

• Definition (Abstract Vectors)

A Vector Space over real numbers is a set S and

- ① addition is defined for any two elements in the set
- ② scalar multiplication is defined for a scalar in \mathbb{R} and any element in the set. *multiply a real number*
- ③ the set is closed under these two operations,

meaning that

- 1) the sum of two elements is still in the set S
- 2) scalar multiplication is still in the set S
- ④ elements in this set is called (abstract) vectors.

Example: the following are all (abstract) vector spaces

$$\mathbb{R} = \{\text{all real numbers}\}$$

elementary vectors

$$\mathbb{R}^2 = \left\{ \begin{bmatrix} x \\ y \end{bmatrix}, \forall x, y \in \mathbb{R} \right\}$$

$$\mathbb{R}^3 = \left\{ \begin{bmatrix} a \\ b \\ c \end{bmatrix}, \forall a, b, c \in \mathbb{R} \right\}$$

$$\mathbb{R}^4 = \left\{ \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}, \forall a, b, c, d \in \mathbb{R} \right\}$$

$$\mathbb{R}^{1 \times 3} = \{ [x \ y \ z] : \forall x, y, z \in \mathbb{R} \}$$

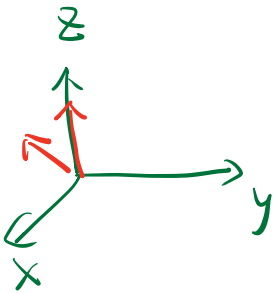
$$\mathbb{R}^{2 \times 2} = \left\{ \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \forall a, b, c, d \in \mathbb{R} \right\}$$

$\mathbb{R}^{m \times n}$

$$P_2(\mathbb{R}) = \{ ax^2 + bx + c : \forall a, b, c \in \mathbb{R} \}$$

S is a set of vectors, $\text{Span}(S)$ denotes the set of all possible linear combination of vectors in S .

Example: ① $\text{Span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\} = \left\{ a \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \forall a \in \mathbb{R} \right\}$ is a line.



② $\text{Span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\} = \left\{ a \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + b \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \forall a, b \in \mathbb{R} \right\}$

③ $\text{Span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\} = \mathbb{R}^3$

$$\begin{bmatrix} a \\ 0 \\ a+b \end{bmatrix}$$

Ex: $\text{Span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}$ is a Vector Space

• There is always a zero vector $\vec{0}$ s.t. $\vec{v} + \vec{0} = \vec{v}$

1) For \mathbb{R}^n , $\vec{0} = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$

2) For $\mathbb{R}^{m \times n}$, $\vec{0}$ is the zero matrix of size $m \times n$.

3) For $P_2(\mathbb{R})$, $\vec{0}$ is the zero polynomial $P(x) = 0$.

• Theorem: Let V be an abstract vector space.

1) $0 \cdot \vec{v} = \vec{0}, \forall \vec{v} \in V$

2) closedness $\Rightarrow \vec{0} = 0 \cdot \vec{v} \in V$.

- Definition: If V is a vector space, $W \subseteq V$,
↓
is a subset of
 and W is also a vector space, then W
 is called a subspace of V .

- Example: ① $\text{Span}\left\{\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}\right\} = \left\{a \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \forall a \in \mathbb{R}\right\}$ is
 a subspace of \mathbb{R}^3 .

② $\text{Span}\left\{\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}\right\}$ is a subspace of \mathbb{R}^3 .

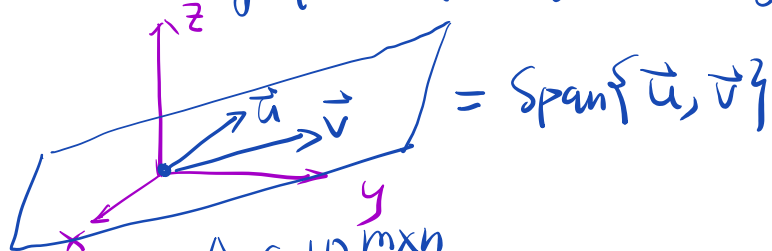
- Theorem: $\forall S \subseteq V$, $\text{Span}(S)$ is a subspace of V .

