

## Conditional Expectation

Some quick recap on concepts and notation...

Consider a probability space  $(\Omega, \mathcal{F}_0, \mathbb{P})$ . Fix a sub  $\sigma$ -algebra  $\mathcal{F} \subset \mathcal{F}_0$ . Let  $X$  be a random variable on  $\mathcal{F}_0$ , i.e.,  $X : \Omega \rightarrow \mathbb{R}$  measurable w.r.t.  $\mathcal{F}_0$  ( $X^{-1}(A) \in \mathcal{F}_0$  for all Borel set  $A$ , i.e., for all  $A \in \mathcal{R}$ ). Also assume  $X \in L^1$ , i.e.,  $\mathbb{E}[|X|] < \infty$ .

We define the **conditional expectation of  $X$  given  $\mathcal{F} \subset \mathcal{F}_0$**  to be any r.v.  $Y$  such that:

- (i)  $Y \in \mathcal{F}$  (measurable:  $Y^{-1}(A) \in \mathcal{F}$  for all  $A \in \mathcal{R}$ ), and
- (ii)  $\int_A Y \, d\mathbb{P} = \int_A X \, d\mathbb{P}$  for all  $A \in \mathcal{F}$ . (The RHS is  $\mathbb{E}[X \mathbf{1}[A]]$ , a well-defined expectation).

In order for this definition to make sense, we will prove the following statements:

- (1) Existence: there exists  $Y$  satisfying (i) and (ii).
- (2) Integrability: any  $Y$  that satisfies (i) and (ii) is  $L^1$  (integrable).
- (3) Uniqueness: if  $Y$  and  $Y'$  both satisfy the conditions, then  $Y = Y'$  a.s.

With these properties established, we denote any satisfying variable by  $\mathbb{E}[X|\mathcal{F}]$ . *This can also be viewed as an equivalence class in the sense that any member inside this equivalence class has the same  $\int_A Y \, d\mathbb{P}$ . Each element is called a version.*

*Proof.* We first prove *integrability* (D4.1.1). If  $Y$  satisfies (i) and (ii) then  $Y \in L^1$ . Let  $A := \{Y > 0\} = \{\omega \in \Omega : Y(\omega) > 0\}$ . Note  $A \in \mathcal{F}$  since  $A = Y^{-1}((0, \infty))$  and consequently  $A^c \in \mathcal{F}$  by closure of complement. Then

$$\int_A |Y| \, d\mathbb{P} = \int_A Y \, d\mathbb{P} = \int_A X \, d\mathbb{P} \leq \int_A |X| \, d\mathbb{P}$$

and

$$\int_{A^c} |Y| \, d\mathbb{P} = \int_{A^c} -Y \, d\mathbb{P} = - \int_{A^c} Y \, d\mathbb{P} = - \int_{A^c} X \, d\mathbb{P} = \int_{A^c} (-X) \, d\mathbb{P} \leq \int_{A^c} |X| \, d\mathbb{P}.$$

Adding these two gives  $\int_{\Omega} |Y| \, d\mathbb{P} \leq \int_{\Omega} |X| \, d\mathbb{P} < \infty$ .

For *uniqueness*: suppose  $Y, Y'$  both satisfy (i) and (ii). Let

$$A_n := \{Y - Y' \geq 1/n\}.$$

Since  $Y, Y'$  are  $\mathcal{F}$ -measurable, so is their difference, so  $A_n \in \mathcal{F}$  for all  $n$ . Then

$$\begin{aligned} 0 &= \int_{A_n} (X - X) \, d\mathbb{P} = \int_{A_n} X \, d\mathbb{P} - \int_{A_n} X \, d\mathbb{P} = \int_{A_n} Y \, d\mathbb{P} - \int_{A_n} Y' \, d\mathbb{P} \\ &= \int_{A_n} (Y - Y') \, d\mathbb{P} \geq \int_{A_n} n^{-1} \, d\mathbb{P} = n^{-1} \mathbb{P}(A_n) \geq 0. \end{aligned}$$

Note subtraction is well-defined since both integrals involving  $Y, Y'$  are finite by  $L^1$  assumption. But then  $\mathbb{P}(A_n) = 0$ , so  $\mathbb{P}(Y > Y') = \mathbb{P}(\cup A_n) \leq \sum \mathbb{P}(A_n) = 0$ . Identical proof for the other direction.

We'll (informally) prove existence later. □

**Theorem: (D4.1.2)**

If  $X_1 = X_2$  on  $B \in \mathcal{F}$ , then  $\mathbb{E}[X_1|\mathcal{F}] = \mathbb{E}[X_2|\mathcal{F}]$  on  $B$ .

