

I Hilbert spaces, closed subspaces and orthogonal projections

II Conditional expectations

III Martingales

IV Applications to mathematical finance

I Hilbert spaces, closed subspaces and orthogonal projections

V normed linear space

$f: V \rightarrow V$ linear

Is every linear f necessarily continuous?

Example

$V = \mathbb{R}^{(\mathbb{N})} = \{ (a_n) \in \mathbb{R}^{\mathbb{N}} \mid a_n \neq 0 \text{ for only finitely many } n \}$

$$\|a_n\| = \left(\sum_{n=1}^{\infty} a_n^2 \right)^{\frac{1}{2}}$$

$e_n = (0, \dots, 0, 1, 0, \dots)$ basis for V

$f: V \rightarrow V$

$$e_n \mapsto n^2 e_n$$

$$x_n := \frac{1}{n} e_n, \|x_n - 0\| = \|x_n\| = \frac{1}{n} \rightarrow 0 \Rightarrow x_n \rightarrow 0$$

$$\text{but } \|f(x_n)\| = \|n^2 \cdot \frac{1}{n} e_n\| = n \rightarrow \infty$$

①

f is not continuous!

Definition

$f: V \rightarrow V$ linear

f is called bounded $\Leftrightarrow \exists K > 0$ such that

$$\|f(x)\| \leq K \cdot \|x\| \quad \forall x \in V$$

$\|f\| := \sup_{\|x\|=1} \|f(x)\|$ in the norm of the linear operator f

Theorem

$f: V \rightarrow V$ linear

f is continuous $\Leftrightarrow f$ is bounded

(see lecture by Kathrin Spindler)

Proof:

" \Rightarrow " f continuous $\Rightarrow f$ continuous at 0

$\exists \delta > 0$ such that $\|x\| \leq \delta \rightarrow \|f(x)\| \leq 1$

Let $x \neq 0$,

$$\begin{aligned} \|f(x)\| &= \left\| \|x\| f\left(\frac{x}{\|x\|}\right) \right\| = \|x\| \cdot \left\| f\left(\frac{x}{\|x\|}\right) \right\| \\ &\leq \frac{\|x\|}{\delta} \underbrace{\left\| f\left(\delta \frac{x}{\|x\|}\right) \right\|}_{\leq 1} \\ &\leq \frac{1}{\delta} \|x\| \end{aligned}$$

" \Leftarrow " f bounded

$$x_n \rightarrow x$$

$$\|f(x_n) - f(x)\| = \|f(x_n - x)\|$$

$$\leq \|f\| \cdot \|x_n - x\| \rightarrow 0 \quad \checkmark$$

Remark V finite-dim. Then every liner $f: V \rightarrow V$ is bounded.

Remark

V finite-dim. vector space with scalar product $\langle \cdot, \cdot \rangle$

$W \subseteq V$ subspace

V can be decomposed into

$$V = W \oplus W^\perp$$

$$x \in V$$

$$x = x_w + x_\perp$$

$P: V \rightarrow W$ orthog. projection onto W
 $x \mapsto x_w$

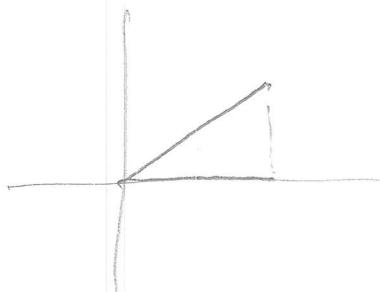
$1-P: V \rightarrow W^\perp$ orthog. projection onto W^\perp
 $x \mapsto x_\perp$

Example

$$V = \mathbb{R}^3$$

$$f : V \rightarrow V$$

$(x_1, x_2, x_3) \mapsto (x_1, 0, 0)$ orth. projection
onto the first coordinate



What about orth. projections in infinite-dim. vector spaces?

