

Given a Hilbert space H (like $L^2(\Omega, \mathcal{F}, P)$),
 and a closed subspace $K \subseteq H$, we have the
 orthogonal projection:

$$P_K : H \rightarrow K$$

a linear transformation, with $\|P_K(x)\| \leq \|x\|$,
 characterized by

- $P_K(x)$ is the unique $z \in K$ minimizing $\|x - z\|$
- $P_K(x)$ is the unique $z \in K$ s.t. $x - z \perp K$

↳ I.e. $\langle P_K(x), y \rangle = \langle x, y \rangle \quad \forall y \in K$

If $H = L^2(\Omega, \mathcal{F}, P)$

Recall: if $\mathcal{G} = \sigma(\{A_n\}_{n=1}^\infty)$ for a partition $\{A_n\}_{n=1}^\infty$ of Ω , then

$$\mathbb{E}[E_{\mathcal{G}}[X|Y]] = \mathbb{E}[XY] \quad \forall Y \in L^2(\Omega, \mathcal{G}, P)$$

Prop: Let (Ω, \mathcal{F}, P) be a probability space,

and let $\mathcal{G} \subseteq \mathcal{F}$ be a sub- σ -field.

Then $L^2(\Omega, \mathcal{G}, P) \subseteq L^2(\Omega, \mathcal{F}, P)$ is a closed subspace.

Pf. We're identifying $L^2(\Omega, \mathcal{G}, P) = \{X \in L^2(\Omega, \mathcal{F}, P) : X \text{ is } \mathcal{G}\text{-measurable}\}$

Subspace:

Closed:

Def: If $X \in L^2(\Omega, \mathcal{F}, P)$ and $\mathcal{G} \subseteq \mathcal{F}$ is any sub- σ -field,

the conditional expectation $E_{\mathcal{G}}[X]$ is the random variable

Eg. $\Omega = \{1, 2, 3\}$, $\mathcal{F} = 2^{\Omega}$

$$\mathcal{G} = \sigma(\{1, 2\}, \{3\})$$

If X is \mathcal{G} -measurable, then $X^{-1}\{t\} \in \mathcal{G}$
for each $t \in \mathbb{R}$. In particular, with $t = X(1)$

In fact, for any P , $L^2(\Omega, \mathcal{G}, P) = \mathbb{R}^{\{1, 2, 3\}}$
 $L^2(\Omega, \mathcal{G}, P)$

Take $P = \text{Unif}\{1, 2, 3\}$. Then $E[X] = \frac{1}{3}(X(1)\nu(1) + X(2)\nu(2) + X(3)\nu(3))$

What is $\mathbb{E}_g[X]$? It is the \mathcal{G} -measurable r.v.
that is closest to X

$$\|X - \mathbb{E}_g[X]\|_{L^2} = \min_{Y \in L^2(\Omega, \mathcal{G}, P)} \|X - Y\|$$

I.e. it is the best guess for X , using only the information in \mathcal{G} .

Question: does it only make sense for L^2 ?

Lemma: If $X \in L^2(\Omega, \mathcal{F}, P)$, then $\mathbb{E}[|\mathbb{E}_g[X]|] \leq \mathbb{E}[|X|]$.

Pf: Set $Y = \mathbb{E}_g[X]$. Take $Z = \text{sgn } Y = \begin{cases} 1, & Y > 0 \\ 0, & Y = 0 \\ -1, & Y < 0 \end{cases}$

$\therefore Z \in L^2(\Omega, \mathcal{G}, P)$, and so

$$\mathbb{E}[|Y|] = \mathbb{E}[Y \cdot \text{sgn } Y]$$

The lemma shows that $\|\mathbb{E}_g[X]\|_{L^1} \leq \|X\|_{L^1} \quad \forall X \in L^2$.

Q: if we equip $L^2(\Omega, \mathcal{F}, P)$ with the L^1 -norm,

\mathbb{E}_g is still Lipschitz.

Note that $L^2 \subseteq L^1$ is dense: given $X \in L^1$,

$X \mathbb{1}_{|X| \leq n}$ is bounded and \mathbb{P} -a.s. in L^2 , and $\|X - X \mathbb{1}_{|X| \geq n}\|_{L^1} = \mathbb{E}[|X| \mathbb{1}_{|X| \geq n}] \xrightarrow{n \rightarrow \infty} 0$

Def: If $X \in L^1(\Omega, \mathcal{F}, P)$ and $\mathcal{G} \subseteq \mathcal{F}$, define $\mathbb{E}_{\mathcal{G}}[X]$ as follows:

↪ Take any sequence $X_n \in L^2(\Omega, \mathcal{G}, P)$ s.t. $\|X_n - X\|_{L^1} \rightarrow 0$

↪ Define $\mathbb{E}_{\mathcal{G}}[X] := L^1\text{-}\lim_{n \rightarrow \infty} \mathbb{E}_{\mathcal{G}}[X_n] = L^1\text{-}\lim_{n \rightarrow \infty} P_{L^2(\Omega, \mathcal{G}, P)}(X_n)$

