

Last time

- Solving a system of linear equations by Gaussian elimination

$$\begin{cases} a_{11}x_1 + \dots + a_{1n}x_n = b_1 \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n = b_m \end{cases} \text{ is equivalent to } (a_{ij}) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$$

Even more succinctly:  $(*) A \vec{x} = \vec{b}$

- In terms of maps,  $(*)$  says  $\vec{x}$  solves  $(*) \Leftrightarrow T_A(\vec{x}) = \vec{b}$   
where  $T_A \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} := \begin{pmatrix} \sum a_{1i}x_i \\ \vdots \\ \sum a_{mi}x_i \end{pmatrix}$

- We checked that  $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is linear:  $\forall \lambda, \mu \in \mathbb{R}$   
 $\forall \vec{x}, \vec{y} \in \mathbb{R}^n$

$$T_A(\lambda \vec{x} + \mu \vec{y}) = \lambda T_A(\vec{x}) + \mu T_A(\vec{y})$$

Note  $f: \mathbb{R} \rightarrow \mathbb{R}$   $f(x) = ax + b$ ,  $a, b \in \mathbb{R}$ ,  $a, b \neq 0$ , is not linear. Why not?

Today: Vector spaces (over  $\mathbb{R}$ ).

"Recall" We have two operations on elements of  $\mathbb{R}^n = \left\{ \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mid x_i \in \mathbb{R} \right\}$ :

1) multiplication by scalars

$$\lambda \cdot \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} := \begin{pmatrix} \lambda x_1 \\ \vdots \\ \lambda x_n \end{pmatrix} \quad \forall \lambda \in \mathbb{R}, \forall \vec{x} \in \mathbb{R}^n$$

2) addition

$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} := \begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \end{pmatrix}$$

These two operations (which are just functions  $\mathbb{R} \times \mathbb{R}^n \rightarrow \mathbb{R}^n$  and  $\mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^n$ ) has several properties, We single out 8:

1)  $+$  is commutative:  $\vec{x} + \vec{y} = \vec{y} + \vec{x}$  for all  $\vec{x}, \vec{y} \in \mathbb{R}^n$  2.2

2)  $+$  is associative:  $(\vec{x} + \vec{y}) + \vec{z} = \vec{x} + (\vec{y} + \vec{z})$  for all  $\vec{x}, \vec{y}, \vec{z} \in \mathbb{R}^n$ .

3) There is a vector  $\vec{0}$  (zero vector) so that

$$\vec{x} + \vec{0} = \vec{x} = \vec{0} + \vec{x} \quad \text{for all } \vec{x} \in \mathbb{R}^n$$

4) additive inverses:  $\forall \vec{x} \in \mathbb{R}^n \exists -\vec{x} \in \mathbb{R}^n$  s.t.  $\vec{x} + (-\vec{x}) = \vec{0}$ .

5) multiplication by 1 doesn't do anything:  $1 \cdot \vec{x} = \vec{x} \quad \forall \vec{x} \in \mathbb{R}^n$

6)  $\lambda \cdot (\mu \cdot \vec{x}) = (\lambda\mu) \cdot \vec{x} \quad \forall \lambda, \mu \in \mathbb{R}, \vec{x} \in \mathbb{R}^n$

7)  $\lambda \cdot (\vec{x} + \vec{y}) = \lambda \cdot \vec{x} + \lambda \cdot \vec{y} \quad \forall \lambda \in \mathbb{R}, \forall \vec{x}, \vec{y} \in \mathbb{R}^n$

8)  $(\lambda + \mu) \cdot \vec{x} = (\lambda \cdot \vec{x}) + (\mu \cdot \vec{x}) \quad \forall \lambda, \mu \in \mathbb{R}, \vec{x} \in \mathbb{R}^n$ .