Example

$$V = l_2 = \left\{ (a_n) \in \mathbb{R}^N \mid \sum_{n=1}^{\infty} a_n^2 < \infty \right\}$$

$$\langle a_n, b_n \rangle = \sum_{n=1}^{\infty} a_n \cdot b_n$$

$$\text{Let } W = \left\{ (a_n) \in \mathbb{R}^N \mid a_n \neq 0 \text{ for only finitely many } n \right\}$$

W is a linear subspace of l_2 .

Take any $x = (a_n) \in l_2$.

Define $\tilde{a}_n := (a_1, \dots, a_n, 0, \dots)$.

$$(\tilde{a}_n) \in W \quad (\tilde{a}_n) \rightarrow x.$$

④

W is dense in ℓ_2 .

Is there an orthogonal projection $P: V \rightarrow W$?

Take $x \in V$.

Assume there is a decomposition $x = x_w + x_{\perp}$
 $\in W \quad \in W^{\perp}$

$$\|x\|^2 = \|x_w\|^2 + \|x_{\perp}\|^2 \Rightarrow \|x_w\| \leq \|x\|$$

$$\|Px\| = \|x_w\| \leq \|x\|$$

Such P would be a bounded operator.

$$1: V \rightarrow V \quad \text{bounded}$$
$$x \mapsto x$$

$$1-P : V \rightarrow W^{\perp}$$
$$x \mapsto x^{\perp} \quad \text{bounded}$$

Consider $(\tilde{a}_n) \in W$.

$$(\tilde{a}_n) \rightarrow (a_n)$$

$$(1-P)(\tilde{a}_n) = 0 \quad \forall n \in \mathbb{N}$$

$$(1-P)x \neq 0 \quad \begin{array}{l} \text{if } x \in V \setminus W \\ \in W \end{array}$$

$$(\tilde{a}_n) \rightarrow x$$

$$\lim (1-P)(\tilde{a}_n) \not\rightarrow (1-P)x \quad \begin{array}{l} \text{if } W \text{ does not have} \\ \text{an orth. complement } W^{\perp} \end{array} \quad (5)$$

Theorem V Hilbert space, W closed subspace.

Then W has an orth. complement W^\perp , and $V = W \oplus W^\perp$.

Proof: see lecture by Karlheinz Sprinck

II Conditional expectations

Example

tossing of a coin, infinitely often

$$H = \text{heads} \quad T = \text{tails} \quad P(H) = P(T) = \frac{1}{2}$$

$$\mathcal{S}_0 = \{H, T\}$$

$$\mathcal{S} = \{(\omega_k)_{k \in \mathbb{N}} \mid \omega_k \in \mathcal{S}_0\}$$

$$\mathcal{A} = \mathcal{P}(\mathcal{S}) \text{ o-field}$$

We want to describe what happens during the first n experiments.

$$\mathcal{A}_n = \{A_1 \times \dots \times A_n \times \mathcal{S}_0 \times \mathcal{S}_0 \times \dots \mid A_j \subseteq \mathcal{S}_0 \text{ for } 1 \leq j \leq n\}$$

Sub o-field describing the first n experiments

$$\mathcal{A}_n \subseteq \mathcal{A}_m \text{ if } n \leq m$$

play A and B

A receives 1 from B if "heads"
-1 from B if "tails"

gain/loss for t after n experiments

$$S_n = \sum_{k=1}^n (1_H(w_k) - 1_T(w_k))$$

$$S := \lim_{n \rightarrow \infty} S_n$$

S is well-defined for each $w = (w_k)$, potentially $+\infty$ or $-\infty$

S_n is \mathcal{A}_n -measurable

Can we describe S in terms of S_n in a measure-theoretic sense?

Remark

(Ω, \mathcal{A}, P) probability space

$\mathcal{A}_0 \subseteq \mathcal{A}$ sub σ -field

consider $L^2(\Omega) = \{X: \Omega \rightarrow \mathbb{R}, X \text{ } \mathcal{A}\text{-measurable} \mid \int X^2 dP < \infty\}$

$L_0^2(\Omega) = \{X \in L^2(\Omega) \mid X \text{ } \mathcal{A}_0\text{-measurable}\}$

$L_0^2(\Omega)$ is a linear subspace of $L^2(\Omega)$.