$\text{Span}(S) \subseteq V$?

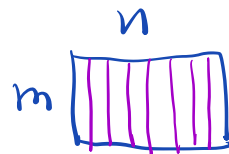
- If $\vec{0} \notin W$, then W is not a subspace.

Example: Any plane that does not pass the origin
 cannot be a subspace of \mathbb{R}^3 .

Ex: Is any plane passing the origin a subspace?



- Definition. $A \in \mathbb{R}^{m \times n}$

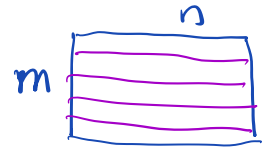


① $\text{Span}\{\text{all cols of } A\}$ is called the column
 space of A , denoted as $\text{Col}(A) \subseteq \mathbb{R}^m$.

Column Space of A

② Span { all rows of A } is called the row space of A , denoted as $\text{Row}(A) \subseteq \mathbb{R}^{1 \times n}$.

Row space of A



Example:
$$\begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -3 \\ -2 & -3 & 7 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2 \\ 8 \\ 10 \end{pmatrix}$$

$A \vec{x} = \vec{b}$

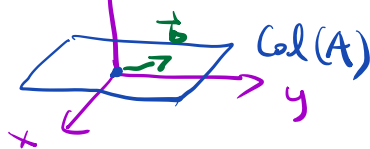
$$\text{Col}(A) = \left\{ a \begin{bmatrix} 2 \\ 4 \\ -2 \end{bmatrix} + b \begin{bmatrix} 4 \\ 9 \\ -3 \end{bmatrix} + c \begin{bmatrix} -2 \\ -3 \\ 7 \end{bmatrix}, \forall a, b, c \in \mathbb{R} \right\} \subseteq \mathbb{R}^3$$

$$\text{Row}(A) =$$

$$\left\{ a [2 \ 4 \ -2] + b [4 \ 9 \ -3] + c [-2 \ -3 \ 7], \forall a, b, c \in \mathbb{R} \right\} \subseteq \mathbb{R}^{1 \times 3}$$

- $A\vec{x} = \vec{b}$ has at least one sol if and only if $\vec{b} \in \text{Col}(A) \subseteq \mathbb{R}^3$

$$\begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -3 \\ -2 & -3 & 7 \end{pmatrix} \begin{pmatrix} x_0 \\ y_0 \\ z_0 \end{pmatrix} = \vec{b} \Leftrightarrow \vec{b} = x_0 \begin{pmatrix} 2 \\ 4 \\ -2 \end{pmatrix} + y_0 \begin{pmatrix} 4 \\ 9 \\ -3 \end{pmatrix} + z_0 \begin{pmatrix} -2 \\ -3 \\ 7 \end{pmatrix}$$



Proof: "if" Assume $\vec{b} \in \text{Col}(A)$, then there are $a_0, b_0, c_0 \in \mathbb{R}$ s.t.

$$\vec{b} = a_0 \begin{bmatrix} 2 \\ 4 \\ -2 \end{bmatrix} + b_0 \begin{bmatrix} 4 \\ 9 \\ -3 \end{bmatrix} + c_0 \begin{bmatrix} -2 \\ -3 \\ 7 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 2 & 4 & -2 \\ 4 & 9 & -3 \\ -2 & -3 & 7 \end{bmatrix} \begin{bmatrix} a_0 \\ b_0 \\ c_0 \end{bmatrix} = \vec{b}$$

$$\Rightarrow \begin{bmatrix} a_0 \\ b_0 \\ c_0 \end{bmatrix} \text{ is a sol to } A\vec{x} = \vec{b}$$

