

H spazio con prodotto scalare

È uno spazio vettoriale dotato di un prodotto scalare, verificante

$$i) (x, y) = (y, x) \quad \forall x, y \in H$$

ii) bilineare (cioè lineare rispetto a ognuno dei due vettori)

$$iii) (x, x) \geq 0 \quad \forall x \in H$$

$$iv) (x, x) = 0 \iff x = 0$$

OSS uno spazio con prodotto scalare è anche normato, con
la norma $\|x\| = \sqrt{(x, x)}$

DEF Uno spazio con prodotto scalare si dice **spazio di Hilbert** se è completo come spazio metrico

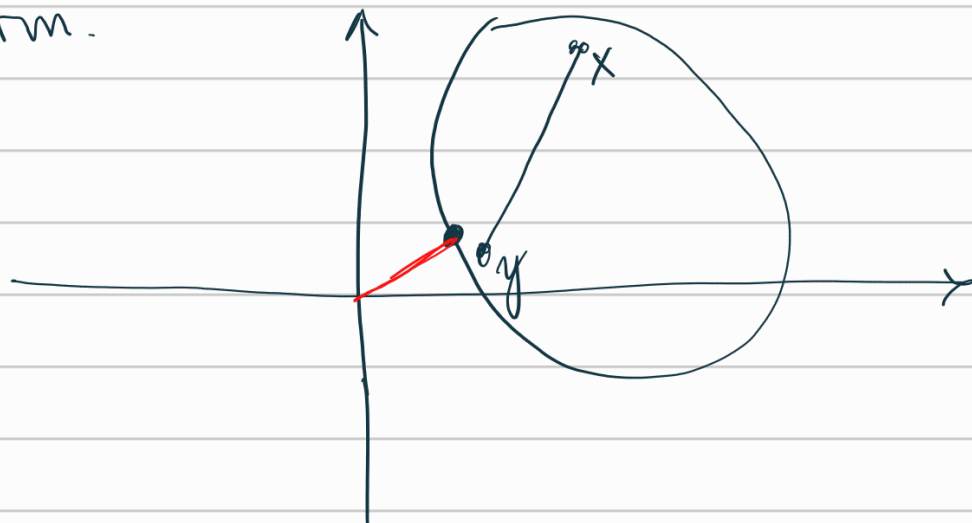
ossia: $\{x_n\}$ di Cauchy $\Rightarrow x_n$ converge a un certo x
 \Downarrow def

$$\forall \varepsilon > 0 \exists \bar{n} \text{ t.c. } \|x_n - x_m\| < \varepsilon \quad \forall n, m \geq \bar{n}$$

THEOREM. H Hilbert space

let $E \subseteq H$ be a closed and convex subset.

Then E admits a unique point x of minimal norm.



E convex means:

$$\forall x, y \in E \quad \forall t \in (0, 1) \quad tx + (1-t)y \in E$$

E closed means:

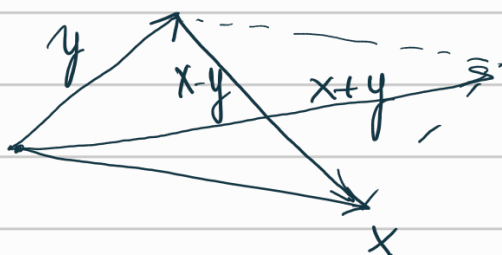
$$\text{If } \begin{array}{l} x_n \in E \\ x_n \rightarrow x \end{array} \quad \Bigg| \Rightarrow \quad x \in E$$

Proof of Thm.

LEMMA Parallelogram identity.

$\forall x, y \in H$ space with scalar product

$$\|x+y\|^2 + \|x-y\|^2 = 2\|x\|^2 + 2\|y\|^2$$



Proof of lemma. It is enough to expand all norms,
for instance

$$\|x+y\|^2 = (x+y, x+y) = \underbrace{(x, x)}_{\|x\|^2} + \underbrace{(y, y)}_{\|y\|^2} + 2(x, y)$$

Proof of theorem.

