

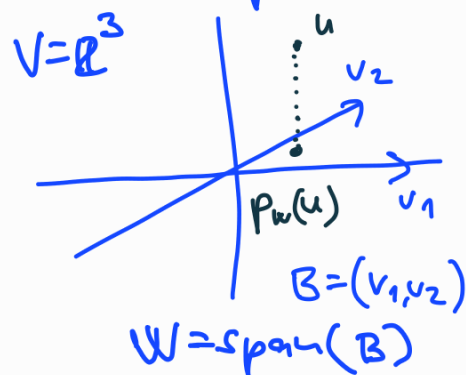
Review:

LA2 3/3/2026

Def: Let W be a subspace of VS V ,

$B = (v_1, \dots, v_n)$ an orthonormal basis of W .

The mapping $p_W(u) = \sum_{i=1}^n \langle u | v_i \rangle \cdot v_i$ from V to W is orthogonal projection of V onto the subspace W .



Lemma: $\forall j=1, \dots, n: (u - p_W(u)) \perp v_j$

Lemma: $p_W(u)$ is the unique nearest vector from W to u .

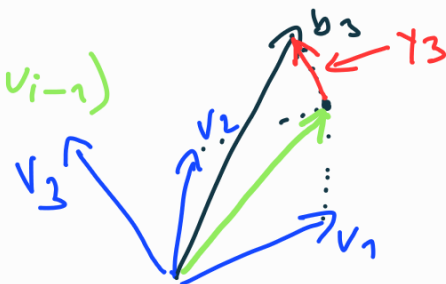
GRAM-SCHMIDT ORTHOGONALIZATION

given a basis (b_1, \dots, b_n) , it constructs an ON basis (v_1, \dots, v_n) .

for $i=1, \dots, n$

do projection of b_i onto $\text{span}(v_1, \dots, v_{i-1})$

$$(*) \quad \begin{cases} y_i := b_i - \sum_{j=1}^{i-1} \langle b_i | v_j \rangle \cdot v_j \\ v_i = \frac{y_i}{\|y_i\|} \end{cases}$$



Corollary 1: Every finite dim VS has ON basis.

Corollary 2: In a finite dim VS V : if W is a subspace of V , then every ON basis of W can be extended into an ON basis of V .

Proof: first extend the given ON basis, say b_1, \dots, b_k , of W , into a basis $b_1, \dots, b_k, z_1, \dots, z_r$ of V by Steinitz theorem.

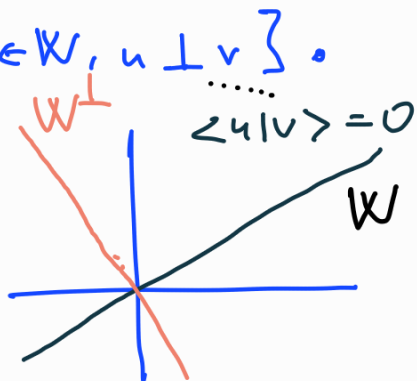
Then Gram-Schmidt. Note: $\forall i=1, \dots, k: v_i = b_i \dots$ no change!

Def: Let W be set of vectors from a VS V with inner product. Then the orthogonal complement of W is the set $W^\perp = \{v \in V : \forall u \in W, u \perp v\}$.

Eg. • $V = \mathbb{R}^2$, standard inner product

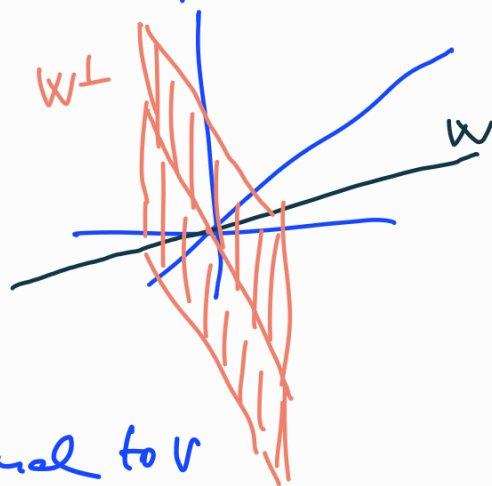
W ... line through the origin

W^\perp ... perpendicular line through origin



• $V = \mathbb{R}^3$, W ... line through the origin

W^\perp ... plane perpendicular to W , through origin



Eye $U \subseteq V \Rightarrow U^\perp \supseteq V^\perp$

proof: obvious: "everything orthogonal to V is orthogonal to U as well".

$$x \in V^\perp \Leftrightarrow \forall v \in V, x \perp v \Rightarrow \forall v \in U, x \perp v \Leftrightarrow x \in U^\perp \quad \square$$

Eye Consider a homogeneous system of lin. eq. $Ax = 0$.

Then $\text{Null}(A) = (R(A))^\perp$ standard inner prod. $A \in \mathbb{R}^{m \times n}$

$$\{x \mid Ax = 0\} \quad \swarrow \text{row space of } A$$

proof: \supseteq ... by previous Eye i.e. $z = \sum_i \lambda_i A_{i*}$

\subseteq ... consider $y \in \text{Null}(A)$ and any $z \in R(A)$.

We need to check that $\langle z | y \rangle = 0$:

$$\langle \sum_i \lambda_i A_{i*} | y \rangle \stackrel{L1,2}{=} \sum_i \lambda_i \langle A_{i*} | y \rangle = 0 \Rightarrow y \in R(A)^\perp \quad \square$$

Lemma: If $B = (b_1, \dots, b_n)$ is a basis of subspace W , then $v \in W^\perp \Leftrightarrow \forall i = 1, \dots, n, v \perp b_i$.

