

# Lecture 19

188 200

Discrete Mathematics and Linear Algebra

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# Overview

## Topics for today.

- ▶ Subspace
- ▶ Subspace of spanned sets
- ▶ Column Space
- ▶ Null Space
- ▶ Basis for Subspace
- ▶ Coordinate system
- ▶ Dimension
- ▶ Rank

**Reference :** Section 2.8-2.9

## Subspace of $\mathbb{R}^n$

Focus on sets of vectors in  $\mathbb{R}^n$  that is called **subspaces**.

**Definition:** A **subspace** of  $\mathbb{R}^n$  is any set  $H$  in  $\mathbb{R}^n$  that has three properties:

1. The zero vector is in  $H$ .
2. For each  $\vec{u}$  and  $\vec{v}$  in  $H$ , the sum  $\vec{u} + \vec{v}$  is in  $H$ .
3. For each  $\vec{u}$  in  $H$  and each scalar  $c$ ,  $c\vec{u}$  is in  $H$ .

**Note:** This means a subspace is **closed** under addition and scalar multiplication.

**Example 1:** If  $\vec{v}_1$  and  $\vec{v}_2$  are in  $\mathbb{R}^n$  and  $H = \text{Span}\{\vec{v}_1, \vec{v}_2\}$ , then  $H$  is a subspace of  $\mathbb{R}^n$ . Verify the statement.

**Solution:** Check if  $H$  has the three properties.

► Is  $\vec{0}$  in  $H$ ?

• Yes, because  $0\vec{v}_1 + 0\vec{v}_2$ .

► For all  $\vec{u}$  and  $\vec{v}$  in  $H$ , is  $\vec{u} + \vec{v}$  in  $H$ ?

• For any  $\vec{u}$  and  $\vec{v}$  in  $H$  let  $\vec{u} = s_1\vec{v}_1 + s_2\vec{v}_2$  and  $\vec{v} = t_1\vec{v}_1 + t_2\vec{v}_2$ .

• Hence we have  $\vec{u} + \vec{v} = (s_1 + t_1)\vec{v}_1 + (s_2 + t_2)\vec{v}_2$ . So  $\vec{u} + \vec{v}$  is in  $H$ .

► For any scalar  $c$ , is  $c\vec{u}$  in  $H$ ?

•  $c\vec{u} = c(s_1\vec{v}_1 + s_2\vec{v}_2) = (cs_1)\vec{v}_1 + (cs_2)\vec{v}_2$  which is in  $H$ .

## Subspace of Spanned Sets

- ▶ If  $\vec{v}_1, \dots, \vec{v}_p$  are in  $\mathbb{R}^n$ , the set of all linear combinations of  $\vec{v}_1, \dots, \vec{v}_p$  is a subspace of  $\mathbb{R}^n$ .
  - The set of all linear combination of  $\vec{v}_1, \dots, \vec{v}_p$  is  $\text{Span}\{\vec{v}_1, \dots, \vec{v}_p\}$ .
    - From Example 1,  $\text{Span}\{\vec{v}_1, \dots, \vec{v}_p\}$  is a subspace of  $\mathbb{R}^n$ .
    - Henceforth, we shall now refer to  $\text{Span}\{\vec{v}_1, \dots, \vec{v}_p\}$  as the **subspace spanned** (or **generated**) by  $\vec{v}_1, \dots, \vec{v}_p$ .
- ▶  $\mathbb{R}^n$  is a **subspace of itself**.
- ▶ The set consist of only **zero vector** is also a subspace in  $\mathbb{R}^n$ .
  - This set is called the **zero subspace**.

## Column Space

**Definition:** The **column space** of matrix  $A$  is the set  $\text{Col } A$  of all linear combinations of the columns of  $A$ .

**Example 2:** Let  $A = \begin{bmatrix} 1 & -3 & -4 \\ -4 & 6 & -2 \\ -3 & 7 & 6 \end{bmatrix}$  and  $\vec{b} = \begin{bmatrix} 3 \\ 3 \\ -4 \end{bmatrix}$ .

