

Lecture 5

29/4/2021

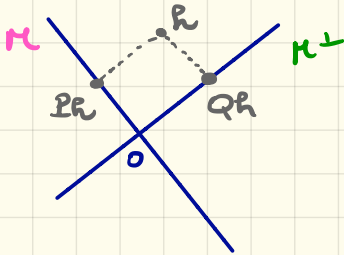
Recall: $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ Hilbert space

orthogonality: $x, y \in \mathcal{H} \quad x \perp y \Leftrightarrow \langle x, y \rangle = 0$

$$A \subseteq \mathcal{H} \quad A^\perp = \{x \in \mathcal{H} : \langle x, a \rangle = 0 \quad \forall a \in A\}$$

\uparrow closed vector subspace

GOAL: Define orthogonal projections:



$$R = PR + QR$$

\uparrow
unique way

Prop $K \subseteq \mathcal{H}$ closed convex set

$$\forall R \in \mathcal{H} \quad \exists! k_0 \in K \quad \|R - k_0\| = \inf \{ \|R - k\| : k \in K \}$$

$\underbrace{\hspace{10em}}_{\substack{\text{"} \\ d(R, K)}}$

Theorem Let $M \subseteq H$ be a closed vector subspace

$$h \in H$$

↑ it is convex

(i) Let $f_0 \in M$ be the unique element of M st

$$\|h - f_0\| = \text{dist}(h, M)$$

↑ it exists because of the previous proposition

Then: $h - f_0 \perp M$

(ii) Conversely if $f_0 \in M$ st $h - f_0 \perp M$

$$\Rightarrow \|h - f_0\| = \text{dist}(h, M)$$

proof (i) $f_0 \in M$ $\|h - f_0\| = d(h, M) = \inf \{ \|h - f\| : f \in M \}$

Take any $f \in M \Rightarrow f_0 + f \in M$ ← M is a vector subspace

$$\|h - f_0\|^2 \leq \|h - \underbrace{(f_0 + f)}_{\in M}\|^2 = \|(h - f_0) - f\|^2 =$$

$$\uparrow = \|h - f_0\|^2 + \|f\|^2 - 2 \operatorname{Re} \langle h - f_0, f \rangle$$

Polan identity

$$\Rightarrow \boxed{2 \operatorname{Re} \langle h - f_0, f \rangle \leq \|f\|^2} \quad \forall f \in M$$

Fix $f \in \mathcal{H}$ and

$$\text{look at } \underbrace{\langle R-f_0, f \rangle}_{\in \mathbb{C}} = \rho e^{i\bar{\theta}} \quad \rho \geq 0 \quad \bar{\theta} \in \mathbb{R} \quad (*)$$

and apply previous inequality to $t e^{i\bar{\theta}}$ $t \in \mathbb{R}$

$$\underbrace{2 \operatorname{Re} \langle R-f_0, t e^{i\bar{\theta}} f \rangle}_{\text{"}} \leq t^2 \|f\|^2$$

$$2 \operatorname{Re} t e^{-i\bar{\theta}} \underbrace{\langle R-f_0, f \rangle}_{\rho e^{i\bar{\theta}}} = 2 t \rho$$

$$\forall t \in \mathbb{R}, \quad \underbrace{2 t \rho}_{\text{"}} \leq \underbrace{t^2 \|f\|^2}_{\text{"}}$$

Cancel as $t \rightarrow 0+$

$$\uparrow \rho > 0 \quad 2\rho \leq t \|f\|^2$$

$$\Rightarrow \rho \leq 0 \quad \text{but we know } \rho \geq 0$$

$$\Rightarrow \rho = 0$$

$$\Rightarrow \text{look at } (*) \quad \langle R-f_0, f \rangle = 0$$

$$\Rightarrow \text{this is true } \forall f \in \mathcal{H} \quad \Rightarrow R-f_0 \perp \mathcal{H}_0$$

(iv) $f_0 \in M$ $h - f_0 \perp M$ $(\Leftrightarrow \langle h - f_0, f \rangle = 0 \quad \forall f \in M)$

Let $f \in M$

$$h - f_0 \perp \underbrace{f_0 - f}_{\substack{D \\ M}} \leftarrow \text{vector subspace}$$

$$\|h - f\|^2 = \|(h - f_0) + (f_0 - f)\|^2 =$$

$$\Rightarrow = \|h - f_0\|^2 + \underbrace{\|f_0 - f\|^2}_{\geq 0} \geq \|h - f_0\|^2$$

