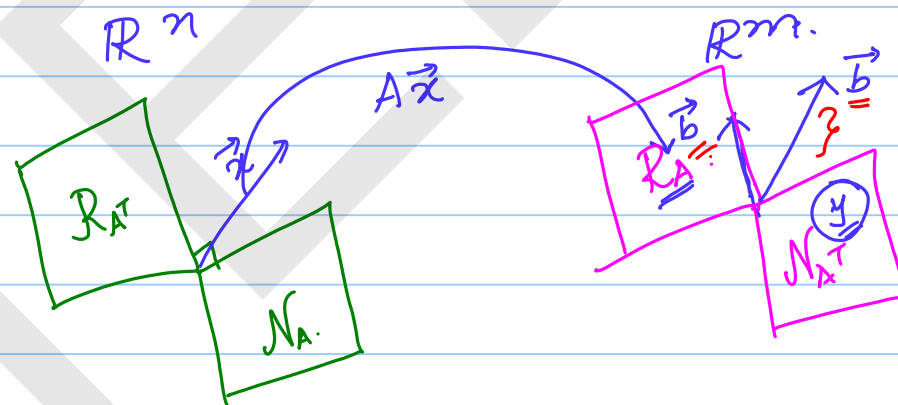


Recall: $R_{A^T} \perp N_A$ in \mathbb{R}^n

$R_A \perp N_{A^T}$ in \mathbb{R}^m .

for any matrix $A \in \mathbb{R}^{m \times n}$.

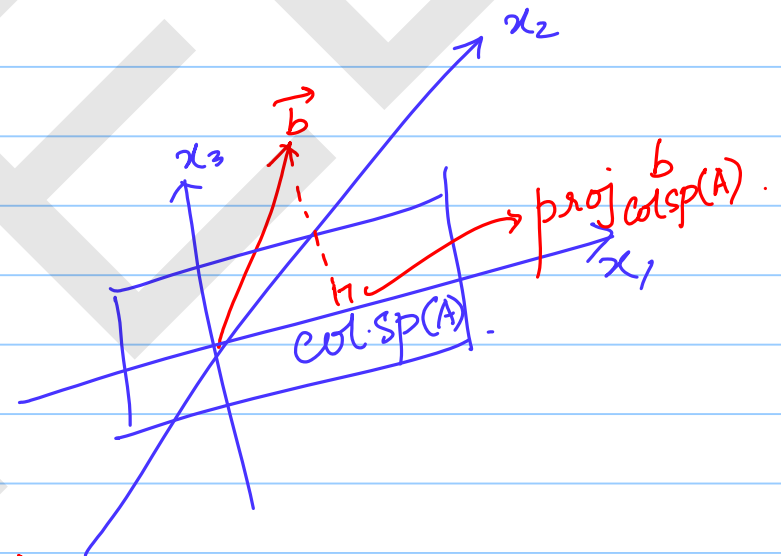


Suppose \vec{b} is not in the
 $\text{Col. Sp}(A) \Rightarrow$ No scalars

$x_1, x_2 \dots x_n$ exist s.t

$$\vec{b} = x_1 A_1 + \dots + x_n A_n$$

A_i : i^{th} col of $A^{m \times n}$



\vec{b} does not belong to the
 $\text{Col. Sp}(A)$.

$\vec{b} \in \mathbb{R}^m$, $\vec{b} \notin \text{Col. Sp}(A)$.

Since \vec{b} does not belong to the $\text{Col. Sp}(A)$, we would like to approx \vec{b} in the $\text{Col. Sp}(A)$

\Rightarrow We want to find that vector \vec{b}^* which best approximates \vec{b} in the $\text{Col. Sp}(A)$.

