=1}^n x_i^2}{\sum_{i=1}^n x_i^2} + \sum_{i=1}^n u_i k_i \end{aligned}$$

$$E(\hat{\beta}_1) = E[\beta_1 + \sum_{i=1}^n u_i k_i] = \beta_1 + \underbrace{E[\sum_{i=1}^n u_i k_i]}_{\text{what assumption is required for this? SLR4} \star}$$

$$\text{SLR4: } E(u_i | x_i) = 0$$

This assumption take the value of  $x$  as given (fixed)  
So, we can treat  $x$  as a constant

$$E(\hat{\beta}_1) = \beta_1 + \sum_{i=1}^n k_i E(u_i)$$

function of  $x$       expected term of error is 0

$$E(\hat{\beta}_1) = \beta_1$$