Definition A (real) vector space (or a vector space over the field of real numbers) is a set  $V$  together with

$$\begin{array}{ll} \text{two functions} & \mathbb{R} \times V \rightarrow V & +: V \times V \rightarrow V \\ & (\lambda, \vec{x}) \mapsto \lambda \cdot \vec{x} & (\vec{x}, \vec{y}) \mapsto \vec{x} + \vec{y} \end{array}$$

and an element  $\vec{0} \in V$  so that properties (1)–(8) above hold for all  $\lambda, \mu \in \mathbb{R}$  and all  $\vec{x}, \vec{y}, \vec{z} \in V$ .

Ex  $\mathbb{R}^n$  ( $n=0, 1, 2, \dots$ ) are vector spaces

Note  $\mathbb{R}^0 = \{0\}$   $+: \{0\} \times \{0\} \rightarrow \{0\}$  is the only map  
and  $\mathbb{R} \times \{0\} \rightarrow \{0\}$  is the only map

Ex  $\mathbb{R}[x] = \{a_0 + a_1x + \dots + a_nx^n \mid n \geq 0, a_0, \dots, a_n \in \mathbb{R}\}$

= polynomials with real coefficients

is a vector space over  $\mathbb{R}$ , under the usual addition of polynomials and multiplication by scalars

Ex  $C^0([0, 1]) = \{f: [0, 1] \rightarrow \mathbb{R} \mid f \text{ is continuous}\}$

is a vector space under the usual addition of functions and multiplication by scalars.

Nonexample  $V = \left\{ \begin{pmatrix} x \\ y \end{pmatrix} \in \mathbb{R}^2 \mid x, y \geq 0 \right\}$

is not a vector space: while  $V \times V \rightarrow V$

$+$ :  $\begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} x' \\ y' \end{pmatrix} \mapsto \begin{pmatrix} x+x' \\ y+y' \end{pmatrix}$  makes sense

$\mathbb{R} \times V \rightarrow V$   $\lambda \cdot \begin{pmatrix} x \\ y \end{pmatrix} := \begin{pmatrix} \lambda x \\ \lambda y \end{pmatrix}$  does not:

$$(-1) \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} -1 \\ -2 \end{pmatrix} \notin V.$$

Lemma 2.1 In a vector space  $V$ , the zero vector is unique: if  $0' + \vec{v} = \vec{v}$  for all  $\vec{v} \in V$ , then  $0' = \vec{0}$ .

Proof  $0' + \vec{0} = \vec{0}$  by definition of  $0'$

$0' + \vec{0} = 0'$  by definition of  $\vec{0}$ .

$$\Rightarrow \vec{0}' = 0' + \vec{0} = \vec{0}.$$

□

Lemma 2.2 Let  $V$  be a vector space. For any  $\vec{v} \in V$   
 $0 \cdot \vec{v} = \vec{0}$ .

Proof  $0 \cdot \vec{v} = (0+0) \cdot \vec{v} = 0 \cdot \vec{v} + 0 \cdot \vec{v}$ .

Now add  $-(0 \cdot \vec{v})$  to both sides ( $-(0 \cdot \vec{v})$  exists by (4))

We get  $0 \cdot \vec{v} + (-(0 \cdot \vec{v})) = 0 \cdot \vec{v} + (0 \cdot \vec{v} + (-(0 \cdot \vec{v})))$

$$\Rightarrow \vec{0} = 0 \cdot \vec{v} + \vec{0} = 0 \cdot \vec{v}$$

□

Lemma 2.3 Let  $V$  be a vector space.

1) Additive inverses are unique: if  $\vec{u} + \vec{v} = \vec{0}$ , then  $\vec{u} = -\vec{v}$ .

2)  $(-1) \cdot \vec{v} = -\vec{v}$ .

Proof 1)  $-\vec{v} \stackrel{(3)}{=} \vec{0} + (-\vec{v}) \stackrel{\text{assumption}}{=} (\vec{u} + \vec{v}) + (-\vec{v}) \stackrel{\text{associativity}}{=} \vec{u} + (\vec{v} + (-\vec{v})) \stackrel{(4)}{=} \vec{u} + \vec{0} \stackrel{(3)}{=} \vec{u}$

2)  $\vec{v} + (-1) \cdot \vec{v} = 1 \cdot \vec{v} + (-1) \cdot \vec{v} = (1 + (-1)) \vec{v} = 0 \cdot \vec{v} = \vec{0}$  (by 2.2)

□

Definition (subspace) A subset  $W$  of a vector space  $V$  is a subspace if  $W$  is a vector space under the operations of  $+$  and scalar multiplication that are defined on  $V$ .

