

From last time:

- 1) Any spanning set which is linearly dependent can be reduced (without changing the span), by removing a vector which is in the span of the rest.
- 2) Any linearly independent set in  $W$ , which does not span  $W$  can be made into a larger linearly independent set in  $W$ , by throwing in a vector which is not in the span of the set.

Example: the set  $\{(1, 2, 1, 1), (1, 3, 5, 6)\}$  is LI. let's find a bigger

LI containing them:

$$\begin{aligned} \text{span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 3 \\ 5 \\ 6 \end{pmatrix} \right\} &= \left\{ a \begin{pmatrix} 1 \\ 2 \\ 1 \\ 1 \end{pmatrix} + b \begin{pmatrix} 1 \\ 3 \\ 5 \\ 6 \end{pmatrix} \mid a, b \in \mathbb{R} \right\} \\ &= \left\{ \begin{pmatrix} a+b \\ 2a+3b \\ a+5b \\ a+6b \end{pmatrix} \mid a, b \in \mathbb{R} \right\} \end{aligned}$$

we have to choose a vector  $\vec{v} = (x_1, x_2, x_3, x_4)$ , which does not belong to this spanning set. If we try  $(1, 0, 0, 0)$  for instance, we see that  $\{(1, 2, 1, 1), (1, 3, 5, 6), (1, 0, 0, 0)\}$  is LI

Example:  $\{(1, 0), (0, 1)\}$  is LI &  $\text{span} \{(1, 0), (0, 1)\} = \mathbb{R}^2$ , so, there is no  $\vec{v} \in \mathbb{R}^2$ , such that  $\{(1, 0), (0, 1), \vec{v}\}$  be LI.

Fact: Any set of 3 or more vectors in  $\mathbb{R}^2$  is LD.

Note: Any LI set in  $\mathbb{R}^2$  has at most two vectors, and any spanning set of  $\mathbb{R}^2$  has at least two vectors. Therefore, every linearly independent spanning set has exactly two vectors!

So far:

If  $S$  is a spanning set, then any bigger set containing  $S$  has to be LD. And if instead  $S$  is a linearly independent set, then no proper subset could span the whole space.

\*Theorem (9.1): (LI sets are never bigger than spanning sets).

If a vector space  $V$  can be spanned by  $n$  vectors, then any linearly independent subset has at most  $n$  vectors.

or,  
if  $V$  has a subset of  $m$  linearly independent vectors, then any spanning set has at least  $m$  vectors.

or,  
the size of any linearly independent set in  $V \leq$  the size of any spanning set of  $V$ ,

or,  
suppose a vector space  $V$  can be spanned by  $n$  vectors. If any set of  $m$  vectors in  $V$  is linearly independent, then  $m \leq n$ .

Examples :

(1)  $\mathbb{R}^3$  is spanned by three vectors, (e.g.  $(1,0,0), (0,1,0), (0,0,1)$ ).

So, any set in  $\mathbb{R}^3$  with 4 or more vectors is LD.

(2)  $M_{2 \times 2}(\mathbb{R})$  is spanned by 4 vectors. So, any set of 5 or more  $2 \times 2$  matrices is LD.

(3) The set of  $2 \times 2$  diagonal matrices is spanned by two vectors, so any set of 3 or more  $2 \times 2$  diagonal matrices is LD.

⊗ The theorem talks about the maximum number of linearly independent vectors in any vector space, including subspaces.

\* Basis of a vector space :

Definition: A set  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  of vectors in  $V$  is called a "basis" of  $V$  if:

(1)  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  is linearly independent, and

(2)  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$  spans  $V$ .

Equivalently,

- A "basis" is a linearly independent spanning set of  $V$ .
- A "basis" is a biggest possible linearly independent set in  $V$ .
- A "basis" is a smallest possible spanning set of  $V$ .

Examples :

(1)  $\{(1,0), (0,1)\}$  is a basis of  $\mathbb{R}^2$ .

(2)  $\{1, x, x^2\}$  is a basis of  ~~$\mathbb{R}$~~   $\mathbb{P}_2$ .

(55)

(3)  $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$  is a basis of  $M_{2 \times 2}(\mathbb{R})$ .

\* Theorem (9.2): (All bases have the same size).

If  $\{\vec{v}_1, \dots, \vec{v}_m\}$  and  $\{\vec{w}_1, \dots, \vec{w}_k\}$  are two bases of a vector space  $V$ , then  $m = k$  (All bases have the same number of vectors).

