Westminster Business School School of Finance and Accounting 7FNCE044 Predictive Analysis for Decision Making Semesters 1 and 2 Weeks 1 and 2 Dr. Issam Malki

## Seminar Week 1 and 2 Sketch Answers

Linear Regression: Extension I

# **Computer Based Questions:**

You are expected to write R studio (Python) codes to the questions. See codes in Blackboard for more details.

## **Theoretical Questions**

## 1. Some linear Algebra

1.

a) 
$$\mathbf{X}'\mathbf{X} = \begin{bmatrix} \mathbf{1}' \\ \mathbf{x}' \end{bmatrix} \begin{bmatrix} \mathbf{1} & \mathbf{x} \end{bmatrix}$$
.

$$= \begin{bmatrix} 1'1 & 1'x \\ 1'x & x'x \end{bmatrix},$$

Using summation notation, we know:

$$A = \mathbf{1'1} = \sum_{i=1}^{n} 1^{2} = n$$
,

$$B = \mathbf{1}'\mathbf{x} = \sum_{i=1}^{n} 1 \cdot x_i = \sum_{i=1}^{n} x_i$$
.

b) 
$$\mathbf{X}'\mathbf{y} = \begin{bmatrix} \mathbf{1}' \\ \mathbf{x}' \end{bmatrix} \mathbf{y}$$

$$= \begin{bmatrix} \mathbf{1}'\mathbf{y} \\ \mathbf{x}'\mathbf{y} \end{bmatrix}$$

Using summation notation, we know:

$$C = \mathbf{1}'\mathbf{y} = \sum_{i=1}^n y_i ,$$

$$D = \mathbf{x}'\mathbf{y} = \sum_{i=1}^n x_i y_i .$$

c) The rule says that if: 
$$\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$
, then  $\mathbf{A} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$ .

From (a), we know:

$$\mathbf{X'X} = \begin{bmatrix} n & \sum_{i=1}^{n} x_i \\ \sum_{i=1}^{n} x_i & \sum_{i=1}^{n} x_i^2 \end{bmatrix}$$

Therefore:

$$(\mathbf{X}'\mathbf{X})^{-1} = \frac{1}{n\sum_{i=1}^{n} x_i^2 - (\sum_{i=1}^{n} x_i)^2} \begin{bmatrix} \sum_{i=1}^{n} x_i^2 & -\sum_{i=1}^{n} x_i \\ -\sum_{i=1}^{n} x_i & n \end{bmatrix}$$

d) The OLS coefficient vector is given by:

$$\hat{\boldsymbol{\beta}} = \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y}$$

From (b) and (c), we know:

$$\begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} = \frac{1}{n \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2} \begin{bmatrix} \sum_{i=1}^n x_i^2 & -\sum_{i=1}^n x_i \\ -\sum_{i=1}^n x_i & n \end{bmatrix} \begin{bmatrix} \sum_{i=1}^n y_i \\ \sum_{i=1}^n x_i y_i \end{bmatrix}$$

Therefore,

$$\hat{\beta}_{1} = \frac{\sum_{i=1}^{n} x_{i}^{2} \sum_{i=1}^{n} y_{i} - \sum_{i=1}^{n} x_{i} \sum_{i=1}^{n} x_{i} y_{i}}{n \sum_{i=1}^{n} x_{i}^{2} - (\sum_{i=1}^{n} x_{i})^{2}};$$

$$\hat{\beta}_2 = \frac{-\sum_{i=1}^n x_i \sum_{i=1}^n y_i + n \sum_{i=1}^n x_i y_i}{n \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2}.$$

e) We know that  $\mathbf{X}$  must have full rank in order for  $\mathbf{X}'\mathbf{X}^{-1}$  to exist, which means that the rank must be 2 in this case.

In order for  $\mathbf{X}$  to have full rank, one column cannot be a linear combination of the other columns. In this case, there are only two variables, therefore we require that  $\mathbf{x}$  is not proportion to the vector of ones. This means that  $\mathbf{x}$  cannot be a constant, *i.e.* the  $x_i$  cannot all be equal.

### 2. OLS Properties

(a) and (b)

$$E(\hat{\beta}) = E\{(X'X)^{-1}X'Y\} = E\{(X'X)^{-1}X'X\beta + (X'X)^{-1}X'u\} = \beta$$

Find the variance:

$$var(\hat{\beta}) = E\{(\hat{\beta} - E(\hat{\beta}))'(\hat{\beta} - E(\hat{\beta}))\} = E\{(X'X)^{-1}u'u\} = (X'X)^{-1}\sigma^2$$

The distribution, assuming normality, can be described as:

$$\hat{\beta} \sim N(\beta, (X'X)^{-1}\sigma^2)$$