*Proof.* Put  $Y_1 = \mathbb{E}[X_1|\mathcal{F}]$  and  $Y_2$  similarly. Let  $A_n = \{Y_1 - Y_2 \geq 1/n\}$ . Then

$$\begin{aligned} 0 &= \int_{A_n \cap B} (X_1 - X_2) d\mathbb{P} = \int_{A_n \cap B} X_1 d\mathbb{P} - \int_{A_n \cap B} X_2 d\mathbb{P} \\ &= \int_{A_n \cap B} Y_1 d\mathbb{P} - \int_{A_n \cap B} Y_2 d\mathbb{P} = \int_{A_n \cap B} (Y_1 - Y_2) d\mathbb{P} \\ &\geq n^{-1} \mathbb{P}(A_n \cap B). \end{aligned}$$

Hence  $\mathbb{P}(A_n \cap B) = 0$ . Therefore  $\mathbb{P}(Y_1 > Y_2) = \mathbb{P}(\cup(A_n \cap B)) = 0$ . Other direction analogous.  $\square$

*Proof of existence of conditional variables.* This proof quotes results from **Lebesgue-Radon-Nikodym** (525a, Durrett appendix A4). Let  $\mu, \nu$  be two  $\sigma$ -finite measures on  $(\Omega, \mathcal{F})$  (making the entire space the union of countably many finite-measure sets). We assume that  $\nu \ll \mu$  (*absolutely continuous*): if  $\mu(A) = 0$  then  $\nu(A) = 0$ . Then there exists  $f \in \mathcal{F}$  such that for all  $A \in \mathcal{F}$ ,

$$\nu(A) = \int_A f d\mu.$$

*Intuition:*  $f$  is the density of  $\nu$ . More formally, this is called the **Radon-Nikodym derivative**, written  $\frac{d\nu}{d\mu}$ .

Going back to the proof. First suppose  $X \geq 0$ . Let  $\nu(A) := \int_A X d\mathbb{P}$ . By non-negativity assumption of  $X$ ,  $\nu: \mathcal{F} \rightarrow [0, \infty)$ . Finite because  $X \in L^1$ . It follows that

- $\nu$  is a measure:  $\nu(\emptyset) = 0$  and countably additive via either MCT or DCT:  $\nu(\sqcup A_n) = \sum \nu(A_n)$  and  $\int_{\sqcup A_n} X d\mathbb{P} = \sum \int_{A_n} X d\mathbb{P}$ .
- $\nu(\Omega) = \int_{\Omega} X d\mathbb{P} < \infty$ , a finite measure, and
- $\nu \ll \mathbb{P}$ , clearly.

Therefore by L-R-N, there exists a random variable  $Y$  such that for all  $A \in \mathcal{F}$ ,  $\int_A Y d\mathbb{P} = \nu(A) = \int_A X d\mathbb{P}$ .

Now, for general  $X \in L^1$  we sign decompose it into  $X^+ = \max(0, X)$  and  $X^- = \max(0, -X)$ . We know

$$Y^\pm := \mathbb{E}[X^\pm|\mathcal{F}]$$

exist. Let  $Y = Y^+ - Y^-$ . It's clear that (i) is satisfied. To see (ii):

$$\int_A Y d\mathbb{P} = \int_A Y^+ d\mathbb{P} - \int_A Y^- d\mathbb{P} = \int_A X^+ d\mathbb{P} - \int_A X^- d\mathbb{P} = \int_A X d\mathbb{P},$$

where again we used the assumption that everything is  $L^1$ .  $\square$

Some examples of conditionals:

- **(D4.1.3, perfect information)** If  $X \in \mathcal{F}$ , then  $\mathbb{E}[X|\mathcal{F}] = X$ . Both definition checks are trivial.
- **(D4.1.4, no information)** If  $X$  is independent of  $\mathcal{F}$ , i.e.  $\mathbb{P}(X \in B \cap A) = \mathbb{P}(X \in B)\mathbb{P}(A)$  for all  $B \in \mathcal{R}, A \in \mathcal{F}$ ,

then  $\mathbb{E}[X|\mathcal{F}] = \mathbb{E}X$ , a constant variable. Constants are clearly  $\mathcal{F}$ -measurable, since the inverse image is either  $\emptyset$  or  $\Omega$ , so (i) passes. And for (ii), by independence,

$$\int_A X \, d\mathbb{P} = \mathbb{E}[X\mathbf{1}[A]] = \mathbb{E}X \cdot \mathbb{E}[\mathbf{1}[A]] = \int_A \mathbb{E}X \, d\mathbb{P}.$$