

## 7. Exercise: Expected value rule and total expectation theorem

### Exercise: Expected value rule and total expectation theorem

6/8 points (graded)

Let  $\mathbf{X}$ ,  $\mathbf{Y}$ , and  $\mathbf{Z}$  be jointly continuous random variables. Assume that all conditional PDFs and expectations are well defined. E.g., when conditioning on  $\mathbf{X} = \mathbf{x}$ , assume that  $\mathbf{x}$  is such that  $f_{\mathbf{X}}(\mathbf{x}) > 0$ . For each one of the following formulas, state whether it is true for all choices of the function  $g$  or false (i.e., not true for all choices of  $g$ ).

$$1. \mathbf{E}[g(\mathbf{Y}) | \mathbf{X} = \mathbf{x}] = \int g(\mathbf{y}) f_{\mathbf{Y}|\mathbf{X}}(\mathbf{y} | \mathbf{x}) d\mathbf{y}$$

True ▼ ✓ Answer: True

$$2. \mathbf{E}[g(\mathbf{y}) | \mathbf{X} = \mathbf{x}] = \int g(\mathbf{y}) f_{\mathbf{Y}|\mathbf{X}}(\mathbf{y} | \mathbf{x}) d\mathbf{y}$$

False ▼ ✓ Answer: False

$$3. \mathbf{E}[g(\mathbf{Y})] = \int \mathbf{E}[g(\mathbf{Y}) | \mathbf{Z} = \mathbf{z}] f_{\mathbf{Z}}(\mathbf{z}) d\mathbf{z}$$

True ▼ ✓ Answer: True

$$4. \mathbf{E}[g(\mathbf{Y}) | \mathbf{X} = \mathbf{x}, \mathbf{Z} = \mathbf{z}] = \int g(\mathbf{y}) f_{\mathbf{Y}|\mathbf{X},\mathbf{Z}}(\mathbf{y} | \mathbf{x}, \mathbf{z}) d\mathbf{y}$$

True ▼ ✓ Answer: True

$$5. \mathbf{E}[g(\mathbf{Y}) | \mathbf{X} = \mathbf{x}] = \int \mathbf{E}[g(\mathbf{Y}) | \mathbf{X} = \mathbf{x}, \mathbf{Z} = \mathbf{z}] f_{\mathbf{Z}|\mathbf{X}}(\mathbf{z} | \mathbf{x}) d\mathbf{z}$$

False ▼ ✗ Answer: True

$$6. \mathbf{E}[g(\mathbf{X}, \mathbf{Y}) | \mathbf{Y} = \mathbf{y}] = \mathbf{E}[g(\mathbf{X}, \mathbf{y}) | \mathbf{Y} = \mathbf{y}]$$

True ▼ ✓ Answer: True

$$7. \mathbf{E}[g(\mathbf{X}, \mathbf{Y}) | \mathbf{Y} = \mathbf{y}] = \mathbf{E}[g(\mathbf{X}, \mathbf{y})]$$

False ▾

✔ Answer: False

$$8. \mathbf{E}[g(X, Z) | Y = y] = \int g(x, z) f_{X,Z|Y}(x, z | y) dy$$

True ▾

✘ Answer: False

**Solution:**

1. True. This is the usual expected value rule, applied to a conditional model where we are given that  $\mathbf{X} = \mathbf{x}$ .
2. False. Here the quantity inside the expectation,  $g(\mathbf{y})$ , is a number (not a random variable). The left-hand side is a function of  $\mathbf{y}$ , whereas on the right-hand side,  $\mathbf{y}$ , is a dummy variable that gets integrated away. So, the formula is wrong on a purely syntactical basis (the left-hand side depends on  $\mathbf{y}$ , while the right-hand side does not).
3. True. This is the total expectation theorem, where we condition on the events  $\mathbf{Z} = \mathbf{z}$ .
4. True. This is the usual expected value rule, applied to a conditional model where we are given that  $\mathbf{X} = \mathbf{x}$  and  $\mathbf{Z} = \mathbf{z}$ .
5. True. This is the same total expectation theorem as in the third part, except that everything is calculated within a conditional model in which event  $\mathbf{X} = \mathbf{x}$  is known to have occurred.
6. True. When we condition on  $\mathbf{Y} = \mathbf{y}$ , we know the value of  $\mathbf{Y}$ , and we can replace  $g(\mathbf{X}, \mathbf{Y})$  by  $g(\mathbf{X}, \mathbf{y})$ .
7. False. Given that  $\mathbf{Y} = \mathbf{y}$ , we need to somehow take into account the conditional distribution of  $\mathbf{X}$ , whereas the right-hand side is determined by the unconditional PDF of  $\mathbf{X}$ .
8. False. The left-hand side is a function of  $\mathbf{y}$ , whereas the right-hand side (after  $\mathbf{y}$  is integrated out) is a function of  $\mathbf{x}$  and  $\mathbf{z}$ . The correct form (expected value rule, in a conditional model) is:

$$\mathbf{E}[g(\mathbf{X}, \mathbf{Z}) | \mathbf{Y} = \mathbf{y}] = \int \int g(\mathbf{x}, \mathbf{z}) f_{\mathbf{X}, \mathbf{Z} | \mathbf{Y}}(\mathbf{x}, \mathbf{z} | \mathbf{y}) d\mathbf{x} d\mathbf{z}.$$

提交

You have used 1 of 1 attempt

❗ Answers are displayed within the problem

讨论

显示讨论