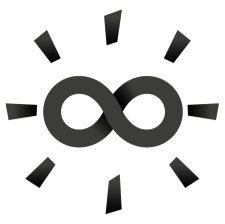


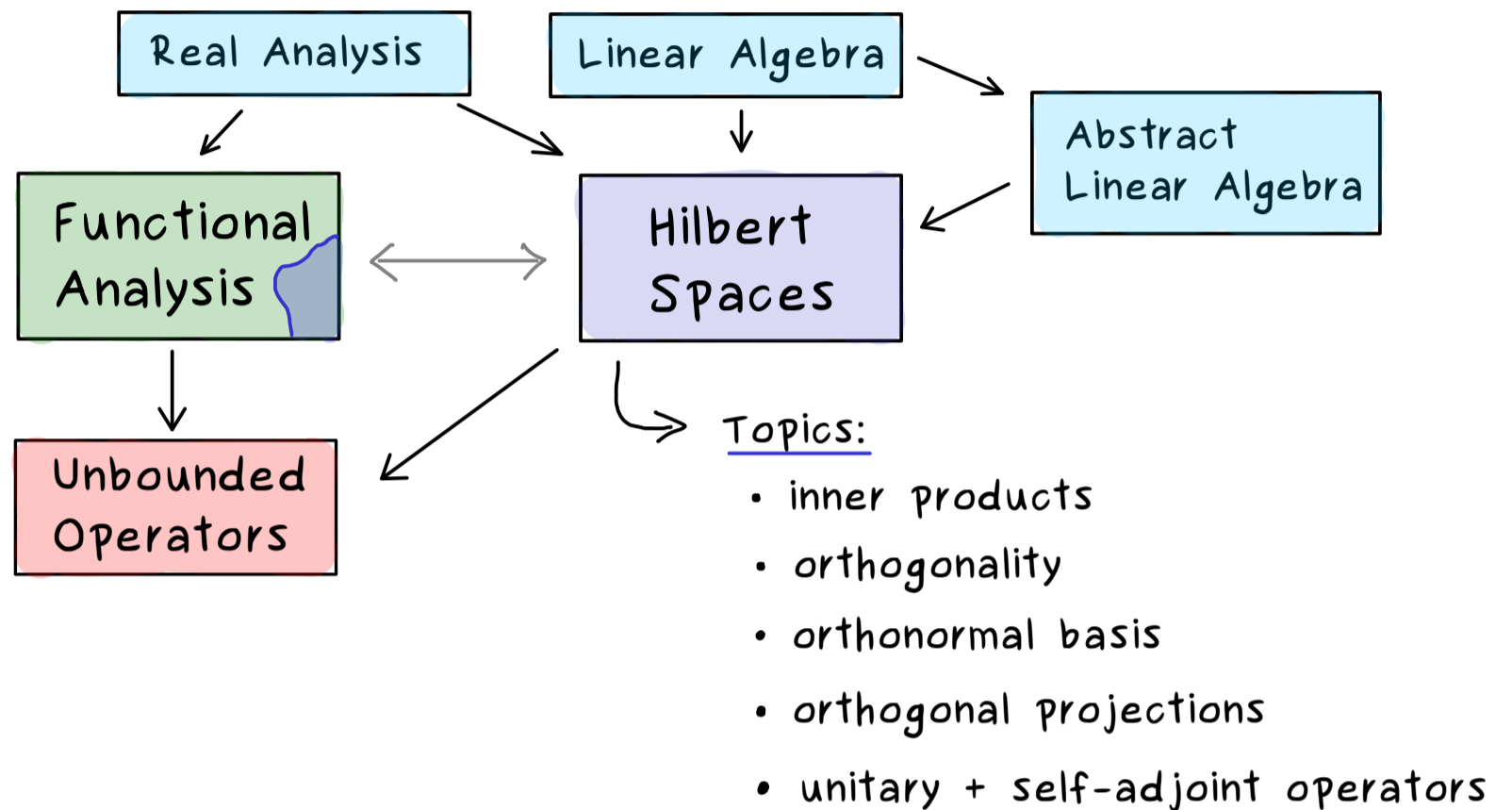
## **The Bright Side of Mathematics**

The following pages cover the whole Hilbert Spaces course of the Bright Side of Mathematics. Please note that the creator lives from generous supporters and would be very happy about a donation. See more here: <https://tbsom.de/support>

Have fun learning mathematics!



