

1)

$$a) \hat{\beta}_2 = \frac{\sum X_i Y_i}{\sum X_i^2} = \frac{319,943.18}{364,023.30} = 0.8789 \# \quad \hat{\beta}_1 = \bar{Y} - \hat{\beta}_2 \bar{X} = 69.1478 - (0.8789 \times 86.0826) = -6.5101 \#$$

$$\therefore \hat{Y}_i = -6.5101 + 0.8789 X_i$$

The intercept of the model is -6.5101 and the slope is 0.8789.

$$b) R^2 = 1 - \frac{\sum \hat{u}_i^2}{\sum (Y_i - \bar{Y})^2} = 1 - \frac{2,610.9211}{94,525.1748} = 0.9724 \#$$

R^2 is a measure for the goodness of fit and 97.24% of the variation in Y_i can be explained by the model.

$$c) \hat{Y}_i = -6.5101 + 0.8789(60) = 46.2239 \#$$

It means that when $X_i = 60$, the average of \hat{Y}_i will be 46.2239.

$$d) \sigma^2 = \frac{\sum \hat{u}_i^2}{n-k} = \frac{\sum (Y_i - \hat{Y}_i)^2}{n-k} = \frac{2610.9211}{46-2} = 59.3391 \#$$

$$\sigma_{\hat{\beta}_1}^2 = \text{var}(\hat{\beta}_1) = \frac{\sigma^2}{n} = \frac{59.3391}{46} = 1.29 \#$$

$$\sigma_{\hat{\beta}_2}^2 = \text{var}(\hat{\beta}_2) = \frac{\sigma^2}{\sum X_i^2} = \frac{59.3391}{364,023.30} = 1.63 \times 10^{-4} \#$$

$$e) \hat{\beta}_2 \pm t_{\frac{\alpha}{2}} \cdot \sigma_{\hat{\beta}_2} = 0.8789 \pm 0.681 = 1.63 \times 10^{-4}$$

$$0.8789 - 2.021 \times \sqrt{1.63 \times 10^{-4}} \leq \beta_2 \leq 0.8789 + 2.021 \times \sqrt{1.63 \times 10^{-4}}$$

$$0.8789 - 0.0258 \leq \beta_2 \leq 0.8789 + 0.0258$$

$$0.8530 \leq \beta_2 \leq 0.9047 \#$$

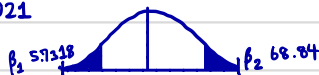
\therefore It's 95% certain that β_2 would be between 0.8532 and 0.9048.

f) Null hypothesis: $\beta_1 = 0, \beta_2 = 0$ / Alternative hypothesis: $\beta_1 \neq 0, \beta_2 \neq 0$

$$\beta_1: t_{cal} = \frac{\hat{\beta}_1 - \beta_1}{\sigma_{\hat{\beta}_1}} = \frac{-6.5101 - 0}{\sqrt{1.29}} = -5.7318$$

$$\beta_2: t_{cal} = \frac{\hat{\beta}_2 - \beta_2}{\sigma_{\hat{\beta}_2}} = \frac{0.8789 - 0}{\sqrt{1.63 \times 10^{-4}}} = 68.84$$

$$t_{\frac{\alpha}{2}} = \pm 2.021$$



We can be 95% certain that β_1 and β_2 is not 0.

2)

- a) No, we can't because an SRF represents a relationship between several sets of multiple independent variables Y and multiple dependent variables X , but a function showing a relationship between multiple data points cannot be created with only one data point.
- b) No, they aren't because β_2 shows that X and Y are related but in some cases, the relationship between X and Y is just coincidence. For example, even though income is related to consumption, we can't assume that an increase in income causes an increase in consumption.
- c) If the level of significant is 95% and β_2 is different from 0, we can't say for sure that β_2 is not 0 and we can't 100% conclude that β_2 is 0 as well.
- d) Interval estimation shows extra information about the estimation. It has the accuracy of the estimation based on significant level. So, it's more accurate than estimate.

3)

$$\begin{aligned} \text{a) } \ln \widehat{\text{wage}} &= 7.658062 + 0.0318017 \text{ hours} \\ \ln \widehat{\text{wage}} &= 7.658082 \\ \widehat{\text{wage}} &= e^{7.658082} \\ \widehat{\text{wage}} &= 2117.6718 \text{ \#} \end{aligned}$$

$$\begin{aligned} \text{b) } \ln \widehat{\text{wage}} &= 7.658082 + 0.0318017 \text{ hours} \\ \frac{d \ln \widehat{\text{wage}}}{d \text{ hour}} &= 0.0318017 \end{aligned}$$

$$\frac{d \widehat{\text{wage}}}{\widehat{\text{wage}}} = 0.0318017 \text{ d hour}$$

$$\Delta \% \text{ wage} = 3.18017 \text{ d hour}$$

$$\therefore \text{wage increase } 3.18017 \% \text{ \#}$$

$$\begin{aligned} \text{c) } \text{new } \hat{\sigma}_{\beta_2} &= 0.079488 \quad \text{new } \hat{\beta}_2 = 0.7632408 \\ \text{confidence interval} &= \hat{\beta}_2 \pm t_{\frac{\alpha}{2}} \cdot \hat{\sigma}_{\beta_2} \\ \text{upper} &= 0.7632408 + 1.984 (0.079488) = 0.9209 \\ \text{lower} &= 0.7632408 - 1.984 (0.079488) = 0.6055 \end{aligned}$$

