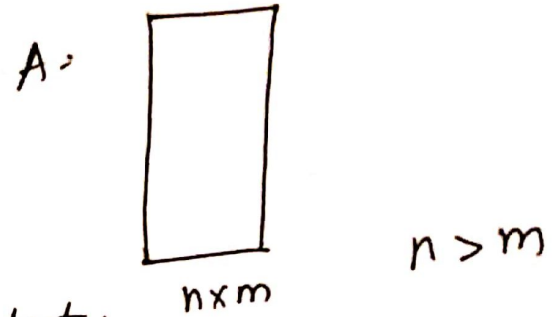


Module 3 - Lecture 5

Nov 26, 2019



Today: Finish Least Squares.

Is ATA invertible?

→ Columns of matrix $A \rightarrow$ linearly independent.

- Example ATA non invertible
- Orthogonal Matching Pursuit.

Last lecture: Thm: $\underbrace{\text{Null}(ATA)}_{\text{invertibility of } ATA} = \underbrace{\text{Null}(A)}_{\text{Null}(A) \text{ is non-trivial if columns of } A \text{ are linearly dependent.}}$

if $A\vec{x} = \vec{0}$, $\vec{x} \neq \vec{0}$.
linear comb. of columns.

Property of transposes.

$$A: n \times m$$

$$B: m \times k$$

$$(AB)^T: k \times n$$

$$AB: n \times k$$

$$A^T: m \times n$$

$$B^T: k \times m$$

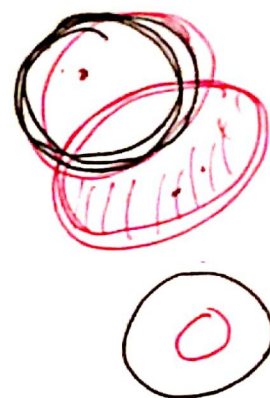
$$(AB)^T = B^T A^T$$

$$B^T A^T: k \times n.$$

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

Proof: (1) If $\vec{v} \in \text{Null}(A)$, then $\vec{v} \in \text{Null}(ATA)$
 (2) If $\vec{w} \in \text{Null}(ATA)$, then $\vec{w} \in \text{Null}(A)$



(1) $\vec{v} \in \text{Null}(A)$
 $A \cdot \vec{v} = \vec{0}$

Want: $\vec{v} \in N(ATA)$
 $ATA \vec{v} = \vec{0}$

Mult. by A^T .

$A^T A \cdot \vec{v} = A^T \cdot \vec{0} = \vec{0}$
 $\Rightarrow \vec{v} \in \text{Null}(ATA)$

(2) $\vec{w} \in N(ATA)$

$ATA\vec{w} = \vec{0}$

Want: $\vec{w} \in N(A)$
 $A\vec{w} = \vec{0}$

Consider: $\|A\vec{w}\|^2 = \langle A\vec{w}, A\vec{w} \rangle$
 $= (A\vec{w})^T A\vec{w} = (\vec{w}^T A^T) A\vec{w} = \vec{w}^T (A^T A \vec{w})$
 $= \vec{w}^T \underline{ATA\vec{w}}$
 $= 0 \checkmark$

~~$\langle x, y \rangle = x^T y$~~

The only way this is possible is if $A\vec{w} = \vec{0} = \vec{w} \in N(A)$.

• If $\|\vec{x}\|^2 = \underbrace{x_1^2}_{\geq 0} + \underbrace{x_2^2}_{\geq 0} + \dots + \underbrace{x_n^2}_{\geq 0} = 0$.

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$x_1 = 0, x_2 = 0, \dots, x_n = 0.$$

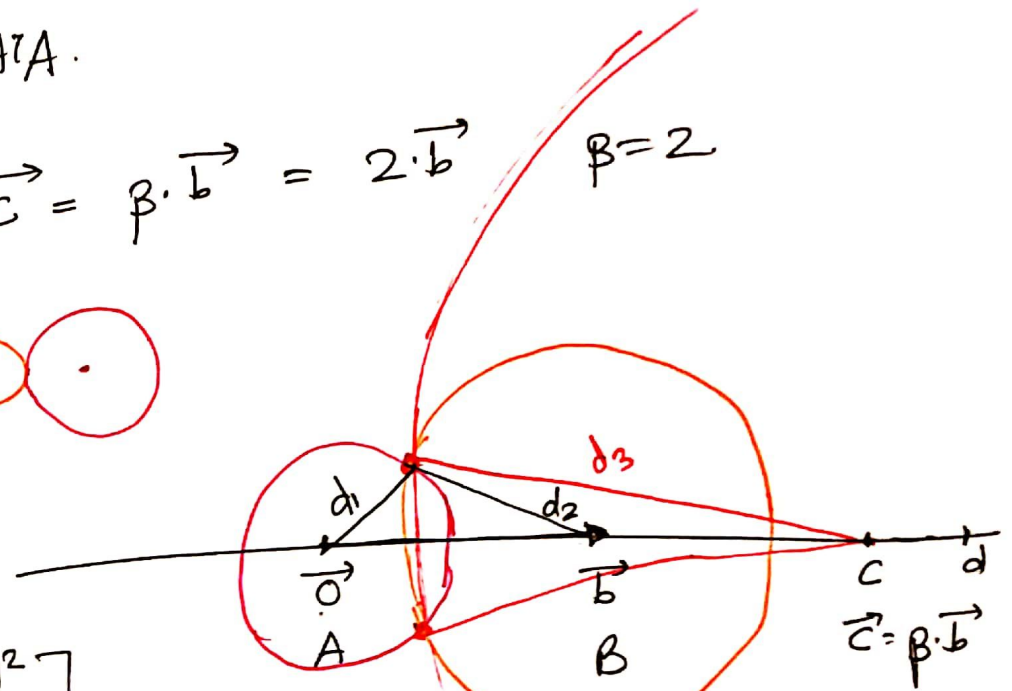
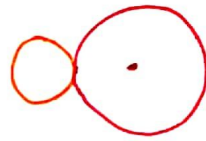
$$x_1 = (-1), x_1^2 = +1.$$

