

# Reminders

Let  $\mathbf{V}$  be a vector space over  $\mathbf{F}$ , let  $v_1, \dots, v_n \in \mathbf{V}$  be vectors.

## Definition

For any  $\alpha_1, \dots, \alpha_n \in \mathbf{F}$ , the vector

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n$$

is a **linear combination** of  $v_1, \dots, v_n$ .

Let  $S \subseteq \mathbf{V}$  be a set of vectors.

## Definition

The **linear span** of  $S$  (denoted by  $\text{span}(S)$ ) is the set of all linear combinations of elements of  $S$ .

We say that  $S$  **generates** a subspace  $\mathbf{U}$  if  $\mathbf{U} = \text{span}(S)$ .

## Lemma

Let  $S$  be a subset of a vector space  $\mathbf{V}$ . If  $v \in \text{span}(S)$ , then

$$\text{span}(S \cup \{v\}) = \text{span}(S).$$

## Proof.

If  $v \in \text{span}(S)$ , then

$$v = \alpha_1 v_1 + \dots + \alpha_n v_n$$

for some  $v_1, \dots, v_n \in S$ .

Clearly,  $\text{span}(S) \subseteq \text{span}(S \cup \{v\})$ . If  $x \in \text{span}(S \cup \{v\})$ , then w.l.o.g.

$$x = \beta v + \beta_1 v_1 + \dots + \beta_n v_n.$$

Hence,

$$x = (\beta_1 + \beta\alpha_1)v_1 + \dots + (\beta_n + \beta\alpha_n)v_n \in \text{span}(S).$$

## Spans generating a subspace

$$\begin{aligned}\{(x, y, z) : 3x - 3y + z = 0\} &= \text{span}(\{(x, y, z) : 3x - 3y + z = 0\}) \\ &= \text{span}((1, 1, 0), (1, 2, 3), (3, 4, 3)) \\ &= \text{span}((1, 1, 0), (1, 2, 3))\end{aligned}$$

# Non-minimality

In

$$\text{span}((1, 1, 0), (1, 2, 3), (3, 4, 3)),$$

the vector  $(3, 4, 3)$  is redundant—a linear combination of other vectors.

$$(3, 4, 3) = 2(1, 1, 0) + (1, 2, 3).$$

Equivalently,

$$(3, 4, 3) - 2(1, 1, 0) - (1, 2, 3) = \mathbf{0}.$$

A linear combination

$$\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n \text{ is}$$

- **trivial** if  $\alpha_1 = \dots = \alpha_n = 0$ ,
- **non-trivial** otherwise.

Let  $\mathbf{V}$  be a vector space.

### Definition

A set  $S \subseteq \mathbf{V}$  is **linearly independent** if no non-trivial linear combination of elements of  $S$  is equal to  $\mathbf{o}$ .

# Examples

- The set  $\{(1, 1, 0), (1, 2, 3), (3, 4, 3)\}$  in  $\mathbf{R}^3$  is **not** linearly independent, since

$$(3, 4, 3) - 2(1, 1, 0) - (1, 2, 3) = \mathbf{o}.$$

- The set  $\{x^2 + x + 1, x^2 + 2x + 3, x^2 + 2.5x + 4, x + 2\}$  in  $\mathcal{P}$  is **not** linearly independent, since

$$-(x^2 + x + 1) + 3(x^2 + 2x + 3) - 2(x^2 + 2.5x + 4) = \mathbf{o}.$$

- The set  $\{(1, 1, 0), (1, 2, 3)\}$  in  $\mathbf{R}^3$  is linearly independent.
- If  $\mathbf{o} \in S$ , then  $S$  is **not** linearly independent, since

$$1\mathbf{o} = \mathbf{o}.$$

- The empty set is linearly independent.

# Testing linear independence

## Problem

Is  $\{(0, 1, 0, 1), (1, 0, 1, 0), (1, 2, 3, 4)\}$  linearly independent?

Decide whether there exist  $\alpha_1, \alpha_2, \alpha_3 \neq 0, 0, 0$  such that

$$\alpha_1(0, 1, 0, 1) + \alpha_2(1, 0, 1, 0) + \alpha_3(1, 2, 3, 4) = (0, 0, 0, 0).$$

Equivalently

$$\begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 2 \\ 0 & 1 & 3 \\ 1 & 0 & 4 \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\text{RREF} \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 2 \\ 0 & 1 & 3 \\ 1 & 0 & 4 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix},$$

the only solution is  $(\alpha_1, \alpha_2, \alpha_3) = (0, 0, 0)$ . The vectors are **linearly independent**.

