MATH 304 Linear Algebra

Lecture 12: Subspaces of vector spaces.

#### **Vector space**

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

$$V imes V 
i (\mathbf{x}, \mathbf{y}) \mapsto \mathbf{x} + \mathbf{y} \in V$$

## and scalar multiplication

$$\mathbb{R} imes V 
i (r, \mathbf{x}) \mapsto r\mathbf{x} \in V$$
,

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 **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  $\mathbf{u}_1, \mathbf{u}_2, \ldots, \mathbf{u}_k \in V$ .

• 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  $r_1, r_2, \dots, r_k \in \mathbb{R}$ ,  

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

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

#### 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  $(x_1, x_2, \dots)$ ,  $x_i \in \mathbb{R}$
- $\{0\}$ : the trivial vector space
- $F(\mathbb{R})$ : the set of all functions  $f:\mathbb{R}\to\mathbb{R}$
- $C(\mathbb{R})$ : all continuous functions  $f:\mathbb{R}\to\mathbb{R}$
- $C^1(\mathbb{R})$ : all continuously differentiable functions  $f: \mathbb{R} \to \mathbb{R}$
- $C^{\infty}(\mathbb{R})$ : all smooth functions  $f:\mathbb{R}\to\mathbb{R}$
- $\mathcal{P}$ : all polynomials  $p(x) = a_0 + a_1 x + \cdots + a_n x^n$

## 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  $f : \mathbb{R} \to \mathbb{R}$
- $C(\mathbb{R})$ : all continuous functions  $f : \mathbb{R} \to \mathbb{R}$  $C(\mathbb{R})$  is a subspace of  $F(\mathbb{R})$ .
- $\mathcal{P}$ : polynomials  $p(x) = a_0 + a_1 x + \cdots + a_k x^k$
- $\mathcal{P}_n$ : 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  $p(x) = a_0 + a_1 x + \cdots + a_n x^n$
- $P_n^*$ : 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.,

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

*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,  $-\mathbf{x} = (-1)\mathbf{x} \in S$ . Thus  $\mathbf{0}$  and  $-\mathbf{x}$  in S are the same as in V. *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),  $t \in \mathbb{R}$ .  $(t, t) + (s, s) = (t + s, t + s) \implies$  closed under addition  $r(t, t) = (rt, rt) \implies$  closed under scaling

The parabola y = x<sup>2</sup> is not a subspace of ℝ<sup>2</sup>.
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 ⇒ 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 **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:

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

Any solution  $(x_1, x_2, \ldots, x_n)$  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  $\mathbf{0} = (0, 0, ..., 0)$  is a solution only if all equations are homogeneous.

"if": a system of homogeneous linear equations is equivalent to a matrix equation  $A\mathbf{x} = \mathbf{0}$ , where A is the coefficient matrix of the system and all vectors are regarded as column vectors.  $A\mathbf{0} = \mathbf{0} \implies \mathbf{0}$  is a solution  $\implies$  solution set is not empty. If  $A\mathbf{x} = \mathbf{0}$  and  $A\mathbf{y} = \mathbf{0}$  then  $A(\mathbf{x} + \mathbf{y}) = A\mathbf{x} + A\mathbf{y} = \mathbf{0}$  $\implies$  solution set is closed under addition. If  $A\mathbf{x} = \mathbf{0}$  then  $A(r\mathbf{x}) = r(A\mathbf{x}) = \mathbf{0}$  $\implies$  solution set is closed under scaling.

 $\implies$  solution set is closed under scaling.

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:  $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ .

Let V be a vector space and  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in V$ . Consider the set L of all linear combinations  $r_1\mathbf{v}_1 + r_2\mathbf{v}_2 + \dots + r_n\mathbf{v}_n$ , where  $r_1, r_2, \dots, r_n \in \mathbb{R}$ .

#### **Theorem** L is a subspace of V.

*Proof:* First of all, *L* is not empty. For example,  $\mathbf{0} = 0\mathbf{v}_1 + 0\mathbf{v}_2 + \cdots + 0\mathbf{v}_n$  belongs to *L*.

The set L is closed under addition since

$$(r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_n\mathbf{v}_n)+(s_1\mathbf{v}_1+s_2\mathbf{v}_2+\cdots+s_n\mathbf{v}_n)=$$
  
= (r\_1+s\_1)\mathbf{v}\_1+(r\_2+s\_2)\mathbf{v}\_2+\cdots+(r\_n+s\_n)\mathbf{v}\_n.

The set L is closed under scalar multiplication since

$$t(r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_n\mathbf{v}_n)=(tr_1)\mathbf{v}_1+(tr_2)\mathbf{v}_2+\cdots+(tr_n)\mathbf{v}_n.$$