

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 $\dim(\text{Im}(L)) \leq m$

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

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, $\text{rank}(BA) = h$. However, $\text{rank}(BA) \leq \text{rank}(B) \leq \ell$.

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,

$$\text{rank}(A) = \text{rank}(A^T A). \quad (1)$$

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

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

Replacing $\text{rank}(A^T A)$ with $\text{rank}(A)$ gives:

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

Apply this again for $\text{rank}(A^T) = \text{rank}(A A^T)$ (start from (1)) gives:

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

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 $\text{rank}(C_i) \leq 1$, $\forall i \in \{1, \dots, k\}$.
(Hint, use 2, and 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 1×1 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)

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