

MATH 304  
Linear Algebra

**Lecture 12:**  
**Subspaces of vector spaces.**

## Vector space

A *vector space* is a set  $V$  equipped with two operations, **addition**

$$V \times V \ni (\mathbf{x}, \mathbf{y}) \mapsto \mathbf{x} + \mathbf{y} \in V$$

and **scalar multiplication**

$$\mathbb{R} \times V \ni (r, \mathbf{x}) \mapsto r\mathbf{x} \in V,$$

that have the following properties:

## Properties of addition and scalar multiplication

**A1.**  $\mathbf{a} + \mathbf{b} = \mathbf{b} + \mathbf{a}$  for all  $\mathbf{a}, \mathbf{b} \in V$ .

**A2.**  $(\mathbf{a} + \mathbf{b}) + \mathbf{c} = \mathbf{a} + (\mathbf{b} + \mathbf{c})$  for all  $\mathbf{a}, \mathbf{b}, \mathbf{c} \in V$ .

**A3.** There exists an element of  $V$ , called the *zero vector* and denoted  $\mathbf{0}$ , such that  $\mathbf{a} + \mathbf{0} = \mathbf{0} + \mathbf{a} = \mathbf{a}$  for all  $\mathbf{a} \in V$ .

**A4.** For any  $\mathbf{a} \in V$  there exists an element of  $V$ , denoted  $-\mathbf{a}$ , such that  $\mathbf{a} + (-\mathbf{a}) = (-\mathbf{a}) + \mathbf{a} = \mathbf{0}$ .

**A5.**  $r(\mathbf{a} + \mathbf{b}) = r\mathbf{a} + r\mathbf{b}$  for all  $r \in \mathbb{R}$  and  $\mathbf{a}, \mathbf{b} \in V$ .

**A6.**  $(r + s)\mathbf{a} = r\mathbf{a} + s\mathbf{a}$  for all  $r, s \in \mathbb{R}$  and  $\mathbf{a} \in V$ .

**A7.**  $(rs)\mathbf{a} = r(s\mathbf{a})$  for all  $r, s \in \mathbb{R}$  and  $\mathbf{a} \in V$ .

**A8.**  $1\mathbf{a} = \mathbf{a}$  for all  $\mathbf{a} \in V$ .

- Associativity of addition implies that a multiple sum  $\mathbf{u}_1 + \mathbf{u}_2 + \cdots + \mathbf{u}_k$  is well defined for any  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k \in V$ .

- **Subtraction** in  $V$  is defined as usual:  
 $\mathbf{a} - \mathbf{b} = \mathbf{a} + (-\mathbf{b})$ .

- Addition and scalar multiplication are called **linear operations**.

Given  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k \in V$  and  $r_1, r_2, \dots, r_k \in \mathbb{R}$ ,

$$\boxed{r_1\mathbf{u}_1 + r_2\mathbf{u}_2 + \cdots + r_k\mathbf{u}_k}$$

is called a **linear combination** of  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ .

## Additional properties of vector spaces

- The zero vector is unique.
- For any  $\mathbf{a} \in V$ , the negative  $-\mathbf{a}$  is unique.
- $\mathbf{a} + \mathbf{b} = \mathbf{c} \iff \mathbf{a} = \mathbf{c} - \mathbf{b}$  for all  $\mathbf{a}, \mathbf{b}, \mathbf{c} \in V$ .
- $\mathbf{a} + \mathbf{c} = \mathbf{b} + \mathbf{c} \iff \mathbf{a} = \mathbf{b}$  for all  $\mathbf{a}, \mathbf{b}, \mathbf{c} \in V$ .
- $0\mathbf{a} = \mathbf{0}$  for any  $\mathbf{a} \in V$ .
- $(-1)\mathbf{a} = -\mathbf{a}$  for any  $\mathbf{a} \in V$ .

## Examples of vector spaces

- $\mathbb{R}^n$ :  $n$ -dimensional coordinate vectors
- $\mathcal{M}_{m,n}(\mathbb{R})$ :  $m \times n$  matrices with real entries
- $\mathbb{R}^\infty$ : infinite sequences  $(x_1, x_2, \dots)$ ,  $x_i \in \mathbb{R}$
- $\{\mathbf{0}\}$ : the trivial vector space
- $F(\mathbb{R})$ : the set of all functions  $f : \mathbb{R} \rightarrow \mathbb{R}$
- $C(\mathbb{R})$ : all continuous functions  $f : \mathbb{R} \rightarrow \mathbb{R}$
- $C^1(\mathbb{R})$ : all continuously differentiable functions  $f : \mathbb{R} \rightarrow \mathbb{R}$
- $C^\infty(\mathbb{R})$ : all smooth functions  $f : \mathbb{R} \rightarrow \mathbb{R}$
- $\mathcal{P}$ : all polynomials  $p(x) = a_0 + a_1x + \dots + a_nx^n$

## Subspaces of vector spaces

*Definition.* A vector space  $V_0$  is a **subspace** of a vector space  $V$  if  $V_0 \subset V$  and the linear operations on  $V_0$  agree with the linear operations on  $V$ .

*Examples.*

- $F(\mathbb{R})$ : all functions  $f : \mathbb{R} \rightarrow \mathbb{R}$
- $C(\mathbb{R})$ : all continuous functions  $f : \mathbb{R} \rightarrow \mathbb{R}$

$C(\mathbb{R})$  is a subspace of  $F(\mathbb{R})$ .

- $\mathcal{P}$ : polynomials  $p(x) = a_0 + a_1x + \cdots + a_kx^k$
- $\mathcal{P}_n$ : polynomials of degree less than  $n$

$\mathcal{P}_n$  is a subspace of  $\mathcal{P}$ .

