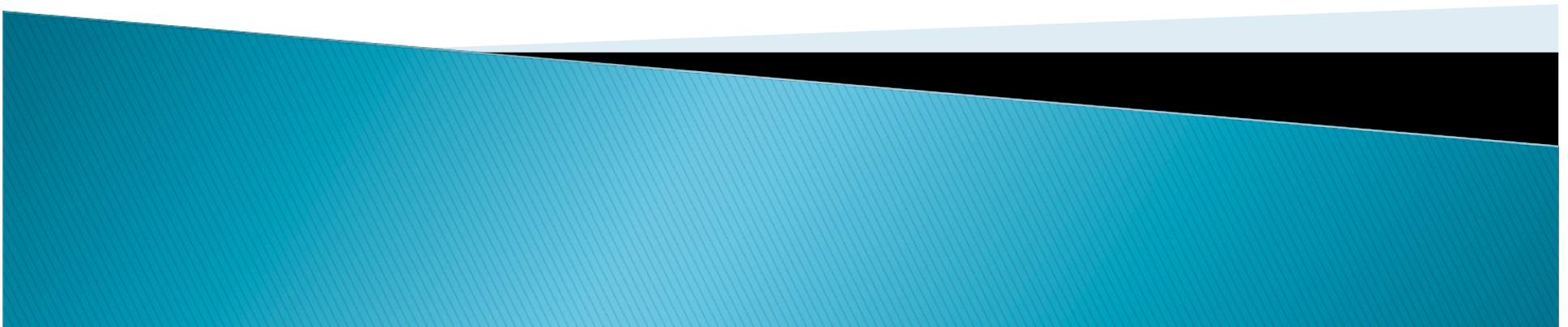


## 3.4. Subspaces, Linear independence



# Subspace

- ▶ A subspace is a set one can do scalar multiplication and addition and not leave the set.

**Definition 3.4.1** A nonempty set of vectors in  $R^n$  is called a *subspace* of  $R^n$  if it is closed under scalar multiplication and addition.

1. A subspace is usually given by conditions.
2. We need to verify the conditions after scalar multiplications or additions.



- ▶  $\{0\}$  is a subspace
- ▶ Every subspace contains  $0$ . Why?
- ▶  $W = \{(x, y) \in \mathbb{R}^2 \mid x > 0, y > 0\}$  is not a subspace. Why?
- ▶  $W = \{(x, y, 0) \in \mathbb{R}^3\}$  is a subspace.
- ▶  $W$  in  $\mathbb{R}^n$  given by  $x_2 = 1, x_3 = -1$  a subspace?
- ▶ Let  $v_1, v_2, \dots, v_s$  is given in  $\mathbb{R}^n$ .
  - Let  $W = \{c_1 v_1 + c_2 v_2 + \dots + c_s v_s \mid c_i \in \mathbb{R}\}$ .
  - That is  $W$  is the set of all linear combinations of given vectors  $v_1, v_2, \dots, v_s$ .
  - Then  $W$  is a subspace.
- ▶ We write  $W = \text{span}\{v_1, v_2, \dots, v_s\}$



**Theorem 3.4.2** If  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s$  are vectors in  $R^n$ , then the set of all linear combinations

$$\mathbf{x} = t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + \cdots + t_s\mathbf{v}_s \quad (3)$$

is a subspace of  $R^n$ .

- ▶ Example 2:  $\text{Span}\{\mathbf{0}\} = \{\mathbf{0}\}$ .
- ▶ Example 3:  $\text{Span}\{(1, 1, 2, 0)\}$  is a line.
- ▶ Example 4.
  - A subspace in  $R^1$ : itself or  $\{\mathbf{0}\}$ .
  - A subspace in  $R^2$ : itself, a line through  $\mathbf{0}$ ,  $\{\mathbf{0}\}$ .
  - A subspace in  $R^3$ : itself, a plane through  $\mathbf{0}$  ( $Ax + By + Cz = 0$ ), a line through  $\mathbf{0}$ ,  $\{\mathbf{0}\}$
  - A subspace in  $R^n$ : itself, a subspace  $\approx R^i$ ,  $\{\mathbf{0}\}$ .



# Solution space of a linear system

**Theorem 3.4.3** *If  $A\mathbf{x} = \mathbf{0}$  is a homogeneous linear system with  $n$  unknowns, then its solution set is a subspace of  $\mathbb{R}^n$ .*

- ▶ **Proof:**  $W = \{\mathbf{x} | A\mathbf{x} = \mathbf{0}\}$ .
  - If  $\mathbf{x}_0$  is a solution, then  $k\mathbf{x}_0$  is a solution.
  - If  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are solutions, then  $\mathbf{x}_1 + \mathbf{x}_2$  is a solution.
  - Thus  $W$  is closed under scalar multiplications and additions.  
Thus  $W$  is a subspace.
- ▶ If one has an inhomogeneous system, then the solution space is not a subspace.
- ▶ See Example \*.



### Theorem 3.4.4

- (a) If  $A$  is a matrix with  $n$  columns, then the solution space of the homogeneous system  $A\mathbf{x} = \mathbf{0}$  is all of  $R^n$  if and only if  $A = 0$ .
- (b) If  $A$  and  $B$  are matrices with  $n$  columns, then  $A = B$  if and only if  $A\mathbf{x} = B\mathbf{x}$  for every  $\mathbf{x}$  in  $R^n$ .

▶ Philosophy:  $A$  is determined by  $A\mathbf{x}$ 's.

