

Conditional expectation

(X, M, μ) : σ -finite measure space

$M' \subset M$ sub σ -algebra. (s.t. (X, M', μ) still σ -fin)

$\Rightarrow L^p(X, M', \mu) \subset L^p(X, M, \mu)$ Banach subsp.

Formal requirement (characterization) of conditional expectation

linear map $E : L^p(X, M, \mu) \rightarrow L^p(X, M', \mu)$

- $\|E\| \leq 1$ i.e. $\|E(f)\|_p \leq \|f\|_p$

- $E(f) = f$ for $f \in L^p(X, M', \mu)$

Motivation : μ probability measure

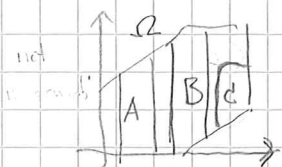
\Rightarrow so we are looking at a probability space.

(Ω, M, \mathbb{P})
 \uparrow set of samples \uparrow encodes probability

X, Y, \dots random variables (measurable funcs. on Ω)
 (not the "first X " as space)

M : encoding what can be "measured" with all rand. vars.

$\Rightarrow M'$: encoding "limited information" with restricted number of rand. vars.



$A, B \in M'$ but not C, \dots

detectable direction. (Ω' below)

$E[X | M'], E[Y | M'], \dots$: M' -measurable

approximation of X, Y, \dots

Rough idea : $\Omega' = \Omega / M'$ $\Omega \rightarrow \Omega'$

$E[X | M'](\omega') = \int_{\{\omega : \pi(\omega) = \omega'\}} X(\omega) d\mathbb{P}(\omega)$
 \leftarrow OK if this has pos. meas

Formally : (back to σ -finite (X, M', μ) , $M' \subset M$.)

construction of $E : L^q(X, M, \mu) \rightarrow L^p(X, M', \mu)$

for $1 < q \leq \infty$ ($\frac{1}{p} + \frac{1}{q} = 1$ for $1 \leq p < \infty$)

Recall : $L^q(X, M, \mu) \cong L^p(X, M, \mu)^*$

$$g \longleftrightarrow \phi_g : f \mapsto \int f \cdot g d\mu$$

$L^p(X, M, \mu)$

Same for M' .

Given $g \in L^q(X, M, \mu)$ take $\psi = \phi_g|_{L^p(X, M', \mu)}$

Claim ψ is a bounded linear functional

on $L^p(X, M', \mu)$, $\|\psi\| \leq \|\phi_g\| = \|g\|_q$

- linearity is obvious from def.

- boundedness : $\|\psi\| = \sup_{\substack{f \in L^p(X, M', \mu) \\ \|f\|_p \leq 1}} |\psi(f)| \leq \sup_{\substack{f \in L^p(X, M, \mu) \\ \|f\|_p \leq 1}} |\phi_g(f)| = \|\phi_g\|$

$\leadsto \psi = \phi_{E(g)}$ for some $E(g) \in L^q(X, M', \mu)$

Then $g \mapsto E(g)$ is linear, contractive $\|E\| \leq 1$.

- linearity $E(\tau g_1 + g_2)$ represents

$$\begin{aligned} \phi_{\tau g_1 + g_2}|_{L^p(X, M', \mu)} &= (\tau \phi_{g_1} + \phi_{g_2})|_{L^p(X, M', \mu)} \\ &= \tau \phi_{E(g_1)} + \phi_{E(g_2)} \end{aligned}$$

$$\text{i.e. } \phi_{E(\tau g_1 + g_2)} = \phi_{\tau E(g_1) + E(g_2)}$$

$$\Rightarrow E(\tau g_1 + g_2) = \tau E(g_1) + E(g_2)$$

- norm estimate $\|E(g)\|_q = \|\psi\| \leq \|\phi_g\| = \|g\|_q$