Determine if  $\vec{b}$  is in  $\text{Col } A$ .

**Solution:**

**Note:**

- ▶  $\text{Col } A$  of an  $m \times n$  matrix is a subspace of  $\mathbb{R}^m$ .
  - Since if  $A = [\vec{a}_1 \dots \vec{a}_n]$ ,  $\text{Col } A = \text{Span}\{\vec{a}_1, \dots, \vec{a}_n\}$ .
- ▶ Example 2 shows that  $\text{Col } A$  in  $A\vec{x} = \vec{b}$  is the set of all  $\vec{b}$  for which **the system has a solution**.

# Null Space

**Definition:** The **null space** of a matrix  $A$  is the set  $\text{Nul } A$  of all solutions to  $A\vec{x} = \vec{0}$ .

**Theorem 12:**  $\text{Nul } A$  of an  $m \times n$  matrix  $A$  is a **subspace of  $\mathbb{R}^n$** . Equivalently, the set of **all the solutions** to  $A\vec{x} = \vec{0}$  of  $m$  homogeneous linear equations in  $n$  unknown is a **subspace of  $\mathbb{R}^n$** .

## Note:

- ▶ To check if any  $\vec{v}$  is in  $\text{Nul } A$ , compute  $A\vec{v}$  to see if it is  $\vec{0}$ .
- ▶  $\text{Nul } A$  is defined **implicitly** since  $\text{Nul } A$  is defined by a condition that must be checked for each  $\vec{v}$ .
- ▶  $\text{Col } A$  is defined **explicitly** since  $\text{Col } A$  can be constructed from the columns of  $A$ .

**Example 3:** Let  $A = \begin{bmatrix} 1 & -1 & 5 \\ 2 & 0 & 7 \\ -3 & -5 & -3 \end{bmatrix}$  and  $\vec{b} = \begin{bmatrix} -7 \\ 3 \\ 2 \end{bmatrix}$ .

Determine if  $\vec{b}$  is in  $\text{Nul } A$ .

**Solution:**

## Basis For a Subspace

Instead of looking at the whole subspace, it is usually better to work with some subsets of subspace. We consider **the smallest set of vectors that spans subspace** that is called **basis**.

**Definition:** A **basis** for a subspace  $H$  of  $\mathbb{R}^n$  is a **linearly independent set** in  $H$  that spans  $H$ .

**Example 4:** Find a basis for  $\mathbb{R}^n$ .

**Solution:**

## How to Identify a Basis

**Example 5:** Find a **basis** for the **null space** of the matrix

$$A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}$$

**Solution:**

## Finding a Basis For Col $A$

**Example 6:** Find a **basis** for **Col  $A$**  of the matrix

$$A = \begin{bmatrix} 1 & 0 & -3 & 5 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

**Solution:**

## Linear Dependence Relationships

The columns of  $A$  have **exactly the same linear dependence relationships** as the columns of  $B$  where  $B$  is row reduced from  $A$ .

- ▶ The linear dependence relation among the columns of  $A$  can be express as the form  $A\vec{x} = \vec{0}$  for some  $\vec{x}$ .

**Example 7:** It can be verified that the matrix

$$B = \begin{bmatrix} 1 & 3 & 3 & 2 & -9 \\ -2 & -2 & 2 & -8 & 2 \\ 2 & 3 & 0 & 7 & 1 \\ 3 & 4 & -1 & 11 & -8 \end{bmatrix}$$

is row equivalent to the matrix  $A$  in Example 5. **Find a basis for Col  $A$ .**

From the previous two examples, we obtain the following theorem.

**Theorem 13:** The **pivot columns** of a matrix  $A$  form a basis for the column space of  $A$ .

**Note:** To use theorem 13, make sure to **use the pivot columns of  $A$  itself** for the basis of  $\text{Col } A$ . This implies that you should not use the pivot columns of an echelon form matrix of  $A$  since they are not often in  $\text{Col } A$ .

