

## Lecture 1.2: Linear independence and spanning sets

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# Linear independence

## Definition (recall)

A **vector space** consists of a **set**  $V$  (of “vectors”) and a **set**  $\mathbb{F}$  (of “scalars”; usually  $\mathbb{R}$  or  $\mathbb{C}$ ) that is:

- closed under **addition**:  $v, w \in V \implies v + w \in V$
- closed under **scalar multiplication**:  $v \in V, c \in \mathbb{F} \implies cv \in V$

In general, we are not allowed to multiply vectors.

## Definition

A set  $S \subseteq V$  is **linearly independent** if for any  $v_1, \dots, v_n \in S$ :

$$a_1 v_1 + \dots + a_n v_n = 0 \implies a_1 = a_2 = \dots = a_n = 0.$$

If  $S$  is not linearly independent, then it is **linearly dependent**.

## Intuition

$S \subseteq V$  is **linearly independent** if none of the vectors in  $S$  can be expressed as a linear combination of the others.

# Linear independence

## Definition (recall)

A set  $S \subseteq V$  is **linearly independent** if for any  $v_1, \dots, v_n \in S$ :

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## Example 1

Let  $V = \mathbb{R}^3$ , and  $S \subseteq V$ .

1. The set  $S = \{v_1\}$  is linearly independent iff  $v_1 \neq 0$ .
2. The set  $S = \{v_1, v_2\}$  is linearly independent iff  $v_1$  and  $v_2$  don't lie on the same line.
3. The set  $S = \{v_1, v_2, v_3\}$  is linearly independent iff  $v_1, v_2, v_3$  don't lie on the same plane.
4. The set  $S = \{v_1, v_2, v_3, v_4\}$  is *never* linearly independent in  $\mathbb{R}^3$ .

## Example 2

Let  $V = \mathbb{R}_3[x]$ , and  $S \subseteq V$ .

1. The set  $S = \{1, x, x^2\}$  is linearly independent.
2. The set  $S = \{1, x, x^2, x^3\}$  is linearly independent.
3. The set  $S = \{1, x, x^2, 1 + 3x - 4x^2\}$  is linearly dependent.
4. The set  $S = \{1, x, x^2, x^3 + x + 1\}$  is linearly independent.

# Linear independence

## Definition (recall)

A set  $S \subseteq V$  is **linearly independent** if for any  $v_1, \dots, v_n \in S$ :

$$a_1 v_1 + \dots + a_n v_n = 0 \implies a_1 = a_2 = \dots = a_n = 0.$$

## Example 3

Let  $V = C^\infty(\mathbb{C})$ , and  $S \subseteq V$ .

1.  $S = \{\cos t, \sin t\}$  is linearly independent.

Reason: If  $C_1 \cos t + C_2 \sin t = 0$ , then  $C_1 = C_2 = 0$ .

2.  $S = \{e^{2t}, e^{3t}\}$  is linearly independent.

Reason: If  $C_1 e^{2t} + C_2 e^{3t} = 0$ , then  $C_1 = C_2 = 0$ .

3.  $S = \{e^{2it}, e^{-2it}, \cos 2t\}$  is linearly dependent.

Reason:  $\cos 2t = \frac{1}{2}e^{2it} + \frac{1}{2}e^{-2it}$ .

4.  $S = \{e^{2t}, e^{-2t}, \cosh 2t\}$  is linearly dependent.

Reason:  $\cosh 2t = \frac{1}{2}e^{2t} + \frac{1}{2}e^{-2t}$ .

# Spanning sets and bases

## Definition

A subset  $S \subseteq V$  **spans**  $V$  if every  $v \in V$  can be written as  $v = a_1v_1 + \cdots + a_nv_n$  where  $v_i \in S$ ,  $a_i \in \mathbb{F}$ .

Moreover, if  $S$  is also **linearly independent** then  $S$  is a **basis** of  $V$ .

## Intuition

- “ $S$  spans  $V$ ” means “ $S$  generates all of  $V$ ”
- “ $S$  is a basis for  $V$ ” means “ $S$  is a minimal set that generates  $V$ .”