Let $\delta = \inf \{\|x\|, x \in E\}$

We want to check that $\exists! x \in E$ s.t. $\|x\| = \delta$

Consequence of the parallelogram identity

$$\begin{aligned} \forall x, y \in E \\ \|x-y\|^2 &= 2\|x\|^2 + 2\|y\|^2 - \underbrace{\| \frac{x+y}{2} \|^2}_{\substack{E \text{ by convexity} \\ \leq \delta^2}} \leq \\ &\leq 2\|x\|^2 + 2\|y\|^2 - 4\delta^2 \end{aligned}$$

Uniqueness of the point x.

Assume that $x, y \in E$ satisfy $\|x\| = \|y\| = \delta$

$$\|x-y\|^2 \leq 2\underbrace{\|x\|^2}_{\delta^2} + 2\underbrace{\|y\|^2}_{\delta^2} - 4\delta^2 = 0 \Rightarrow x=y.$$

Existence Let $\{x_n\}$ be a "minimizing sequence", i.e.

$$x_n \in E, \quad \|x_n\| \rightarrow \delta.$$

Then this sequence is Cauchy.

$$\|x_n - x_m\|^2 \leq 2\underbrace{\|x_n\|^2}_{\delta^2} + 2\underbrace{\|x_m\|^2}_{\delta^2} - 4\delta^2 \xrightarrow{\text{for } n, m \text{ large}} 0$$

H is complete, $x_n \rightarrow x \in H$

E is closed $\Rightarrow x \in E$.

If we show that $\|\cdot\|$ is a continuous function

$$\|x_n\| \rightarrow \delta = \inf \{ \|y\|, y \in E \}$$



$$\|x\|$$

$\Rightarrow \|x\| = \delta$ as required.

The norm is continuous, indeed, by the triangle inequality

$$\|x\| = \|x - y + y\| \leq \|x - y\| + \|y\|$$

$$\Rightarrow \|x\| - \|y\| \leq \|x - y\|$$

exchange x and y

$$\|y\| - \|x\| \leq \|x - y\|$$

$$x_n \rightarrow x$$



$$\|x_n\| \rightarrow \|x\|$$

$$|\|x\| - \|y\|| \leq \|x - y\| \Rightarrow \|x\| \text{ is a continuous function on } H.$$

Orthogonal projections.

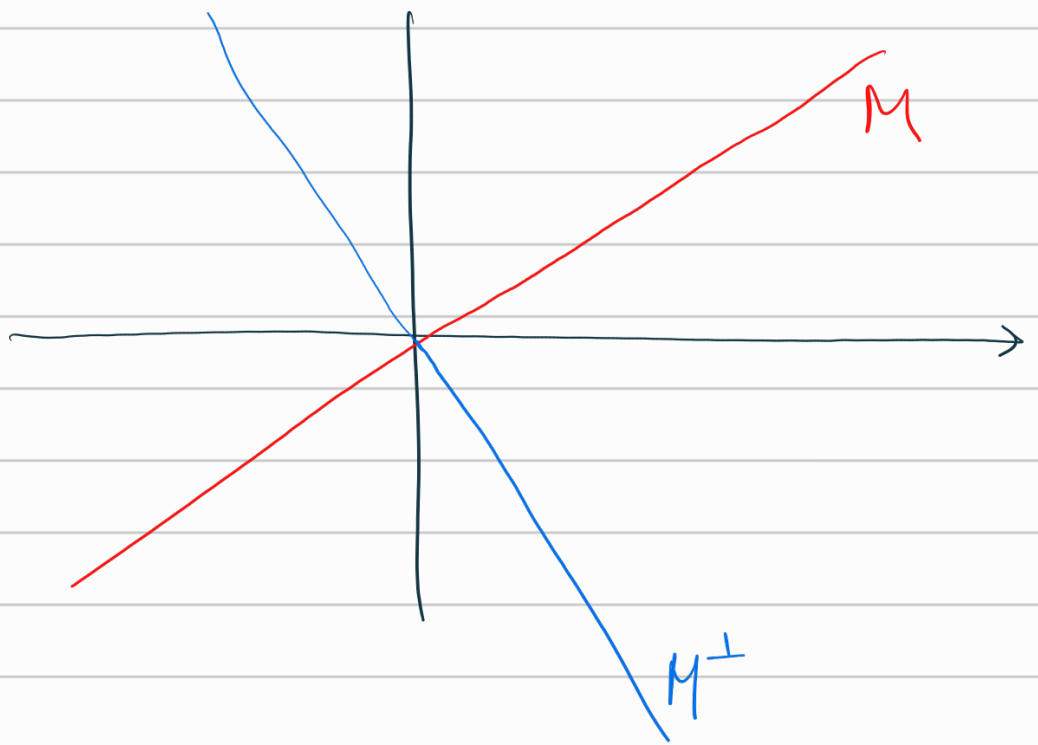
Def $x, y \in H$. We say that they are orthogonal,

