



EE325 Introductory Econometrics, Semester 1/2019 (Section 046402)

Due Date: Thursday 27th February 2020 by 09.30 via Assignment Submission in Moodle.

Instruction: Do all questions with your own handwriting and your own attempt.

Use 4 decimal places for numerical answers

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

Table 1

$\bar{y} = 3.2125$ $\bar{x} = 77.625$

Student	Y_i	X_i	$x_i - \bar{x}$	$y_i - \bar{y}$	$(x_i - \bar{x})^2$	$(x_i - \bar{x})(y_i - \bar{y})$
1	2.8	63	-14.625	-0.4125	213.8906	6.0328
2	3.4	72	-5.625	0.1875	31.6406	-1.0547
3	3.0	78	0.375	-0.2125	0.1406	-0.0797
4	3.5	81	3.375	0.2875	11.3906	0.9703
5	3.6	87	9.375	0.3875	87.8906	3.6328
6	3.0	75	-2.625	-0.2125	6.8906	0.5578
7	2.7	75	-2.625	-0.5125	6.8906	1.3453
8	3.7	90	12.375	0.4875	153.1406	6.0328
					$\sum = 511.8748$	$\sum = 17.4347$

1.1 Now consider the two-variable model $Y_i = \beta_1 + \beta_2 X_i + u_i$, $u_i \sim NIID(0, \sigma^2)$

Use OLS to find the estimator of β_1 and β_2 . Interpret the regression.

1.2 Find \hat{Y}_i and \hat{u}_i and show that $\sum_{i=1}^n \hat{u}_i \approx 0$

1.3 Find $var(\hat{u}_i)$, $var(\hat{\beta}_1)$, and $var(\hat{\beta}_2)$

1.1 Now consider the two-variable model $Y_i = \beta_1 + \beta_2 X_i + u_i$, $u_i \sim NIID(0, \sigma^2)$

Use OLS to find the estimator of β_1 and β_2 . Interpret the regression.

$$\beta_1 = \bar{Y}_i - \beta_2 \bar{X}$$

$$\beta_2 = \frac{Cov(X_i, Y_i)}{Var(X_i)} = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sum (X_i - \bar{X})^2} = \frac{17.4347}{511.8748} = 0.0341$$

$$\beta_1 = \bar{Y}_i - \beta_2 \bar{X}$$

$$\beta_1 = 3.2125 - (0.0341)(77.625)$$

$$\beta_1 = 0.5455$$

$$\hat{Y}_i = 0.5455 + 0.0341 X_i + u_i$$

1.2 Find \hat{Y}_i and \hat{u}_i and show that $\sum_{i=1}^n \hat{u}_i \approx 0$

1.3 Find $var(\hat{u}_i)$, $var(\hat{\beta}_1)$, and $var(\hat{\beta}_2)$

$\bar{Y} = 3.2125$ $\bar{X} = 77.625$

Student	Y_i	X_i	$X_i - \bar{X}$	$Y_i - \bar{Y}$	$(X_i - \bar{X})^2$	$(X_i - \bar{X})(Y_i - \bar{Y})$	X_i^2
1	2.8	63	-14.625	-0.4125	213.8906	6.0328	3969
2	3.4	72	-5.625	0.1875	31.4406	-1.0597	5184
3	3.0	78	0.375	-0.2125	0.1406	-0.0797	6084
4	3.5	81	3.375	0.2875	11.3906	0.9703	6561
5	3.6	87	9.375	0.3875	87.8906	3.6328	7569
6	3.0	75	-2.625	-0.2125	6.8906	0.5578	5625
7	2.7	75	-2.625	-0.5125	6.8906	1.3453	5625
8	3.7	90	12.375	0.4875	153.1406	6.0328	8100
						$\sum = 17.4347$	$\sum = 48,717$