## Hilbert Spaces - Part 1



Definition:  $\mathbb{F} \in \{\mathbb{R}, \mathbb{C}\}$ . An  $\mathbb{F}$ -vector space  $X$  with inner product  $\langle \cdot, \cdot \rangle : X \times X \rightarrow \mathbb{F}$ , which means

(1)  $\langle x, x \rangle \geq 0$  for all  $x \in X$  (positive definite)

and  $\langle x, x \rangle = 0 \implies x = 0$  (zero vector)

(2)  $\langle y, x + \tilde{x} \rangle = \langle y, x \rangle + \langle y, \tilde{x} \rangle$  for all  $x, \tilde{x}, y \in X$

$\langle y, \lambda \cdot x \rangle = \lambda \cdot \langle y, x \rangle$  for all  $\lambda \in \mathbb{F}, x, \tilde{x}, y \in X$

(linear in the second argument)

(3)  $\langle x, y \rangle = \overline{\langle y, x \rangle}$  for all  $x, y \in X$  (conjugate symmetric)

is called an inner-product space. (pre-Hilbert space)

Cauchy-Schwarz inequality: For an inner product space  $(X, \langle \cdot, \cdot \rangle)$ , we have:

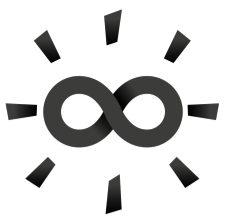
$$|\langle y, x \rangle|^2 \leq \langle x, x \rangle \langle y, y \rangle \quad \text{for all } x, y \in X$$

Proof: For  $y \neq 0$ :

$$\begin{aligned} 0 &\leq \left\langle x - \frac{\langle y, x \rangle}{\langle y, y \rangle} y, x - \frac{\langle y, x \rangle}{\langle y, y \rangle} y \right\rangle \\ &= \langle x, x \rangle - \frac{\overline{\langle y, x \rangle}}{\langle y, y \rangle} \langle y, x \rangle - \frac{\langle y, x \rangle}{\langle y, y \rangle} \langle x, y \rangle \\ &\quad + \frac{\overline{\langle y, x \rangle}}{\langle y, y \rangle} \frac{\langle y, x \rangle}{\langle y, y \rangle} \langle y, y \rangle \\ &= \langle x, x \rangle - \frac{|\langle y, x \rangle|^2}{\langle y, y \rangle} \end{aligned} \quad \square$$

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

Definition: An inner product space  $(X, \langle \cdot, \cdot \rangle)$  is called a Hilbert space if  $(X, \|\cdot\|)$  is complete.



## Hilbert Spaces - Part 2

Definition (Hilbert space):  $(X, \langle \cdot, \cdot \rangle)$   $\mathbb{F}$ -vector space  
 $\langle \cdot, \cdot \rangle : X \times X \rightarrow \mathbb{F}$  inner product

where  $(X, \|\cdot\|)$  is a Banach space

with respect to the norm  $\|x\| := \sqrt{\langle x, x \rangle}$

Example: (a)  $\mathbb{C}^N$  with standard inner product  
 (b)  $\mathbb{R}^n$  with given inner product

} ← (finite-dimensional normed vector spaces are always complete)

(c)  $\ell^2(\mathbb{N}, \mathbb{C}) := \left\{ \underset{\substack{\parallel \\ x}}{(x_n)_{n \in \mathbb{N}}} \mid x_n \in \mathbb{C} \text{ and } \sum_{n=1}^{\infty} |x_n|^2 < \infty \right\}$

with inner product:  $\langle y, x \rangle = \sum_{n=1}^{\infty} \overline{y_n} \cdot x_n$  (convergent series!)