if $(x, y) = 0$, and we will write $x \perp y$

Def. Let M be a vector subspace of H .

We define the orthogonal complement M^\perp of M as

$$M^\perp = \{ y \in H \text{ s.t. } y \perp x \quad \forall x \in M \}.$$



Rem M^\perp is also a subspace.

$$\begin{array}{l} y \in M^\perp \\ z \in M^\perp \end{array} \Bigg| \begin{array}{l} ? \\ \rightarrow \end{array} y+z \in M^\perp$$

$$\text{Let } x \in M \Rightarrow (y+z, x) = \underbrace{(y, x)}_{=0} + \underbrace{(z, x)}_{=0} = 0$$

Similarly

$$\begin{array}{l} y \in M^\perp \\ \lambda \in \mathbb{R} \end{array} \Bigg| \Rightarrow \lambda y \in M^\perp$$

So M^\perp is a subspace.

It is also a closed subspace.

$$\begin{array}{l} y_n \in M^\perp \\ y_n \rightarrow y \end{array} \Bigg| \begin{array}{l} ? \\ \Rightarrow \end{array} y \in M^\perp$$

$$\text{let } x \in M, 0 = (y_n, x) \xrightarrow{?} (y, x) = 0$$

$$2) \|x - P(x)\| = \min \{\|x - y\| : y \in M\}$$

$$3) \|x\|^2 = \|P(x)\|^2 + \|Q(x)\|^2 \quad (\text{Pythagoras's theorem}).$$

4) P, Q are linear.

Proof. Let $x \in H$. We consider the set

$$x + M = \{x + y : y \in M\} \quad \text{is convex and closed.}$$

So, by the previous theorem,

$\exists!$ element of $x + M$ of minimal norm.

This I will call $Q(x)$, and I define

$$P(x) = x - Q(x).$$

I must prove that $P(x) \in M$
 $Q(x) \in M^\perp$

$$Q(x) \in x + M \Rightarrow Q(x) = x + y, \quad y \in M$$

$$P(x) = x - Q(x) = \cancel{x} - (\cancel{x} + y) = -y \in M.$$

$$(Q(x), y) \stackrel{?}{=} 0 \quad \forall y \in M^\perp$$

I can always assume $\|y\| = 1$, otherwise replace y by $\frac{y}{\|y\|}$

let $z = Q(x)$. By minimality

$$\cancel{\|z\|^2} \leq \|z + \alpha y\|^2 = (z + \alpha y, z + \alpha y) =$$

$$= \|z\|^2 + \alpha^2 \|y\|^2 + 2\alpha(z, y)$$

$$\alpha^2 \|y\|^2 + 2\alpha(z, y) \geq 0 \quad \forall \alpha \in \mathbb{R}$$

$$\Rightarrow 0 \geq \Delta = 4(z, y)^2 \Rightarrow (z, y) = 0 \Rightarrow Q(x) \in M^\perp$$

