

Assignment #1

Instructions:

- For all questions, answer up to 4 decimal places.
- This assignment is due on **Tuesday, Mar 9, 2021 before class (11.00)**.
- Write your answer in either digital or ordinary paper. For digital paper, export pages into a single PDF file. For ordinary paper, take photos of your writing and convert them into a single PDF file as well.
- There is no need to rewrite the question. Assign number item, ie. 1 a., clearly before your answer is sufficient.
- Submit your assignment into Moodle.
- Name your file as StudentID_Nickname (in Thai) such as 123456789_๑๑

1. Given this information

$$\begin{array}{l} n = 30 \qquad \sum_{i=1}^n X_i = 366 \qquad \sum_{i=1}^n Y_i = 631 \qquad \bar{X} = 12.20 \qquad \bar{Y} = 21.03 \\ \sum_{i=1}^n (X_i)^2 = 5,564 \qquad \sum_{i=1}^n X_i Y_i = 7,524 \qquad \sum_{i=1}^n (X_i - \bar{X})^2 = 1098.8 \qquad \sum_{i=1}^n (Y_i - \bar{Y})^2 = 882.97 \\ \sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y}) = -174.20 \qquad \sum_{i=1}^n \hat{u}_i^2 = 873.14 \end{array}$$

Answer the following questions. Show your work.

- From regression model: $Y_i = \beta_1 + \beta_2 X_i + u_i$, $u_i \sim NIID(0, \sigma^2)$ (normally, identically and independently distributed), find the estimators of β_1 and β_2 with OLS method and explain the meaning of the model.
- Find r^2 and explain its meaning.
- If $X_i = 5$, estimate the value of \hat{Y}_i and explain its meaning.
- Find the estimators of $\text{var}(u_i)$, $\text{var}(\hat{\beta}_1)$ and $\text{var}(\hat{\beta}_2)$
- Test the hypothesis whether coefficients are different from zero at 0.05 level of significance.
- Test the hypothesis whether coefficients are less than zero at 0.01 level of significance.

$$\begin{aligned} b_1 &= \frac{\sum X_i^2 \sum Y_i - \sum X_i \sum X_i Y_i}{n \sum X_i^2 - (\sum X_i)^2} \\ &= \bar{Y} - b_2 \bar{X} \end{aligned}$$

(1.a) From the formula : $\hat{\beta}_2 = \frac{n\sum xy - \sum x \sum y}{n\sum x^2 - (\sum x)^2}$

$$\hat{\beta}_2 = \frac{(30)(7,524) - (366)(631)}{(30)(5,564) - (366)^2}$$

$$\hat{\beta}_2 = \frac{225,720 - 230,946}{166,920 - 133,956}$$

$$\hat{\beta}_2 = \frac{-5,226}{32,964}$$

$$\hat{\beta}_2 = -0.1585$$

From the formula : $\hat{\beta}_1 = \bar{Y} - \hat{\beta}_2 \bar{x}$

$$\hat{\beta}_1 = 21.03 - (-0.1585)(12.20)$$

$$\hat{\beta}_1 = 22.9637$$

Therefore, $Y_i = 22.9637 - 0.1585X_i$

- $E(Y_i | x_i = 0) = 22.9637$, assume other things being equal ($x_i = 0$). The average or condition mean of $Y_i = 22.9637$ units

- slope = $\hat{\beta}_2 = -0.1585$, if x increase by 1 unit, y will decrease by 0.1585 units

(1.b) To find r^2 using $r^2 = 1 - \frac{\sum \hat{u}_i^2}{\sum (y_i - \bar{y})^2}$

$$r^2 = 1 - \frac{873.14}{882.97}$$

$$r^2 = 0.0111$$

Hence, r^2 measures how goodness of fit of the fitted regression line comparing to an estimator x can explain by 0.0111, which is the explain part. In this case, r^2 is equal to 0.0111 which is close to 0.

