

§ Gram-Schmidt Process

Recall: $S = \{\vec{v}_1, \dots, \vec{v}_k\} \implies S$ lin. indep.

orthogonal, $\vec{v}_i \neq \vec{0}$
for all i

Gram-Schmidt
orthogonalization process

E.g. $S = \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$ lin. indep.
(\mathbb{R}^2)



$$S = \{\vec{w}_1, \vec{w}_2\} \xrightarrow{\text{GS}} S' = \{\vec{w}_1, \underbrace{\vec{w}_2 - c\vec{w}_1}_{\vec{v}_1 \perp \vec{v}_2}\}$$

lin. indep.

Note: (1) $\text{Span } S = \text{Span } S'$

$$(2) \vec{v}_1 \perp \vec{v}_2 \Leftrightarrow \langle \vec{w}_1, \vec{w}_2 - c\vec{w}_1 \rangle = 0$$

$$\Leftrightarrow \langle \vec{w}_1, \vec{w}_2 \rangle - \bar{c} \langle \vec{w}_1, \vec{w}_1 \rangle = 0$$

$$\Leftrightarrow c = \frac{\langle \vec{w}_2, \vec{w}_1 \rangle}{\langle \vec{w}_1, \vec{w}_1 \rangle}$$

$$\vec{v}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\vec{v}_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} - \frac{1}{1^2} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

orthogonal.

E.g. (3 vectors)

$$(\mathbb{R}^3) \quad S = \{ \vec{w}_1, \vec{w}_2, \vec{w}_3 \}$$

$$\vec{w}_1 = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}, \quad \vec{w}_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \quad \vec{w}_3 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

Idea: Make them \perp "one vector at a time."

$$\vec{v}_1 = \vec{w}_1 = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}$$

$$\vec{v}_2 = \vec{w}_2 - \frac{\langle \vec{w}_2, \vec{v}_1 \rangle}{\langle \vec{v}_1, \vec{v}_1 \rangle} \vec{v}_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{3}{5} \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} = \begin{pmatrix} 2/5 \\ -1/5 \\ 1 \end{pmatrix} \perp \vec{v}_1.$$

$$\vec{v}_3 = \vec{w}_3 - \underbrace{\frac{\langle \vec{w}_3, \vec{v}_1 \rangle}{\langle \vec{v}_1, \vec{v}_1 \rangle} \vec{v}_1}_{\perp \vec{v}_1} - \underbrace{\frac{\langle \vec{w}_3, \vec{v}_2 \rangle}{\langle \vec{v}_2, \vec{v}_2 \rangle} \vec{v}_2}_{\perp \vec{v}_2}$$

$$\frac{\|\cdot\|}{\frac{1}{5}(2^2 + 1^2 + 5^2)^{1/2}} = \frac{\sqrt{30}}{5}$$

$$= \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} - \frac{1}{5} \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} - \frac{2/5}{\sqrt{30}/5} \begin{pmatrix} 2/5 \\ -1/5 \\ 1 \end{pmatrix} = \dots \dots$$

Note: $\text{span}\{\vec{v}_1, \vec{v}_2\} = \text{span}\{\vec{w}_1, \vec{w}_2\}$

Span remains the same

$$\text{span}\{\vec{v}_1, \vec{v}_2, \vec{v}_3\} = \text{span}\{\vec{w}_1, \vec{w}_2, \vec{w}_3\}$$

after each step.

Thm: (Gram-Schmidt Process)

Given a lin. indep. $S = \{\vec{w}_1, \vec{w}_2, \dots, \vec{w}_k\}$ in $(V, \langle \cdot, \cdot \rangle)$,

$\Rightarrow \exists$ orthogonal subset $S' = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$

s.t. $\text{Span} \underbrace{\{\vec{w}_1, \dots, \vec{w}_i\}}_{\subseteq S} = \text{Span} \underbrace{\{\vec{v}_1, \dots, \vec{v}_i\}}_{\subseteq S'}$ for ALL i .

More explicitly,

$$\left\{ \begin{array}{l} \vec{v}_1 = \vec{w}_1 \\ \vec{v}_2 = \vec{w}_2 - \frac{\langle \vec{w}_2, \vec{v}_1 \rangle}{\langle \vec{v}_1, \vec{v}_1 \rangle} \vec{v}_1 \\ \vdots \quad \vdots \\ \vec{v}_k = \vec{w}_k - \sum_{j=1}^{k-1} \frac{\langle \vec{w}_k, \vec{v}_j \rangle}{\langle \vec{v}_j, \vec{v}_j \rangle} \vec{v}_j \end{array} \right.$$

To produce O.N.B.:

Any ordered basis \xrightarrow{GS} orthogonal basis $\xrightarrow{\text{normalize}}$ O.N.B.