This decomposition is unique: indeed

$$\text{Assume that } x = \underbrace{P(x)}_M + \underbrace{Q(x)}_{M^\perp} = \underbrace{x_1}_M + \underbrace{x_2}_{M^\perp}$$

$$\Rightarrow \underbrace{P(x)}_M - x_1 = x_2 - \underbrace{Q(x)}_{M^\perp} \Rightarrow \text{since } M \cap M^\perp = \{0\}$$

$$P(x) = x_1, \quad Q(x) = x_2 \Rightarrow \text{Uniqueness.}$$

Now we show 4) (Linearity of P and Q).

let $x, y \in H, \alpha, \beta \in \mathbb{R}$.

we want to show that

$$\underbrace{P(\alpha x + \beta y)}_Q = \alpha \underbrace{P(x)}_Q + \beta \underbrace{P(y)}_Q$$

$$x = P(x) + Q(x)$$

$$y = P(y) + Q(y)$$

$$\alpha x + \beta y = \alpha P(x) + \alpha Q(x) + \beta P(y) + \beta Q(y)$$

||

$$\alpha x + \beta y = P(\alpha x + \beta y) + Q(\alpha x + \beta y)$$

$$\Rightarrow \underbrace{\alpha P(x) + \beta P(y) - P(\alpha x + \beta y)}_{\substack{\text{L.H.S.} \\ \text{M}^\perp}} = \underbrace{Q(\alpha x + \beta y) - \alpha Q(x) - \beta Q(y)}_{\substack{\text{R.H.S.} \\ \text{M}^\perp}}$$

so both r.h.s. and l.h.s. must be zero \Rightarrow linearity.

let us prove 3) Pythagoras's identity

$$\begin{aligned} \|x\|^2 &= \|(P(x) + Q(x), P(x) + Q(x))\| \\ &= \|P(x)\|^2 + \|Q(x)\|^2 + 2 \underbrace{(P(x), Q(x))}_0 \end{aligned}$$

2) is trivial, since

$$\begin{aligned} \|x - P(x)\| &= \min \{ \|x + y\|, y \in M \} && \text{I write} \\ &= \min \{ \|x - \tilde{y}\|, \tilde{y} \in M \} && y = -\tilde{y} \\ &&& \text{with } \tilde{y} \in M \end{aligned}$$

Fourier series in Hilbert spaces

V vector space, $u_1, u_2, \dots, u_k \in V$

DEF u_1, u_2, \dots, u_k are linearly independent if

$$u = c_1 u_1 + c_2 u_2 + \dots + c_k u_k = 0 \quad (c_1, c_2, \dots, c_k \in \mathbb{R})$$

\Downarrow

$$c_1 = c_2 = \dots = c_k = 0$$

Now, if H is a vectorspace with scalar product.

DEF u_1, u_2, \dots, u_k are orthonormal if

$$(u_i, u_j) = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

Remark

u_1, \dots, u_k orthonormal $\Rightarrow u_1, \dots, u_k$ linearly independent.

Assume that u_1, \dots, u_k are orthonormal.

If

$$0 = c_1 u_1 + c_2 u_2 + \dots + c_k u_k = \sum_{i=1}^k c_i u_i$$

$$0 = (0, 0) = \left(\sum_{i=1}^k c_i u_i, \sum_{j=1}^k c_j u_j \right) = (\text{bilinearity})$$

$$= \sum_{i=1}^k \sum_{j=1}^k c_i c_j (u_i, u_j) = \sum_{i=1}^k c_i^2 \cdot 1$$

$$\Rightarrow c_i = 0 \quad \forall i = 1, \dots, k.$$

PROP $\{u_1, \dots, u_k\}$ orthonormal set of H .