$$1.2 \hat{Y} = \hat{\beta}_1 + \hat{\beta}_2 X_i$$

$$\hat{Y}_1 = 0.5455 + 0.0341(63) = 2.7138$$

$$\hat{Y}_2 = 0.5455 + 0.0341(72) = 3.0207$$

$$\hat{Y}_3 = 0.5455 + 0.0341(78) = 3.2253$$

$$\hat{Y}_4 = 0.5455 + 0.0341(81) = 3.3276$$

$$\hat{Y}_5 = 0.5455 + 0.0341(87) = 3.5322$$

$$\hat{Y}_6 = 0.5455 + 0.0341(75) = 3.123$$

$$\hat{Y}_7 = 0.5455 + 0.0341(75) = 3.123$$

$$\hat{Y}_8 = 0.5455 + 0.0341(90) = 3.6345$$

$$\hat{u}_i = Y_i - \hat{Y}_i$$

$$\hat{u}_1 = 2.8 - 2.7138 = 0.0862$$

$$\hat{u}_2 = 3.4 - 3.0207 = 0.3793$$

$$\hat{u}_3 = 3.0 - 3.2253 = -0.2253$$

$$\hat{u}_4 = 3.5 - 3.3276 = 0.1724$$

$$\hat{u}_5 = 3.6 - 3.5322 = 0.0678$$

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

$$\hat{u}_7 = 2.7 - 3.123 = -0.423$$

$$\hat{u}_8 = 3.7 - 3.6345 = 0.0655$$

$n=8$

$$\sum_{i=1}^8 \hat{u}_i = -0.0001 \approx 0$$

$$(\hat{u}_1)^2 = 0.0074 \quad (\hat{u}_5)^2 = 0.0046$$

$$(\hat{u}_2)^2 = 0.1439 \quad (\hat{u}_6)^2 = 0.0151$$

$$(\hat{u}_3)^2 = 0.0508 \quad (\hat{u}_7)^2 = 0.1789$$

$$(\hat{u}_4)^2 = 0.0296 \quad (\hat{u}_8)^2 = 0.0043$$

$$\sum_{i=1}^8 (\hat{u}_i)^2 = 0.4347$$

1.3 Find $\text{var}(\hat{u}_i)$, $\text{var}(\hat{\beta}_1)$, and $\text{var}(\hat{\beta}_2)$

$$\text{var}(\hat{u}_i) = \frac{SbR}{n-2} = \frac{\sum_{i=1}^{n=8} u_i^2}{n-2} = \frac{0.4347}{6} = 0.0725$$

$$SbR = \sum (y_i - \hat{y}_i)^2$$

$$\text{var}(\hat{\beta}_1) = \frac{\sum_{i=1}^n \sigma_u^2 x_i^2}{n \sum_{i=1}^n x_i^2} = \frac{4.8717 \cdot 0.0725}{8(511.8748)} = \frac{3529.5467}{4,094.9984} = 0.8619$$

$$\text{var}(\hat{\beta}_2) = \frac{\sigma_u^2}{\sum_{i=1}^n x_i^2} = \frac{0.0725}{511.8748} = 0.0001$$

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 $Y_i = \beta_1 + \beta_2 X_i + u_i, u_i \sim NIID(0, \sigma^2)$

Find estimators of β_1 and β_2 from the OLS method and interpret the meaning.

2.2 Find the value of \hat{Y}_i and \hat{u}_i . Show that $\sum \hat{u}_i \approx 0$

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

2.4 If $X_i = 18$, what is the predicted Y?

2.5 Find $var(\hat{u}_i), var(\hat{\beta}_1), var(\hat{\beta}_2)$

3. Consider the below regression function: consider the two-variable model

$$Y_i = \beta_1 + \beta_2 X_i + u_i, u_i \sim NIID(0, \sigma^2)$$

Find an OLS estimator of β_1 . Then, provide a proof that this is an unbiased estimator.

Please state the assumption(s) of CLRM when used (pages 66-75 in Gujarati).

“Practice makes Perfect.”

2.1 From the simple regression model $Y_i = \beta_0 + \beta_1 X_i + u_i, u_i \sim NID(0, \sigma^2)$

Find estimators of β_0 and β_1 from the OLS method and interpret the meaning.

2.2 Find the value of $\hat{\beta}_0$ and $\hat{\beta}_1$. Show that $\sum \hat{u}_i = 0$

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

2.4 If $X_i = 18$, what is the predicted \hat{Y} ?

2.5 Find $var(\hat{\beta}_0)$, $var(\hat{\beta}_1)$, $var(\hat{\beta}_2)$

X_i	Y_i	$X_i - \bar{X}$	$Y_i - \bar{Y}$	$(X_i - \bar{X})^2$	$(Y_i - \bar{Y})^2$	$(X_i - \bar{X})(Y_i - \bar{Y})$	X_i^2
10	0	-10	-9.1	100	81	-91	100
12	2	-8	-7.1	64	50.41	56.8	144
14	5	-6	-4.1	36	16.81	24.6	196
16	6	-4	-3.1	16	9.61	12.4	256
18	7	-2	-2.1	4	4.41	4.2	324
22	10	2	0.9	4	0.81	1.8	484
24	10	4	0.9	16	0.81	3.6	576
26	15	6	5.9	36	34.81	35.4	676
28	16	8	6.9	64	47.61	55.2	784
30	20	10	10.9	100	118.81	109	900
				$\Sigma = 440$		$\Sigma = 394$	
							$\Sigma = 4440$

