

Linear Algebra Tutorial8

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Plus/Minus Theorem

V is a vector space, $S \subset V$

- If S is an independent set, and $\mathbf{v} \in V, \mathbf{v} \notin S$, then $S \cup \{\mathbf{v}\}$ is also an **independent set**

proof by contradiction, suppose that $S \cup \{\mathbf{v}\}$ is linear dependent

$$\Rightarrow \mathbf{v} = \text{span}(S)$$

- If $\mathbf{v} \in S$, and \mathbf{v} can be written as a linear combination of other vectors in S , then $\text{span}(S - \mathbf{v}) = \text{span}(S)$

$\mathbf{v} \in S$, WLOG, take $\mathbf{v} = v_1$, consider $\forall \mathbf{w} \in \text{span}(S)$, can be written as linear combination of $\mathbf{v}_2, \dots, \mathbf{v}_n$

coordinate

$n \geq 1, \dim(V) = n, S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a basis of V

- 1. for any vector set $M = \{\mathbf{w}_1, \dots, \mathbf{w}_r\} \subset V$
 M is an independent set $\Leftrightarrow [\mathbf{v}_1]_S, \dots, [\mathbf{v}_r]_S$ are independent set

proof: set $[\mathbf{v}_i] = (a_{i1}, \dots, a_{in})$, then $\mathbf{v}_i = a_{i1}\mathbf{v}_1 + \dots + a_{in}\mathbf{v}_n$

- 2. for vector set $M = \{\mathbf{w}_1, \dots, \mathbf{w}_n\} \subset V$
 M is the basis of $V \Leftrightarrow [\mathbf{v}_1]_S, \dots, [\mathbf{v}_n]_S$ is the basis of \mathbb{R}^n
 $\Leftrightarrow [\mathbf{v}_1]_S, \dots, [\mathbf{v}_n]_S$ is the standard basis of \mathbb{R}^n

from 1., we know that M is independent $\Rightarrow [\mathbf{v}_1]_S, \dots, [\mathbf{v}_n]_S$ is independent, so
we just need to prove that $\text{span}\{[\mathbf{w}_1]_S, \dots, [\mathbf{w}_n]_S\} = \mathbb{R}^n$

example

例子：(2022年线性代数考试题)

Let $p_1(x) = 1 + 3x, p_2(x) = 2 + 4x, p_3(x) = -4x^2$ be three polynomials in P_2 .

1. Let $M = \{1, x, x^2\}$ be the standard basis of P_2 . Let $A = \begin{bmatrix} [p_1(x)]_M & [p_2(x)]_M & [p_3(x)]_M \end{bmatrix}$ be the matrix such that its columns are $[p_1(x)]_M, [p_2(x)]_M$ and $[p_3(x)]_M$. Compute the adjoint matrix of A .
2. Prove that $S = \{p_1(x), p_2(x), p_3(x)\}$ is a basis of P_2 .

basis' theorem

V is a vector space, $\dim(V) = n$, $S = \{\mathbf{v}_1, \dots, \mathbf{v}_m\} \subset V$

- $\text{span}(S) = V$, $m > n$, then we can delete some of the vectors in S to get a basis of V
- if S is a linear independent set, $m < n$, then we can add some vectors to S to get a basis of V

basis' theorem

V is a vector space, $\dim(V) = n$, $W \subset V$ is a subspace.

- let $m = \dim(W)$, then $m \leq n$
- $W = V$ iff $m = n$

Change of basis

V is the vector space, B, B' are two bases of V

- $B = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$
- $B' = \{\mathbf{v}'_1, \dots, \mathbf{v}'_n\}$

If we have $(\mathbf{v})_B = (c_1, \dots, c_n)$

how could we find $(\mathbf{v})_{B'} = (c'_1, \dots, c'_n)$?

as we know that $[\mathbf{v}]_B, [\mathbf{v}]_{B'}$ has the unique expression

Change of basis

- transition matrix(过渡矩阵/转移矩阵) P
 P is invertible, P^{-1} is called the transition matrix from B to B'
- transition matrix from B to B'
 $P_{B \rightarrow B'}$ or $P_{B' \leftarrow B}$
- transition matrix from B' to B
 $P_{B' \rightarrow B}$ or $P_{B \leftarrow B'}$
- $P_{B \leftarrow B'} P_{B' \leftarrow B} = I$

notice that the definition of the transition matrix may be different with some of the Chinese textbooks!!