If $x = c_1 u_1 + \dots + c_k u_k = \sum_{i=1}^k c_i u_i$, then

1) $c_i = (x, u_i)$ ← Fourier coefficients $\forall i = 1, \dots, k$

2) $\|x\|^2 = \sum_{i=1}^k c_i^2$ Pythagoras' theorem.

Proof of 1) $(x, u_i) = \left(\sum_{j=1}^k c_j u_j, u_i \right) =$
 $= \sum_{j=1}^k c_j (u_j, u_i) = c_i$

Proof of 2)

$$\|x\|^2 = (x, x) = \left(\sum_{i=1}^k c_i u_i, \sum_{j=1}^k c_j u_j \right) = \text{as before}$$
$$= \sum_{i=1}^k c_i^2 \quad \square$$

Let $\{u_1, \dots, u_k\}$ be an orthonormal set in H Hilbert space

I want to approximate a generic point of H , say $x \in H$, with linear combinations of $\{u_1, \dots, u_k\}$.

I need to find constant c_1, \dots, c_k such that the linear combination $\sum_{i=1}^k c_i u_i$ best approximates x , i.e. such that

$$\left\| x - \sum_{i=1}^k c_i u_i \right\| \text{ is minimal.}$$

So we want to use orthogonal projections on the subspace

$$M = \text{span}\{u_1, \dots, u_k\} = \left\{ \sum_{i=1}^k c_i u_i, c_i \in \mathbb{R} \right\}$$

It is clear that M is a subspace

We must show that M is closed.

So assume that $\{x^{(n)}\} \in M$, $x^{(n)} \rightarrow x$.

We must show that $x \in M$

$\{x^{(n)}\}$ converges $\Rightarrow \{x^{(n)}\}$ is a Cauchy sequence

$$\Rightarrow \forall \varepsilon > 0 \exists \bar{n} \text{ s.t. } \forall n, m > \bar{n} \quad \|x^{(m)} - x^{(n)}\| < \varepsilon$$

$$\varepsilon^2 > \|x^{(m)} - x^{(n)}\|^2 = \sum_{i=1}^k (c_i^{(m)} - c_i^{(n)})^2 \geq (c_j^{(m)} - c_j^{(n)})^2$$

$\sum_{i=1}^k c_i^{(m)} u_i \quad \parallel \quad \sum_{i=1}^k c_i^{(n)} u_i$ $\forall j$

$\Rightarrow \{c_j^{(n)}\}$ is a Cauchy sequence in \mathbb{R}

$\Rightarrow c_j^{(n)} \rightarrow c_j \quad \forall j = 1, \dots, k$

Now we set $y = \sum_{j=1}^k c_j u_j$

$$\|x^{(n)} - y\|^2 = \sum_{j=1}^k (c_j^{(n)} - c_j)^2 \xrightarrow{n \rightarrow \infty} 0$$

$$\begin{array}{l} x^{(n)} \rightarrow y \\ x^{(n)} \rightarrow x \end{array} \Bigg| \Rightarrow y = x$$

\square

Therefore $M = \text{span}\{u_1, \dots, u_k\}$ is a closed subspace of a Hilbert space.

\Rightarrow I can take orthogonal projections

$\forall x \in H \quad \exists! P_x \in M, Q_x \in M^\perp$ st.

$x = \underbrace{P_x}_{x_0} + Q_x$. Moreover

$$\|x - x_0\| = \min_{y \in M} \|x - y\| = \text{dist}(x, M)$$

$$x - x_0 \in M^\perp$$

$x - P_x$

Q_x

$$\Rightarrow x - x_0 \perp u_i \quad \forall i = 1, \dots, k,$$

$$(x - x_0, u_i) = 0$$

$$(x, u_i) = (x_0, u_i) = c_i$$

$$\Rightarrow x_0 = \sum_{i=1}^k c_i u_i = \sum_{i=1}^k (x, u_i) u_i$$

$$\begin{aligned} & \text{Red annotations:} \\ & \text{Under } c_i \text{ in the previous equation: } (x_0, u_i) \\ & \text{Under } (x, u_i) \text{ in the previous equation: } (x, u_i) \end{aligned}$$

Summary If $\{u_1, \dots, u_k\}$ is an orthonormal set in H Hilbert space, and $x \in H$, the best approximation of x by an element of $M = \text{span}[u_1, \dots, u_k]$ is the following

$$x_0 = \sum_{i=1}^k (x, u_i) u_i$$

It means that the norm $\|x - x_0\|$ is minimal.

