



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

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

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 $var(\hat{\beta}_2)$

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).

| Student | Y_i | X_i | $(x_i - \bar{x})$ | $(y_i - \bar{y})$ | $x_i y_i$ |
|----------|-------|-------|-------------------|-------------------|--------------------|
| 1 | 2.8 | 63 | -14.625 | -0.4125 | 0.0328125 |
| 2 | 3.4 | 72 | -5.625 | 0.1875 | -1.0546875 |
| 3 | 3.0 | 78 | 0.375 | -0.2125 | -0.0796875 |
| 4 | 3.5 | 81 | 3.375 | 0.2875 | 0.9703125 |
| 5 | 3.6 | 87 | 9.375 | 0.3875 | 3.6328125 |
| 6 | 3.0 | 75 | -2.625 | -0.2125 | 0.5578125 |
| 7 | 2.7 | 75 | -2.625 | -0.5125 | 1.3453125 |
| 8 | 3.7 | 90 | 12.375 | 0.4875 | 6.0328125 |
| Σ | 25.7 | 621 | | | $\Sigma = 17.4375$ |

$$1.1 \quad \hat{\beta}_1 = \bar{y} - \hat{\beta}_2 \bar{x}$$

$$\hat{\beta}_1 = 3.2125 - 0.0340659 (77.625) \\ = 0.568132$$

$$\bar{y} = 3.2125 \\ \bar{x} = 77.625 \\ \Sigma x_i^2 = 511.875$$

$$\hat{y}_i = 0.568132 + 0.0340659 (x_i)$$

$$\hat{\beta}_2 = \frac{\Sigma (x_i - \bar{x})(y_i - \bar{y})}{\Sigma (x_i - \bar{x})^2} \\ = \frac{17.4375}{511.875} = 0.0340659$$

with the exam score increases by 1 unit, GPA will increase by 0.0340659 units. With no score on exam, student will get GPA equal to 0.568132

1.2

$$\begin{aligned} \hat{y}_1 &= 2.914284 & \hat{y}_2 &= 3.020877 & \hat{y}_3 &= 3.225452 & \hat{y}_4 &= 3.32965 & \hat{y}_5 &= 3.532045 & \hat{y}_6 &= 3.123255 \\ \hat{u}_1 &= 0.085716 & \hat{u}_2 &= 0.379123 & \hat{u}_3 &= -0.225452 & \hat{u}_4 &= 0.17235 & \hat{u}_5 &= 0.067955 & \hat{u}_6 &= -0.123255 \\ \hat{y}_7 &= 3.123255 & \hat{y}_8 &= 3.634243 & & & & & & & \Sigma \hat{u}_i &= 0.001061 \\ \hat{u}_7 &= -0.423255 & \hat{u}_8 &= 0.065757 & & & & & & & & \end{aligned}$$

$$1.3 \quad \text{Var}(\hat{u}_i) = \frac{\Sigma u_i^2}{n-2} = \frac{0.4348931}{6} = 0.0724822 = \sigma^2$$

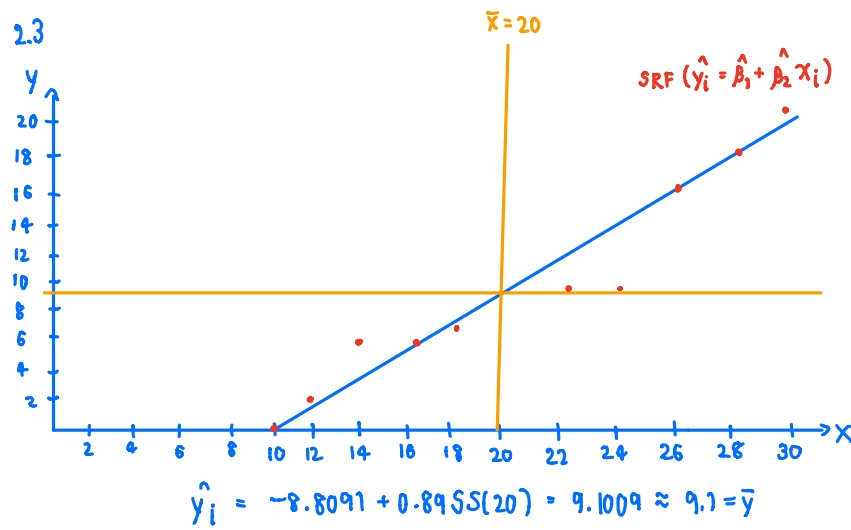
$$\text{Var}(\hat{\beta}_1) = \frac{\Sigma x_i^2}{n \Sigma x_i^2} \sigma^2 = \frac{48797}{8(511.875)} (0.0724822) = 0.862299$$