## Subspaces of vector spaces

*Counterexamples.*

- $\mathbb{R}^n$ :  $n$ -dimensional coordinate vectors
- $\mathbb{Q}^n$ : vectors with rational coordinates

$\mathbb{Q}^n$  is not a subspace of  $\mathbb{R}^n$ .

$\sqrt{2}(1, 1, \dots, 1) \notin \mathbb{Q}^n \implies \mathbb{Q}^n$  is not a vector space  
(scaling is not well defined).

- $\mathcal{P}$ : polynomials  $p(x) = a_0 + a_1x + \dots + a_nx^n$
- $P_n^*$ : polynomials of degree  $n$  ( $n > 0$ )

$P_n^*$  is not a subspace of  $\mathcal{P}$ .

$-x^n + (x^n + 1) = 1 \notin P_n^* \implies P_n^*$  is not a vector space  
(addition is not well defined).



If  $S$  is a subset of a vector space  $V$  then  $S$  inherits from  $V$  addition and scalar multiplication. However  $S$  need not be closed under these operations.

**Proposition** A subset  $S$  of a vector space  $V$  is a subspace of  $V$  if and only if  $S$  is **nonempty** and **closed under linear operations**, i.e.,

$$\begin{aligned} \mathbf{x}, \mathbf{y} \in S &\implies \mathbf{x} + \mathbf{y} \in S, \\ \mathbf{x} \in S &\implies r\mathbf{x} \in S \text{ for all } r \in \mathbb{R}. \end{aligned}$$

*Proof:* “only if” is obvious.

“if”: properties like associative, commutative, or distributive law hold for  $S$  because they hold for  $V$ . We only need to verify properties A3 and A4. Take any  $\mathbf{x} \in S$  (note that  $S$  is nonempty). Then  $\mathbf{0} = 0\mathbf{x} \in S$ . Also,  $-\mathbf{x} = (-1)\mathbf{x} \in S$ .

*Example.*  $V = \mathbb{R}^2$ .

- The line  $x - y = 0$  is a subspace of  $\mathbb{R}^2$ .

The line consists of all vectors of the form  $(t, t)$ ,  $t \in \mathbb{R}$ .

$$(t, t) + (s, s) = (t + s, t + s) \implies \text{closed under addition}$$
$$r(t, t) = (rt, rt) \implies \text{closed under scaling}$$

- The parabola  $y = x^2$  is not a subspace of  $\mathbb{R}^2$ .

It is enough to find one explicit counterexample.

*Counterexample 1:*  $(1, 1) + (-1, 1) = (0, 2)$ .

$(1, 1)$  and  $(-1, 1)$  lie on the parabola while  $(0, 2)$  does not  
 $\implies$  not closed under addition

*Counterexample 2:*  $2(1, 1) = (2, 2)$ .

$(1, 1)$  lies on the parabola while  $(2, 2)$  does not  
 $\implies$  not closed under scaling

*Example.*  $V = \mathbb{R}^3$ .

- The plane  $z = 0$  is a subspace of  $\mathbb{R}^3$ .
- The plane  $z = 1$  is not a subspace of  $\mathbb{R}^3$ .
- The line  $t(1, 1, 0)$ ,  $t \in \mathbb{R}$  is a subspace of  $\mathbb{R}^3$  and a subspace of the plane  $z = 0$ .
- The line  $(1, 1, 1) + t(1, -1, 0)$ ,  $t \in \mathbb{R}$  is not a subspace of  $\mathbb{R}^3$  as it lies in the plane  $x + y + z = 3$ , which does not contain  $\mathbf{0}$ .
- In general, a straight line or a plane in  $\mathbb{R}^3$  is a subspace if and only if it passes through the origin.

System of linear equations:

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\ \dots\dots\dots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m \end{cases}$$

Any solution  $(x_1, x_2, \dots, x_n)$  is an element of  $\mathbb{R}^n$ .

**Theorem** The solution set of the system is a subspace of  $\mathbb{R}^n$  if and only if all  $b_i = 0$ .

**Theorem** The solution set of a system of linear equations in  $n$  variables is a subspace of  $\mathbb{R}^n$  if and only if all equations are homogeneous.

*Proof:* “only if”: the zero vector  $\mathbf{0} = (0, 0, \dots, 0)$  is a solution only if all equations are homogeneous.

“if”: a system of homogeneous linear equations is equivalent to a matrix equation  $A\mathbf{x} = \mathbf{0}$ .

$A\mathbf{0} = \mathbf{0} \implies \mathbf{0}$  is a solution  $\implies$  solution set is not empty.

If  $A\mathbf{x} = \mathbf{0}$  and  $A\mathbf{y} = \mathbf{0}$  then  $A(\mathbf{x} + \mathbf{y}) = A\mathbf{x} + A\mathbf{y} = \mathbf{0}$ .

If  $A\mathbf{x} = \mathbf{0}$  then  $A(r\mathbf{x}) = r(A\mathbf{x}) = \mathbf{0}$ .

Examples of subspaces of  $\mathcal{M}_{2,2}(\mathbb{R})$ :  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$

- diagonal matrices:  $b = c = 0$
- upper triangular matrices:  $c = 0$
- lower triangular matrices:  $b = 0$
- symmetric matrices ( $A^T = A$ ):  $b = c$
- anti-symmetric (or skew-symmetric) matrices ( $A^T = -A$ ):  $a = d = 0, c = -b$
- matrices with zero trace:  $a + d = 0$   
(trace = the sum of diagonal entries)
- matrices with zero determinant,  $ad - bc = 0$ ,

**do not** form a subspace:  $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ .