

## Recitation #3 (Section 03)

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for Data Science



## Rank

Let  $x_1, \dots, x_k \in \mathbb{R}^n$  we define the rank as

$$\text{rank}(x_1, \dots, x_k) = \dim(\text{Span}(x_1, \dots, x_k))$$

Informally: rank = "the number of linearly independent vectors among  $x_1, \dots, x_k$ "

# practice: the Rank

## Exercise 1

Calculate the rank of

$$A = \begin{pmatrix} 7 & 3 \\ 19 & 19 \\ -1 & 5 \end{pmatrix}$$

- ① Gaussian elimination
- ②  $\text{rank}(A) = \# \text{ col/rows that are non zero}$

$$A \rightarrow \begin{pmatrix} * & * \\ 0 & * \\ 0 & 0 \end{pmatrix} \quad \text{rank}(A) = 2$$

## Exercise 2

Let  $A, B \in \mathbb{R}^{n \times n}$  be two matrices then

$$\text{rank}(A + B) = \text{rank}(A) + \text{rank}(B)$$

True or false? ✓True ✗False

$$\begin{matrix} 1 \times 1 & & 1 \times 1 \\ A = (1) & B = (-1) \end{matrix}$$

$$A + B = (0)$$

$$\text{rank}(A + B) = 0$$

$$\text{rank}(A) = \text{rank}(B) = 1 \quad \left\{ \begin{array}{l} 0 \neq 1 \end{array} \right.$$

# Rank nullity theorem

## Rank nullity theorem

Let  $L : \mathbb{R}^m \rightarrow \mathbb{R}^n$  be a linear transformation then

$$\text{rank}(L) + \dim(\ker(L)) = m$$



$$\tilde{L} = n \times m \quad m \text{ in } \mathbb{R}^m$$

$$\begin{matrix} n \\ \text{in } \mathbb{R}^n \end{matrix} \uparrow \left( \begin{array}{c|c|c|c|c} | & | & | & | & | \\ \hline \end{array} \right)$$

# practice: the Rank

## Exercise 3

Let  $A \in \mathbb{R}^{m \times n}$  be a matrix. Prove that  $\text{rank}(A) = \text{rank}(A^T)$ .  
(That is "the rank by columns = the rank by rows")

We will proceed by steps:

① Prove that  $x^T A^T A x \geq 0$  for all  $x \in \mathbb{R}^n$

② When is  $x^T A^T A x = 0$ ?

③ Prove that  $\ker(A) = \ker(A^T A)$

④ Use this to show  $\text{rank}(A) = \text{rank}(A^T A)$

⑤ Show that  $\text{rank}(A) = \text{rank}(A^T)$

$$\begin{array}{l} x=0 \\ A x=0 \\ x \in \ker(A) \end{array}$$

# Exercise 3.1: Prove that $x^T A^T A x \geq 0$ for all $x \in \mathbb{R}^n$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \quad \begin{array}{l} A \text{ } m \times n \\ x \text{ } n \times 1 \\ x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \end{array}$$

When is  $x^T A^T A x = 0$ ?  $x = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$   $x$  is generic

$$x^T A^T A x = \underbrace{(Ax)^T}_{x} (Ax) = (y_1 \dots y_m) \begin{pmatrix} y_1 \\ \vdots \\ y_m \end{pmatrix} = y_1^2 + \dots + y_m^2 = \sum_{i=1}^m y_i^2 \geq 0$$

$$Ax = \begin{pmatrix} \sum_j a_{1j} x_j \\ \vdots \\ \sum_j a_{mj} x_j \end{pmatrix} = \begin{pmatrix} y_1 \\ \vdots \\ y_m \end{pmatrix}$$