2.1 $\beta_1 = \bar{Y} - \beta_2 \bar{X}$

$$\beta_2 = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sum (X_i - \bar{X})^2} = \frac{394}{440} = 0.8955$$

$$\beta_1 = 9.1 - 0.8955(20) = -8.81$$

$$Y_i = -8.81 + 0.8955 X_i + u_i$$

2.2 $\hat{Y}_i = \beta_1 + \beta_2 \cdot X_i$

$$\hat{Y}_1 = -8.81 + 0.8955(10) = 0.145$$

$$\hat{Y}_2 = -8.81 + 0.8955(12) = 1.936$$

$$\hat{Y}_3 = -8.81 + 0.8955(14) = 3.727$$

$$\hat{Y}_4 = -8.81 + 0.8955(16) = 5.518$$

$$\hat{Y}_5 = -8.81 + 0.8955(18) = 7.309$$

$$\hat{u}_i = (Y_i - \hat{Y}_i)$$

$$\hat{u}_1 = 0 - 0.145 = -0.145$$

$$\hat{u}_2 = 2 - 1.936 = 0.064$$

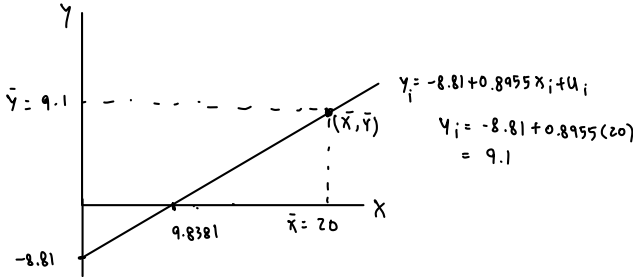
$$\hat{u}_3 = 5 - 3.727 = 1.273$$

$$\hat{u}_4 = 6 - 5.518 = 0.482$$

$$\hat{u}_5 = 7 - 7.309 = -0.309$$

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

2.3



\therefore This regression function passes through (\bar{X}, \bar{Y})

2.4 If $X_i = 18$, what is the predicted Y ?

2.5 Find $var(\hat{u}_i)$, $var(\hat{\beta}_1)$, $var(\hat{\beta}_2)$

$$2.4 \quad Y_i = -8.81 + 0.8955(x_i)$$

$$Y_i = -8.81 + 0.8955(18)$$

$$Y_i = 7.309$$

$$2.5 \quad var(\hat{u}_i) = \frac{\sum \hat{u}_i^2}{n-2} = \frac{14.0909}{8} = 1.7614$$

$$var(\hat{\beta}_1) = \frac{\sum (X_i)^2 \cdot \sigma_u^2}{n \sum (X_i - \bar{x})^2} = \frac{1.7614}{10 \cdot 440} = \frac{4440 \cdot 1.7614}{4400} = 1.7614$$

$$var(\hat{\beta}_2) = \frac{\sigma_u^2}{\sum X_i^2} = \frac{1.7614}{440} = 0.004$$

$$y_i = \beta_1 + \beta_2 X_i + u_i$$

$$\beta_2 = \frac{\beta_1 + u_i - y_i}{-X_i}$$

$$\bar{y} = \beta_1 + \beta_2 \bar{x}$$

3. Consider the below regression function: consider the two-variable model

$$Y_i = \beta_1 + \beta_2 X_i + u_i, u_i \sim N(0, \sigma^2)$$

Find an OLS estimator of β_1 . Then, provide a proof that this is an unbiased estimator.

Please state the assumption(s) of CLRM when used (pages 66-75 in Gujarati).

$$\text{OLS} \quad \hat{y}_i = \hat{\beta}_1 + \hat{\beta}_2 \hat{x}_i + \hat{u}_i$$

$$\min \sum \hat{u}_i^2 \quad \hat{u}_i = y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i$$

$$\downarrow$$

$$E \hat{u}_i = E(y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i)$$

$$\downarrow$$

$$\text{F.O.C. :}$$

$$-2 \sum (y_i - \hat{\beta}_1 - \hat{\beta}_2 x_i) = 0$$

$$E y_i - E \hat{\beta}_1 - \beta_2 E x_i = 0$$

$$E \hat{\beta}_1 = E y_i - \beta_2 E x_i$$

$$n \hat{\beta}_1 = \sum (y_i - \beta_2 x_i) \text{ wrt } n.$$

$$\hat{\beta}_1 = \bar{y} - \beta_2 \bar{x}$$

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

$$\hat{\beta}_1 = \bar{y} - \beta_2 \bar{x}$$

$$\hat{\beta}_1 = \beta_1 + \beta_2 \bar{x} - \beta_2 \bar{x}$$

$$\hat{\beta}_1 = \beta_1$$

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

$$\begin{matrix} \hat{y}_i \\ \hat{x}_i \end{matrix} = \begin{matrix} \beta_1 \\ \beta_2 \end{matrix} + \begin{matrix} u_i \\ x_i \end{matrix}$$