# LECTURE-4 Basis and Dimension

# **Basis**

Spanning

Sets

Bases

# Basis :

V: a vector space

$$S = \{\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n\} \subseteq V$$

1) S spans V (i.e., span(S) = V)



 $\Rightarrow$  S is called a basis for V

# Notes:

A basis S must have enough vectors to span V, but not so many vectors that one of them could be written as a linear combination of the other vectors in S

Linearly

Independent

Sets

### ■ Notes:

(1) the **standard basis** for  $R^3$ :

$$\{i, j, k\}$$
, for  $i = (1, 0, 0), j = (0, 1, 0), k = (0, 0, 1)$ 

(2) the **standard basis** for  $R^n$ :

$$\{\mathbf{e}_1, \mathbf{e}_2, ..., \mathbf{e}_n\}$$
, for  $\mathbf{e}_1 = (1,0,...,0)$ ,  $\mathbf{e}_2 = (0,1,...,0)$ , ...,  $\mathbf{e}_n = (0,0,...,1)$ 

Ex: For  $\mathbb{R}^4$ , {(1,0,0,0), (0,1,0,0), (0,0,1,0), (0,0,0,1)}

RExpress any vector in  $R^n$  as the linear combination of the vectors in the standard basis: the coefficient for each vector in the standard basis is the value of the corresponding component of the examined vector,

e.g., (1, 3, 2) can be expressed as  $1 \cdot (1, 0, 0) + 3 \cdot (0, 1, 0) + 2 \cdot (0, 0, 1)$ 

# (3) the **standard basis** for $m \times n$ matrix space:

$$\{E_{ij} \mid 1 \le i \le m, 1 \le j \le n\}$$
, and in  $E_{ij}$  
$$\begin{cases} a_{ij} = 1 \\ \text{other entries are zero} \end{cases}$$

Ex:  $2 \times 2$  matrix space:

$$\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$$

(4) the **standard basis** for  $P_n(x)$ :

$$\{1, x, x^2, ..., x^n\}$$

Ex: 
$$P_3(x) = \{1, x, x^2, x^3\}$$

• Ex 4.1: The nonstandard basis for  $R^2$ 

Show that  $S = \{\mathbf{v}_1, \mathbf{v}_2\} = \{(1, 1), (1, -1)\}$  is a basis for  $\mathbb{R}^2$ 

(1) For any 
$$\mathbf{u} = (u_1, u_2) \in \mathbb{R}^2$$
,  $c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 = \mathbf{u} \implies \begin{cases} c_1 + c_2 = u_1 \\ c_1 - c_2 = u_2 \end{cases}$ 

Because the coefficient matrix of this system has a **nonzero determinant**, the system has a unique solution for each  $\mathbf{u}$ . Thus you can conclude that S spans  $R^2$ 

(2) For 
$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 = \mathbf{0} \implies \begin{cases} c_1 + c_2 = 0 \\ c_1 - c_2 = 0 \end{cases}$$

Because the coefficient matrix of this system has a **nonzero determinant**, you know that the system has only the trivial solution. Thus you can conclude that *S* is linearly independent

• Theorem 4.1: Uniqueness of basis representation for any vectors If  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a basis for a vector space V, then every vector in V can be written in one and only one way as a linear combination of vectors in S

Pf:

∴ S is a basis ⇒ 
$$\begin{cases} (1) \operatorname{span}(S) = V \\ (2) S \text{ is linearly independent} \end{cases}$$
∴ span(S) = V Let  $\mathbf{v} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_n \mathbf{v}_n$ 

$$\mathbf{v} = b_1 \mathbf{v}_1 + b_2 \mathbf{v}_2 + \dots + b_n \mathbf{v}_n$$
⇒  $\mathbf{v} + (-1)\mathbf{v} = \mathbf{0} = (c_1 - b_1)\mathbf{v}_1 + (c_2 - b_2)\mathbf{v}_2 + \dots + (c_n - b_n)\mathbf{v}_n$ 
∴ S is linearly independent ⇒ with only trivial solution
$$\Rightarrow \operatorname{coefficients} \text{ for } \mathbf{v}_i \text{ are all zero}$$
⇒  $c_1 = b_1$ ,  $c_2 = b_2$ ,...,  $c_n = b_n$  (i.e., unique basis representation).

