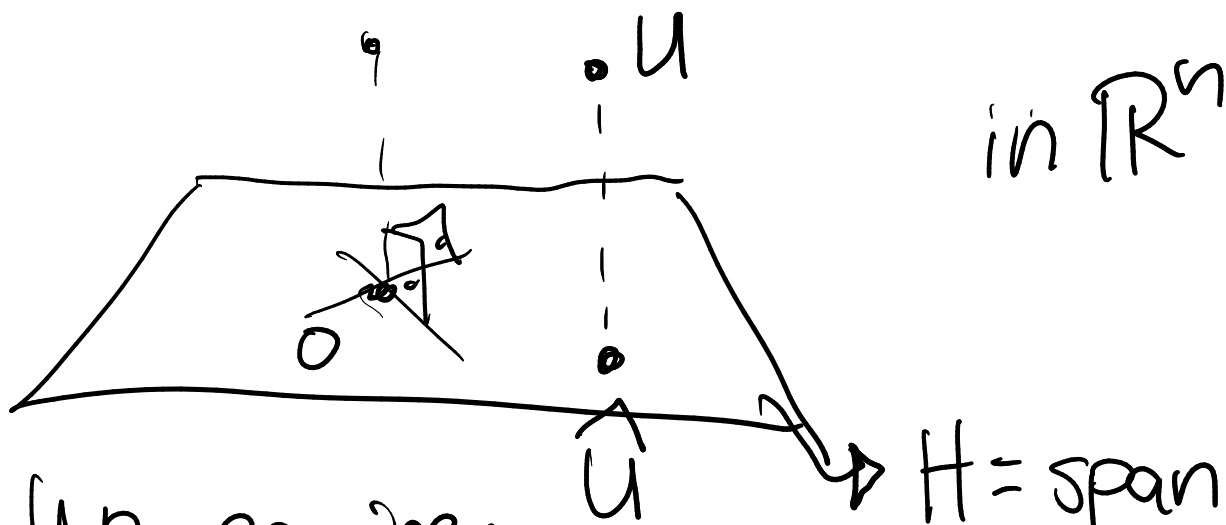


Orthogonal projection to a subspace



We require: $\hat{u} \rightarrow H = \text{span}(a_1, \dots, a_k)$

$$(a) \hat{u} \in H \Leftrightarrow \hat{u} = Ax \text{ for some } x_{k \times 1}$$

where $A = (a_1 \mid \dots \mid a_k)_{n \times k}$.

$$(b) u - \hat{u} \perp H = \text{col}(A)$$

$$\Leftrightarrow u - \hat{u} \perp a_i, \forall i.$$

$$\Leftrightarrow A^T(u - \hat{u}) = 0$$

Thus $A^T u = A^T A x$.

Suppose $A^T A$ is invertible.

$$x = (A^T A)^{-1} A^T u \text{ \&}$$

$$\hat{u} = Ax = \underbrace{A(A^T A)^{-1} A^T}_{P_{n \times n}} u$$

What we did is essentially the following:
 Given u , we express u as

$$u = \hat{u} + (u - \hat{u})$$

$$= \text{proj}_H u + \text{proj}_{H^\perp} u$$

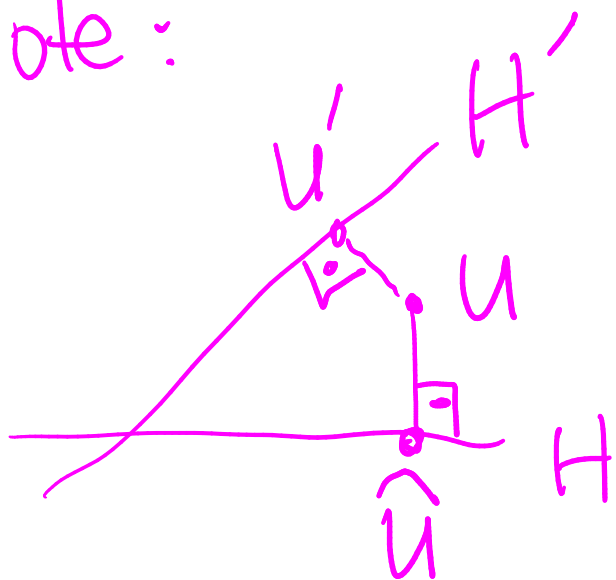
where H^\perp is the orthogonal complement of H .

- $\hat{u} \perp u - \hat{u}$.

- $\text{proj}_{H^\perp} u = u - \hat{u}$:

(a) $u - \hat{u} \in H^\perp$; (b) $u - (u - \hat{u}) = \hat{u} \in H$
 $\hat{u} \perp H^\perp$

Note:



If $H' \neq H^\perp$,
 $u \neq \hat{u} + u'$

ex: In \mathbb{R}^4 , $H = \text{span}(a_1, a_2, a_3, a_4)$ with
 $a_1 = \begin{pmatrix} 1 \\ 0 \\ 2 \\ 1 \end{pmatrix}$, $a_2 = \begin{pmatrix} 2 \\ -1 \\ 1 \\ 0 \end{pmatrix}$, $a_3 = \begin{pmatrix} 3 \\ -1 \\ 3 \\ 1 \end{pmatrix}$, $a_4 = \begin{pmatrix} 4 \\ -1 \\ 5 \\ 2 \end{pmatrix}$.

Find the orthog. proj. of $u = \begin{pmatrix} 3 \\ 2 \\ 0 \\ 1 \end{pmatrix}$ on H .
 $A = (a_1, \dots, a_4)_{4 \times 4}$

$$\text{rank } A = 2$$

$$\Leftrightarrow \dim \text{null } A = 4 - 2 = 2$$

$$\begin{pmatrix} 1 & * \\ 0 & 1 & * \\ \hline 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Take $x \in \text{null}(A)$; i.e. $Ax = 0$.

Then $(A^T A) \underset{k \times k}{x} \underset{k \times 1}{=} \underset{k \times 1}{\vec{0}} \Rightarrow x \in \text{null } A^T A$

So $\text{null}(A) \subset \text{null}(A^T A)$.

In particular $2 = \dim \text{null}(A) \leq \dim \text{null}(A^T A)$

Hence $\text{rank}(A^T A) \leq 2$.

so that $A^T A$ is not invertible.

Instead, define $B = \begin{pmatrix} 1 & 2 \\ 0 & -1 \\ 2 & 1 \\ 1 & 0 \end{pmatrix}_{4 \times 2} = (a_1; a_2)$

$$B^T B = \begin{pmatrix} 1 & 0 & 2 & 1 \\ 2 & -1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 0 & -1 \\ 2 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 6 & 4 \\ 4 & 6 \end{pmatrix}$$

$$(B^T B)^{-1} = \frac{1}{20} \begin{pmatrix} 6 & -4 \\ -4 & 6 \end{pmatrix}$$

$$P = B (B^T B)^{-1} B^T$$

$$= \begin{pmatrix} 1 & 2 \\ 0 & -1 \\ 2 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{20} \begin{pmatrix} 6 & -4 \\ -4 & 6 \end{pmatrix} B^T$$

$$= \frac{1}{20} \begin{pmatrix} -2 & 8 \\ 4 & -6 \\ 8 & -2 \\ 6 & -4 \end{pmatrix} B^T = \frac{1}{20} \begin{pmatrix} 14 & -8 & 4 & -2 \\ -8 & +6 & 2 & 4 \\ 4 & +2 & 14 & 8 \\ -2 & +4 & 8 & 6 \end{pmatrix}$$

Then $\hat{u} = P \cdot u$.

facts.

① Above P is symmetric. But this is always the case:

$$\begin{aligned} P^T &= (A(A^T A)^{-1} A^T)^T \\ &= A((A^T A)^{-1})^T A^T \\ &= A((A^T A)^T)^{-1} A^T = A(A A^T)^{-1} A^T = P \end{aligned}$$

② Moreover $P^2 = P$:

$$\begin{aligned} P^2 &= \cancel{A(A^T A)^{-1}} \cancel{A^T} \cancel{A(A^T A)^{-1}} \cancel{A^T} \\ &= A(A^T A)^{-1} A^T = P \end{aligned}$$

① & ②: P is a projection matrix.

③ If a_1, \dots, a_k are linearly independent then $A^T A$ is invertible

proof: $\text{null}(A) = \{0\}$.

We prove $\text{null}(\underbrace{A^T A}_{\text{square}}) = \{0\}$ too.

We've shown

$$\text{null}(A) \subseteq \text{null}(A^T A)$$

If we show the equality then we're done. So take $x \in \text{null}(A^T A)$

$$\text{i.e. } (A^T A)x = \vec{0}.$$

$$\text{Then } x^T (A^T A)x = 0 \Leftrightarrow (x^T A^T)(Ax) = 0 \\ \Rightarrow (Ax)^T (Ax) = \|Ax\|^2 = 0.$$

$$\text{Hence } Ax = 0.$$

Rank-Nullity Thm part III. \mathbb{R}^n

Multiplying with $A_{p \times q}$ sends $\text{row}(A)$ to $\text{col}(A) \subset \mathbb{R}^p$ in a 1-1 onto fashion.

defn: A 1-1 onto linear transform.
[is called a linear isomorphism.

proof of part II.