• Exists: $\|\mathbb{E}_{\mathcal{G}}[X_n] - \mathbb{E}_{\mathcal{G}}[X_m]\|_{L^1}$

• Well-defined: if $X_n, Y_n \rightarrow X$ in L^1 , then

$$\|\mathbb{E}_{\mathcal{G}}[X_n] - \mathbb{E}_{\mathcal{G}}[Y_n]\|_{L^1}$$

Prop: (Averaging property / characterization)

For $X \in L^1(\Omega, \mathcal{F}, P)$ and $\mathcal{G} \subseteq \mathcal{F}$, $E_{\mathcal{G}}[X]$ is the unique $L^1(\Omega, \mathcal{G}, P)$ random variable with the property:

$$(\star) \quad E[E_{\mathcal{G}}[X]Y] = E[XY] \quad \forall Y \in \mathcal{B}(\Omega, \mathcal{G}).$$

Pf. If $X \in L^2$, (\star) holds by defⁿ of orth-proj. (b/c $\mathcal{B}(\Omega, \mathcal{G}) \subseteq L^2(\Omega, \mathcal{G}, P)$).

In general, $E_{\mathcal{G}}[X] = L^1 - \lim_{n \rightarrow \infty} E_{\mathcal{G}}[X_n]$ for any $X_n \rightarrow X$ in L^1

$$\begin{aligned} \therefore E_{\mathcal{G}}[X_n]Y &\rightarrow E_{\mathcal{G}}[X]Y \\ X_nY &\rightarrow XY \end{aligned} \quad \text{in } L^1$$

$$\therefore E[E_{\mathcal{G}}[X]Y - XY] = \lim_{n \rightarrow \infty} E[E_{\mathcal{G}}[X_n]Y - X_nY]$$

Conversely, if $Z_1, Z_2 \in L^1(\Omega, \mathcal{G}, P)$ each satisfy

$$E[Z_1Y] = E[Z_2Y] = E[XY] \quad \forall Y \in \mathcal{B}(\Omega, \mathcal{G})$$

then $E[(Z_1 - Z_2)Y] = 0$.

Theorem: (Main Properties of Conditional Expectation)

Let (Ω, \mathcal{F}, P) be a probability space, and $\mathcal{G} \subseteq \mathcal{F}$ a sub- σ -field. The linear transformation

$$E_{\mathcal{G}} : L^1(\Omega, \mathcal{F}, P) \rightarrow L^1(\Omega, \mathcal{G}, P)$$

satisfies:

1. (Monotonicity) if $X \leq Y$ a.s. [P] then $E_{\mathcal{G}}[X] \leq E_{\mathcal{G}}[Y]$ a.s. [P].

2. (Delta-meq.) $|E_{\mathcal{G}}[X]| \leq E_{\mathcal{G}}[|X|]$ a.s. [P].

✓ 3. (Averaging) $E[E_{\mathcal{G}}[X|Y]] = E[XY] \quad \forall Y \in \mathcal{B}(\Omega, \mathcal{G})$

4. (Product Rule) If $Y \in \mathcal{B}(\Omega, \mathcal{G})$, $E_{\mathcal{G}}[XY] = E_{\mathcal{G}}[X] \cdot Y$.

5. (Tower Property) If $\mathcal{H} \subseteq \mathcal{G} \subseteq \mathcal{F}$ are all σ -fields, then

$$E_{\mathcal{G}}[E_{\mathcal{H}}[X]] = E_{\mathcal{H}}[E_{\mathcal{G}}[X]] = E_{\mathcal{H}}[X].$$

Pf. 1. $E[E_{\mathcal{G}}[X|Y]] = E[X|Y] \leq E[Y|Y] = E[E_{\mathcal{G}}[Y|Y]] \quad \forall Y \in \mathcal{G}$.

So, suffices to show that, if $Z \in L^1(\Omega, \mathcal{G}, P)$ and $E[Z|Y] \geq 0 \quad \forall Y \in \mathcal{G}$ then $Z \geq 0$ a.s. [P].

2. $X \leq |X|$ and $-X \leq |X|$

$\therefore \mathbb{E}_Y[X] \leq \mathbb{E}_Y[|X|]$ and

$\mathbb{E}_Y[-X] \leq \mathbb{E}_Y[|X|]$ a.s.

3. ✓

4. Let $Y, Z \in \mathcal{B}(\Omega, \mathcal{F})$. Then

$$\mathbb{E}[\mathbb{E}_Y[X]Y \cdot Z] = \mathbb{E}[\mathbb{E}_Y[X] \cdot YZ] = \mathbb{E}[XYZ]$$

$$\therefore \mathbb{E}[(Y\mathbb{E}_Y[X] - \mathbb{E}_Y[XY])Z]$$

5. $\mathbb{E}_Y[\mathbb{E}_{\mathcal{H}}[X]] = \mathbb{E}_{\mathcal{H}}[\mathbb{E}_Y[X]] = \mathbb{E}_{\mathcal{H}}[X]$ holds for $X \in L^2$ by orth.proj. thm.

Now approximate $X \in L^2$ by $X_n \in L^2$,
and be careful.