Proof: \Rightarrow obvious, by Eye

\Leftarrow we have to check that $\forall u \in W, \langle u | v \rangle = 0$

$$u = \sum_{i=1}^n \alpha_i b_i \text{ for some } \alpha_i. \quad \langle u | v \rangle = \langle \sum_i \alpha_i b_i | v \rangle \stackrel{L1,2}{=} \sum_i \alpha_i \langle b_i | v \rangle \stackrel{\text{Eye}}{=} 0. \quad \square$$

Theorem (Properties of orthogonal complement).

Let W be a subspace of VS V of finite dimension.

- Then
- W^\perp is a subspace of V .
 - $\dim(W) + \dim(W^\perp) = \dim(V)$
 - $(W^\perp)^\perp = W$
 - $W^\perp \cap W = \{\emptyset\}$.

Proof: i) we check that W^\perp is closed under $+$, \cdot :

$$\bullet u, v \in W^\perp : \forall x \in W, \langle u+v, x \rangle = \underbrace{\langle u, x \rangle}_{=0} + \underbrace{\langle v, x \rangle}_{=0} = 0, \\ \Rightarrow u+v \in W^\perp \quad \text{as } u, v \in W^\perp$$

$$\bullet \alpha \in W^\perp, \alpha \in \mathbb{T} : \forall x \in W, \langle \alpha u, x \rangle = \alpha \langle u, x \rangle = \alpha \cdot 0 = 0 \Rightarrow \alpha u \in W^\perp$$

ii) let $B = (b_1, \dots, b_k)$ be an ON basis of W . We extend it to an ON basis of V : $H = (b_1, \dots, b_k, c_1, \dots, c_l)$.

Now we verify that $C = (c_1, \dots, c_l)$ is a basis of W^\perp .

Consider any $v \in W^\perp$. Its coordinates w.r.t. H are given by the Fourier coefficients:

$$\left. \begin{array}{l} \langle v, b_i \rangle \dots \text{for the vectors } b_i \\ \langle v, c_j \rangle \dots \text{for the vectors } c_j \end{array} \right\} \Rightarrow v \in \text{span}(C) \Rightarrow C \text{ is basis of } W^\perp.$$

iii) By lemma, $v \in (W^\perp)^\perp \Leftrightarrow \forall i=1, \dots, l, v \perp c_i, \Leftrightarrow$

$$\forall i=1, \dots, l, \langle v, c_i \rangle = 0 \Leftrightarrow v \in \text{span}(B) = W.$$

Fourier coeff. w.r.t. the basis H of V

iv) If $\exists u \in W^\perp \cap W, u \neq \emptyset$, then $\exists \alpha_1, \dots, \alpha_k, \beta_1, \dots, \beta_l$ s.t. $\sum \alpha_i b_i = \sum \beta_j c_j \Rightarrow H$ is not a basis of V - contradiction. \square

Or: by proof of ii): $\forall v \in V \exists u \in W, z \in W^\perp$ s.t. $v = u + z$

$$\Rightarrow W + W^\perp = V$$

$$\text{As } \dim(X+Y) + \dim(X \cap Y) = \dim(X) + \dim(Y),$$

$$\dim(W \cap W^\perp) = (k+l) - k - l = 0.$$

LEAST SQUARES METHOD

Consider an inconsistent (=no solution) system of linear equations $Ax=b$, $A \in \mathbb{R}^{m \times n}$. meaning?

Goal: find $x \in \mathbb{R}^n$ that violates the constraints the least.

Recall: $Ax=b$ - find a linear combination of columns of A that equals b

Let $W = C(A)$. No solution of $Ax=b$: $b \notin W$



Idea: 1) compute the projection $p_W(b)$ of b to W
 2) and solve $Ax = p_W(b)$.

Then • there is a solution ... denote it \bar{x} standard norm
↓

• moreover: $\|A\bar{x} - b\| = \|p_W(b) - b\| = \min_{y \in W} \|y - b\| = \min_{x \in \mathbb{R}^n} \|Ax - b\|$

measure of constraints violation

i.e., \bar{x} is such that the norm of $A\bar{x} - b$ is the least, as required.

Why least squares?

Note: the optima of $\min_{x \in \mathbb{R}^n} \|Ax - b\|$ and $\min_{x \in \mathbb{R}^n} \|Ax - b\|^2 =$
 are the same (by monotonicity of z^2)
 $= \min_{x \in \mathbb{R}^n} \sum_i (A_{i \cdot} x - b_i)^2$
least squares

How to solve it? aim: avoid projection

Let $\mathcal{O} = \{x \in \mathbb{R}^n \mid Ax = p_W(b)\}$.

Theorem: $\mathcal{O} = \{x \in \mathbb{R}^n \mid A^T Ax = A^T b\}$.

Proof: as $(b - p_W(b)) \perp W$, and $W = C(A) = R(A^T)$,

we have $b - p_W(b) \in \text{Null}(A^T)$.

Consider $x \in \mathcal{O}$: $Ax = p_W(b) \Rightarrow A^T(Ax - p_W(b)) = 0 \Leftrightarrow Ax - p_W(b) \in \text{Null}(A^T) \Leftrightarrow$

$\Leftrightarrow \underbrace{Ax - p_W(b)}_{\in C(A)} + \underbrace{(p_W(b) - b)}_{\in C(A)} \in \text{Null}(A^T) \Leftrightarrow A^T(Ax - b) = 0 \Leftrightarrow A^T Ax = A^T b$.

The other inclusion: note that $Ax - p_W(b) \in C(A)$, and $Ax - p_W(b) \perp C(A)$
 $\Rightarrow Ax - p_W(b) = 0$, i.e., $x \in \mathcal{O}$.

\Rightarrow A way to solve it.