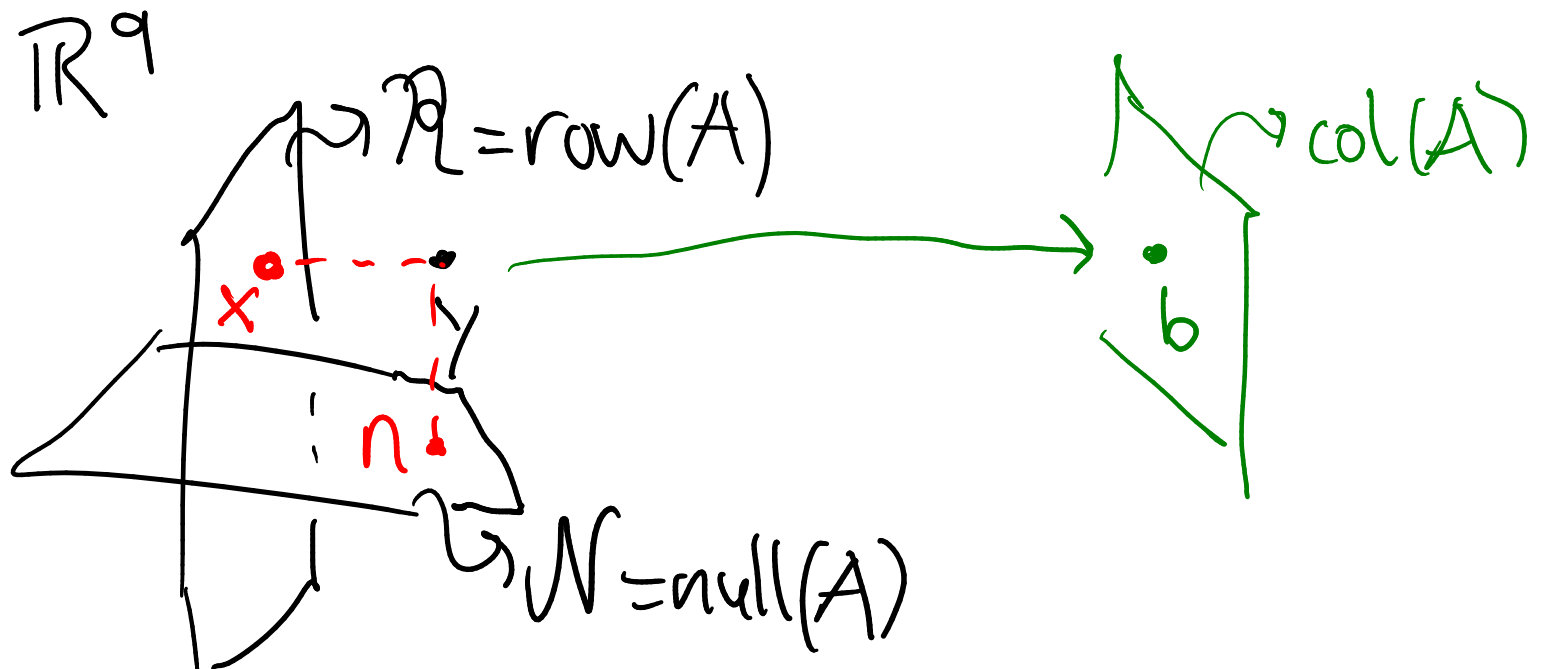
1-1. Let $x, y \in \text{row}(A)$ & $Ax = Ay$.

Then $0 = Ax - Ay = A(\underbrace{x-y})$

$\Rightarrow x-y \in \text{null}(A)$ too. $\text{row}(A)$

$\Rightarrow x-y = 0$.

onto. Given any $b \in \text{col}(A)$, there
is $x \in \text{row}(A)$ such that $Ax = b$.



Let $x = \text{proj}_U y$ & $n = \text{proj}_W y$.

We know $y = x + n$. Then

$$b = Ay = A(x + n) = Ax + \underbrace{An}_0 = Ax.$$

RANK-NULLITY THM (final version)

- $\text{row}(A)$ & $\text{null}(A_{p \times q})$ are orthog. complements in \mathbb{R}^q .
- $\text{col}(A)$ & $\text{null}(A^T)$ are orthog. complements in \mathbb{R}^p .
- $\text{rank}(A) = \dim \text{row}(A) = \dim \text{col}(A)$.
- $\text{row}(A)$ & $\text{col}(A)$ are linearly isomorphic via multiplication with A .