Theorem 4.2: Bases and linear dependence

If  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a basis for a vector space V, then every set containing more than n vectors in V is linearly dependent (In other words, every linearly independent set contains at most n vectors)

Pf:

Let  $S_1 = \{\mathbf{u}_1, \mathbf{u}_2, ..., \mathbf{u}_m\}$ , be another set such that m > n, we want to show that  $S_1$  is linearly dependent.

If 
$$V = \text{span}(S_1)$$

$$\mathbf{u}_1 = c_{11}\mathbf{v}_1 + c_{21}\mathbf{v}_2 + \dots + c_{n1}\mathbf{v}_n$$
 And 
$$\mathbf{u}_i \in V \Longrightarrow \quad \mathbf{u}_2 = c_{12}\mathbf{v}_1 + c_{22}\mathbf{v}_2 + \dots + c_{n2}\mathbf{v}_n$$
 :

$$\mathbf{u}_m = c_{1m} \mathbf{v}_1 + c_{2m} \mathbf{v}_2 + \dots + c_{nm} \mathbf{v}_n$$

Consider  $k_1\mathbf{u}_1 + k_2\mathbf{u}_2 + \dots + k_m\mathbf{u}_m = \mathbf{0}$  (For L.I) (if  $k_i$ 's are not all zero,  $S_1$  is linearly dependent)

$$\Rightarrow d_1 \mathbf{v}_1 + d_2 \mathbf{v}_2 + \dots + d_n \mathbf{v}_n = \mathbf{0} \ (d_i = c_{i1} k_1 + c_{i2} k_2 + \dots + c_{im} k_m)$$

:. 
$$S ext{ is L.I.} \Rightarrow d_i = 0 \quad \forall i$$
 i.e.,  $c_{11}k_1 + c_{12}k_2 + \dots + c_{1m}k_m = 0$  
$$c_{21}k_1 + c_{22}k_2 + \dots + c_{2m}k_m = 0$$
 
$$\vdots$$
 
$$c_{n1}k_1 + c_{n2}k_2 + \dots + c_{nm}k_m = 0$$

- : If the homogeneous system has fewer equations (n equations) than variables ( $k_1, k_2, ..., k_m$ ), then it must have infinitely many solutions
- $\therefore m > n \Rightarrow k_1 \mathbf{u}_1 + k_2 \mathbf{u}_2 + \dots + k_m \mathbf{u}_m = \mathbf{0}$  has nontrivial (nonzero) solution

$$\Rightarrow$$
  $S_1$  is linearly dependent

- Theorem 4.3: Number of vectors in a basis
   If a vector space V has one basis with n vectors, then every basis for V has n vectors
- Pf:  $\times$  According to Thm. 4.2, every linearly independent set contains at most n vectors of a vector space if there is a basis of n vectors spanning that vector space

$$S = \{\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n\}$$

$$S' = \{\mathbf{u}_1, \mathbf{u}_2, ..., \mathbf{u}_m\}$$
 are two bases with different sizes for  $V$ 

 $\Re$  For  $R^3$ , since the standard basis  $\{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$  can span this vector space, you can infer any basis that can span  $R^3$  must have exactly 3 vectors

# **Dimension**

# Dimension:

The dimension of a vector space V is defined to be the number of vectors in a basis for V

V: a vector space S: a basis for V

 $\Rightarrow$  dim(V) = #(S) (the number of vectors in a basis S)

# • Finite dimensional:

A vector space V is finite dimensional if it has a basis consisting of a finite number of elements

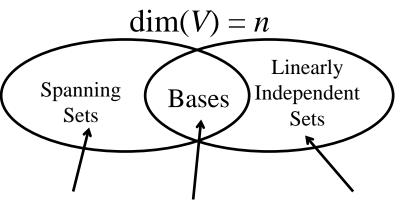
# • Infinite dimensional:

If a vector space V is not finite dimensional, then it is called infinite dimensional

# Notes:

 $(1) \dim(\{\mathbf{0}\}) = 0$ 

(If *V* consists of the zero vector alone, the dimension of *V* is defined as zero)



(2) Given 
$$\dim(V) = n$$
, for  $S \subseteq V \#(S) \ge n \#(S) = n \#(S) \le n$ 

S: a spanning set  $\Rightarrow$  #(S)  $\geq$  n (Ex 3.2 on Slides 6 and 7 previous lec.)