(d)  $(\Omega, \mathcal{A}, \mu)$  measure space

$\mathcal{L}^2(\Omega, \mu) := \left\{ f : \Omega \rightarrow \mathbb{C} \text{ measurable} \mid \int_{\Omega} |f|^2 d\mu < \infty \right\}$

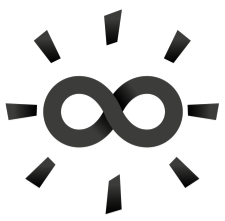
$\|f\| := \sqrt{\int_{\Omega} |f|^2 d\mu}$  not a norm in general!

$L^2(\Omega, \mu) := \mathcal{L}^2(\Omega, \mu) / \mathcal{N}$  where  $\mathcal{N} := \left\{ f : \Omega \rightarrow \mathbb{C} \text{ measurable} \mid \|f\| = 0 \right\}$

$\|[f]\| := \|f\|$  well-defined  $\leadsto$  norm on  $L^2(\Omega, \mu)$

We get a Hilbert space with the following inner product:

$$\langle [g], [f] \rangle := \int_{\Omega} \overline{g(\omega)} f(\omega) d\mu(\omega)$$



## Hilbert Spaces - Part 3

$(X, \langle \cdot, \cdot \rangle)$  inner product space ( $\mathbb{F}$ -vector space + inner product)

$$\Rightarrow (X, \|\cdot\|) \text{ normed space with } \|x\| := \sqrt{\langle x, x \rangle}$$

*norm induced by inner product*

Polarization identity: (for case  $\mathbb{F} = \mathbb{C}$ )

$(X, \langle \cdot, \cdot \rangle)$  inner product space with induced norm  $\|\cdot\|$ . Then, for all  $x, y \in X$ :

$$\langle x, y \rangle = \frac{1}{4} \left( \|x+y\|^2 - \|x-y\|^2 - i \|x+iy\|^2 + i \|x-iy\|^2 \right) \quad \text{inner product is linear in the second argument}$$

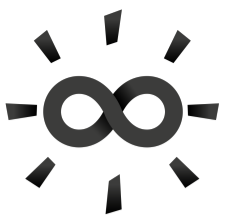
Proof:

$$\begin{aligned} \|x+y\|^2 &= \langle x+y, x+y \rangle = \langle x, x \rangle + \langle y, x \rangle + \langle x, y \rangle + \langle y, y \rangle \\ -\|x-y\|^2 &= -\langle x-y, x-y \rangle = -\langle x, x \rangle + \langle y, x \rangle + \langle x, y \rangle - \langle y, y \rangle \\ -i \|x+iy\|^2 &= -i \langle x+iy, x+iy \rangle = -i \langle x, x \rangle - \langle y, x \rangle + \langle x, y \rangle - i \langle y, y \rangle \\ i \|x-iy\|^2 &= i \langle x-iy, x-iy \rangle = i \langle x, x \rangle - \langle y, x \rangle + \langle x, y \rangle + i \langle y, y \rangle \end{aligned}$$

□

Polarization identity: (for case  $\mathbb{F} = \mathbb{R}$ )

$$\langle x, y \rangle = \frac{1}{4} \left( \|x+y\|^2 - \|x-y\|^2 \right) \quad \text{for all } x, y \in X.$$



## Hilbert Spaces - Part 4

$(X, \langle \cdot, \cdot \rangle)$  inner product space ( $\mathbb{F}$ -vector space + inner product)

$$\|x\|_{\langle \cdot, \cdot \rangle} := \sqrt{\langle x, x \rangle} \quad \text{induced norm}$$

