

## 3.4. Subspaces and Linear Independence

### Definition

If  $W$  is a nonempty set of vectors in  $\mathbb{R}^n$ , then

1.  $W$  is **closed under scalar multiplication** if  $c\mathbf{v} \in W$  for any scalar  $c$  and any vector  $\mathbf{v} \in W$ .
2.  $W$  is **closed under addition** if  $\mathbf{v}_1 + \mathbf{v}_2 \in W$  for any vectors  $\mathbf{v}_1, \mathbf{v}_2 \in W$ .
3. A nonempty set  $W$  of vectors in  $\mathbb{R}^n$  is called a **subspace** of  $\mathbb{R}^n$  if it is closed under scalar multiplication and addition.

### Examples

- ▶ lines through the origin in  $\mathbb{R}^2$
- ▶ planes through the origin in  $\mathbb{R}^3$

# Properties of subspaces

- ▶  $\{\mathbf{0}\}$  is a subspace of  $\mathbb{R}^n$ , called **zero subspace** or **trivial subspace**.
- ▶  $\mathbb{R}^n$  itself is a subspace of  $\mathbb{R}^n$ .
- ▶ Every subspace of  $\mathbb{R}^n$  must contain  $\mathbf{0}$ .

# Properties of subspaces

## Theorem

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

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

*is a subspace of  $\mathbb{R}^n$  called the **span** of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s$  and denoted by **span**  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s\}$ .*

# Properties of subspaces

## Example

$$\{\mathbf{0}\} = \text{span}\{\mathbf{0}\}.$$

## Example

$$\mathbb{R}^n = \text{span}\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}.$$

# Properties of subspaces

## Example

List all subspaces of  $\mathbb{R}^2$ .

## Example

List all subspaces of  $\mathbb{R}^3$ .

# Solution space of a linear system

## Theorem

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

# Solution space of a linear system

## Theorem

1. If  $A$  is a matrix with  $n$  columns, then the solution space of the homogeneous system  $A\mathbf{x} = \mathbf{0}$  is all of  $\mathbb{R}^n$  if and only if  $A = 0$ .
2. 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  $\mathbb{R}^n$ .

# Linear independence

## Definition

A nonempty set of vectors  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s\}$  in  $\mathbb{R}^n$  is said to be **linearly independent** if the only scalar  $c_1, c_2, \dots, c_n$  satisfying the equation

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_s \mathbf{v}_s = \mathbf{0}$$

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**.



# Linear independence

## Example

A vector  $\mathbf{v}$  is linearly independent if and only if it is not the zero vector.

## Example

A nonempty set of vectors in  $\mathbb{R}^n$  containing the zero vector is linearly dependent.

# Linear independence

## Theorem

*A set  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_s\}$  in  $\mathbb{R}^n$  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 other vectors in  $S$ .*

# Linear independence

## Example

Two vectors in  $\mathbb{R}^n$  are linearly dependent if they are colinear and linearly independent if they are not.

## Example

Three vectors in  $\mathbb{R}^n$  are linearly dependent if they lie in a plane through the origin and are linearly independent if they are not.

# Linear independence

## Theorem

*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.*

# Linear independence

## Example

Determine whether the following vectors are linearly independent or not.

$$\mathbf{v}_1 = (1, 2, 1), \mathbf{v}_2 = (2, 5, 0), \mathbf{v}_3 = (3, 3, 8)$$

# Linear independence

## Example

Determine whether the following vectors are linearly independent or not.

$$\mathbf{v}_1 = (2, -4, 6), \mathbf{v}_2 = (0, 7, -5), \mathbf{v}_3 = (6, 9, 8), \mathbf{v}_4 = (5, 0, 1)$$

## Theorem

*A set with more than  $n$  vectors in  $\mathbb{R}^n$  is linearly dependent.*

# The unifying theorem

## Theorem

*If  $A$  is an  $n \times n$  matrix, then the followings are equivalent.*

- 1. The reduced row echelon form of  $A$  is  $I_n$ .*
- 2.  $A$  is expressible as a product of elementary matrices.*
- 3.  $A$  is invertible.*
- 4.  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.*
- 5.  $A\mathbf{x} = \mathbf{b}$  is consistent for every vector  $\mathbf{b}$  in  $\mathbb{R}^n$ .*
- 6.  $A\mathbf{x} = \mathbf{b}$  has exactly one solution for every vector  $\mathbf{b}$  in  $\mathbb{R}^n$ .*
- 7. The column vectors of  $A$  are linearly independent.*
- 8. The row vectors of  $A$  are linearly independent.*