$$\text{Var}(\hat{\beta}_2) = \frac{\sigma^2}{\Sigma x_i^2} = \frac{0.0724822}{511.875} = 0.0001416$$

$$2.1 \quad \begin{aligned} \hat{\beta}_1 &= \bar{y} - \hat{\beta}_2 \bar{x} \\ &= 9.1 - (17.909091) \\ &= -8.809091 \\ \hat{y}_i &= -8.8091 + 0.8955 (x_i) \end{aligned} \quad \begin{aligned} \hat{\beta}_2 &= \frac{\Sigma x_i y_i}{\Sigma x_i^2} \\ &= \frac{394}{440} = 0.895455 \end{aligned}$$

2.2

| | |
|--------------------------|-------------------------------------|
| $\hat{y}_1 = 0.1459$ | $\hat{u}_1 = -0.1459$ |
| $\hat{y}_2 = 1.9369$ | $\hat{u}_2 = 0.0631$ |
| $\hat{y}_3 = 3.7279$ | $\hat{u}_3 = 1.2721$ |
| $\hat{y}_4 = 5.5189$ | $\hat{u}_4 = 0.4811$ |
| $\hat{y}_5 = 7.3099$ | $\hat{u}_5 = -0.3099$ |
| $\hat{y}_6 = 10.8919$ | $\hat{u}_6 = -0.8919$ |
| $\hat{y}_7 = 12.6829$ | $\hat{u}_7 = -2.6829$ |
| $\hat{y}_8 = 14.4739$ | $\hat{u}_8 = 0.5261$ |
| $\hat{y}_9 = 16.2649$ | $\hat{u}_9 = -0.2649$ |
| $\hat{y}_{10} = 18.0559$ | $\hat{u}_{10} = 1.9441$ |
| | $\sum \hat{u}_i = -0.009 \approx 0$ |



2.4 $x_i = 18, \hat{y}_i = 7.3099$ from 2.2

2.5 $\text{Var}(\hat{u}_i) = \sigma^2 = \frac{\sum \hat{u}_i^2}{n-2} = \frac{14.09092}{8} = 1.761365$

$\text{Var}(\hat{\beta}_2) = \frac{\sigma^2}{\sum x_i^2} = \frac{1.761365}{440} = 0.004$

$\text{Var}(\hat{\beta}_1) = \frac{\sum x_i^2}{n} \cdot \frac{\sigma^2}{\sum x_i^2} = \frac{444}{10} (0.004) = 1.776$

3. As $\hat{\beta}_1$ is an estimator of β_1 , $E(\hat{\beta}_1)$ should be equal to β_1

$\hat{\beta}_1 = \bar{y} - \hat{\beta}_2 \bar{x}, \hat{\beta}_2 = \frac{\sum x_i y_i}{\sum x_i^2}$ let $k = \frac{x_i}{\sum x_i^2}$

$\hat{\beta}_1 = \bar{y} - \sum k y_i \bar{x}$ and $y_i = \beta_1 + \beta_2 x_i + u_i$

$$\begin{aligned} \hat{\beta}_1 &= \frac{\sum y_i}{n} - \sum k y_i \bar{x} = \sum \left(\frac{1}{n} - k \bar{x} \right) y_i \\ &= \sum \left(\frac{1}{n} - k \bar{x} \right) (\beta_1 + \beta_2 x_i + u_i) \\ &= \sum \left(\frac{\beta_1}{n} + \frac{\beta_2 x_i}{n} + \frac{u_i}{n} - k \bar{x} \beta_1 - k \bar{x} \beta_2 x_i - k \bar{x} u_i \right) \\ &= \sum \frac{\beta_1}{n} + \beta_2 \bar{x} + \frac{\sum u_i}{n} - \beta_2 \bar{x} - \bar{x} \sum k_i u_i \end{aligned}$$

take $E(\cdot) \rightarrow E(\hat{\beta}_1) = E(\beta_1) + \bar{x} E(\sum k_i u_i) \rightarrow$ assumption 3: zero mean value of disturbance
treat X as given
 $E(u_i | x_i)$

$E(\hat{\beta}_1) = E(\beta_1) + \bar{x} \sum k_i \overset{0}{E(u_i)}$

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

$\hat{\beta}_1$ is an unbiased estimator of actual β_1