

CONCORDIA UNIVERSITY
Department of Economics

ECON 222/4 SECTIONS A, B and BB
STATISTICAL METHODS II
WINTER 2016 – MIDTERM 2 (SOLUTIONS)
Sunday, March 20, 1:00pm – 3:00pm

1. (34 marks) A population is assumed to follow the behaviour $y_i = \beta_0 + \beta_1 x_i + e_i$, where β_0 and β_1 are unknown population parameters and $e \sim iid N(0, \sigma^2)$. Use the following sample of five observations to answer the following questions.

x	1	2	3	4	5
y	3	7	5	11	14

- a. (2 marks) Calculate b_0 .

$$b_0 = \bar{y} - b_1 \bar{x} = 8 - 2.6 \cdot 3 = 0.2$$

- b. (2 marks) Calculate b_1 .

$$b_1 = \frac{\sum_{i=1}^n x_i y_i - n \bar{x} \bar{y}}{\sum_{i=1}^n x_i^2 - n \bar{x}^2} = \frac{1 \cdot 3 + 2 \cdot 7 + 3 \cdot 5 + 4 \cdot 11 + 5 \cdot 14 - 5 \cdot 3 \cdot 8}{1^2 + 2^2 + 3^2 + 4^2 + 5^2 - 5 \cdot 3^2} = \frac{146 - 120}{55 - 45} = 2.6$$

- c. (2 marks) **Briefly** interpret b_0 .

When $x = 0$, y should equal 0.2.

- d. (2 marks) **Briefly** interpret b_1 .

When x changes by one unit, y should change by 2.6 units.

- e. (2 marks) Calculate SST.

$$SST = \sum_{i=1}^n (y_i - \bar{y})^2 = \sum_{i=1}^n y_i^2 - n \bar{y}^2 = 3^2 + 7^2 + 5^2 + 11^2 + 14^2 - 5 \cdot 8^2 = 400 - 320 = 80$$

- f. (2 marks) Calculate SSR.

$$SSR = \sum_{i=1}^n (\hat{y}_i - \bar{y})^2 = \sum_{i=1}^n \hat{y}_i^2 - n \bar{y}^2 = 2.8^2 + 5.4^2 + 8^2 + 10.6^2 + 13.2^2 - 5 \cdot 8^2 = 387.6 - 320 = 67.6$$

g. (2 marks) Calculate SSE.

$$SSE = \sum_{i=1}^n \hat{u}_i^2 = \sum_{i=1}^n (y_i - \hat{y}_i)^2 = (3 - 2.8)^2 + (7 - 5.4)^2 + (5 - 8)^2 + (11 - 10.6)^2 + (14 - 13.2)^2 = 12.4$$

h. (2 marks) Calculate the coefficient of determination, R^2 .

$$R^2 = \frac{SSR}{SST} = \frac{67.6}{80} = 0.845$$

i. (2 marks) **Briefly** interpret R^2 .

84.5 percent of the variation in y can be explained by the estimated regression.

j. (2 marks) Calculate the correlation coefficient, r .

$$r = (\text{sign of } b_1) \sqrt{R^2} = \sqrt{0.845} = 0.9192$$

k. (2 marks) **Briefly** interpret r .

There is an almost perfect positive linear relationship between x and y.

l. (2 marks) Calculate the standard error of the estimate, $\hat{\sigma}$.

$$\hat{\sigma} = \sqrt{\frac{SSE}{n-k-1}} = \sqrt{\frac{12.4}{5-1-1}} = 2.0331$$

m. (2 marks) Calculate $\text{var}(b_1)$.

$$\text{var}(b_1) = \frac{\hat{\sigma}^2}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{4.13}{10} = 0.413$$

n. (2 marks) Calculate $\text{var}(b_0)$.

$$\text{var}(b_0) = \frac{\hat{\sigma}^2 \sum_{i=1}^n x_i^2}{n \sum_{i=1}^n (x_i - \bar{x})^2} = \frac{4.13 \cdot 55}{5 \cdot 10} = 4.546$$

o. (2 marks) Calculate $\text{cov}(b_0, b_1)$.

$$\text{cov}(b_0, b_1) = \frac{-\bar{x}\hat{\sigma}^2}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{-3 \cdot 4.13}{100} = -1.24$$

p. (2 marks) Construct a 99-percent confidence interval for β_0 , using $t_{0.005,3} = \pm 5.841$.

$$b_0 \pm t_c \cdot \text{se}(b_0) = 0.2 \pm 5.841 \cdot \sqrt{4.546} \Rightarrow [-12.2547, 12.6547]$$

q. (2 marks) Test, at the 1-percent level of significance, whether $\beta_0 = 10$. Clearly state the null and alternative hypotheses, the test statistic and your conclusion. Use a critical value of $t_{0.005,3} = \pm 5.841$.

$$H_0: \beta_0 = 10 \text{ and } H_1: \beta_0 \neq 10$$

Since the t -statistic, $t_s = \frac{b_0 - \beta_0}{\text{se}(b_0)} = \frac{0.2 - 10}{\sqrt{4.546}} = -4.596$, does not lie beyond the lower critical value,

$t_{0.005,3} = -5.841$, we cannot reject the null hypothesis. Therefore, we conclude that there is insufficient evidence to reject that $\beta_0 = 10$.