Ex  $V = \mathbb{R}^2$   $W = \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mid x_1 = x_2 \right\}$  is a subspace: 2.4  
 $= \left\{ \begin{pmatrix} x \\ x \end{pmatrix} \mid x \in \mathbb{R} \right\}$

Check: (i)  $\forall \lambda \in \mathbb{R} \quad \lambda \cdot \begin{pmatrix} x \\ x \end{pmatrix} = \begin{pmatrix} \lambda x \\ \lambda x \end{pmatrix} \in W$

So we have  $\mathbb{R} \times W \rightarrow W$

(ii)  $\forall \begin{pmatrix} x \\ x \end{pmatrix}, \begin{pmatrix} y \\ y \end{pmatrix} \in W \quad \begin{pmatrix} x \\ x \end{pmatrix} + \begin{pmatrix} y \\ y \end{pmatrix} = \begin{pmatrix} x+y \\ x+y \end{pmatrix} \in W$

So we have  $+$ ;  $W \times W \rightarrow W$ .

(iii)  $\vec{0} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \in W$

(iv)  $\forall \begin{pmatrix} x \\ x \end{pmatrix} \in W, -\begin{pmatrix} x \\ x \end{pmatrix} = \begin{pmatrix} -x \\ -x \end{pmatrix} \in W$

Note that  $+$  is automatically commutative and associative

Also  $\forall \vec{x} \in W \quad 1 \cdot \vec{x} = \vec{x}$  since it's true  $\forall \vec{x} \in \mathbb{R}^2$

Similarly (6), (7), (8) hold automatically.

Non-example  $W = \mathbb{R}^2 - \{0\}$  is not a subspace of  $\mathbb{R}^2$ :  $\vec{0} \notin W$ .

Non-example  $W = \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2 \mid x_1 + x_2 = 1 \right\}$  is not a subspace of  $\mathbb{R}^2$ : it's not closed under  $+$   
 $\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \in W$  by  $\begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \notin W$ .

Ex  $W$  For each  $n \geq 0$ ,  $P_n = \{a_0 + a_1x + \dots + a_nx^n \mid a_0, \dots, a_n \in \mathbb{R}\}$

$\equiv \{p(x) \in \mathbb{R}[x] \mid \deg p \leq n\}$  is a subspace of  $\mathbb{R}[x]$ :

$P_n$  is closed under  $+$  and scalar multiplication,

the zero polynomial is in  $P_n$  and if  $p(x) \in P_n$ ,  $-p(x) \in P_n$  as well.

Definition A map  $T: V \rightarrow W$  between two vector spaces is linear if  $T$  preserves  $+$  and scalar multiplication!

$\forall \lambda, \mu \in \mathbb{R}, \vec{x}, \vec{y} \in V$

$$T(\lambda \vec{x} + \mu \vec{y}) = \lambda T(\vec{x}) + \mu T(\vec{y})$$

Example We have seen that any  $m \times n$  matrix  $A = (a_{ij})$  2.5  
defines a linear map  $T_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$T_A \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} := \begin{pmatrix} \sum_i a_{1i} x_i \\ \vdots \\ \sum_i a_{mi} x_i \end{pmatrix}$$

Ex  $\frac{d}{dx}: \mathbb{R}[x] \rightarrow \mathbb{R}[x]$  is linear

$$\frac{d}{dx} \left( \sum_{i=0}^n a_i x^i \right) = \sum_{i=0}^{n-1} i a_i x^{i-1} \text{ is linear.}$$

Question Suppose  $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear map.

Is there an  $m \times n$  matrix  $A = (a_{ij})$  st

$$T(\vec{x}) = T_A(\vec{x}) \text{ for all } \vec{x} \in \mathbb{R}^n?$$