Note that for whatever x , we have

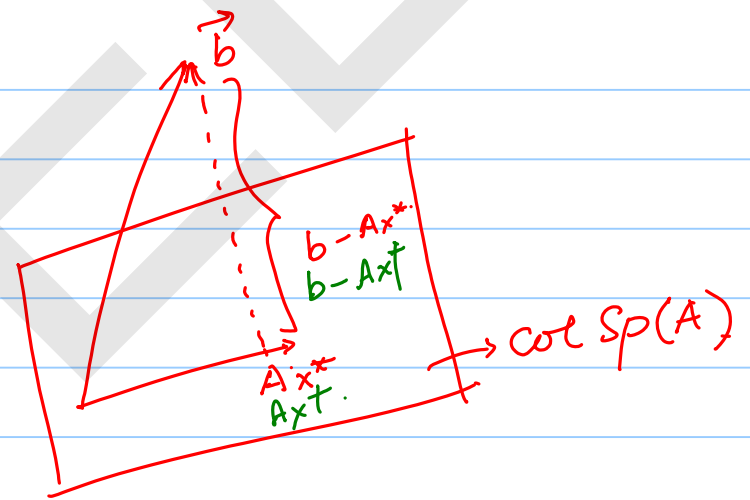
$\|\vec{b} - A\vec{x}\|$ will never be 0.

because no (x_1, \dots, x_n) exists

s.t. $A\vec{x} = \vec{b}$.

So we want to make $\|\vec{b} - A\vec{x}\|$ as small as possible.

If we can find a unique vector x^* that minimizes the length of $b - Ax$, we call x^* to be the best optimal soln to $Ax = b$ in the sense of **LEAST SQUARES**.



$\|b - Ax\|$ is the smallest when

Ax^* is the orthogonal proj of \vec{b} onto the $\text{col-sp}(A)$.

$(b - Ax^*)$ is orthogonal to the
col $Sp(A)$.

We want $(b - Ax^*)$ to be orthog.
to every vector in the col. $Sp(A)$.

any vector in the col. $Sp(A) = Ay$
for $y \in \mathbb{R}^n$.

$$\Rightarrow (b - Ax^*) \cdot Ay = 0.$$

$$= (Ay)^T (b - Ax^*) = 0$$

$$= y^T A^T (b - Ax^*) = 0.$$

$$= y^T (\underbrace{A^T b}_{\uparrow} - \underbrace{A^T A x^*}_{\uparrow}) = 0.$$

for every $\vec{y} \in \mathbb{R}^n$.

Since the only vector which
is orthogonal to every vector
in \mathbb{R}^n is the zero vector, we have

$$A^T b - A^T A x^* = 0.$$

$$\Rightarrow A^T A x^* = A^T b.$$

$$A \in \mathbb{R}^{m \times n} \quad A^T \in \mathbb{R}^{n \times m}$$

$A^T A$: Square matrix of dim. $n \times n$.

Suppose the cols of A are all linearly indep \Rightarrow Rank of $A = n$.

$A^T A$ will be invertible.

$$\Rightarrow \boxed{x^* = (A^T A)^{-1} A^T b.}$$

$x^* = (A^T A)^{-1} A^T b$ is the best optimal soln to $Ax = b$.

$$\boxed{x^\dagger = (A^T A)^{-1} A^T b}$$

$x^\dagger = x$ dagger

We wanted to solve

$$Ax = b.$$

But if $b \notin \text{colSp}(A)$ no soln exists. Therefore we solve

for $Ax^\dagger = b^*$, where

b^* = ^{Orthog} Projection of b onto $\text{colSp}(A)$.

$$x^\dagger = (A^T A)^{-1} A^T b. \rightarrow \textcircled{1}$$

Suppose we have A which is an invertible matrix then given $Ax = b$, we get

$$x = A^{-1} b. \rightarrow \textcircled{2}$$

$(A^T A)^{-1} A^T \rightarrow$ Pseudo inverse of A .

$$x^\dagger = (A^T A)^{-1} A^T b.$$

look at

$$A x^\dagger = \underbrace{(A (A^T A)^{-1} A^T)}_{\text{Projection of } \vec{b} \text{ onto the column space of } A} b$$

Projection of \vec{b} onto the column space of A .

\therefore . The matrix

$A(A^T A)^{-1} A^T$ is the

matrix that projects every

\vec{b} onto the col. $\text{sp}(A)$.

Question: How do we prove
that $A(A^T A)^{-1} A^T$ is actually
a projection matrix?

Question 2: What is a projection
matrix?

Question 3: What are the properties
of a projection matrix?