# Recitation #3 (Section 03)

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DS-GA 1014: Optimization and Computational Linear Algebra for Data Science



## recall: the Rank

#### Rank

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

$$\mathsf{rank}(x_1,..,x_k) = \mathsf{dim}(\mathit{Span}(x_1,..,x_k))$$

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

# practice: the Rank

## Exercise 1

Calculate the rank of

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

# practice: the Rank

#### Exercise 2

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

$$rank(A + B) = rank(A) + rank(B)$$

# Rank nullity theorem

### Rank nullity theorem

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

$$rank(L) + dim(ker(L)) = m$$

## practice: the Rank

#### Exercise 3

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

We will proceed by steps:

- **1** Prove that  $x^T A^T A x \ge 0$  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  $rank(A) = rank(A^T A)$
- **3** Show that  $rank(A) = rank(A^T)$

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

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

When is  $x^T A^T A x = 0$ ?

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

Exercise 3.4: Use this to show  $rank(A) = rank(A^T A)$ 

# Exercise 3.5: Show that $rank(A) = rank(A^T)$

#### Solution

- Prove  $rank(A) \ge rank(A^T)$ Recall that  $rank(T_1T_2) \le min(rank(T_1), rank(T_2))$ . Then, starting from  $rank(A) = rank(A^TA)$  we can say that  $rank(A) = rank(A^TA) \le min(rank(A^T), rank(A)) \le rank(A^T)$
- Prove  $rank(A^T) \ge rank(A)$ Same as in 1) but interchanging the roles of A and  $A^T$ . Note that  $(A^T)^T = A$ . Strating from  $rank(A) = rank(A^TA)$  and switching roles we have  $rank(A^T) = rank(AA^T)$ . Now, same as before

$$rank(A^T) = rank(AA^T) \le min(rank(A), rank(A^T)) \le rank(A)$$

And we are done.

## 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.

## Exercise 5

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

- **1** The columns of *A* are linearly independent
- 2 The rows of A are linearly independent
- rank(A) = m
- $\bullet$  rank(A) = n
- **5** The equation Ax = 0 has one solution
- The equation Ax = b has at least one solution for all b (assuming  $b \neq 0$ )
- $\mathbf{0}$   $Im(A^T) = \mathbb{R}^n$
- The linear transformation corresponding to A is injective (one-to-one)
- $\odot$  The linear transformation corresponding to A is exhaustive (onto)
- **1**  $ker(A) = \{0\}$
- **2** The span of the columns of A is  $\mathbb{R}^m$

## Exercise 5

#### Solution

- Equivalent to rank(A) = n
- 2 Equivalent to rank(A) = m
- rank(A) = m
- $\bullet$  rank(A) = n
- Equivalent to  $ker(A) = \{0\}$ , by rank nullity theorem equivalent to rank(A) = n
- This means dim(Im(A)) = m so equivalent to rank(A) = m
- **1** Equivalent to rank(A) = n
- **3** Equivalent to rank(A) = m
- ① This means  $ker(A) = \{0\}$  so by rank nullity theorem is equivalent to rank(A) = n
- This means dim(Im(A) = m) so equivalent to rank(A) = m
- **1** By rank nullity theorem equivalent to rank(A) = n