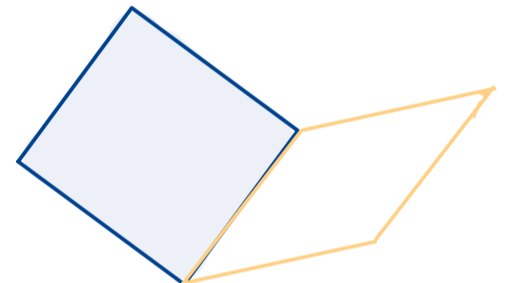
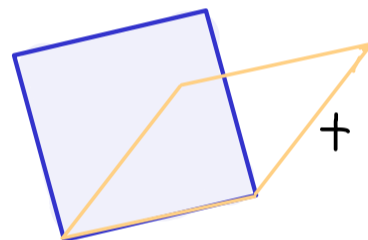
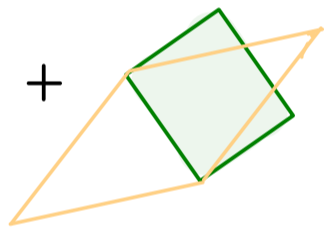
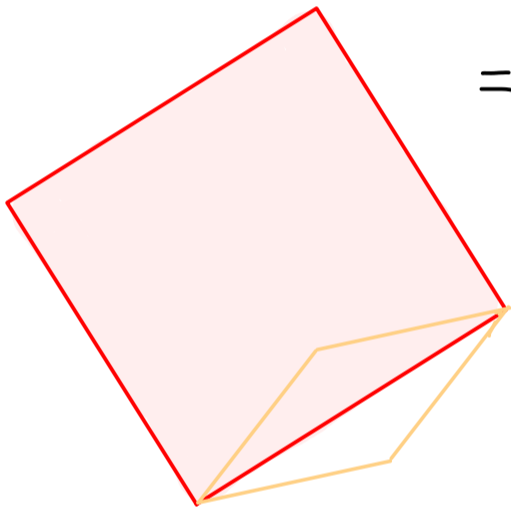
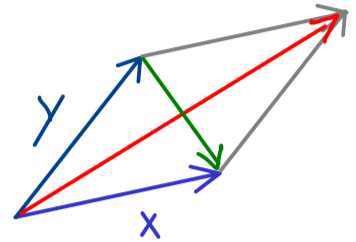
We get:  $\|x+y\|_{\langle \cdot, \cdot \rangle}^2 + \|x-y\|_{\langle \cdot, \cdot \rangle}^2$

$$= \langle x+y, x+y \rangle + \langle x-y, x-y \rangle$$

$$= \langle x, x \rangle + \langle y, x \rangle + \langle x, y \rangle + \langle y, y \rangle$$

$$+ \langle x, x \rangle - \langle y, x \rangle - \langle x, y \rangle + \langle y, y \rangle$$

$$= 2 \cdot \|x\|_{\langle \cdot, \cdot \rangle}^2 + 2 \cdot \|y\|_{\langle \cdot, \cdot \rangle}^2 \quad \text{(Parallelogram law)}$$



$$\|x+y\|_{\langle \cdot, \cdot \rangle}^2 + \|x-y\|_{\langle \cdot, \cdot \rangle}^2 = 2 \cdot \|x\|_{\langle \cdot, \cdot \rangle}^2 + 2 \cdot \|y\|_{\langle \cdot, \cdot \rangle}^2$$

Proposition: Let  $(X, \|\cdot\|)$  be a normed space. Then:

the parallelogram law is satisfied  $(\forall x, y \in X: \|x+y\|^2 + \|x-y\|^2 = 2 \cdot \|x\|^2 + 2 \cdot \|y\|^2)$

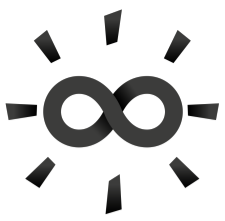
$\Leftrightarrow$   $\|\cdot\|$  is induced by an inner product on  $X$  ( $\|\cdot\|_{\langle \cdot, \cdot \rangle} = \|\cdot\|$ )  
next video

In this case:  $\langle x, y \rangle := \frac{1}{4} (\|x+y\|^2 - \|x-y\|^2)$  for  $\mathbb{F} = \mathbb{R}$

$$\langle x, y \rangle := \frac{1}{4} (\|x+y\|^2 - \|x-y\|^2 - i \|x+iy\|^2 + i \|x-iy\|^2)$$

gives the inner product on  $X$ . for  $\mathbb{F} = \mathbb{C}$

Remember: A Hilbert space is a Banach space where the parallelogram law holds.