Change of basis

- We can represent the transition matrix as $P_{B \rightarrow B'}$ or $P_{B' \leftarrow B}$
- $[v]_{B'} = P_{B' \leftarrow B} [v]_B$
- $[v]_B = P_{B \leftarrow B'} [v]_{B'}$
- method to get the transition matrix
 $[B' | B] \Rightarrow [I | P_{B' \leftarrow B}]$

example of transition matrix

例子：考虑 \mathbb{R}^3 的两组基底 $B = \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ 与 $B' = \{\mathbf{u}'_1, \mathbf{u}'_2, \mathbf{u}'_3\}$ ：

$$\mathbf{u}_1 = \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}, \mathbf{u}_2 = \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}, \mathbf{u}_3 = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix};$$

$$\mathbf{u}'_1 = \begin{bmatrix} 3 \\ 1 \\ -5 \end{bmatrix}, \mathbf{u}'_2 = \begin{bmatrix} 1 \\ 1 \\ -3 \end{bmatrix}, \mathbf{u}'_3 = \begin{bmatrix} -1 \\ 0 \\ 2 \end{bmatrix}.$$

我们利用之前介绍的计算步骤求解转移矩阵 $P_{B' \leftarrow B}$ 。

Row space, Column space and Null space

A is a $m \times n$ matrix

- row space 行空间

$$\text{row}(A) = \text{span}\{\mathbf{r}_1, \dots, \mathbf{r}_m\}$$

- column space 列空间

$$\text{col}(A) = \text{span}\{\mathbf{c}_1, \dots, \mathbf{c}_n\}$$

- null space 零空间

$$\text{null}(A) = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} = \mathbf{0}\}$$

- left null space 左零空间

$$\text{null}(A^T) = \{\mathbf{x} \in \mathbb{R}^m : A^T\mathbf{x} = \mathbf{0}\}$$

Fundamental Matrix Spaces

Definition 4.31. 对 $m \times n$ -矩阵 A , 以下六个向量空间被称为 A 的基本空间(*the fundamental spaces of A*):

- A 的行空间 $Row(A)$,
 - A 的列空间 $Col(A)$,
 - A^T 的行空间 $Row(A^T)$,
 - A^T 的列空间 $Col(A^T)$,
 - A 的零空间 $Null(A)$,
 - A^T 的零空间 $Null(A^T)$ 。
-
- 行空间和零空间互为正交补
 - 列空间和左零空间互为正交补

正交补(Orthogonal Complements)