▶ Proof:

- (a)  $\rightarrow$ )  $A=0$ .  $A\mathbf{x}=0$ .
- $\leftarrow$ )  $A\mathbf{x}=0$  for all  $\mathbf{x}$ .  $A\mathbf{e}_1=0, A\mathbf{e}_2=0, \dots, A\mathbf{e}_n=0$ .
  - $A=A\mathbf{I}=A[\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n]=[A\mathbf{e}_1, A\mathbf{e}_2, \dots, A\mathbf{e}_n]=0$ .
  - Thus all columns of  $A$  are zero.
- (b)  $A\mathbf{x}=B\mathbf{x}$  for all  $\mathbf{x}$ .  $A\mathbf{x}-B\mathbf{x}=0$ .  $(A-B)\mathbf{x}=0$  for all  $\mathbf{x}$ .  $A-B=0$ .  
 $A=B$ .



# Linear independence

- ▶ How can we find a good way to describe a subspaces...
  - Find equations... See as solutions spaces
  - Find parameters... Write a vector as a linear combination of vectors in unique way for a fixed set of vectors. These should be the least in number.
  - So we want to avoid “linearly dependent set of vectors”: when some of the vectors in the set can be written as a linear combination of some others.
  - In such cases, the number can be reduced by eliminating these.



**Definition 3.4.5** A nonempty set of vectors  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s\}$  in  $R^n$  is said to be *linearly independent* if the only scalars  $c_1, c_2, \dots, c_s$  that satisfy the equation

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_s\mathbf{v}_s = \mathbf{0} \quad (9)$$

are  $c_1 = 0, c_2 = 0, \dots, c_s = 0$ . If there are scalars, not all zero, that satisfy this equation, then the set is said to be *linearly dependent*.

- ▶  $\{\mathbf{0}\}$  is linearly dependent.  $c\mathbf{0}=\mathbf{0}$  for all  $c$ .
- ▶  $\{\mathbf{v}\}$   $\mathbf{v}$  nonzero is linearly independent.  $c\mathbf{v}=\mathbf{0}$  iff  $c=0$ .





**Theorem 3.4.6** A set  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s\}$  in  $R^n$  with two or more vectors is linearly dependent if and only if at least one of the vectors in  $S$  is expressible as a linear combination of the other vectors in  $S$ .

- ▶ **Proof:**  $\rightarrow$ )  $0 = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_s\mathbf{v}_s$ .
  - Not all  $c_i$  are zero. Say  $c_i$  is not.
  - Then  $c_i\mathbf{v}_i = c_1\mathbf{v}_1 + \dots + c_{(i-1)}\mathbf{v}_{(i-1)} + c_{(i+1)}\mathbf{v}_{(i+1)} + \dots + c_s\mathbf{v}_s$ .
  - $\mathbf{v}_i = (c_1/c_i)\mathbf{v}_1 + \dots + (c_{(i-1)}/c_i)\mathbf{v}_{(i-1)} + (c_{(i+1)}/c_i)\mathbf{v}_{(i+1)} + \dots + (c_s/c_i)\mathbf{v}_s$ .
- ▶  $\leftarrow$ )  $\mathbf{v}_i = d_1\mathbf{v}_1 + \dots + d_{(i-1)}\mathbf{v}_{(i-1)} + d_{(i+1)}\mathbf{v}_{(i+1)} + \dots + d_s\mathbf{v}_s$ .
  - Thus,  $d_1\mathbf{v}_1 + \dots + d_{(i-1)}\mathbf{v}_{(i-1)} + (-1)\mathbf{v}_i + d_{(i+1)}\mathbf{v}_{(i+1)} + \dots + d_s\mathbf{v}_s = 0$



- ▶ Example 10. two vectors in  $\mathbb{R}^n$ .
- ▶ Example 11. three vectors in  $\mathbb{R}^n$  is dependent if one is a linear combination of the other two.
  - Thus, the three vectors lie in a common plane or a common plane or  $\{O\}$ .
  - Three vectors are linearly independent if there are no such planes, lines.



# Linear independence and homogeneous linear systems

- ▶ Given  $v_1, v_2, \dots, v_s$ , write  $A = [v_1, v_2, \dots, v_s]$ .
- ▶ We write  $c_1v_1 + c_2v_2 + \dots + c_s v_s = 0$  as

$$[v_1, v_2, \dots, v_s] \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_s \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

**Theorem 3.4.7** *A homogeneous linear system  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution if and only if the column vectors of  $A$  are linearly independent.*

- ▶ See Examples 12.



**Theorem 3.4.8** *A set with more than  $n$  vectors in  $R^n$  is linearly dependent.*

**Theorem 3.4.9** *If  $A$  is an  $n \times n$  matrix, then the following statements are equivalent.*

- (a) *The reduced row echelon form of  $A$  is  $I_n$ .*
- (b)  *$A$  is expressible as a product of elementary matrices.*
- (c)  *$A$  is invertible.*
- (d)  *$A\mathbf{x} = \mathbf{0}$  has only the trivial solution.*
- (e)  *$A\mathbf{x} = \mathbf{b}$  is consistent for every vector  $\mathbf{b}$  in  $R^n$ .*
- (f)  *$A\mathbf{x} = \mathbf{b}$  has exactly one solution for every vector  $\mathbf{b}$  in  $R^n$ .*
- (g) *The column vectors of  $A$  are linearly independent.*
- (h) *The row vectors of  $A$  are linearly independent.*

Proof: (d)(g) equivalent by Th.3.4.7.

(g) $\rightarrow$ (h): (g) $\rightarrow$ (c).  $A^T$  is invertible. Use (g) for  $A^T$ . (h) follows

(h) $\rightarrow$ (g): (g) for  $A^T$  holds.  $A^T$  is invertible.  $\rightarrow A$  is invertible  $\rightarrow$  (g).

# Ex. Set. 3.4.

- ▶ 1-8 Span problem
- ▶ 9,10 independence
- ▶ 13-16 span problem
- ▶ 17-22 linear independence
- ▶ 23-26 Subspaces