$$(1.c) \quad Y_j = 22.9637 - 0.1585X_j$$

$$\text{when } x_j = 5 : Y_j = 22.9637 - 0.1585(5)$$

$$Y_j = 22.1712$$

Therefore, when X_j is equal to 5 unit, Y_j will equal to 22.1712 units

$$(1.d) \quad \text{A formula to calculate } \text{var}(u_j) = \frac{\sum \hat{u}_i^2}{n-k}$$

$$= \frac{873.14}{30-2}$$

$$= 31.1836$$

$$\text{A formula to calculate } \text{var}(\hat{\beta}_1) = \frac{\sum x_i^2}{n \sum x_i^2} \sigma^2$$

$$= \frac{5,564}{30(5,564)} (31.1836)$$

$$= 1.0395$$

$$\text{A formula to calculate } \text{var}(\hat{\beta}_2) = \frac{\sigma^2}{\sum x_i^2}$$

$$= \frac{31.1836}{5,564}$$

$$= 0.0056$$

(1.e) From $Y_i = 22.9637 - 0.1585(X_i) + u_i$

- Prove β_1

step 1 : State your hypothesis

$H_0 : \beta_1 = 0$ - null hypothesis

$H_a : \beta_1 \neq 0$ - Alternative hypothesis

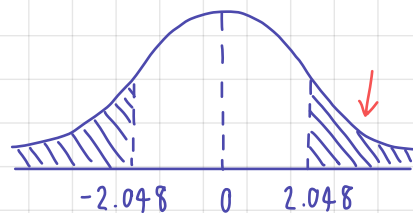
step 2 : Calculate test statistics

$$t_{cal} = \frac{\hat{\beta}_1 - \beta_1}{\hat{\sigma}_{\hat{\beta}_1}} = \frac{22.9637 - 0}{1.0196} = 22.5222$$

step 3 : state your decision rule , when $\alpha = 0.05$ use \pm to create CI around $\beta_1 = 0$ based on $d.f = 30 - 2 = 28$

The lower bound : $t_{\frac{\alpha}{2}} = -2.048$

The upper bound : $t_{\frac{\alpha}{2}} = 2.048$



step 4 : t_{cal} is 22.5222 which lies beyond 2.048. Hence, we can reject null hypothesis , at the significance level of 95%.

Moreover, β_1 is not zero 95 out of 100 times when we sample

- prove β_2

1. $H_0 : \beta_2 = 0$ - null hypothesis

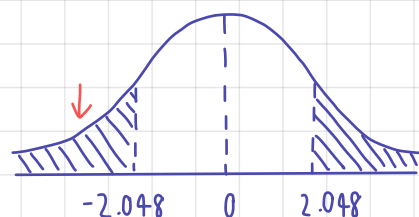
$H_a : \beta_2 \neq 0$ - alternative hypothesis

$$2. t_{cal} = \frac{\hat{\beta}_2 - \beta_2}{\hat{\sigma}_{\hat{\beta}_2}} = \frac{-0.1585 - 0}{0.0748} = -2.1190$$

3. state your decision rule , when $\alpha = 0.05$ use \pm to create CI around $\beta_2 = 0$ based on $d.f = 30 - 2 = 28$

- The lower bound : $t_{\frac{\alpha}{2}} = -2.048$

- The upper bound : $t_{\frac{\alpha}{2}} = 2.048$



4. t_{cal} is -2.1190 which lies beyond CI. Hence, we can reject null hypothesis , at the significance level of 95%.

Moreover, β_2 is not zero 95 out of 100 times when we sample

(1.f) - Prove β_1

step 1 : State your hypothesis

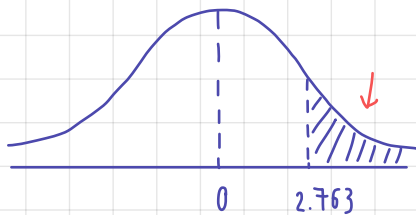
$$H_0 : \beta_1 \leq 0 \quad \text{null hypothesis}$$

$$H_a : \beta_1 > 0 \quad \text{alternative hypothesis}$$

$$\text{step 2} : t_{cal} = \frac{\hat{\beta}_1 - \beta_1}{\hat{\sigma}_{\hat{\beta}_1}} = \frac{22.9637 - 0}{1.0196} = 22.5222$$

step 3 : when $\alpha = 0.01$ use to create one tail acceptance region when $\beta_1 \leq 0$ based on d.f. = $30 - 2 = 28$

- The upper bound : $t_{\frac{\alpha}{2}} = 2.763$



step 4 t_{cal} is 22.5222 which lie beyond 2.763, we can reject the null hypothesis, at the significant level of 99%. we cannot sure that β_1 is less than 0 99 out of 100 times when we sample.

