

ECONOMICS 241B EXERCISE 1 PROPOSED SOLUTIONS
 CONDITIONAL EXPECTATION FUNCTIONS AND SPECIFICATION OF
 CONDITIONAL EXPECTATION FUNCTIONS

1.

- a. If $\mathbb{E}(y|x) = \beta_1 x + \beta_0$, find $\mathbb{E}(yx)$ as a function of the moments of x .

First, observe that $y = \mathbb{E}(y|x) + e = \beta_1 x + \beta_0 + e$. Multiply both sides by x

$$\mathbb{E}(yx) = \beta_1 \mathbb{E}(x^2) + \beta_0 \mathbb{E}(x) + \mathbb{E}(ex).$$

Observe that $\mathbb{E}(ex) = \mathbb{E}(x\mathbb{E}(e|x)) = 0$ by construction.

- b. Suppose the random variables y and x take only the values 0 and 1 and have the following joint probability distribution

	$x = 0$	$x = 1$
$y = 0$	a	c
$y = 1$	b	d

To satisfy the properties of a joint distribution, what must be true of (a, b, c, d) ? Find $\mathbb{E}(y|x)$, $\mathbb{E}(y^2|x)$, and $\text{Var}(y|x)$ for $x = 0$ and $x = 1$.

The values (a, b, c, d) must all be non-negative and sum to 1. The conditional moments are formed as

$$\mathbb{E}(y|x = 0) = 0 \cdot \mathbb{P}(y = 0|x = 0) + 1 \cdot \mathbb{P}(y = 1|x = 0) = \mathbb{P}(y = 1|x = 0).$$

The conditional probability is

$$\mathbb{P}(y = 1|x = 0) = \frac{\mathbb{P}(y = 1, x = 0)}{\mathbb{P}(x = 0)} = \frac{b}{a + b}.$$

Importantly, note that $\mathbb{P}(y = 1|x = 0) \neq b$. Similarly

$$\mathbb{P}(y = 1|x = 1) = \frac{d}{c + d}.$$

Here $\mathbb{E}(y^2|x) = \mathbb{E}(y|x)$ and

$$\begin{aligned} \text{Var}(y|x = 0) &= \mathbb{E}(y|x = 0)(1 - \mathbb{E}(y|x = 0)) = \frac{b}{a + b} \cdot \frac{a}{a + b} \\ \text{Var}(y|x = 1) &= \frac{d}{c + d} \cdot \frac{c}{c + d}. \end{aligned}$$

2. Assume $\mathbb{E} |g(x) y| < \infty$.

Prove

$$\mathbb{E}(g(x) y | x) = g(x) \mathbb{E}(y | x).$$

Proof. We have

$$\begin{aligned} \mathbb{E}(g(x) y | x) &= \int g(x) y f_{y|x}(y | x) dy \\ &= g(x) \int y f_{y|x}(y | x) dy \\ &= g(x) \mathbb{E}(y | x). \end{aligned}$$

3. If $y = x\beta + u$, $x \in \mathbb{R}$, then for each of the following statements either establish that they are true or provide a counterexample:

i) $\mathbb{E}(u | x) = 0$ implies $\mathbb{E}(x^2 u) = 0$,

True: $\mathbb{E}(x^2 u) = \mathbb{E}(x^2 \mathbb{E}(u | x)) = 0$,

ii) $\mathbb{E}(xu) = 0$ implies $\mathbb{E}(x^2 u) = 0$,

False: Consider the joint distribution for $(x, u) = \begin{cases} (-1, 1) & \text{with probability } .5 \\ (1, 1) & \text{with probability } .5 \end{cases}$,
so that $\mathbb{E}(xu) = -1(.5) + 1(.5) = 0$ but $\mathbb{E}(x^2 u) = 1(.5) + 1(.5) = 1$,

iii) $\mathbb{E}(u | x) = 0$ implies $\mathbb{E}(y | x) = x\beta$,

True: $\mathbb{E}(y | x) = x\beta + \mathbb{E}(u | x)$,

iv) $\mathbb{E}(xu) = 0$ implies $\mathbb{E}(y | x) = x\beta$,

False: Consider $y = x\beta + x^2\gamma + e$ with $\mathbb{E}x^3 = 0$, so that $u = x^2\gamma + e$. Here $\mathbb{E}(y | x) = x\beta + x^2\gamma$ and $\mathbb{E}(xu) = \mathbb{E}x^3\gamma + \mathbb{E}(xe) = 0$. (Recall $\mathbb{E}(e | x) = 0$ by construction.)

4. Recall that the conditional variance is $\sigma^2(x) = \text{Var}(y | x) = \mathbb{E}((y - \mathbb{E}(y | x))^2 | x)$. Show that the conditional variance can be written as

$$\sigma^2(x) = \mathbb{E}(y^2 | x) - (\mathbb{E}(y | x))^2.$$

We have

$$\begin{aligned} \mathbb{E}((y - \mathbb{E}(y | x))^2 | x) &= \mathbb{E}((y^2 - 2y\mathbb{E}(y | x) + \mathbb{E}(y | x)^2) | x) \\ &= \mathbb{E}(y^2 | x) - 2\mathbb{E}(y\mathbb{E}(y | x) | x) + \mathbb{E}(\mathbb{E}(y | x)^2 | x) \\ &= \mathbb{E}(y^2 | x) - (\mathbb{E}(y | x))^2 \end{aligned}$$