Corollary: Any $(V, \langle \cdot, \cdot \rangle)$ has an O.N.B. $\beta = \{\vec{v}_1, \dots, \vec{v}_n\}$

so, then any $\vec{v} \in V$ can be expressed

$$\vec{v} = \underbrace{\langle \vec{v}, \vec{v}_1 \rangle}_{\text{Fourier Coefficients}} \vec{v}_1 + \underbrace{\langle \vec{v}, \vec{v}_2 \rangle}_{\text{Fourier Coefficients}} \vec{v}_2 + \dots + \underbrace{\langle \vec{v}, \vec{v}_n \rangle}_{\text{Fourier Coefficients}} \vec{v}_n$$

E.g. (Legendre polynomials)

$$V = \underbrace{P_2(\mathbb{R})}_{\substack{\text{poly. of} \\ \deg \leq 2 \\ (\mathbb{R}-\text{coeff.})}} \subseteq C([-1, 1]) \quad , \quad \langle f, g \rangle := \int_{-1}^1 f(t) g(t) dt$$

Start with std basis for $P_2(\mathbb{R})$

$$\beta = \{1, x, x^2\} \quad \text{lin. indep.}$$

$\vec{w}_1, \vec{w}_2, \vec{w}_3$ BUT NOT O.N.B. w.r.t. $\langle \cdot, \cdot \rangle$

Apply Gram-Schmidt,

$$\vec{v}_1 = \vec{w}_1 = 1$$

Compute inner products,

$$\langle \vec{w}_2, \vec{v}_1 \rangle = \langle x, 1 \rangle := \int_{-1}^1 x dx = 0$$

$$\langle \vec{v}_1, \vec{v}_1 \rangle = \langle 1, 1 \rangle := \int_{-1}^1 1 dx = 2$$

$$\Rightarrow \vec{v}_2 = \vec{w}_2 - \frac{\langle \vec{w}_2, \vec{v}_1 \rangle}{\langle \vec{v}_1, \vec{v}_1 \rangle} \vec{v}_1 = x - \frac{0}{2} \cdot 1 = x$$

Compute inner products.

$$\langle \vec{w}_3, \vec{v}_1 \rangle = \langle x^2, 1 \rangle = \int_{-1}^1 x^2 dx = \frac{2}{3}$$

$$\langle \vec{w}_3, \vec{v}_2 \rangle = \langle x^2, x \rangle = \int_{-1}^1 x^3 dx = 0$$

$$\langle \vec{v}_2, \vec{v}_2 \rangle = \langle x, x \rangle = \int_{-1}^1 x^2 dx = \frac{2}{3}$$

$$\Rightarrow \vec{v}_3 = \vec{w}_3 - \frac{\langle \vec{w}_3, \vec{v}_1 \rangle}{\langle \vec{v}_1, \vec{v}_1 \rangle} \vec{v}_1 - \frac{\langle \vec{w}_3, \vec{v}_2 \rangle}{\langle \vec{v}_2, \vec{v}_2 \rangle} \vec{v}_2$$

$$= x^2 - \frac{2/3}{2} \cdot 1 - 0 = x^2 - \frac{1}{3}$$

$\therefore \{1, x, x^2 - \frac{1}{3}\}$ orthogonal basis

↓ normalize

$$\beta' = \left\{ \underbrace{\frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}}x, \sqrt{\frac{5}{8}}(3x^2 - 1)}_{\text{Legendre polynomials}} \right\} \text{ O.N.B.}$$

Legendre polynomials

§ Orthogonal Complement

$(V, \langle \cdot, \cdot \rangle)$: inner product space / \mathbb{R} or \mathbb{C}
 ($\dim V$ could be $+\infty$)