## Hilbert Spaces - Part 5

Jordan-von Neumann Theorem: Let  $(X, \|\cdot\|)$  be a normed space. Then:

the parallelogram law is satisfied  $(\forall x, y \in X: \|x+y\|^2 + \|x-y\|^2 = 2\|x\|^2 + 2\|y\|^2)$

$\Rightarrow \|\cdot\|$  is induced by an inner product on  $X$

(there is an inner product  $\langle \cdot, \cdot \rangle$  on  $X$  such that  $\|x\| := \sqrt{\langle x, x \rangle}$ )

In this case:  $\langle x, y \rangle := \frac{1}{4} (\|x+y\|^2 - \|x-y\|^2)$  for  $\mathbb{F} = \mathbb{R}$

$\langle x, y \rangle := \frac{1}{4} (\|x+y\|^2 - \|x-y\|^2 - i\|x+iy\|^2 + i\|x-iy\|^2)$

gives the inner product on  $X$ . for  $\mathbb{F} = \mathbb{C}$

Proof: Consider case  $\mathbb{F} = \mathbb{R}$ . So we define:  $\langle x, y \rangle := \frac{1}{4} (\|x+y\|^2 - \|x-y\|^2)$ .

To show three properties: (1) positive definite

(2) linear in the second argument

(3) symmetry

$$(1): \langle x, x \rangle = \frac{1}{4} (\|x+x\|^2 - \|x-x\|^2) = \frac{1}{4} \|2x\|^2 = \|x\|^2 \geq 0$$

$$\text{and } \langle x, x \rangle = 0 \Rightarrow x = 0$$

$$(3): \langle y, x \rangle = \frac{1}{4} (\|y+x\|^2 - \|y-x\|^2) = \frac{1}{4} (\|x+y\|^2 - \|x-y\|^2) = \langle x, y \rangle$$

(2) linearity:

$$\text{we will use: } \|x+y\|^2 + \|x-y\|^2 = 2\|x\|^2 + 2\|y\|^2$$

First step:  $\langle w, z \rangle = \frac{1}{4} \left( \|w+z\|^2 - \|w-z\|^2 \right)$

$$= \frac{1}{4} \left( \underbrace{\|w+z\|^2}_{x+y} + \underbrace{\|w\|^2}_{x-y} - \left( \underbrace{\|w\|^2}_{\tilde{x}+\tilde{y}} + \underbrace{\|w-z\|^2}_{\tilde{x}-\tilde{y}} \right) \right)$$

$\hookrightarrow x := w + \frac{1}{2}z$        $\tilde{x} := w - \frac{1}{2}z$   
 $y := \frac{1}{2}z$                        $\tilde{y} := \frac{1}{2}z$

Parallelogram law  $\swarrow$

$$= \frac{1}{4} \left( 2 \cdot \|x\|^2 + 2 \cdot \|y\|^2 - \left( 2 \cdot \|\tilde{x}\|^2 + 2 \cdot \|\tilde{y}\|^2 \right) \right)$$

$$= \frac{1}{2} \left( \|x\|^2 - \|\tilde{x}\|^2 \right) = \frac{1}{2} \left( \|w + \frac{1}{2}z\|^2 - \|w - \frac{1}{2}z\|^2 \right)$$

$$= 2 \cdot \left\langle w, \frac{1}{2}z \right\rangle$$

First result:  $\frac{1}{2} \langle w, z \rangle = \langle w, \frac{1}{2}z \rangle$  induction  $n \in \mathbb{N}$   $\rightarrow$  
 $\frac{1}{2^n} \langle w, z \rangle = \langle w, \frac{1}{2^n}z \rangle$  (\*)

