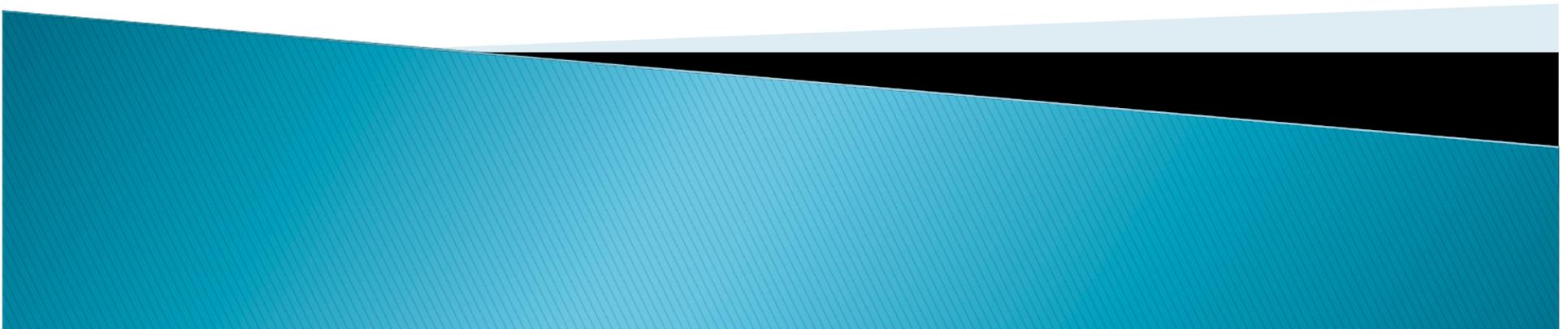


3.4. Subspaces, Linear independence



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 *.



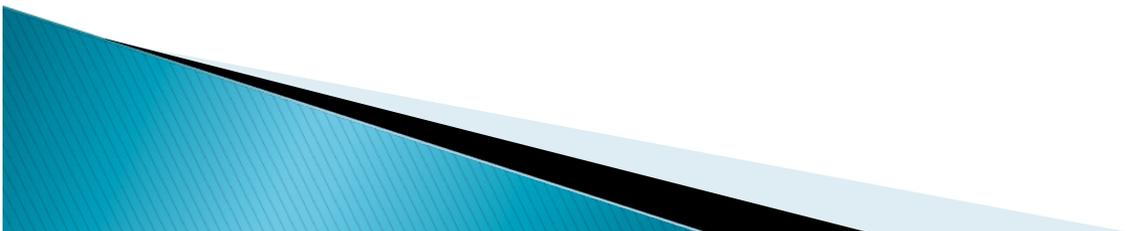
Theorem 3.4.4

- (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