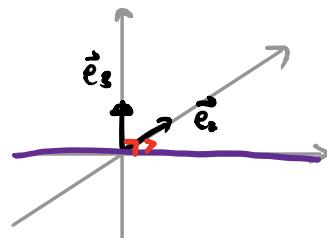
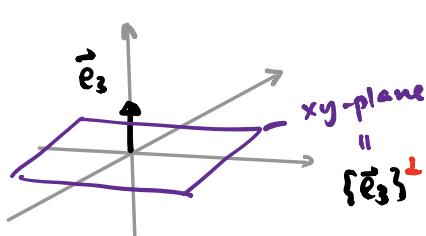
Defⁿ: Take any subset $\phi \neq S \subseteq V$, define

$$S^\perp := \left\{ \vec{x} \in V : \langle \vec{x}, \vec{v} \rangle = 0 \quad \forall \vec{v} \in S \right\}$$

orthogonal complement of S

Ex: Trivial: $\{\vec{0}\}^\perp = V$, $V^\perp = \{\vec{0}\}$

\mathbb{R}^3 : $\{\vec{e}_3\}^\perp = xy\text{-plane}$ $\{\vec{e}_2, \vec{e}_3\}^\perp = x\text{-axis}$



Remarks: (1) $S^\perp = (\text{span } S)^\perp$ $\therefore \text{WLOG, assume } S \text{ is a subspace}$

(2) S^\perp is always a subspace! (even when S is not)

Theorem: If $W \subseteq (V, \langle \cdot, \cdot \rangle)$ is a subspace, $\dim W < +\infty$.

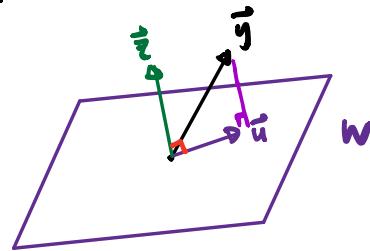
then

$$V = W \oplus W^\perp \quad \left(\begin{array}{l} \text{Remark: true for} \\ \dim V = +\infty \end{array} \right)$$

i.e. any $\vec{y} \in V$ can be uniquely decomposed as

$$\vec{y} = \underbrace{\vec{u}}_{\substack{\in \\ W}} + \underbrace{\vec{z}}_{\substack{\in \\ W^\perp}}$$

"orthogonal projection of \vec{y} onto W "



Proof: "Existence": $V = W + W^\perp$

W finite dim $\Rightarrow \beta = \{\vec{v}_1, \dots, \vec{v}_k\}$ O.N.B. for W

For any $\vec{y} \in V$, define

$$\vec{u} := \sum_{i=1}^k \langle \vec{y}, \vec{v}_i \rangle \vec{v}_i \in W$$

Note:
 $\langle \vec{u}, \vec{v}_i \rangle = \langle \vec{y}, \vec{v}_i \rangle$
since β O.N.B.

Check: $\vec{z} := \vec{y} - \vec{u} \in W^\perp \checkmark$

$$\langle \vec{z}, \vec{x} \rangle = \langle \vec{z}, \sum_{i=1}^k a_i \vec{v}_i \rangle = \sum_{i=1}^k \overline{a_i} \underbrace{\langle \vec{z}, \vec{v}_i \rangle}_{\substack{\parallel \\ \langle \vec{y} - \vec{u}, \vec{v}_i \rangle = 0}} = 0$$

any vector
in W

"Uniqueness": $W \cap W^\perp = \{\vec{0}\}$

Take any $\vec{y} \in W \cap W^\perp$,

$$\langle \vec{y}, \vec{y} \rangle = 0 \Rightarrow \vec{y} = \vec{0}.$$

$\overset{\text{def}}{W} \quad \overset{\text{def}}{W^\perp}$

Geometrically, $\vec{y} = \underset{W}{\vec{u}} + \underset{W^\perp}{\vec{v}}$

then " \vec{u} is the unique vector in W which is closest to \vec{y} "

i.e. $\forall \vec{x} \in W, \|\vec{y} - \vec{x}\| \geq \|\vec{y} - \vec{u}\| \quad \text{--- (*)}$

\Rightarrow Pythagoras Thm: $\|x+y\|^2 = \|x\|^2 + \|y\|^2 \quad \text{if } \langle x, y \rangle = 0.$

Ex: Prove (*) using .

Corollary: Assume $\dim V < +\infty$.

O.N.B.

(a) $\{\vec{v}_1, \dots, \vec{v}_k\} \subseteq W \xrightarrow{\text{extend}} \underbrace{\{\vec{v}_1, \dots, \vec{v}_k, \vec{v}_{k+1}, \dots, \vec{v}_n\}}_{W} \subseteq V$

O.N. basis $\overset{\text{O.N.B.}}{W^\perp}$

(b) $\dim V = \dim W + \dim W^\perp$