Additivity:  $\langle w, z \rangle + \langle w, \hat{z} \rangle$

$$= \frac{1}{4} \left( \|w+z\|^2 - \|w-z\|^2 \right) + \frac{1}{4} \left( \|w+\hat{z}\|^2 - \|w-\hat{z}\|^2 \right)$$

$$= \frac{1}{4} \left( \left\| w + \frac{z+\hat{z}}{2} + \frac{z-\hat{z}}{2} \right\|^2 + \left\| w + \frac{z+\hat{z}}{2} - \frac{z-\hat{z}}{2} \right\|^2 - \left( \left\| w - \frac{z+\hat{z}}{2} + \frac{z-\hat{z}}{2} \right\|^2 + \left\| w - \frac{z+\hat{z}}{2} - \frac{z-\hat{z}}{2} \right\|^2 \right) \right)$$

Parallelogram law  $\swarrow$

$$= \frac{1}{4} \left( 2 \cdot \left\| w + \frac{z+\hat{z}}{2} \right\|^2 + 2 \cdot \left\| \frac{z-\hat{z}}{2} \right\|^2 - \left( 2 \cdot \left\| w - \frac{z+\hat{z}}{2} \right\|^2 + 2 \cdot \left\| \frac{z-\hat{z}}{2} \right\|^2 \right) \right)$$

$$= \frac{1}{2} \left( \left\| w + \frac{z+\hat{z}}{2} \right\|^2 - \left\| w - \frac{z+\hat{z}}{2} \right\|^2 \right) = 2 \left\langle w, \frac{z+\hat{z}}{2} \right\rangle$$

(\*)  $= \langle w, z + \hat{z} \rangle$



Homogeneity:

$$\langle w, z \rangle + \langle w, z \rangle \stackrel{\text{additivity}}{=} \langle w, z + z \rangle \\ \stackrel{\text{"}}{=} 2 \cdot \langle w, z \rangle \qquad \qquad \qquad \stackrel{\text{"}}{=} \langle w, 2 \cdot z \rangle$$

induction

$\xrightarrow{k \in \mathbb{N}}$

$$k \cdot \langle w, z \rangle = \langle w, k z \rangle$$

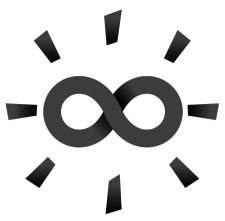
combining with (\*) :  $\frac{k}{2^n} \langle w, z \rangle = \langle w, \frac{k}{2^n} z \rangle$  for all  $k, n \in \mathbb{N}$

$$0 \cdot \langle w, z \rangle = \langle w, 0 \cdot z \rangle$$

$$(-1) \cdot \langle w, z \rangle = \langle w, (-1) \cdot z \rangle$$

all positive  
real numbers  
can be approximated

□



## Hilbert Spaces - Part 6

$$(X, \langle \cdot, \cdot \rangle)$$

gives geometry to vector space  $X$

we can measure lengths:  $\|x\| := \sqrt{\langle x, x \rangle}$

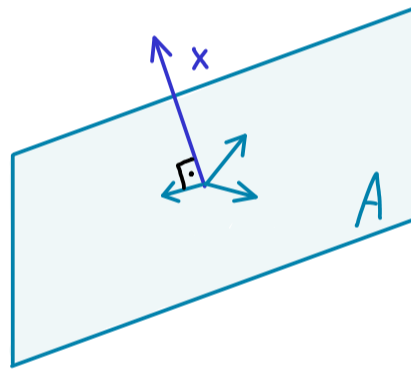
we can measure angles / orthogonality

Definition:  $(X, \langle \cdot, \cdot \rangle)$  inner product space.

(1)  $x \in X$  is orthogonal to  $y \in X$  if  $\langle x, y \rangle = 0$ . Write  $x \perp y$ .

(2)  $x \in X$  is called orthogonal to  $A \subseteq X$  if  $\langle x, a \rangle = 0$  for all  $a \in A$ .

