

# Recitation 3

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# Rank Nullity Theorem

## Theorem (Rank-Nullity Theorem (!!!))

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

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

- ▶ **One of the most important theorems in linear algebra.**
- ▶ You should be able to state and prove this theorem (with no notes).
- ▶ The ‘conservation of dimension’ theorem
  - ▶ The main rank inequality:
    - ▶  $\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$
- ▶ If  $n > m$ , then the  $\max(\dim(\text{Im}(L))) = m$

# Questions: Rank-Nullity Theorem

Let  $A \in \mathbb{R}^{l \times h}$  and  $B \in \mathbb{R}^{h \times l}$ , and  $h > l$ .

Prove or give a counterexample to the following statements.

1.  $\exists A, B$  s.t  $AB$  is invertible.
2.  $\exists A, B$  s.t.  $BA$  is invertible.

# Questions: Rank-Nullity Theorem

Let  $A \in \mathbb{R}^{l \times h}$  and  $B \in \mathbb{R}^{h \times l}$ , and  $h > l$ .

Prove or give a counterexample to the following statements.

## Solution

1.  $\exists A, B$  s.t  $AB$  is invertible. **True**

$$\text{Consider } A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}. AB = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

2.  $\exists A, B$  s.t.  $BA$  is invertible. **False.**

In order for  $BA$  to be invertible,  $\text{rank}(BA) = 3$ . However,  $\text{rank}(BA) \leq \text{rank}(B) = 2$ .

# Symmetric Matrices: That's cute!

- ▶ Symmetric Matrices are not just “cute”...
  - ▶ They are actually DEEPLY LINKED to many topics in linear algebra.
- ▶ Concepts involving Symmetric Matrices
  - ▶ Orthogonal Projections (Lec 4) are symmetric.
  - ▶ Spectral Theorem (Lec 7) “eigenvectors of symmetric matrices are orthogonal”.
  - ▶ PCA: Covariance matrix is symmetric
  - ▶ Concavity: Hessian Matrix (matrix of second derivative) is symmetric
- ▶ But, we will see most of this later. For now, just trust me!

# Questions: Symmetric Matrices

Let  $A \in \mathbb{R}^{k \times n}$ .

Prove/answer the following statements.

1. Show that  $\forall x \in \mathbb{R}^n, x^T A^T A x \geq 0$
2. When is  $x^T A^T A x = 0$ ?
3. Show that  $\text{Ker}(A) = \text{Ker}(A^T A)$
4. Use this to show  $\text{rank}(A) = \text{rank}(A^T A)$
5. Now, show that  $\text{rank}(A) = \text{rank}(A^T)$

# Solutions: Symmetric Matrices

Let  $A \in \mathbb{R}^{k \times n}$ .

## Solution

1. Show that  $\forall x \in \mathbb{R}^n, x^T A^T A x \geq 0$

Let  $y = Ax$ , with  $y = \begin{bmatrix} y_1 & \dots & y_n \end{bmatrix}^T$

Then  $x^T A^T A x = (Ax)^T (Ax) = y^T y = \sum_{i=1}^n y_i^2$ .

Since  $y_i$  is a real number,  $\sum_{i=1}^n y_i^2 \geq 0$ .

2. What happens when  $x^T A^T A x = 0$ ?

$\sum_{i=1}^n y_i^2 = 0 \iff y_i = 0 \quad \forall i \in \{1, \dots, n\}$

So,  $x^T A^T A x = 0 \iff x \in \text{Ker}(A)$

# Solutions: Symmetric Matrices

Let  $A \in \mathbb{R}^{k \times n}$ .

3. Show that  $\text{Ker}(A) = \text{Ker}(A^T A)$

## Solution

$\text{Ker}(A) \subset \text{Ker}(A^T A)$  is trivial.

We now show  $\text{Ker}(A^T A) \subset \text{Ker}(A)$ .

Let  $x \in \text{Ker}(A^T A)$ .

Then  $A^T A x = 0$ .

Then  $x^T (A^T A x) = x^T 0 = 0$

By the previous question, then  $x \in \text{Ker}(A)$ .



# Solutions: Symmetric Matrices

Let  $A \in \mathbb{R}^{k \times n}$ .

Prove or give a counter example to the following statements.

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

## Solution

*By the rank nullity theorem,*

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

*Now,  $A^T A \in \mathbb{R}^{n \times n}$ . So by the rank nullity theorem,*

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

*Setting these equations equal to each other yields:*

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

*And since  $\text{Ker}(A) = \text{Ker}(A^T A)$ , then*

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

# Solutions: Symmetric Matrices

Let  $A \in \mathbb{R}^{k \times n}$ .

5. Now show  $\text{rank}(A) = \text{rank}(A^T)$

## Solution

*By the previous question,*

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

*and similarly,*

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

*Recall that  $\text{rank}(T_1 T_2) \leq \min(\text{rank}(T_1), \text{rank}(T_2))$ , and therefore,*

$$\text{rank}(T_1 T_2) \leq \text{rank}(T_1).$$

*Applying this to  $\text{rank}(A^T A)$  and  $\text{rank}(A A^T)$  yields:*

$$\text{rank}(A^T) \geq \text{rank}(A^T A) \quad \text{and} \quad \text{rank}(A) \geq \text{rank}(A A^T).$$

*Replacing  $\text{rank}(A^T A)$  and  $\text{rank}(A A^T)$ , we get that*

$$\text{rank}(A^T) \geq \text{rank}(A) \quad \text{and} \quad \text{rank}(A) \geq \text{rank}(A^T).$$

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

# Solutions: Matrix Products

Let  $x, y \in \mathbb{R}^{n \times 1}$ .

1. What is the shape and rank of  $x^T y$ ?
2. What is the shape and rank of  $xy^T$ ?
3. Let  $A \in \mathbb{R}^{m \times k}$  and  $B \in \mathbb{R}^{k \times n}$ . Show that the matrix product  $AB$  can be expressed as:  $AB = C_1 + \cdots C_k$  s.t  $rank(C_i) \leq 1$ ,  $\forall i \in \{1, \dots, k\}$ .  
(Hint, try manually calculating for small values of  $m, k, n$ )

# Questions: Matrix Products

Let  $x, y \in \mathbb{R}^{n \times 1}$  both have rank 1.

## Solution

1. What is the shape and rank of  $x^T y$ ?  
Shape is  $1 \times 1$  and rank is 1 (or 0).
2. What is the shape and rank of  $xy^T$ ?  
Shape is  $n \times n$  and rank is 1 (or 0).
3. Let  $A \in \mathbb{R}^{m \times k}$  and  $B \in \mathbb{R}^{k \times n}$ . Show that the matrix product  $AB$  can be expressed as:  $AB = C_1 + \cdots C_k$  s.t  $\text{rank}(C_i) \leq 1$ ,  $\forall i \in \{1, \dots, k\}$ .  
(Hint, use 2. and manually calculating for small values of  $m, k, n$ )

$$\text{Let } A = \begin{bmatrix} | & \dots & | \\ a_1 & \dots & a_k \\ | & \dots & | \end{bmatrix} \text{ and } B = \begin{bmatrix} - & b_1 & - \\ \vdots & \vdots & \vdots \\ - & b_k & - \end{bmatrix}$$