In formulas.

$$1) \left\| x - \sum_{i=1}^k (x, u_i) u_i \right\| \leq \left\| x - \sum_{i=1}^k \lambda_i u_i \right\|$$

$$\forall \lambda_1, \dots, \lambda_k \in \mathbb{R}$$

$$2) \sum_{i=1}^k (x, u_i)^2 \leq \|x\|^2 \quad (\text{Bessel inequality}).$$

Proof

$$\begin{aligned} \|x\|^2 &= \|P(x)\|^2 + \overbrace{\|Q(x)\|^2}^{\geq 0} \geq \|P(x)\|^2 = \\ &= \left\| \sum_{i=1}^k (x, u_i) u_i \right\|^2 \\ &= \sum_{i=1}^k (x, u_i)^2 \end{aligned}$$

Application $H = L^2(-\pi, \pi)$ Hilbert space.

$$u_0(t) = \frac{\sqrt{2}}{2}$$

$$u_k(t) = \cos(kt) \quad k = 1, 2, \dots$$

$$v_k(t) = \sin(kt) \quad k = 1, 2, \dots$$

Let us check that these functions are orthonormal

On $L^2(-\pi, \pi)$ we use the scalar product

$$(f, g) = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t) g(t) dt$$

$$\|u_0\|^2 = (u_0, u_0) = \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} dt = 1$$

$$\|u_k\|^2 = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2(kt) dt = \quad k=1,2.$$

$$\cos^2 s = \frac{1 + \cos(2s)}{2}$$

$$= \frac{1}{\pi} \int_{-\pi}^{\pi} \left(\frac{1}{2} + \frac{1}{2} \cos(2kt) \right) dt = 1.$$

the integral is zero

$$\sin^2 s = \frac{1 - \cos(2s)}{2}$$

$$\|v_k\|^2 = \frac{1}{\pi} \int_{-\pi}^{\pi} \sin^2(kt) dt = 1$$

$$(u_0, u_k) = \quad k=1,2,\dots$$

$$= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{\sqrt{2}}{2} \cos(kt) dt = 0$$

$$(u_0, v_k) = \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{\sqrt{2}}{2} \sin(kt) dt = 0$$

$$(u_k, u_h) = \quad k \neq h$$

$$= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(kt) \cos(ht) dt =$$

$$\int_{-\pi}^{\pi} \cos(kt) \cos(ht) dt = \frac{1}{h} \cos(kt) \sin(ht) \Big|_{-\pi}^{\pi} -$$

$$- \frac{k}{h} \int_{-\pi}^{\pi} \sin(kt) \sin(ht) dt =$$

$$= \frac{k}{h^2} \sin(kt) \cos(ht) \Big|_{-\pi}^{\pi} + \frac{k^2}{h^2} \int_{-\pi}^{\pi} \cos(kt) \cos(ht) dt$$

$$\Rightarrow \underbrace{\left(1 - \frac{k^2}{h^2}\right)}_{\neq 0} \int_{-\pi}^{\pi} \cos(kt) \cos(ht) dt = 0$$

and so on

Fix n , and consider a trigonometric polynomial of the form.

$$a_0 u_0 + \sum_{k=1}^n (a_k u_k(t) + b_k v_k(t))$$

For a fixed function f in $L^2(-\pi, \pi)$

I want to choose a_0, a_k, b_k such that the

distance from f to the trig. polynomial is minimal

By the theory just exposed, we must take

$$a_k = (f, u_k) \quad , \quad b_k = (f, v_k)$$