2. (4 marks) In the regression, $\hat{y}_i = \beta_0 + \beta_1 x_i$, use the fact that

$$b_1 = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} \text{ to prove}$$

a. (2 marks) b_1 is an unbiased estimator of β_1 .

$$\begin{aligned} b_1 &= \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})y_i}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})(\beta_0 + \beta_1 x_i + e_i)}{\sum_{i=1}^n (x_i - \bar{x})^2} \\ &= \frac{\beta_0 \sum_{i=1}^n (x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2} + \frac{\beta_1 \sum_{i=1}^n (x_i - \bar{x})^2}{\sum_{i=1}^n (x_i - \bar{x})^2} + \frac{\sum_{i=1}^n (x_i - \bar{x})e_i}{\sum_{i=1}^n (x_i - \bar{x})^2} = \beta_1 + \frac{\sum_{i=1}^n (x_i - \bar{x})e_i}{\sum_{i=1}^n (x_i - \bar{x})^2} \Rightarrow E(b_1) = \beta_1 + \frac{\sum_{i=1}^n (x_i - \bar{x})E(e_i)}{\sum_{i=1}^n (x_i - \bar{x})^2} = \beta_1 \end{aligned}$$

b. (2 marks) $\text{var}(b_1) = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$.

$$\text{var}(b_1) = \text{var} \left(\frac{\sum_{i=1}^n (x_i - \bar{x}) e_i}{\sum_{i=1}^n (x_i - \bar{x})^2} \right) = \frac{\text{var} \left(\sum_{i=1}^n (x_i - \bar{x}) e_i \right)}{\left(\sum_{i=1}^n (x_i - \bar{x})^2 \right)^2} = \frac{\left(\sum_{i=1}^n (x_i - \bar{x})^2 \text{var}(e_i) \right)}{\left(\sum_{i=1}^n (x_i - \bar{x})^2 \right)^2} = \frac{\sigma^2 \sum_{i=1}^n (x_i - \bar{x})^2}{\left(\sum_{i=1}^n (x_i - \bar{x})^2 \right)^2} = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

3. (6 marks) Price elasticity of demand is given by $\varepsilon = \frac{\Delta q}{\Delta p} \frac{p}{q}$. In the regression $\hat{q}_i = b_0 - b_1 p_i$, elasticity

can be calculated for any price as $\varepsilon = \frac{-b_1 p_i}{b_0 - b_1 p_i}$. Calculate price elasticity of demand from each of the following estimated regressions.

a. (2 marks) $\hat{q}_i = b_0 - b_1 \ln p_i$

$$\frac{\Delta q}{\Delta \ln p} = \frac{\Delta q}{(\Delta p/p)} = -b_1 \Rightarrow \frac{\Delta q}{\Delta p} \frac{p}{q} = \frac{-b_1}{b_0 - b_1 \ln p_i}$$

b. (2 marks) $\widehat{\ln q}_i = b_0 - b_1 p_i$

$$\frac{\Delta \ln q}{\Delta p} = \frac{(\Delta q/q)}{\Delta p} = -b_1 \Rightarrow \frac{\Delta q}{\Delta p} \frac{p}{q} = \frac{-b_1}{p_i}$$

c. (2 marks) $\widehat{\ln q}_i = b_0 - b_1 \ln p_i$

$$\frac{\Delta \ln q}{\Delta \ln p} = \frac{(\Delta q/q)}{(\Delta p/p)} = -b_1$$

4. (6 marks) Consider two regression models, $y_i = b_0 + b_1 x_i + e_i$ and $z_i = a_0 + a_1 w_i + v_i$, where $z_i = c y_i$ and $w_i = h x_i$.

a. (2 marks) Derive expressions that relate b_0 to a_0 , b_1 to a_1 and e_i to v_i .

$$z_i = a_0 + a_1 w_i + v_i \Rightarrow c\hat{y}_i = a_0 + a_1 h x_i + v_i \Rightarrow y_i = \frac{a_0}{c} + \frac{h a_1}{c} x_i + \frac{v_i}{c} \Rightarrow b_0 = \frac{a_0}{c}, b_1 = \frac{h a_1}{c}, e_i = \frac{v_i}{c}$$

b. (2 marks) Derive expressions that relate $\text{var}(b_0)$ to $\text{var}(a_0)$, $\text{var}(b_1)$ to $\text{var}(a_1)$ and $\text{var}(e_i)$ to $\text{var}(v_i)$.

$$b_0 = \frac{a_0}{c} \Rightarrow \text{var}(b_0) = \frac{\text{var}(a_0)}{c^2}, b_1 = \frac{h a_1}{c} \Rightarrow \text{var}(b_1) = \frac{h^2 \text{var}(a_1)}{c^2}, e_i = \frac{v_i}{c} \Rightarrow \text{var}(e_i) = \frac{\text{var}(v_i)}{c^2}$$

c. (2 marks) **Briefly** compare the t -statistics for a_0 to a_1 , $t_{a_0} = \frac{a_0}{\text{se}(a_0)}$ and $t_{a_1} = \frac{a_1}{\text{se}(a_1)}$, to the t -

statistics for b_0 to b_1 , $t_{b_0} = \frac{b_0}{\text{se}(b_0)}$ and $t_{b_1} = \frac{b_1}{\text{se}(b_1)}$.

$$t_{a_0} = \frac{a_0}{\text{se}(a_0)} = \frac{a_0}{\sqrt{\text{var}(a_0)}} = \frac{c b_0}{\sqrt{c^2 \text{var}(b_0)}} = \frac{b_0}{\sqrt{\text{var}(b_0)}} = \frac{b_0}{\text{se}(b_0)} = t_{b_0}$$

$$t_{a_1} = \frac{a_1}{\text{se}(a_1)} = \frac{a_1}{\sqrt{\text{var}(a_1)}} = \frac{c b_1 / h}{\sqrt{c^2 \text{var}(b_0) / h^2}} = \frac{b_1}{\sqrt{\text{var}(b_1)}} = \frac{b_1}{\text{se}(b_1)} = t_{b_1}$$