It is a closed subspace:

Let $(X_n) \in L_0^2$, $X_n \rightarrow X$ (in L^2 -sense).

Then a suitable subsequence converges pointwise to X.

Since limits of \mathcal{A}_0 -measurable functions are \mathcal{A}_0 -measurable,

X is \mathcal{A}_0 -measurable.

Connice the orthogonal projection of a random variable $X \in L^2(\Omega)$ onto $L_0^2(\Omega)$, written as $X_0 = P(X)$.

$$X = X_0 + X_{\perp}$$

Now take $A \in \mathcal{A}_0$. $1_A \in L_0^2(\Omega)$.

$$\langle X, 1_A \rangle = \langle X_0 + X_{\perp}, 1_A \rangle = \underbrace{\langle X_0, 1_A \rangle}_{=0} + \underbrace{\langle X_{\perp}, 1_A \rangle}_{=0}$$

$$\int_X dP = \int_X X \cdot 1_A dP = \langle X, 1_A \rangle = \langle X_0, 1_A \rangle = \int_{A^c} X_0 dP \quad \forall A \in \mathcal{A}_0$$

Definition

(Ω, \mathcal{A}, P) probability space

$\mathcal{A}_0 \subseteq \mathcal{A}$ sub σ -field

$$X : \Omega \rightarrow \mathbb{R} \quad L^2$$

The conditional expectation of X w.r.t. \mathcal{A}_0 is any \mathcal{A}_0 -measurable random variable that satisfies

$$\int_X dP = \int_Y dP \quad \forall A \in \mathcal{A}_0.$$

Y is uniquely defined in L^2 -sense (that is, P almost everywhere). We write $Y = E(X|A_0)$.

Example

$$\Omega = \{1, 2, 3, \dots, 6\}, \quad A = P(\Omega)$$

$$X: \Omega \rightarrow \mathbb{R}$$

$$\begin{aligned} 1 &\mapsto 1 \\ 2 &\mapsto 2 \\ &\vdots \\ 6 &\mapsto 6 \end{aligned}$$

$$A_0 = \{\emptyset, \Omega, \{1, 2, 3\}, \{4, 5, 6\}\}$$

The conditional expectation $E(X|A_0)$ must be of the form

$$\left. \begin{array}{l} 1 \\ 2 \\ 3 \end{array} \right\} \mapsto a \quad \left. \begin{array}{l} 4 \\ 5 \\ 6 \end{array} \right\} \mapsto b \quad , \quad a, b \in \mathbb{R}.$$

$$\int_X dP = \int_{\Omega} X \cdot 1_{\{1, 2, 3\}} dP = 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + 3 \cdot \frac{1}{6} = 1$$

$$\int_{\{1, 2, 3\}} dP = a \cdot \frac{1}{2} \Rightarrow a = 2$$

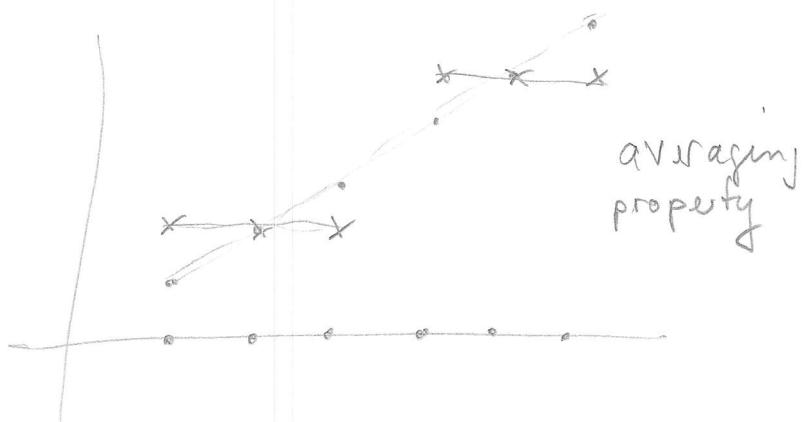
$$\int_{\{4, 5, 6\}} X dP = (4+5+6) \cdot \frac{1}{6} = \frac{5}{2} \quad , \quad \int_{\{4, 5, 6\}} b dP = b \cdot \frac{1}{2}$$

$$\Rightarrow b = 5$$

$E(X|A_0)$ is given by

$$\begin{matrix} 1 \\ 2 \\ 3 \end{matrix} \} \mapsto 2$$

$$\begin{matrix} 4 \\ 5 \\ 6 \end{matrix} \} \mapsto 5$$



Proposition (properties of taking conditional expectation)