- prove β_2

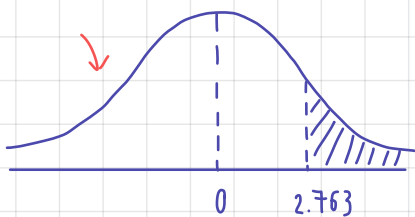
step 1 $H_0 : \beta_2 \leq 0$ null hypothesis

$H_a : \beta_2 > 0$ alternative hypothesis

$$\text{step 2} : t_{cal} = \frac{\hat{\beta}_2 - \beta_2}{\hat{\sigma}_{\hat{\beta}_2}} = \frac{-0.1585 - 0}{0.0748} = -2.1190$$

step 3 : when $\alpha = 0.01$ use to create one tail acceptance region when $\beta_2 \leq 0$ based on d.f. = $30 - 2 = 28$

- The upper bound : $t_{\frac{\alpha}{2}} = 2.763$



step 4 t_{cal} is -2.1190 which lie within any boundary CI, we cannot reject the null hypothesis, at the significant level of 99%. we are sure that β_2 is less than 0 99 out of 100 times when we sample.

2. Given that Y is market price of a car (USD) while X is how long a car aged (years), results of the regression are as follows.

$$\hat{Y}_i = 7,836 - 502.4X_i$$

(52) (411.8)

Given that u_i is normally, identically and independently distributed with zero mean and σ^2 variance, total number of observations is 11,

$$\bar{X} = 7.45,$$

$$\hat{\sigma}^2 = 212,877,$$

$$\sum(X_i - \bar{X})^2 = 78.73,$$

Answer the following questions. Show your work.

- a) Does the sign of $\hat{\beta}_2$ make economic sense? Provide your explanation.
- b) If you are a car expert and someone asks you to estimate how much his car will be **averagely** priced at when his car is 5 years old, how much is the market price range that you would estimate that you can make sure that for 95% of the time, market price will be within the specific range?
- c) If you multiply all the X with 10, report the new SRF with the standard error resulted from the multiplication.
- d) Calculate the elasticity of market price when a car is 10 years old.

(2.a) Yes, it follows economic sense since $\hat{\beta}_2 = -502.4$. The sign of $\hat{\beta}_2$ is negative meaning that when car is older for 1 year, the market price of car will decrease by 502.4 USD

(2.b) From $\hat{Y}_i = 7836 - 502.4x_i$, $x = 5$

$$\hat{Y}_i = 7836 - 502.4(5)$$

$$\hat{Y}_i = 5,324$$

when \hat{Y}_0 is an estimator of $E(Y|X_0)$

$$\begin{aligned} 1. \text{ From } \text{var}(\hat{Y}_0) &= \sigma^2 \left[\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{\sum (x_i - \bar{x})^2} \right] \\ &= (212,877) \left[\frac{1}{11} + \frac{(5 - 7.45)^2}{78.73} \right] \\ &= 35,582.5345 \end{aligned}$$

$$2. \text{ Find } \sigma_{\hat{Y}_0} = \sqrt{35,582.5345} = 188.6333$$

3 Find 95% CI for $E(Y|X_0 = 5)$

$$\text{- upper: } \hat{Y}_0 + \left(t_{\frac{\alpha}{2}} \sigma_{\hat{Y}_0}\right) = 5324 + (2.262 \cdot 188.6333) = 5750.6885$$

$$\text{- lower: } \hat{Y}_0 - \left(t_{\frac{\alpha}{2}} \sigma_{\hat{Y}_0}\right) = 5324 - (2.262 \cdot 188.6333) = 4897.3115$$

- Hence, if we create a CI over the mean value, 95 out of 100 times that the CI will cover true value $E(Y|X_0)$.

- when car is 5 yearolds, the market price is between 4897.3115 to 5750.6885 USD

$$(2.c) \hat{Y}_i = 7836 - 502.4x_i$$

$$\text{New: } \hat{Y}_i = 7836 - 502.4(10)x_i$$

$$\text{So, } \hat{Y}_i = 7836 - 5024x_i$$

when x increase 1 year, the market price of car decreases by 5024 USD

Standard error of the old one is 411.8 and standard error when changed x is $411.8(10) = 4118$

$$\text{Hence, new SER when multiply } x \text{ with } 10 \cdot \hat{Y}_i = 7836 - 5024x_i$$

se (52) (4118)

$$(2.d) \quad \hat{y}_i = 7836 - 502.4X_i \quad \text{when } X=10, \hat{y} = 7836 - 502.4(10)$$

$$\hat{y} = 2812$$

$$\frac{d\hat{y}_i}{dX_i} = -502.4$$

$$\left(\frac{dY}{dX}\right) \cdot \frac{X}{Y} = -502.4$$

$$\begin{aligned} \frac{dY}{dX} \cdot \frac{X}{Y} &= (-502.4) \left(\frac{10}{2812}\right) \\ &= -1.7866 \end{aligned}$$

Ans when car is 10 yearolds the elasticity of market price is -1.7866