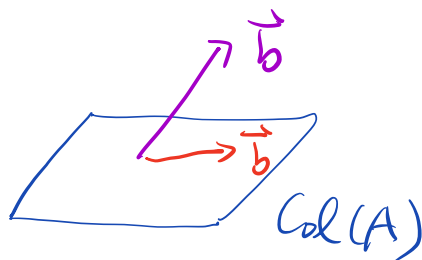
"only if" Assume $A\vec{x} = \vec{b}$ has one sol $\begin{bmatrix} x_0 \\ y_0 \\ z_0 \end{bmatrix}$

$$\Rightarrow \begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -3 \\ -2 & -3 & 7 \end{pmatrix} \begin{pmatrix} x_0 \\ y_0 \\ z_0 \end{pmatrix} = \begin{pmatrix} 2 \\ 8 \\ 10 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 2 \\ 8 \\ 10 \end{pmatrix} = x_0 \begin{pmatrix} 2 \\ 4 \\ -2 \end{pmatrix} + y_0 \begin{pmatrix} 4 \\ 9 \\ -3 \end{pmatrix} + z_0 \begin{pmatrix} -2 \\ -3 \\ 7 \end{pmatrix}$$

$\Rightarrow \vec{b}$ is spanned by cols of A .

- For $A \in \mathbb{R}^{3 \times 3}$, if $\text{Col}(A) = \mathbb{R}^3$, $A\vec{x} = \vec{b}$ always has at least one sol for any \vec{b} .



- Definition: all solutions to $A\vec{x} = \vec{0}$ form a subspace in \mathbb{R}^n , called null space of A , denoted as $\text{Null}(A)$.

$$A \in \mathbb{R}^{m \times n}$$

$$\vec{x} \in \mathbb{R}^n$$

Check closedness: $\forall \vec{u}, \vec{v} \in \text{Null}(A), a \in \mathbb{R}$

$$\begin{aligned} \vec{u} \in \text{Null}(A) &\Rightarrow A\vec{u} = \vec{0} \\ \vec{v} \in \text{Null}(A) &\Rightarrow A\vec{v} = \vec{0} \end{aligned} \Rightarrow A\vec{u} + A\vec{v} = \vec{0} + \vec{0} \\ &\Rightarrow A(\vec{u} + \vec{v}) = \vec{0} \\ &\Rightarrow \vec{u} + \vec{v} \in \text{Null}(A)$$

$$\begin{aligned} A\vec{u} = \vec{0} &\Rightarrow aA\vec{u} = a\vec{0} \Rightarrow A(a\vec{u}) = \vec{0} \\ &\Rightarrow a\vec{u} \in \text{Null}(A) \end{aligned}$$

• Example:

Matrix Form $A\vec{x} = \vec{0}$

$$\begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -3 \\ -2 & -3 & 7 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Augmented Matrix $[A | \vec{0}]$

$$\left(\begin{array}{ccc|c} 2 & 4 & -2 & 0 \\ 4 & 9 & -3 & 0 \\ -2 & -3 & 7 & 0 \end{array} \right)$$

RREF is

$$\left(\begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right)$$

$\Rightarrow \vec{0}$ is the only sol

$$\Rightarrow \text{Null}(A) = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\}$$

• Example: if RREF of $[A | \vec{0}]$ is

$$\left[\begin{array}{ccc|c} 1 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow y = t, \forall t \in \mathbb{R}$$

Solve it backwards

$$\textcircled{1} z = 0$$

$$\textcircled{2} x + 2y = 0 \Rightarrow x = -2t$$

$$\textcircled{3} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} -2t \\ t \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} -2t \\ t \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} + t \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix} = t \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}, \forall t \in \mathbb{R}$$

$$\Rightarrow \text{Null}(A) = \text{Span} \left\{ \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} \right\} = \left\{ t \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}, \forall t \in \mathbb{R} \right\}$$

• Definition (Linear Independence)

A set of (abstract) vectors $S = \{ \vec{v}_1, \dots, \vec{v}_n \}$ is called linearly dependent if there are scalars a_1, \dots, a_n which are not all zeros,

$$\text{sit. } a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_n \vec{v}_n = \vec{0}.$$

Otherwise, S is linearly independent.