## Examples

Let  $V = \mathbb{R}^2$ .

- $S = \{(1, 0), (0, 1)\}$
- $S = \{(3, 1), (1, 1)\}$
- $S = \{(1, 0), (0, 1), (3, 1)\}$
- $S = \{(1, 1)\}$

Spans  $\mathbb{R}^2$ ?

Basis for  $\mathbb{R}^2$ ?

# Spanning sets and bases

## Theorem

Let  $S \subseteq V$ . The following are equivalent:

- $S$  is a **basis** of  $V$ ,
- $S$  is a **minimal spanning set** of  $V$ ,
- $S$  is a **maximal linearly independent set** in  $V$ .

**Example.** Let  $V = \mathbb{R}^3$ ,  $W \subseteq V$  any plane (through  $\mathbf{0}$ ).

*Intuition.* We need two vectors (not collinear) to generate  $W$ .

In fact,  $S = \{v_1, v_2\}$  is a *basis* for  $W$  iff  $v_1$  and  $v_2$  are not collinear.

Let's "go up" a dimension and find a basis for  $V$ .

$S = \{v_1, v_2, v_3\}$  is a basis for  $V$  iff they are not co-planar.

It should be clear how this generalizes to higher dimensions.

(By the way, what do we mean by "*dimension*"?)

# Spanning sets and bases

## Definition

The **dimension** of a vector space is the number of vectors in any basis.

## Examples

- $\dim(\mathbb{R}^n) = n$ :                      Basis:  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$
- $\dim(\mathbb{R}_n[x]) = n + 1$ :                Basis:  $\{1, x, \dots, x^n\}$
- $\dim(\mathbb{R}[x]) = \infty$ :                    Basis:  $\{1, x, x^2, \dots\}$
- $\dim(\text{Per}_{2\pi}) = \infty$ :                Basis:  $\{1, \cos x, \cos 2x, \dots\} \cup \{\sin x, \sin 2x, \dots\}$ .

## Remark

Any subset  $S \subseteq V$  spans a subspace  $W$  of  $V$ . Denote this subspace by  $\text{Span}(S)$ .

## How to construct a basis from a spanning set

### Algorithm

Consider any finite subset  $S \subseteq V$  and let  $W = \text{Span}(S)$ .

We may ask: *If  $S$  a **basis** for  $W$ ?*

If not, then  $S$  is not a minimal spanning set, so we can remove some  $v_1$  to get  $S' = S \setminus \{v_1\}$ , a smaller set that spans  $W$ .

We ask again: *Is  $S'$  a **basis** for  $W$ ?*

If not, then we can remove some  $v_2 \in S'$  to get  $S'' := S' \setminus \{v_2\}$ , a smaller set that spans  $W$ .

Since  $|S| < \infty$ , this process will eventually terminate, and we'll be left with  $\mathcal{B} := S^{(k)}$ , a **basis** for  $W$ .



## How to construct a basis from a spanning set

### Example

Let  $S = \{(1, 0, 0), (0, 1, 0), (1, 1, 0), (3, 1, 0)\} \subseteq \mathbb{R}^3$ .

$W = \text{Span}(S)$  is a plane. Since  $\dim(W) = 2$ , a basis of  $W$  has 2 vectors.

We can remove  $(1, 1, 0)$  and  $(3, 1, 0)$  to get a basis  $\mathcal{B} = \{(1, 0, 0), (0, 1, 0)\}$  of  $W$ .

This means that

$$\begin{aligned} W &= \{C_1(1, 0, 0) + C_2(0, 1, 0) \mid C_1, C_2, \in \mathbb{R}\} \\ &= \{(C_1, C_2, 0) \mid C_1, C_2, \in \mathbb{R}\}. \end{aligned}$$

However, note that  $\{(1, 0, 0), (3, 1, 0)\}$  is also a basis for  $W$ .

This means that

$$\begin{aligned} W &= \{C_1(1, 0, 0) + C_2(3, 1, 0) \mid C_1, C_2, \in \mathbb{R}\} \\ &= \{(C_1 + 3C_2, C_2, 0) \mid C_1, C_2, \in \mathbb{R}\}. \end{aligned}$$