(Ω, \mathcal{A}, P) probability space

$A_0 \subseteq \mathcal{A}$ sub σ -field

$X, X_1, X_2 : \Omega \rightarrow \mathbb{R}$ random variables $\in L^2$

(a) $E(a_1 X_1 + a_2 X_2 | A_0) = a_1 E(X_1 | A_0) + a_2 E(X_2 | A_0)$

(b) Let X be A_0 -measurable.

Then $E(X | A_0) = X$

(c) $A_0 \subseteq A_1 \subseteq \mathcal{A}$ sub σ -fields

Then $E(E(X | A_0) | A_1) = E(E(X | A_1) | A_0)$
 $= E(X | A_0)$

(d) $E(E(X | A_0)) = E(X)$

(e) X independent of A_0

Then $E(X | A_0) = E(X)$

Proof : (a), (b), (c) obvious because of the projection property

$$(d) \int_A E(X|t_0) dP = \int_A X dP \quad \forall A \in t_0$$

Set $A = \Omega$

(e) Let $A \in t_0$.

$$\int_A X dP = E(1_A \cdot X) = E(1_A) \cdot E(X)$$

$$= P(A) \cdot E(X)$$

$$= \int_A E(X) dP$$

Proposition (best approximation property)

$t_0 \subseteq \mathcal{A}$ sub σ -field

Consider $L^2(\Omega)$ and $L_0^2(\Omega) = \{X \in L^2(\Omega) \mid X \text{ is } t_0\text{-meas.}\}$

For each $X \in L^2(\Omega)$ there is uniquely one $Y \in L_0^2(\Omega)$ such that

$$\|X - Y\|_2 = \inf_{Z \in L_0^2} \|X - Z\|_2 .$$

This is $Y = E(X|A_0)$.

Can we define conditional expectations $E(X|A_0)$ for $X \in L^1$ as well?

Proposition

(Ω, \mathcal{A}, P) probability space

Then $L^2(\Omega) \xrightarrow{\text{dense}} L^1(\Omega)$.

This also holds for any $(\Omega, \mathcal{A}, \mu)$ with $\mu(\Omega) < \infty$.

Proof

1st version ("by hand")

$$\int_{\Omega} |X| dP = \int_{\{|X| \geq 1\}} |X| dP + \int_{\{|X| < 1\}} |X| dP$$

$$\leq \int_{\{|X| \geq 1\}} |X|^2 dP + \int_{\{|X| < 1\}} |X| dP$$

$$\Rightarrow \|X\|_1 \leq \left(\|X\|_2 \right)^2 + 1$$

$$\Rightarrow (X \in L^2 \Rightarrow X \in L^1)$$

2nd version (using Hölder's inequality)

$$\|X\|_1 = \|X \cdot 1\|_1 \leq \|X\|_2 \cdot \underbrace{\|1\|_2}_{=1}$$

We still have to prove "dense".

Let $X \in L^1$.

Define $X_n := X \cdot 1_{\{|X| \leq n\}}$, $X_n \in L^2$

Then $X_n \rightarrow X$ pointwise and in L^1

(by the dominated convergence theorem).

Thus L^2 is dense in L^1 .

Construction

Let $X \in L^1$.

Define $X_n := X \cdot 1_{\{|X| \leq n\}} \in L^2$

$Y_n := E(X_n | \mathcal{A}_0)$ is $\in L^2$

Show that (Y_n) is a Cauchy sequence in L^1 .

Let $\gamma := \lim Y_n$ and set $E(X | \mathcal{A}_0) := \gamma$.