We write  $x \perp A$ .

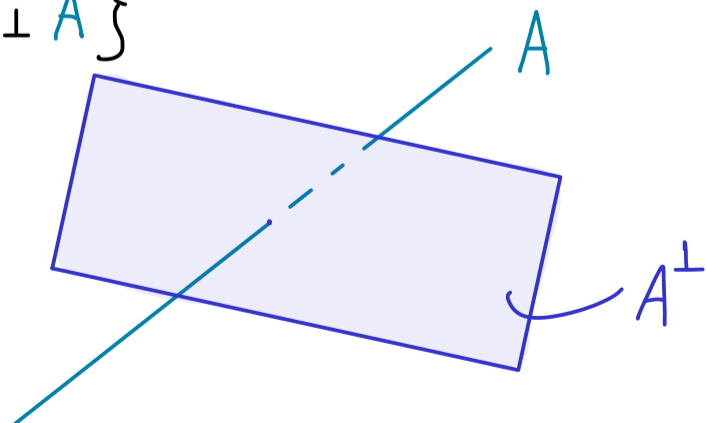


(3)  $B \subseteq X$  is called orthogonal to  $A \subseteq X$  if  $\langle b, a \rangle = 0$  for all  $a \in A$  for all  $b \in B$

We write  $B \perp A$ .

(4) The orthogonal complement of  $A \subseteq X$  is defined by:

$$A^\perp := \{x \in X \mid x \perp A\}$$



Properties:  $(X, \langle \cdot, \cdot \rangle)$  inner product space,  $A \subseteq X$ .

(a)  $A^\perp$  is a subspace in  $X$ .

(b)  $A^\perp$  is closed in  $X$  (complement  $X \setminus A^\perp$  is an open set)

(c)  $A^\perp = \overline{A}^\perp$

(d)  $A^\perp = \text{Span}(A)^\perp$

Proof: (a)  $x, y \in A^\perp, a \in A, \lambda \in \mathbb{F}$

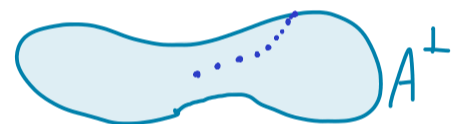
$$\Rightarrow \langle x+y, a \rangle = \langle x, a \rangle + \langle y, a \rangle = 0$$

$$\langle 0, a \rangle = 0$$

$$\langle \lambda \cdot x, a \rangle = \bar{\lambda} \langle x, a \rangle = 0$$

$\Rightarrow A^\perp$  subspace in  $X$ .

(b) Take  $(x_n)_{n \in \mathbb{N}} \subseteq A^\perp$  with  $x_n \xrightarrow{n \rightarrow \infty} x \in X$ .



For any  $a \in A$ :

inner product continuous  
in both arguments

$$0 = \lim_{n \rightarrow \infty} \langle x_n, a \rangle \stackrel{\text{inner product continuous in both arguments}}{=} \langle \lim_{n \rightarrow \infty} x_n, a \rangle = \langle x, a \rangle \Rightarrow x \in A^\perp$$

(c)  $A \subseteq \overline{A} \Rightarrow A^\perp \supseteq \overline{A}^\perp$

Other inclusion? ( $\subseteq$ )  $x \in A^\perp, b \in \overline{A}$ , choose  $(a_n) \subseteq A$  with  $\lim_{n \rightarrow \infty} a_n = b$

$$\langle x, b \rangle = \langle x, \lim_{n \rightarrow \infty} a_n \rangle \stackrel{\text{inner product continuous in both arguments}}{=} \lim_{n \rightarrow \infty} \langle x, a_n \rangle = 0$$