Pythagorean Thm

$$\forall f \in M \quad \|h - f_0\| \leq \|h - f\|$$

$$\Rightarrow \|h - f_0\| \leq \text{dist}(h, M)$$

$$\Rightarrow \|h - f_0\| = \text{dist}(h, M) \quad \square$$

\uparrow
 $f_0 \in M$

Orthogonal projection

M closed vector subspace

$$P: H \rightarrow M$$

$$h \mapsto \frac{P(h)}{h} = f_0$$

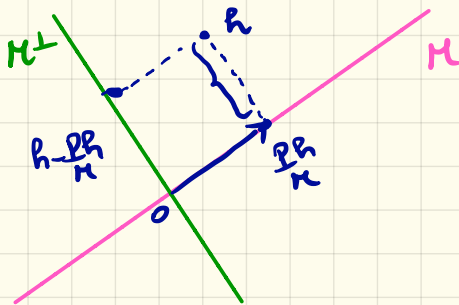
THIS IS WELL-DEFINED BECAUSE f_0 IS UNIQUELY DETERMINED

previous thm

orthogonal projection of h to M

unique element of M resulting $\text{dist}(h, M)$

From previous thm $h - P(h) \perp M$



direct sum of vector subspaces

$$H = M \oplus M^\perp$$

$$h = \frac{P(h)}{h} + (h - \frac{P(h)}{h})$$

Properties of the orthogonal projection

Thm: M closed linear subspace

$P: H \rightarrow M$ orthogonal projection

$h \mapsto P(h) \leftarrow$ the unique element $\overset{of M}{\text{realizing}}$
 $d(h, M)$

a) P is a linear map

b) $\|P(h)\| \leq \|h\| \quad \forall h \in H$

c) $P^2 = P \quad (P^2 = P \circ P)$

d) $\text{ker } P = M^\perp \quad \text{Range}(P) = M$

$\uparrow P(h) = 0 \Leftrightarrow h \in M^\perp$

Keep in mind!

$P - P_h \perp M$ (~~xxx~~)

$\|h - P(h)\| = \text{dist}(h, M)$
(~~xxx~~)

proof

a) $h_1, h_2 \in H \quad \alpha_1, \alpha_2 \in \mathbb{F}$

Show that $P(\alpha_1 h_1 + \alpha_2 h_2) = \alpha_1 P(h_1) + \alpha_2 P(h_2)$

$f \in \mathcal{M}$

$$\langle (\alpha_1 h_1 + \alpha_2 h_2) - (\alpha_1 P(h_1) + \alpha_2 P(h_2)), f \rangle$$

← Linearity of $\langle \cdot, \cdot \rangle$

$$= \alpha_1 \underbrace{\langle h_1 - P(h_1), f \rangle}_{\substack{0 \\ \leftarrow (**)}} + \alpha_2 \underbrace{\langle h_2 - P(h_2), f \rangle}_{\substack{0 \\ \leftarrow (**)}}$$

$$= 0$$

this means

$$\langle (\alpha_1 h_1 + \alpha_2 h_2) - \underbrace{(\alpha_1 P(h_1) + \alpha_2 P(h_2))}_{\substack{\sigma \\ \sim}}, f \rangle \perp \mathcal{M}$$

\Rightarrow by our previous theorem (ii)

$$\|(\alpha_1 h_1 + \alpha_2 h_2) - \sigma\| = \text{dist}(\alpha_1 h_1 + \alpha_2 h_2, \mathcal{M})$$

\Rightarrow by uniqueness σ must be the projection:

$$P(\alpha_1 h_1 + \alpha_2 h_2) = \alpha_1 P(h_1) + \alpha_2 P(h_2) \quad \square$$