### Exercise 3.3: Prove that $\ker(A) = \ker(A^T A)$

$$\textcircled{1} \ker(A) \subseteq \ker(A^T A)$$

$$v \in \ker(A) \rightarrow Av = 0 \rightarrow A^T(Av) = A^T 0 = 0$$

$$(A^T A)v = 0 \rightarrow v \in \ker(A^T A) \quad \square$$

$$\textcircled{2} \ker(A^T A) \subseteq \ker(A)$$

$$v \in \ker(A^T A) \rightarrow (A^T A)v = 0 \rightarrow \underbrace{A^T}_{\neq 0} \underbrace{(Av)}_{=0} = 0$$

$$A \neq 0 \rightarrow Av = 0 \rightarrow v \in \ker(A)$$

$\square$



Exercise 3.4: Use this to show  $\text{rank}(A) = \text{rank}(A^T A)$

$$\text{Ker}(A) = \text{Ker}(A^T A)$$

$$A \text{ } m \times n$$

• Rank nullity theorem:

$$\text{rank}(A) + \dim(\text{Ker}(A)) = n$$

$$\text{rank}(A^T A) + \dim(\text{Ker}(A^T A)) = \textcircled{n}$$

$A^T A \rightarrow m \times m$   
 $\begin{matrix} \parallel & \parallel \\ m \times m & m \times n \end{matrix}$

$$\rightarrow \text{rank}(A) = \text{rank}(A^T A) \quad \square$$

Exercise 3.5: Show that  $\text{rank}(A) = \text{rank}(A^T)$

$$\boxed{\text{rank}(A) = \text{rank}(A^T A)}$$

We know

$$\begin{array}{l} A \text{ } m \times n \\ A^T \text{ } n \times m \end{array}$$

$$\textcircled{1} \text{rank}(A) \leq \text{rank}(A^T A)$$

Recall:  $\text{rank}(T_1 T_2) \leq \min(\text{rank}(T_1), \text{rank}(T_2))$

$$\underline{\text{rank}(A)} = \text{rank}(A^T A) \leq \min(\text{rank}(A), \text{rank}(A^T)) \leq$$

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

$$\textcircled{A^T A} \textcircled{A A^T} \xrightarrow{\text{rank}(A^T)}$$

$$\text{rank}(A^T) \geq \text{rank}(A^T A) \text{ replace rank}(A^T) \text{ by the rank}(A) \rightarrow \text{rank}(A) \geq \text{rank}(A^T) \quad \square$$

## recall: invertible matrix

### Invertible matrix

A matrix  $M \in \mathbb{R}^{n \times n}$  is invertible if there exists another matrix  $M^{-1} \in \mathbb{R}^{n \times n}$  such that

$$M M^{-1} = M^{-1} M = Id_n$$

### Exercise 4

Let  $M \in \mathbb{R}^{n \times n}$  be an invertible matrix, then its transpose  $M^T$  is also invertible.

$$M M^{-1} = Id$$

$$(M M^{-1})^T = Id^T$$

$$M^T (M^{-1})^T = Id$$

$$M^T (M^T)^{-1} = Id$$

$\hookrightarrow M^T$  has an inverse

The same goes for  $M^{-1} M = Id$



## Exercise 5

Which of the following are equivalent for  $A \in \mathbb{R}^{m \times n}$ ?

- ❶ The columns of  $A$  are linearly independent
- ❷ The rows of  $A$  are linearly independent
- ❸  $\text{rank}(A) = m$
- ❹  $\text{rank}(A) = n$
- ❺ The equation  $Ax = 0$  has one solution
- ❻ The equation  $Ax = b$  has at least one solution
- ❼  $\text{Im}(A^T) = \mathbb{R}^n$
- ❽  $\text{Im}(A) = \mathbb{R}^m$
- ❾ The linear transformation corresponding to  $A$  is injective (one-to-one)
- ❿ The linear transformation corresponding to  $A$  is exhaustive (onto)
- ⓫  $\ker(A) = \{0\}$
- ⓬ The span of the columns of  $A$  is  $\mathbb{R}^m$

## Exercise 5

goal

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

Know  $\text{rank}(A) = \text{rank}(A^T A)$  ✓

$$A \leftrightarrow A^T$$

$$\text{rank}(A^T) \neq \text{rank}(A) \quad \checkmark$$

$m \times m$

~~rank~~

$$A A^T$$

$$\begin{aligned} \text{rank}(A^T) &= \text{rank}(A A^T) \leq \min(\text{rank}(A), \text{rank}(A^T)) \\ &\leq \text{rank}(A) \end{aligned}$$

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