

# HW1

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ECON 7910 Econometrics

Due on Sep 23, 2021

## 1 Question 1– 2.1

**Solution:**

1. •

$$\frac{\partial E(y|x_1, x_2)}{\partial x_1} = \beta_1 + \beta_4 x_2$$

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$$\frac{\partial E(y|x_1, x_2)}{\partial x_2} = \beta_2 + 2\beta_3 x_2 + \beta_4 x_1$$

2.  $E(u|x_1, x_2)$  means error term  $u$  and covariates are independent, i.e.,  $E(u|x_1, x_2) = 0$ . For  $E(u|x_1, x_2, x_2^2, x_1 x_2)$ , if we are given previous CE, then in this CE,  $x_2^2$  and  $x_1 x_2$  are both redundant, because they can be expressed by  $x_1$  and  $x_2$ .
3. Given part(b),  $E(u|x_1, x_2)$  doesn't give any additional information about  $Var(u|x_1, x_2)$ . The only thing we can know for sure is this item is nonnegative. But for whether it's constant or relies on other variables, aka, heterogeneity or homogeneity, we have no information.

## 2 Question 2 – 2.2

**Solution:**

1. Since in this question,  $\mu = E(x)$ , thus:

$$\frac{\partial E(y|x)}{\partial x} = \delta_1 + 2\delta_2 x - 2\mu = \delta_1 + 2\delta_2(x - \mu)$$

2. Take expectation on both sides:

$$E\left(\frac{\partial E(y|x)}{\partial x}\right) = \delta_1 + 2\delta_2 E(x - E(x)) = \delta_1 + 0 = \delta_1$$

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3.

$$\begin{aligned}
L(y|1, x) &= L(E(y)|1, x) \\
&= L(\delta_0 + \delta_1(x_1 - \mu) + \delta_2(x_1 - \mu)^2|x) \\
&= \delta_0 + \delta_1(x_1 - \mu) + \delta_2L((x_1 - \mu)^2|1, x) \\
&= \alpha_0 + \delta_1x_1
\end{aligned}$$

For the first equation above, I borrow the result from the Property LP.5 of Appendix 2.A.3 directly. And the last equation, I intend to neglect some extra proof, because this equation is based on what I guess. The last  $L(\cdot)$  would be a constant, so  $\alpha_0 = \delta_0 - \delta_1\mu + L(\cdot)$ , and it's still a constant number, i.e., it doesn't rely on  $x$ .

### 3 Question 3 – 2.4

#### Solution:

Since the expectation of  $u$  given  $x$  and  $v$  is the function of  $x$  and  $v$ , and  $u$ ,  $v$  is uncorrelated with  $x$ , then, it's convenient to kill  $x$  term in this conditional expectation. So the CE becomes:  $E(u|x, v) = E(u|v)$ . We assume that  $E(u|v) = \rho_0 + \rho_1v$ . By LIE,  $E(E(u|v)) = 0 = \rho_0 + \rho_1E(v)$ . From the question,  $E(v) = 0$ , then we can deduce that  $\rho_0 = 0$ . Thus, back to our previous equation,  $E(u|v) = \rho_1v$ .  $\odot$ .

### 4 Question 4 – 2.7

#### Solution:<sup>1</sup>

Since  $E(y|x, z) = g(x) + z\beta$ , we can generalize  $y$  with covariates  $x$  and  $z$ :

$$y = g(x) + z\beta + \epsilon \quad (1)$$

Take expectation on both sides of (1) on the condition of  $x$ :

$$E(y|x) = g(x) + \beta E(z|x) \quad (2)$$

(1)-(2):

$$y - E(y|x) = z\beta - \beta E(z|x) + \epsilon = \beta(z - E(z|x)) + \epsilon \quad (3)$$

Substitute  $\tilde{y}$  and  $\tilde{z}$  into equation (3):

$$\tilde{y} = \beta\tilde{z} + \epsilon \quad (4)$$

Take expectation of equation of (4) on variable  $\tilde{z}$ :

$$E(\tilde{y}|\tilde{z}) = \beta\tilde{z} + 0 = \beta\tilde{z} \quad (5)$$

The reason why we get 0 from equation (5) is that the error term  $\epsilon$  is uncorrelated with  $\tilde{z}$ , thus,  $E(\epsilon|\tilde{z}) = 0$ .

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<sup>1</sup>I partly refer to the online resource about the beginning of this solution, because I forgot to set up a more flexible function of  $y$ , aka, equation (1).