正交: $col(A) \perp null(A)$

补: $col(A) + null(A) = \mathbb{R}^n$

Row space, Column space and Null space

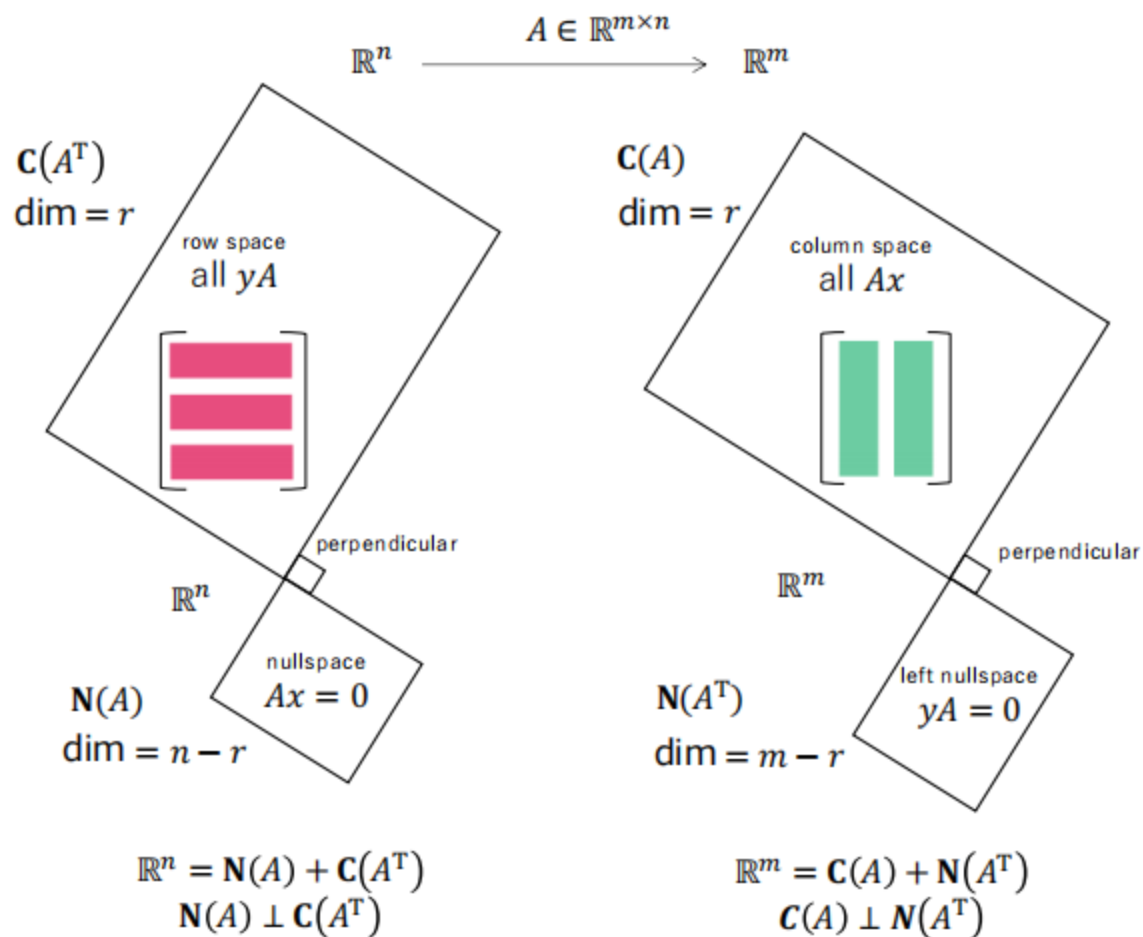
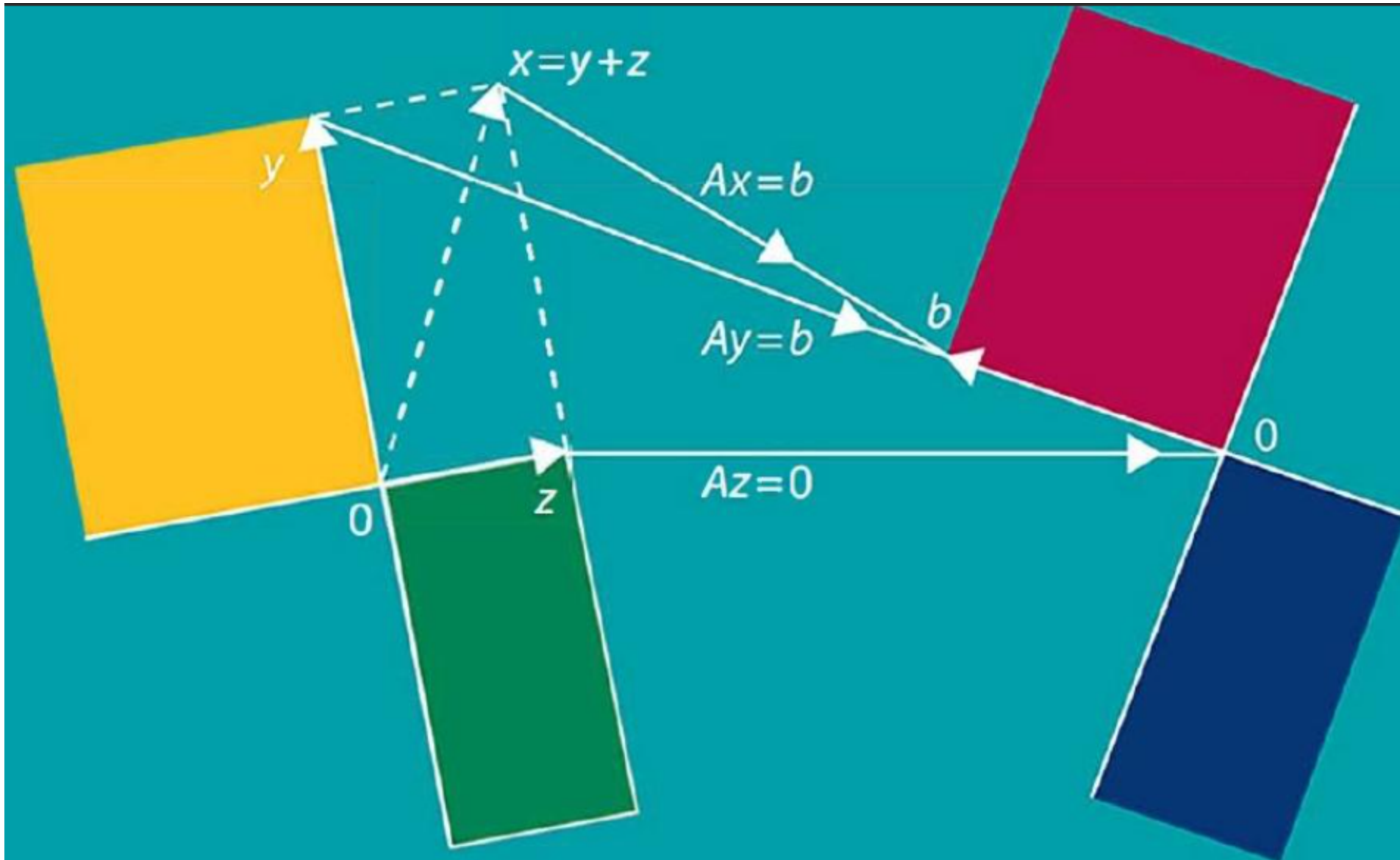


