

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	$Y_i$	$X_i$	$x_i - \bar{x}$	$y_i - \bar{y}$	$(x_i - \bar{x})(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$\hat{y}$	$\hat{u}$
1	2.8	63	-14.625	-0.4125	6.0328	213.8906	2.7143	0.0857
2	3.4	72	-5.625	0.1875	-1.0547	31.6406	3.0209	0.3791
3	3	78	0.375	-0.2125	-0.0797	0.1406	3.2253	-0.2253
4	3.5	81	3.375	0.2875	0.9703	11.3906	3.3275	0.1725
5	3.6	87	9.375	0.3875	3.6328	87.8906	3.5319	0.0681
6	3.0	75	-2.625	-0.2125	0.5578	6.8906	3.1231	-0.1231
7	2.7	75	-2.625	-0.5125	1.3453	6.8906	3.1231	-0.4231
8	3.7	90	12.375	0.4875	6.0328	153.1406	3.6341	0.0659

$\bar{y} = 3.2125$ ,  $\bar{x} = 77.625$        $\Sigma = 17.4374$        $\Sigma = 511.875$

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).

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

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

$$\hat{\beta}_1 = \frac{17.4374}{511.875} = 0.0341$$

$$\hat{\beta}_0 = 3.2125 - (0.0341) 77.625$$

$$= 0.5655$$

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

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

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

$y$	$x$	$\hat{y}$	$\hat{u}$
2.8	63	2.7143	0.0857
3.4	72	3.0209	0.3791
3	78	3.2253	-0.2253
3.5	81	3.3275	0.1725
3.6	87	3.5319	0.0681
3	75	3.1231	-0.1231
2.7	75	3.1231	-0.4231
3.7	90	3.6341	0.0659

1  $\sum_{i=0}^N \hat{u}_i = 0$

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

2. Data is listed in the table

$\bar{x} = 20$	$\bar{y} = 9.1$	$x_i - \bar{x}$	$y_i - \bar{y}$	$(x_i - \bar{x})(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$\hat{y}_i$	$\hat{u}_i$
$X_i$	$Y_i$						
		-10	-9.1	91	100	0.145	-0.145
10	0	-8	-7.1	56.8	64	1.936	0.064
12	2	-6	-4.1	24.6	36	3.727	1.273
14	5	-4	-3.1	12.4	16	5.518	0.482
16	6	-2	-2.1	4.2	4	7.309	-0.309
18	7	2	0.9	1.8	4	10.891	-0.891
22	10	4	0.9	3.6	16	12.682	-2.682
24	10	6	5.9	35.4	36	14.473	0.527
26	15	8	6.9	55.2	64	16.264	-0.264
28	16	10	10.9	109	100	18.055	1.945
30	20			394	440		

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.

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

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

$$\hat{\beta}_1 = \frac{394}{440} = 0.8955$$

$$\hat{\beta}_0 = 9.1 - (0.8955)(20) = -8.81$$

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

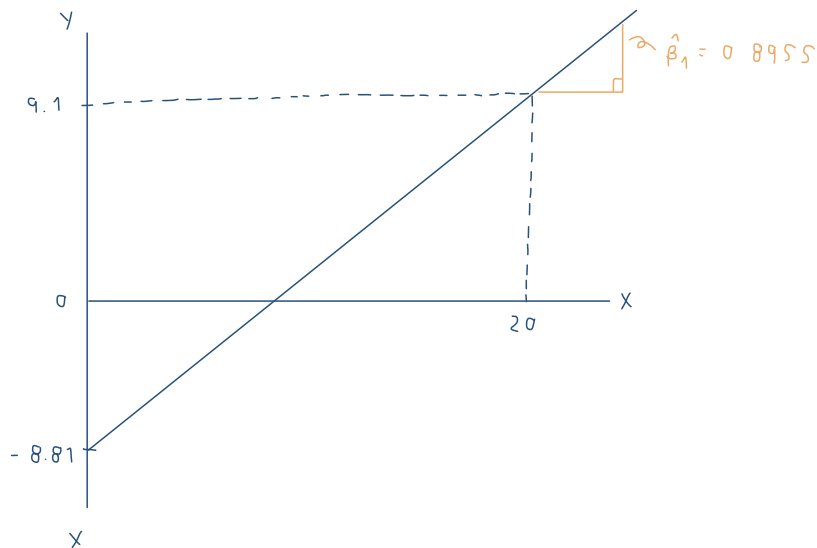
$$\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i$$

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

$x_i$	$y_i$	$\hat{y}_i$	$\hat{u}_i$
10	0	0.145	-0.145
12	2	1.936	0.064
14	5	3.727	1.273
16	6	5.518	0.482
18	7	7.309	-0.309
22	10	10.891	-0.891
24	10	12.682	-2.682
26	15	14.473	0.527
28	16	16.264	-0.264
30	20	18.055	1.945

$$\sum_{i=0}^N \hat{u}_i = 0$$

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



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

$$\text{if } x_i = 16, \hat{y} = 5.518$$

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

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.

$$0 = \sum_i (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)$$

$$= \sum_i y_i - n \hat{\beta}_0 - \hat{\beta}_1 \sum_i x_i$$

$$\hat{\beta}_0 = \frac{\sum_i y_i}{n} - \hat{\beta}_1 \frac{\sum_i x_i}{n}$$

$$\hat{\beta}_0 = 0$$

sub.  $\hat{\beta}_0$  into 3.2, we get

$$0 = \sum_i x_i (y_i - (\bar{y} - \hat{\beta}_1 \bar{x}) - \hat{\beta}_1 x_i)$$

$$= \sum_i x_i y_i - \sum_i x_i \bar{y} + \sum_i \hat{\beta}_1 \bar{x} x_i - \sum_i \hat{\beta}_1 x_i^2$$

$$= \sum_i x_i (y_i - \bar{y}) - \hat{\beta}_1 \sum_i x_i (x_i - \bar{x})$$

$$\hat{\beta}_1 = \frac{\sum_i x_i (y_i - \bar{y})}{\sum_i x_i (x_i - \bar{x})}$$

$$\hat{\beta}_1 = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{\sum_i (x_i - \bar{x})^2}$$

$$\hat{\beta}_1 = \frac{\text{cov}(x, y)}{\text{var}(x)}$$