(b) Take $h \in H = M \oplus M^\perp$

$$h = \underbrace{Ph}_{\in M} + \underbrace{(h - Ph)}_{\in M^\perp}$$

$$\|h\|^2 = \|Ph\|^2 + \|h - Ph\|^2 \geq \|Ph\|^2 \quad \square$$

↑

Pyth. thm

$Ph \perp (h - Ph)$

$\text{dist}(f, M) = 0 \Rightarrow P(f) = f$

c) Take $f \in M \Rightarrow P(f) = f$

Then $\forall h \in H \quad P(h) \in M$ by definition of projector

$$P^2(h) = P(\underbrace{P(h)}_{\in M}) = P(h) \quad \forall h \in H$$

$$\Rightarrow \boxed{P^2 = P}$$

$$\downarrow) \text{ Ker } P = \mathcal{K}^\perp$$

$$\begin{aligned} \cdot \text{ if } P(\mathbf{r}) = \mathbf{0} & \Rightarrow \text{by properties of projection} \\ & \mathbf{r} - \underbrace{P(\mathbf{r})}_{\mathbf{0}} \in \mathcal{K}^\perp \\ & \Rightarrow \mathbf{r} \in \mathcal{K}^\perp \end{aligned}$$

$$\text{Ker } P \subseteq \mathcal{K}^\perp$$

$$\begin{aligned} \cdot \text{ if } \mathbf{r} \in \mathcal{K}^\perp & \text{ then } \mathbf{0} \text{ is the unique vector in } \mathcal{K} \\ & \text{st } \mathbf{r} - \mathbf{0} \perp \mathcal{K} \\ & \Rightarrow P(\mathbf{r}) = \mathbf{0} \end{aligned}$$

The fact that $\text{range}(P) = \mathcal{K}$ is clear from the definition of $P: H \rightarrow \mathcal{K}$

□

Remark It was already in the proof that M was
a closed vector subspace

↳ in infinite dim not all vector subspaces are
closed

Ex. $H = L^2([a, b], \text{leb}) \leftarrow$ Hilbert space

$$\langle f, g \rangle = \int_a^b f \bar{g} \, dx$$

Take $M = C[a, b]$

↑ vector subspace

→ IT IS NOT CLOSED! IT IS DENSE IN $L^2[a, b]$

$$\overline{M} = L^2[a, b]$$

Propositions

I) M is a closed vector subspace of H

$$\Leftrightarrow (M^\perp)^\perp = M$$

II) let $S \subseteq H$ subset of H

$$\Rightarrow (S^\perp)^\perp = \overline{\text{span}(S)}$$

✓ (exercise) using II.

In particular S is dense in $H \Leftrightarrow S^\perp = \{0\}$

Remark: $A \subseteq H$ is a subset of H

$\Rightarrow A^\perp$ is a closed vector subspace of H

→ it was proved last lecture

$A = M^\perp$

proof I) (\Leftarrow) If $M = (M^\perp)^\perp \Rightarrow M$ is a closed vector subspace

$$(\Rightarrow) x \in (M^\perp)^\perp$$

since M is a closed vector subspace, consider P_M orthog. proj on M

$$x = \underbrace{P(x)}_M + \underbrace{(x - P(x))}_{M^\perp}$$

$$\Rightarrow \langle x, y_0 \rangle = \langle x, x \rangle + \langle y_0, y_0 \rangle = \|y_0\|^2 \Rightarrow \|y_0\| = 0$$

$$\Rightarrow y_0 = 0$$

$$\Rightarrow x = P(x) \in M$$

$x \perp M^\perp$
 $x_0 \perp y_0$

$$\Rightarrow x = P(x) \Rightarrow x \in M$$

$$(M^\perp)^\perp \subseteq M$$

$$\text{Observe that } M \subseteq (M^\perp)^\perp \left. \vphantom{M \subseteq (M^\perp)^\perp} \right\} = M = (M^\perp)^\perp$$

↑ exercise (it is essentially the definition)

I) $\overline{\text{Span}(S)}$ ← the smallest closed subspace containing S

observe

$$S \subseteq \overline{\text{Span}(S)}$$

exercise
if $A \subseteq B$
 $\Rightarrow B^\perp \subseteq A^\perp$

$$\Rightarrow (\overline{\text{Span}(S)})^\perp \subseteq S^\perp$$

$$\Rightarrow (S^\perp)^\perp \subseteq (\overline{\text{Span}(S)^\perp})^\perp \stackrel{\text{ITEM (I)}}{=} \overline{\text{Span}(S)}$$

\cup
 S ↑

easy to check as above

$M = \overline{\text{Span}(S)}$

consider a vector subspace containing S being $\overline{\text{Span}(S)}$, the smallest
 $\Rightarrow (S^\perp)^\perp = \overline{\text{Span}(S)}$ ◻