Figure 5: 四个子空间



Null space

For a linear system $A\mathbf{x} = \mathbf{b}$

We can write the solutions as

$$\mathbf{s} = \mathbf{s}_0 + c_1\mathbf{v}_1 + \cdots + c_k\mathbf{v}_k$$

where \mathbf{s}_0 is a particular solution, $\mathbf{v}_1, \cdots, \mathbf{v}_k$ is the basis of the null space of A

- \mathbf{s}_0 : 特解
- $Null(A) = span\{\mathbf{v}_1, \cdots, \mathbf{v}_k\}$
so $\mathbf{v}_1, \cdots, \mathbf{v}_k$ are the basis of the null space of A

Rank, Nullity

- $\text{rank}(A) = \dim(\text{row}(A)) = \dim(\text{col}(A))$
 A 的秩 = A 的行阶梯矩阵的首一个数
 $\Rightarrow \text{rank}(A) \leq \min(n, m)$
- $\text{rank}(A)$ 可看作行阶梯矩阵的首一(非零行/主元) 个数
 $\text{nullity}(A) = \dim(\text{Null}(A))$ 可看作自由元的个数
 $\Rightarrow \text{rank}(A) + \text{nullity}(A) = n$

rank property

- $A \in M_{m \times n}, W = \{A\mathbf{v} : \mathbf{v} \in \mathbb{R}^n\}$
then $W = \text{Col}(A)$
- $\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$
矩阵相乘秩不增
- $\text{rank}(A^T A) = \text{rank}(A)$

Equivalent expression

$$A \in M_{n \times n}$$

- A 可逆;
- $A\mathbf{x} = \mathbf{0}$ 只有平凡解;
- A 的简化阶梯型为单位矩阵;
- A 是一组初等矩阵的乘积;
- $A\mathbf{x} = \mathbf{b}$ 对任何 $n \times 1$ 的列向量 \mathbf{b} 都有解;
- $A\mathbf{x} = \mathbf{b}$ 对任何 $n \times 1$ 的列向量 \mathbf{b} 都有且只有一个解;
- $\det(A) \neq 0$;
- A 的所有 n 个行向量线性无关;
- A 的所有 n 个列向量线性无关;
- $\text{span}(\text{Row}(A)) = \mathbb{R}^n$;
- $\text{span}(\text{Col}(A)) = \mathbb{R}^n$;
- A 的所有 n 个行向量构成 \mathbb{R}^n 的一组基底;
- A 的所有 n 个列向量构成 \mathbb{R}^n 的一组基底;
- $\text{rank}(A) = n$;
- $\text{Null}(A) = \{0\}$.

Geometry review

- detail in tutorial 5

- 平面的一般式

$$Ax + By + Cz + D = 0$$

- 平面的法向量

$$\mathbf{n} = (A, B, C)$$

- 空间中一点 (x_0, y_0, z_0) 到平面的距离

$$d = \frac{|Ax_0 + By_0 + Cz_0 + D|}{\sqrt{A^2 + B^2 + C^2}}$$

- 过空间内一点 (x_0, y_0, z_0) , 法向量为 $\mathbf{v} = (a, b, c)$ 的平面的方程

$$a(x - x_0) + b(y - y_0) + c(z - z_0) = 0$$