

Solutions to *Bruce Hansen's Econometrics*

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Contents

1	Introduction	3
2	Conditional Expectations and Projections	4
3	The Algebra of Least Squares	6

Revision History

Revision	Date	Author(s)	Description
1.0	01-05-2015	TC	created

Chapter 1

Introduction

No exercises in this chapter.

Chapter 2

Conditional Expectations and Projections

Problem 1: Find $\mathbb{E}(\mathbb{E}(\mathbb{E}(Y \mid X_1, X_2, X_3) \mid X_1, X_2) \mid X_1)$.

Solution: By the law of iterated expectations:

$$\mathbb{E}(\mathbb{E}(Y \mid X_1, X_2, X_3) \mid X_1, X_2) = \mathbb{E}(Y \mid X_1, X_2)$$

and

$$\mathbb{E}(\mathbb{E}(Y \mid X_1, X_2) \mid X_1) = \mathbb{E}(Y \mid X_1)$$

Problem 2: If $\mathbb{E}(Y \mid X) = a + bX$, find $\mathbb{E}(YX)$ in terms of the moments of X .

Solution: We know that:

$$\begin{aligned}\mathbb{E}(YX) &= \mathbb{E}(\mathbb{E}(Y \mid X)X) && \text{by LIE} \\ &= \mathbb{E}(aX + bX^2) \\ &= a\mu_X + b(\sigma_X^2 - \mu_X^2)\end{aligned}$$

Problem 3: Prove Theorem 2.8.1.4 (Properties of the CEF error). If $\mathbb{E}(|Y|) < \infty$, then for any function $h(\mathbf{X})$ such that $\mathbb{E}(|h(\mathbf{X})\varepsilon|) < \infty$, $\mathbb{E}(h(\mathbf{X})\varepsilon) = 0$.

Solution: Since $\mathbb{E}(|h(\mathbf{X})\varepsilon|) < \infty$, $h(\mathbf{X})\varepsilon$ is integrable, and by the conditioning theorem,

$$\mathbb{E}(h(\mathbf{X})\varepsilon) = \mathbb{E}(h(\mathbf{X})\mathbb{E}(\varepsilon \mid \mathbf{X}))$$

but by the mean independence of the errors, $\mathbb{E}(\varepsilon \mid \mathbf{X}) = 0$, so

$$= 0$$

Problem 4: Suppose that $X, Y \in \{0, 1\}$ and the joint PDF is

	$X = 0$	$X = 1$
$Y = 0$	0.1	0.2
$Y = 1$	0.4	0.3

Find $\mathbb{E}(Y | X)$, $\mathbb{E}(Y^2 | X)$, $\mathbb{V}(Y | X)$ for $X = 0$ and $X = 1$.

Solution:

$$\begin{aligned}\mathbb{E}(Y | X = 0) &= 0 \times 0.1 + 1 \times 0.4 \\ &= 0.4\end{aligned}$$

$$= \mathbb{E}(Y^2 | X = 0)$$

$$\begin{aligned}\mathbb{E}(Y | X = 1) &= 0 \times 0.2 + 1 \times 0.3 \\ &= 0.3\end{aligned}$$

$$= \mathbb{E}(Y^2 | X = 1)$$

$$\begin{aligned}\mathbb{V}(Y | X = 0) &= \mathbb{E}(Y^2 | X = 0) - (\mathbb{E}(Y | X = 0))^2 \\ &= 0.24\end{aligned}$$

$$\begin{aligned}\mathbb{V}(Y | X = 1) &= \mathbb{E}(Y^2 | X = 1) - (\mathbb{E}(Y | X = 1))^2 \\ &= 0.21\end{aligned}$$

Problem 5: Show that $\sigma^2(\mathbf{X})$ is the best predictor of ε^2 given \mathbf{X} :

- Write down the MSE of a predictor $h(\mathbf{X})$ for ε^2 .
- What does it mean to be predicting ε^2 ?
- Show that $\sigma^2(\mathbf{X})$ minimizes the MSE and is thus the best prediction.

Solution: The MSE of any function h for ε^2 is

$$\mathbb{E}(\varepsilon^2 - h(\mathbf{X}))^2$$

We say we are predicting ε^2 if we are attempting to estimate its value given the information in the observables \mathbf{X} , using the function estimate of h .

Chapter 3

The Algebra of Least Squares