

## Session 2 Review Exercises

### AY2025/26 Term 1

**Question 1.** (Econometric Notes Exercise 3.6) Suppose  $Y$  and  $X$  have the following joint distribution function:

	10	0	0	0	0	0.1
	9	0	0	0	0.1	0
	8	0	0	0.1	0	0
	7	0	0.1	0	0	0
	6	0.1	0	0	0	0
Y	5	0.1	0	0	0	0
	4	0	0.1	0	0	0
	3	0	0	0.1	0	0
	2	0	0	0	0.1	0
	1	0	0	0	0	0.1
		1	2	3	4	5
				X		

- i. Find the marginal distribution of  $X$  and of  $Y$ .
- ii. Find the conditional distribution, conditional mean, and conditional variance of  $Y$  given  $X$ , and of  $X$  given  $Y$ .
- iii. Find  $Cov(X, Y)$ .
- iv. In what way is the conditional distribution of  $Y$  related to  $X$ ?

**Question 2.** (Adapted from Econometric Notes Exercise 3.12) For the simple linear regression with intercept, estimated by OLS on the dataset  $\{X_i, Y_i\}_{i=1}^n$ , define the OLS fitted values  $\{\widehat{Y}_i\}_{i=1}^n$  and residuals  $\{\widehat{\epsilon}_i\}_{i=1}^n$  in the usual way, so that

$$Y_i = \widehat{Y}_i + \widehat{\epsilon}_i, \quad i = 1, \dots, n. \quad (1)$$

To simplify notation, we drop the “ols” subscripts from the residuals and fitted values.

(a) Show that  $\sum_{i=1}^n Y_i^2 = \sum_{i=1}^n \widehat{Y}_i^2 + \sum_{i=1}^n \widehat{\epsilon}_i^2$ .

(b) Show that

$$\underbrace{\sum_{i=1}^n (Y_i - \bar{Y})^2}_{TSS} = \underbrace{\sum_{i=1}^n (\widehat{Y}_i - \bar{Y})^2}_{FSS} + \underbrace{\sum_{i=1}^n \widehat{\epsilon}_i^2}_{RSS}. \quad (2)$$

*Hint: Subtract  $\bar{Y}$  from both sides of (1). Show that  $\bar{Y} = \bar{\widehat{Y}}$ . Square both sides of the equation and sum both sides from  $i = 1$  to  $n$ . Show that  $\sum_{i=1}^n (\widehat{Y}_i - \bar{Y})\widehat{\epsilon}_i = 0$ .*

*Remark: This equation can be described as “Total Sum of Squares = Fitted Sum of Squares + Residual Sum of Squares”, or  $TSS = FSS + RSS$ . Dividing (2) throughout by  $n - 1$ , the equation can also be described as a variance decomposition, specifically:*

$$\text{sample var.}(Y_i) = \text{sample var.}(\widehat{Y}_i) + \text{sample var.}(\widehat{\epsilon}_i)$$

(c) The  $R^2$  is defined as

$$R^2 = 1 - \frac{RSS}{TSS}.$$

It is used as a measure of **goodness-of-fit**, since  $RSS = 0$  when you have a perfect fit. Where you don't have a perfect fit,  $RSS > 0$  and therefore  $R^2 < 1$ .

- i. When will  $R^2 = 0$ ? Can  $R^2$  fall below zero?
- ii. What is  $R^2$  for the case where  $Y_1 = \dots = Y_n = c$  for some constant  $c$ ?