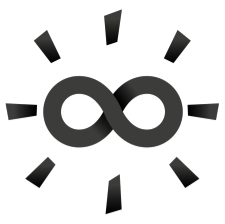
inner product continuous  
in both arguments

$$\Rightarrow x \in \overline{A}^\perp$$

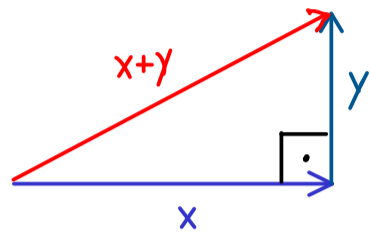
(d)  $A \subseteq \text{Span}(A) \Rightarrow A^\perp \supseteq \text{Span}(A)^\perp$

Other inclusion? ( $\subseteq$ )  $x \in A^\perp, \sum_{j=1}^n \lambda_j \cdot a_j \in \text{Span}(A)$ :

$$\langle x, \sum_{j=1}^n \lambda_j \cdot a_j \rangle = \sum_{j=1}^n \lambda_j \cdot \langle x, a_j \rangle = 0 \Rightarrow x \in \text{Span}(A)^\perp$$



## Hilbert Spaces - Part 7



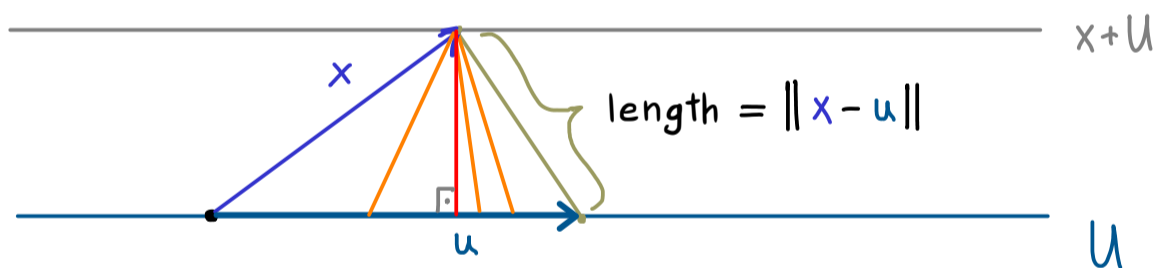
choose  $x, y$  orthogonal:  $\langle x, y \rangle = 0$

Pythagorean theorem:  $(X, \langle \cdot, \cdot \rangle)$  inner product space with induced norm  $\|\cdot\|$ .

For any  $x, y \in X$  with  $x \perp y$ , we have:

$$\begin{aligned} \|x+y\|^2 &= \langle x+y, x+y \rangle = \langle x, x \rangle + \overbrace{\langle y, x \rangle}^{=0} + \overbrace{\langle x, y \rangle}^{=0} + \langle y, y \rangle \\ &= \|x\|^2 + \|y\|^2 \end{aligned}$$

### Approximation Formula

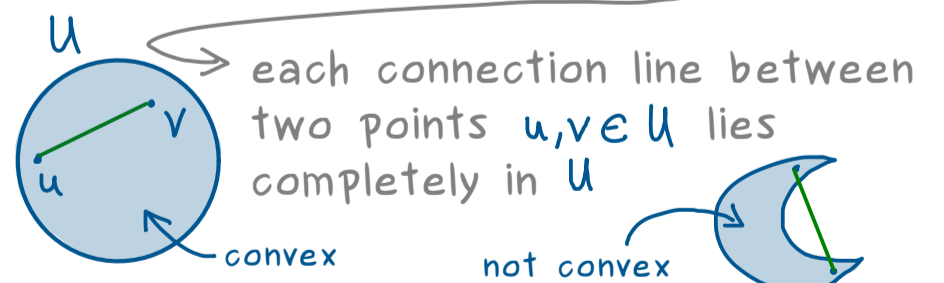


distance between  $x+U$  and  $U$ :  $\inf \{ \|x-u\| \mid u \in U \} =: \text{dist}(x, U)$

Theorem: Let  $(X, \langle \cdot, \cdot \rangle)$  be a Hilbert space,  $U \subseteq X$  be closed and convex.

For every  $x \in X$  there exists a unique best approximation:

$$x|_U \in U$$



This means:  $\|x - x|_U\| = \text{dist}(x, U)$