Remark: As long as one of a_i is not zero, it satisfies the definition

Example: $\textcircled{1} S = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$ is linearly independent

$$a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{cases} a + b = 0 \\ 0 \cdot a + b = 0 \end{cases}$$

$$\Rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Augmented Matrix is $\left[\begin{array}{cc|c} 1 & 1 & 0 \\ 0 & 1 & 0 \end{array} \right]$

RREF is $\left[\begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \end{array} \right]$

$$\Rightarrow \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \text{only zero sol}$$

\Rightarrow linearly independent.

② If two vectors \vec{u}, \vec{v} are parallel in \mathbb{R}^3 ,
then $\{\vec{u}, \vec{v}\}$ is dependent

$$\vec{u} \parallel \vec{v} \Rightarrow \vec{u} = a\vec{v} \text{ for some } a \in \mathbb{R}$$

$$\Rightarrow \vec{u} - a\vec{v} = \vec{0}$$

If $a\vec{u} + b\vec{v} = \vec{0}$
at least one of a, b
is not 0,
assume $a \neq 0$,

$$a\vec{u} = -b\vec{v}$$

$$\vec{u} = -\frac{b}{a}\vec{v}$$

nation of columns of A .

2. (20 pts) For the invertible matrix

$$A = \begin{pmatrix} 0 & 1 & 1 \\ -1 & 3 & 1 \\ -1 & 2 & 1 \end{pmatrix},$$

find suitable elementary matrices so that A^{-1} can be written as a product of them.

$$A^{-1} = E_8 E_7 \dots E_1$$

3. (20 pts) Let $A = \begin{pmatrix} 0 & 2 & 4 \\ -2 & 3 & 1 \\ -4 & 4 & 2 \end{pmatrix}$.

- (a) Determine whether columns of A are linearly independent as follows: assume there are numbers a, b, c s.t.

$$a \begin{pmatrix} 0 \\ -2 \\ -4 \end{pmatrix} + b \begin{pmatrix} 2 \\ 3 \\ 4 \end{pmatrix} + c \begin{pmatrix} 4 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \quad \begin{cases} a \cdot 0 + b \cdot 2 + c \cdot 4 = 0 \\ a \cdot (-2) + b \cdot 3 + c \cdot 1 = 0 \end{cases}$$

which gives three equations about a, b, c . Solve the linear system about a, b, c . If there are nonzero solutions, then the column vectors are linearly dependent. Otherwise, they are linearly independent.

- (b) Determine whether rows of A are linearly independent as follows: assume there are numbers a, b, c s.t.

$$a(0 \ 2 \ 4) + b(-2 \ 3 \ 1) + c(-4 \ 4 \ 2) = (0 \ 0 \ 0) \quad \begin{cases} a \cdot 0 + b \cdot (-2) + c \cdot (-4) \\ a \cdot 2 + b \cdot 3 + c \cdot 4 = 0 \end{cases}$$

which gives three equations about a, b, c . Solve the linear system about a, b, c . If there are nonzero solutions, then the row vectors are linearly dependent. Otherwise, they are linearly independent.

4. (20 pts)

Definition 1 (Transpose matrix). For a matrix A of size $m \times n$, its transpose matrix A^T has size $n \times m$, and the j -th column of A^T is obtained by converting the j -th row of A to a column. For example,

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}, \quad A^T = \begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix}.$$

For a square matrix A , A^T can also be viewed as flipping non-diagonal entries with respect to the diagonal entries. For example:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}, \quad A^T = \begin{pmatrix} a & c \\ b & d \end{pmatrix}.$$