Thm! If $\|\vec{x}\|^2 = 0$,
then $\vec{x} = \vec{0}$

Columns of $A \iff$ Invertibility of $A^T A$.

$$\vec{a} = \vec{0}, \quad \vec{b} = \vec{b}, \quad \vec{c} = \beta \cdot \vec{b} = 2 \cdot \vec{b} \quad \beta = 2$$

$$\vec{b} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$



$$2 \underbrace{\begin{bmatrix} \vec{b}^T - \vec{a}^T \\ \vec{c}^T - \vec{a}^T \end{bmatrix}}_A = \begin{bmatrix} d_1^2 - d_2^2 - \|\vec{a}\|^2 + \|\vec{b}\|^2 \\ d_1^2 - d_2^2 - \|\vec{a}\|^2 + \|\vec{c}\|^2 \end{bmatrix}$$

KNOWN.

$$\begin{bmatrix} \vec{b}^T \\ \vec{c}^T \end{bmatrix} = \begin{bmatrix} \vec{b}^T \\ 2 \vec{b}^T \end{bmatrix} = \begin{bmatrix} b_1 & b_2 \\ 2b_1 & 2b_2 \end{bmatrix}$$

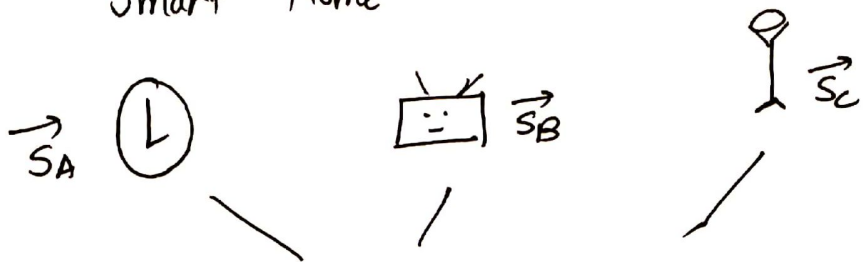
Columns are linearly dep.

~~Is A~~ Are columns of A linearly indep? \rightarrow NO.

\Rightarrow $A^T A$ invertible? \rightarrow NO.

Orthogonal Matching Pursuit

"Smart Home"



How do you figure out which devices are transmitting?

10,000s of devices.

Gold codes:

\vec{s}_A, \vec{s}_B

± 1 Gold codes

$$\langle \vec{s}_A, \vec{s}_A \rangle = \text{large} \\ = N \text{ (length of vector).}$$

$$\langle \vec{s}_A, \vec{s}_B \rangle = \text{small, almost orthogonal.}$$

1023 length
gold codes.

$$\vec{r} = \alpha_1 \vec{s}_1 + \alpha_2 \vec{s}_2 + \dots + \alpha_{10,000} \vec{s}_{10,000}$$

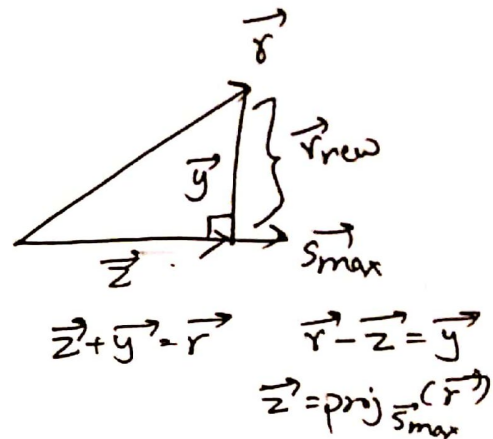
1023

$$\begin{bmatrix} | & | & & | \\ \hline \vec{s}_1 & \vec{s}_2 & \dots & \vec{s}_{10,000} \\ \hline | & | & & | \\ \hline \end{bmatrix} \begin{matrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_{10,000} \end{matrix} = \vec{r}$$

10,000

1023

$$\vec{r} = \vec{s}_A + 2\vec{s}_B$$



How to solve this:

GREEDY.

Consider:

- $\langle \vec{r}, \vec{s}_A \rangle$
- $\langle \vec{r}, \vec{s}_B \rangle$
- \vdots
- $\langle \vec{r}, \vec{s}_C \rangle$
- \vdots
- $\langle \vec{r}, \vec{s}_Z \rangle$
- \vdots
- $\langle \vec{r}, \vec{s}_{10,000} \rangle$

- Consider all the inner products.
 - Find the maximum inner product: \vec{s}_{\max}
- Guess: \vec{s}_{\max} is transmitting.

Project: \vec{r} into \vec{s}_{\max}

$$\text{proj}_{\vec{s}_{\max}}(\vec{r})$$

$$\vec{r}_{\text{new}} = \vec{r} - \text{proj}_{\vec{s}_{\max}}(\vec{r})$$