

Hilbert spaces

X vector space over $\mathbb{K} = \mathbb{R}$ or \mathbb{C}

$\langle \cdot, \cdot \rangle : X \times X \rightarrow \mathbb{K}$ scalar product if for all $x, y, z \in X, \lambda \in \mathbb{K}$:

$$\left. \begin{aligned} \bullet \langle x+y, z \rangle &= \langle x, z \rangle + \langle y, z \rangle \\ \bullet \langle \lambda x, y \rangle &= \lambda \langle x, y \rangle \\ \bullet \langle x, y \rangle &= \overline{\langle y, x \rangle} \end{aligned} \right\} \begin{array}{l} \text{linearity in} \\ \text{first argument} \end{array} \left. \vphantom{\begin{aligned} \bullet \langle x+y, z \rangle &= \langle x, z \rangle + \langle y, z \rangle \\ \bullet \langle \lambda x, y \rangle &= \lambda \langle x, y \rangle \\ \bullet \langle x, y \rangle &= \overline{\langle y, x \rangle} \end{aligned}} \right\} \begin{array}{l} \text{"sesquilinear"} \text{ (bilinear} \\ \text{if } \mathbb{K} = \mathbb{R}) \end{array}$$

$\bullet \langle x, x \rangle \geq 0, \langle x, x \rangle = 0 \iff x = 0$ ← "positive definite"

allows to define angles between vectors + norm = "length"

Cauchy-Schwarz inequality: $|\langle x, y \rangle|^2 \leq \langle x, x \rangle \langle y, y \rangle$

"=" $\iff x = \lambda y$ for some $\lambda \in \mathbb{K}$.

Corollary: $\|x\| := \sqrt{\langle x, x \rangle}$ defines a norm on X

Definition: $(X, \langle \cdot, \cdot \rangle)$ is a Hilbert space if X endowed with the topology from

$\|\cdot\|$ is complete, i.e. every Cauchy sequence converges:

In other words, given a sequence $\{x_n\}_{n=1}^{\infty}$ in X s.t. for all $\epsilon > 0$ there exists N such that for all $n, m \geq N: \|x_n - x_m\| < \epsilon$, then $\|x_n - x\| \xrightarrow{n \rightarrow \infty} 0$ for some $x \in X$.

Examples: $\bullet \mathbb{R}^n, \mathbb{C}^n$ with the usual scalar product

\bullet For $\Omega \subset \mathbb{R}^n, w: \Omega \rightarrow [0, \infty)$: $\int_{\Omega} w dx < \infty$

$$L^2(\Omega, m dx) = \{ f: \Omega \rightarrow \mathbb{K} \text{ "measurable": } \int_{\Omega} |f|^2 m dx < \infty \}$$

$$\langle f, g \rangle := \int_{\Omega} f \bar{g} m dx$$

\bullet For any set A : $l^2(A) = \{ f: A \rightarrow \mathbb{K} : \sum_{a \in A} |f(a)|^2 < \infty \}$

$$\sum_{a \in A} |f(a)|^2 := \sup_{\substack{F \subset A \\ \text{finite}}} \sum_{a \in F} |f(a)|^2 < \infty$$

$$\langle f, g \rangle := \sum_{a \in A} f(a) \bar{g}(a)$$

in particular: $A = \{1, \dots, n\} \Rightarrow l^2(A) = \mathbb{K}^n$
 $A = \mathbb{N} \Rightarrow l^2(A) = \text{square-summable sequences}$

Definition: $x, y \in X$ orthogonal $\iff \langle x, y \rangle = 0$.

$B = \{v_\alpha : \alpha \in A\} \subseteq X$ orthonormal $\iff \langle v_\alpha, v_\beta \rangle = \begin{cases} 1, & \alpha = \beta \\ 0, & \alpha \neq \beta \end{cases}$

e.g. $B_1 = \{e^{inx} : n \in \mathbb{N}\} \subseteq L^2((-\pi, \pi), \frac{dx}{2\pi})$

$B_2 = \{1, \sqrt{2} \cos(nx), \sqrt{2} \sin(nx) : n \in \mathbb{N}\} \subseteq L^2((-\pi, \pi), \frac{dx}{2\pi})$

Given such a B , define the Fourier coefficients of $x \in X$ as

$$\hat{x}(\alpha) := \langle x, v_\alpha \rangle$$

For the above B_1, B_2 , these are precisely the real/complex Fourier coefficients that we have been studying.

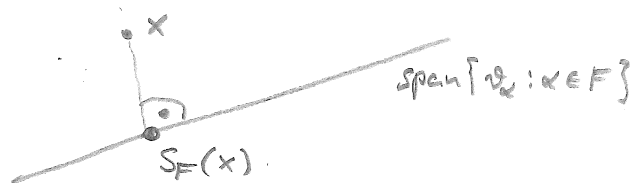
Further example: $X = \mathbb{K}^n$, $B = \{v_1, \dots, v_n\}$ orthonormal basis

$\implies \hat{x}(k) = \langle x, v_k \rangle =$ coefficient of x in the expansion with respect to the basis B , i.e.

$$x = \sum_{k=1}^n \hat{x}(k) v_k$$

FCA finite \implies partial Fourier sum $S_F(x) = \sum_{\alpha \in F} \hat{x}(\alpha) v_\alpha$

"Mean square approximation": $\|x - S_F(x)\| = \min_{u \in \text{span}\{v_\alpha : \alpha \in F\}} \|x - u\|$



Bessel's inequality: $\sum_{\alpha \in A} |\hat{x}(\alpha)|^2 \leq \|x\|^2$

Corollary: The map $X \rightarrow \ell^2(A)$ is continuous and surjective.
 $x \mapsto \hat{x}$

To see surjectivity: Given $f \in \ell^2(A)$, the vector $x = \sum_{\alpha \in A} f(\alpha) v_\alpha$ satisfies $\hat{x} = f$.

Main Theorem 1: The following assertions about an orthonormal set $B = \{v_\alpha : \alpha \in A\}$ are equivalent:

- B is maximal, i.e. whenever $B' \supset B$ is orthonormal $\Rightarrow B' = B$.
- $\overline{\text{span}\{B\}}^{\|\cdot\|} = X$
- $\langle x, v_\alpha \rangle = 0$ for all $\alpha \in A \Rightarrow x = 0$.
- Parseval 1: $\sum_{\alpha \in A} |\hat{x}(\alpha)|^2 = \|x\|^2$ for all $x \in X$.
- Parseval 2: $\sum_{\alpha \in A} \hat{x}(\alpha) \overline{\hat{y}(\alpha)} = \langle x, y \rangle$ for all $x, y \in X$
- $x \mapsto \hat{x}$ is an isometric isomorphism from X to $\ell^2(A)$.

Such a B is said to be an orthonormal basis of X .

Zorn's Lemma implies that any orthonormal set can be extended to an orthonormal basis by adding further vectors.

Corollary: Let X be a Hilbert space. Then

- 1) X has an orthonormal basis.
- 2) There is a set A st. $X \cong \ell^2(A)$.

Main Theorem 2: (Measure theory + Weierstrass approximation theorem)

$B_1 = \{e^{inx} : n \in \mathbb{N}\}$ and $B_2 = \{1, \sqrt{2}\cos(nx), \sqrt{2}\sin(nx) : n \in \mathbb{N}\}$ are orthonormal bases of $L^2((-\pi, \pi), \frac{dx}{2\pi})$.

Apart from the pointwise and uniform convergence theorems, this includes almost everything we have learned about classical Fourier series.