S: a L.I. set  $\Rightarrow$  #(S)  $\leq$  n (from Theorem 4.2)

S: a basis  $\Rightarrow \#(S) = n$  (Since a basis is defined to be a set of L.I.  $\Rightarrow \#(S) = n$  vectors that can spans V,  $\#(S) = \dim(V) = n$ ) (see the above figure)

- (3) Given  $\dim(V) = n$ , if W is a subspace of  $V \Rightarrow \dim(W) \le n$ 
  - X For example, if  $V = R^3$ , you can infer the dim(V) is 3, which is the number of vectors in the standard basis
  - X Considering  $W = R^2$ , which is a subspace of  $R^3$ , due to the number of vectors in the standard basis, we know that the dim(W) is 2, that is smaller than dim(V)=3

- Ex4.2: Find the dimension of a vector space according to the standard basis
  - \* The simplest way to find the dimension of a vector space is to count the number of vectors in the standard basis for that vector space
    - (1) Vector space  $R^n$   $\Rightarrow$  standard basis  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$   $\Rightarrow \dim(R^n) = n$
    - (2) Vector space  $M_{m \times n} \implies \text{standard basis } \{E_{ij} \mid 1 \le i \le m, 1 \le j \le n\}$ and in  $E_{ij}$   $\begin{cases} a_{ij} = 1 \\ \text{other entries are zero} \end{cases}$   $\implies \dim(M_{m \times n}) = mn$
    - (3) Vector space  $P_n(x) \Rightarrow \text{standard basis } \{1, x, x^2, \dots, x^n\}$  $\Rightarrow \dim(P_n(x)) = n+1$
    - (4) Vector space  $P(x) \implies$  standard basis  $\{1, x, x^2, ...\}$

- Ex 4.3: Determining the dimension of a subspace of  $R^3$ 
  - (a)  $W = \{(d, c-d, c): c \text{ and } d \text{ are real numbers}\}$
  - (b)  $W = \{(2b, b, 0): b \text{ is a real number}\}$

Sol: (Hint: find a set of L.I. vectors that spans the subspace, i.e., find a basis for the subspace.)

- (a) (d, c-d, c) = c(0, 1, 1) + d(1, -1, 0)  $\Rightarrow S = \{(0, 1, 1), (1, -1, 0)\} \text{ (S is L.I. and S spans W)}$   $\Rightarrow S \text{ is a basis for } W$  $\Rightarrow \dim(W) = \#(S) = 2$
- (b) :: (2b,b,0) = b(2,1,0)  $\Rightarrow S = \{(2, 1, 0)\} \text{ spans } W \text{ and } S \text{ is L.I.}$   $\Rightarrow S \text{ is a basis for } W$  $\Rightarrow \dim(W) = \#(S) = 1$

• Ex 4.4: Finding the dimension of a subspace of  $M_{2\times2}$ Let W be the subspace of all symmetric matrices in  $M_{2\times2}$ . What is the dimension of W?

Sol:

$$W = \left\{ \begin{bmatrix} a & b \\ b & c \end{bmatrix} | a, b, c \in R \right\}$$

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

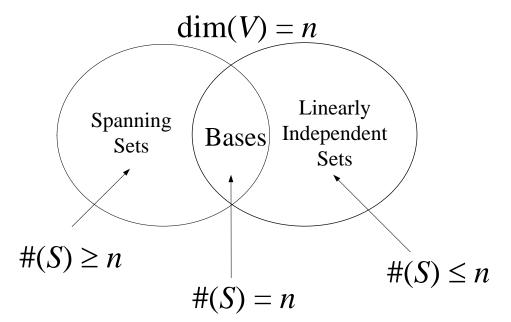
$$\Rightarrow S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \text{ spans } W \text{ and } S \text{ is L.I.}$$

$$\Rightarrow$$
 S is a basis for  $W \Rightarrow \dim(W) = \#(S) = 3$ 

■ Theorem 4.4: Methods to identify a basis in an *n*-dimensional space

Let V be a vector space of dimension n

- (1) If  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a linearly independent set of vectors in V, then S is a basis for V
- (2) If  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  spans V, then S is a basis for V (Both results are due to the fact that #(S) = n)



Solve Problems