## Nul $A$ VS Col $A$

1. Nul  $A$  is a subspace of  $\mathbb{R}^n$ .
2. Nul  $A$  is implicitly defined.
3. Take time to find vectors in Nul  $A$ . Row operations on  $[A \ \vec{0}]$  are required.
4. There is no obvious relation between Nul  $A$  and the entries in  $A$ .
5. A typical vector  $\vec{v}$  in Nul  $A$  has the property that  $A\vec{v} = \vec{0}$ .
6. Easy to check if  $\vec{v}$  is in Nul  $A$ . Compute  $A\vec{v}$ .

1. Col  $A$  is a subspace of  $\mathbb{R}^n$ .
2. Col  $A$  is explicitly defined.
3. Easy to find. The columns of  $A$  are displayed; others are formed from them.
4. There is an obvious relation between Col  $A$  and the entries in  $A$ , since each column of  $A$  is in Col  $A$ .
5. A typical vector  $\vec{v}$  in Col  $A$  has the property that  $A\vec{x} = \vec{v}$  is consistent.
6. Harder to check, compute  $[A \ \vec{v}]$ .

## Nul $A$ VS Col $A$

7.  $\text{Nul } A = \{\vec{0}\}$  iff  $A\vec{x} = \vec{0}$  has only trivial solution.
8.  $\text{Nul } A = \{\vec{0}\}$  iff  $x \mapsto Ax$  is one-to-one.

7.  $\text{Col } A = \mathbb{R}^m$  iff  $A\vec{x} = \vec{b}$  has a solution for every  $\vec{b}$ .
8.  $\text{Col } A = \mathbb{R}^m$  iff  $x \mapsto Ax$  maps  $\mathbb{R}^n$  onto  $\mathbb{R}^m$ .

## Coordinate Systems

We select a basis instead of subspace for  $H$  because each vector in  $H$  can be written in **only one way** as a linear combination of the basis.

**Definition:** Suppose  $\mathcal{B} = \{\vec{b}_1, \dots, \vec{b}_p\}$  is a basis of a subspace  $H$ . For each  $\vec{x}$  in  $H$ , the **coordinates of  $\vec{x}$  relative to  $\mathcal{B}$**  are the weight  $c_1, \dots, c_p$  s.t.  $\vec{x} = c_1\vec{b}_1 + \dots + c_p\vec{b}_p$ , and the vector in  $\mathbb{R}^p$

$$[\vec{x}]_{\mathcal{B}} = \begin{bmatrix} c_1 \\ \vdots \\ c_p \end{bmatrix}$$

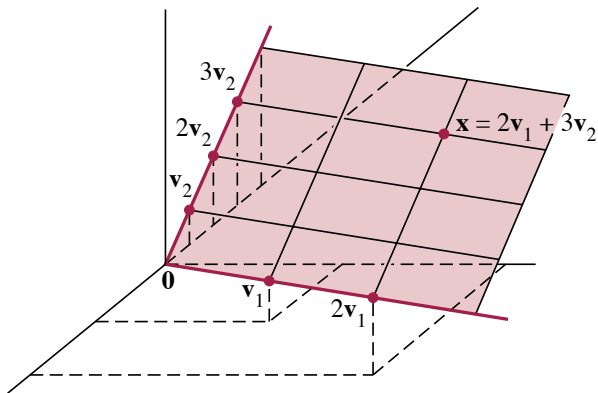
is called the **coordinate vector of  $\vec{x}$  (relative to  $\mathcal{B}$ )** or the  **$\mathcal{B}$ -coordinate vector of  $\vec{x}$** .

**Example 8:** Let  $\vec{v}_1 = \begin{bmatrix} 3 \\ 6 \\ 2 \end{bmatrix}$ ,  $\vec{v}_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ ,  $\vec{x} = \begin{bmatrix} 3 \\ 12 \\ 7 \end{bmatrix}$  and

$\mathcal{B} = \{\vec{v}_1, \vec{v}_2\}$ . Then  $\mathcal{B}$  is a basis for  $H = \text{Span}\{\vec{v}_1, \vec{v}_2\}$  because  $\vec{v}_1$  and  $\vec{v}_2$  are linearly independent. Determine if  $\vec{x}$  is in  $H$ , and if it is, find the coordinate vector of  $\vec{x}$  relative to  $\mathcal{B}$ .