Proof: Since  $\{\vec{v}_1, \dots, \vec{v}_m\}$  spans  $V$  and  $\{\vec{w}_1, \dots, \vec{w}_k\}$  is LI, then  $m \geq k$ . Similarly  $k \geq m$ , then  $m = k$ .  $\square$

Dimension of a vector space:

Definition: If  $V$  has a finite basis  $\{\vec{v}_1, \dots, \vec{v}_n\}$ , then the "dimension" of  $V$  is  $n$ , the number of vectors in that basis.

$$\boxed{\dim(V) = n}$$

we also say that  $V$  is finite-dimensional. If  $V$  doesn't have a finite basis, then  $V$  is infinite-dimensional.

Examples:

(1)  $\dim(\mathbb{R}^2) = 2,$

(2)  $\dim(\mathbb{P}_2) = 3,$

(3)  $\dim(M_{2 \times 2}(\mathbb{R})) = 4,$

(4) The set  $\{(1, 0, 0, \dots, 0), (0, 1, 0, \dots, 0), (0, 0, 1, \dots, 0), \dots, (0, 0, \dots, 1)\}$  of vectors in  $\mathbb{R}^n$  is linearly independent and spans  $\mathbb{R}^n$  and so  $\dim(\mathbb{R}^n) = n$ . (56)

(5) The set  $\{1, x, x^2, \dots, x^n\}$  is linearly independent and spans  $\mathbb{P}_n$  and so  $\dim(\mathbb{P}_n) = n+1$ .

(6) Consider the set of  $m \times n$  matrices  $\{E_{ij} \mid 1 \leq i \leq m, 1 \leq j \leq n\}$ , where  $E_{ij}$  is the matrix with zeros everywhere except for a 1 at the  $(i, j)$ th position. These are linearly independent, since

$$\sum_{i,j} a_{ij} E_{ij}$$

is the matrix with  $a_{ij}$  at entry  $(i, j)$ . The dependence relation has only the trivial solution. They span  $M_{m \times n}(\mathbb{R})$  and so this set is a basis of  $M_{m \times n}(\mathbb{R})$ . Therefore,

$$\dim(M_{m \times n}(\mathbb{R})) = m \times n.$$

(7) The vector space  $\mathbb{P}$  of all polynomial functions is infinite-dimensional since for any  $n$  the set  $\{1, x, x^2, \dots, x^n\}$  is linearly independent and because a basis must be larger than any linearly independent sets, we cannot find a ~~small~~ finite basis for  $\mathbb{P}$ .

(8) The vector space  $F(\mathbb{R})$  is also infinite-dimensional.

(9) consider  $L = \{ A \in M_{2 \times 2}(\mathbb{R}) \mid \text{Tr}(A) = 0 \}$  ~~( $\mathbb{R}$  is a field)~~ (57)

$$L = \left\{ \begin{pmatrix} a & b \\ c & -a \end{pmatrix} \mid a, b, c \in \mathbb{R} \right\} = \text{Span} \left\{ \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \right\}$$

$$\Rightarrow a \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \Leftrightarrow a = b = c = 0.$$

So,  $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} \right\}$  is linearly independent and also spans  $L$ , or it is a basis of  $L \Rightarrow \dim(L) = 3$ .

(10) Find a basis for  $W = \{ 1, \sin x, \cos x \}$ , a subspace of  $F(\mathbb{R})$ .

Actually,  $\{ 1, \sin x, \cos x \}$  is a spanning set for  $W$  and it is also linearly independent, so it is a basis of  $W \Rightarrow \dim(W) = 3$ .

(11)  $U = \{ (x, y, z) \mid x + z = 0 \}$ : find the basis.

$$U = \{ (x, y, -x) \mid x, y \in \mathbb{R} \} = \text{Span} \{ (1, 0, -1), (0, 1, 0) \}$$

So,  $\{ (1, 0, -1), (0, 1, 0) \}$  spans  $U$  and it is LI, so a basis.

Thus,  $\dim(U) = 2$ .

Remark: The zero space  $\{ \vec{0} \}$  has no basis, because no LI set of vectors can contain the zero vector  $\vec{0}$ . Therefore, an "empty" set of vectors is a basis of  $\{ \vec{0} \}$ , and  $\dim \{ \vec{0} \} = 0$ . Thus, the statement that "the dimension of a vector space is the number of vectors in a basis" holds even for the zero space.