**Solution:**

The figure shows a coordinate system on a plane  $H$  in  $\mathbb{R}^3$



- ▶ Notice although  $H$  is in  $\mathbb{R}^3$  they are completely determined by their coordinate which are in  $\mathbb{R}^2$ .
- ▶ The grid on the plane makes  $H$  look like  $\mathbb{R}^2$ .

- ▶ The correspondence  $\vec{x} \mapsto [\vec{x}]_B$  is a **one-to-one** correspondence between  $H$  and  $\mathbb{R}^2$  that preserve linear combination.
- ▶ We call such a correspondence an **isomorphism**, and we say  $H$  is **isomorphic to**  $\mathbb{R}^2$ .
- ▶ In general, if  $B = \{\vec{b}_1, \dots, \vec{b}_p\}$  is a basis for  $H$ , then  $\vec{x} \mapsto [\vec{x}]_B$  is a one-to-one correspondence that makes  $H$  look and act the same as  $\mathbb{R}^p$ , even though  $H$  may have more than  $p$  entries.

# The Dimension of a Subspace

**Definition:** The **dimension** of a nonzero subspace  $H$ , denoted by  $\dim H$ , is the **number of vectors** in **any basis** for  $H$ . The dimension of the zero subspace  $\{\vec{0}\}$  is defined to be zero.

## Note:

- ▶ The zero subspace has no basis because  $\{\vec{0}\}$  is a linearly dependent set.
- ▶ The space  $\mathbb{R}^n$  has dimension  $n \rightarrow$  every basis for  $\mathbb{R}^n$  has  $n$  vectors.
- ▶ A plane through  $\vec{0}$  in  $\mathbb{R}^3$  is two dimensional.
- ▶ A line through  $\vec{0}$  is one dimensional.

## Rank of a Matrix

**Definition:** The **rank** of a matrix  $A$ , denoted by **rank**  $A$ , is the **dimension** of the **column space** of  $A$ .

**Note:** Recall that the pivot columns of  $A$  form a basis for  $\text{Col } A$ .

**Example 9:** Determine the rank of the matrix

$$A = \begin{bmatrix} 2 & 5 & -3 & -4 & 8 \\ 4 & 7 & -4 & -3 & 9 \\ 6 & 9 & -5 & 2 & 4 \\ 0 & -9 & 6 & 5 & -6 \end{bmatrix}$$

**Solution:**

# Theorems

**Theorem: (The Rank Theorem)** If a matrix  $A$  has  $n$  columns, then  $\text{rank } A + \dim \text{Nul } A = n$ .

**Theorem (The Basis Theorem)** Let  $H$  be a  $p$ -dimensional subspace of  $\mathbb{R}^n$ . Any linearly independent set of exactly  $p$  elements in  $H$  is automatically a basis for  $H$ . Also, any set of  $p$  elements of  $H$  that spans  $H$  is automatically a basis for  $H$ .

## IMT (continued)

**Theorem:** The Invertible Matrix Theorem (continued)

Let  $A$  be an  $n \times n$  matrix. Then the following statements are each equivalent to the statement that  $A$  is an invertible matrix.

a.  $A$  is an invertible matrix.

⋮

⋮

m. The columns of  $A$  form a basis of  $\mathbb{R}^n$ .

n.  $\text{Col } A = \mathbb{R}^n$

o.  $\dim \text{Col } A = n$

p.  $\text{rank } A = n$

q.  $\text{Nul } A = \{\vec{0}\}$

r.  $\dim \text{Nul } A = 0$

## Recap

- ▶ Subspace
- ▶ Subspace of spanned sets
- ▶ Column Space
- ▶ Null Space
- ▶ Basis for Subspace
- ▶ Coordinate system
- ▶ Dimension
- ▶ Rank

Next time, we will start start Chapter 5 with Eigenvectors and Eigenvalues.