# Linear independence and RREF

## Lemma

*Vectors  $a_1, \dots, a_k \in \mathbf{F}^{n \times 1}$  are linearly independent if and only if all columns of*

$$RREF(a_1 | a_2 | \dots | a_k)$$

*are basis columns.*



# Linear independence and minimality of span

## Lemma

*A set  $S \subseteq \mathbf{V}$  is linearly independent if and only if for every  $T \subsetneq S$ ,*

$$\text{span}(T) \neq \text{span}(S).$$

## Proof.

$\Rightarrow$  If  $\text{span}(T) = \text{span}(S)$  for some  $T \subsetneq S$ , then consider  $v \in S \setminus T$ . Since  $v \in S \subseteq \text{span}(S) = \text{span}(T)$ , we have

$$v = \alpha_1 v_1 + \dots + \alpha_n v_n$$

for some  $v_1, \dots, v_n \in T$ . But

$$\alpha_1 v_1 + \dots + \alpha_n v_n - v = 0$$

contradicts the independence of  $S$ .

# Linear independence and minimality of span

## Lemma

*A set  $S \subseteq \mathbf{V}$  is linearly independent if and only if for every  $T \subsetneq S$ ,*

$$\text{span}(T) \neq \text{span}(S).$$

## Proof.

⇐ Suppose for a contradiction that

$$\alpha_1 v_1 + \dots + \alpha_n v_n = \mathbf{0}$$

for some  $v_1, \dots, v_n \in S$ , with  $\alpha_1 \neq 0$ . Then

$$v_1 = -\alpha_1^{-1}(\alpha_2 v_2 + \dots + \alpha_n v_n) \in \text{span}(S \setminus \{v_1\}),$$

and thus

$$\text{span}(S \setminus \{v_1\}) = \text{span}(S).$$

Let  $\mathbf{V}$  be a vector space.

## Definition

A set  $S \subseteq \mathbf{V}$  is a **basis** if

- $S$  generates  $\mathbf{V}$ , i.e.,  $\text{span}(S) = \mathbf{V}$ , and
- $S$  is linearly independent.

# Example

Each of the following sets

- $\{(1, 1, 0), (1, 2, 3)\}$
- $\{(1, 1, 0), (3, 4, 3)\}$
- $\{(1, 2, 3), (3, 4, 3)\}$

is a basis of the plane  $\{(x, y, z) : 3x - 3y + z = 0\}$ .

# Standard basis

The vectors

$$(1, 0, 0), (0, 1, 0), (0, 0, 1)$$

form a basis of  $\mathbf{R}^3$ .

More generally, the vectors

$$(1, 0, 0, \dots) = \left( \mathbf{e}_1^{(n)} \right)^T$$

$$(0, 1, 0, \dots) = \left( \mathbf{e}_2^{(n)} \right)^T$$

...

$$(0, 0, \dots, 0, 1) = \left( \mathbf{e}_n^{(n)} \right)^T$$

form the **standard basis** of  $\mathbf{R}^n$ .

# Does every vector space have a basis?

- “It depends.”
- Equivalent to the Axiom of Choice.
- E.g., no constructive way to obtain basis for the space of functions  $\mathbf{R} \rightarrow \mathbf{R}$ .

# Basis in a “nice” space

## Definition

A vector space  $\mathbf{V}$  is **countably generated** if there exists a (finite or infinite) sequence  $v_1, v_2, \dots \in \mathbf{V}$  such that

$$\mathbf{V} = \text{span}(v_1, v_2, \dots).$$

All spaces we considered are countably generated, except for

- the space of functions/continuous functions  $\mathbf{R} \rightarrow \mathbf{R}$
- the space of infinite sequences
- $\mathbf{R}$  as a vector space over  $\mathbf{Q}$

# Basis in a “nice” space

## Lemma

*Every countably generated space has a basis.*

## Proof.

If  $\mathbf{V} = \text{span}(v_1, v_2, \dots)$ , let

$$B = \{v_i : v_i \notin \text{span}(v_1, v_2, \dots, v_{i-1})\}.$$

We have  $\text{span}(B) = \text{span}(v_1, v_2, \dots) = \mathbf{V}$ .

If  $\alpha_1 v_{i_1} + \dots + \alpha_k v_{i_k} = \mathbf{0}$  for some  $v_{i_1}, \dots, v_{i_k} \in B$ , where  $\alpha_k \neq 0$  and  $i_1 < i_2 < \dots < i_k$ , then

$$v_{i_k} = -\alpha_k^{-1}(\alpha_1 v_{i_1} + \dots + \alpha_{k-1} v_{i_{k-1}}) \in \text{span}(v_1, \dots, v_{i_{k-1}}),$$

which contradicts the construction of  $B$ . □



# Transfer lemma

## Lemma (Transfer lemma)

*Suppose that  $S \subset \text{span}(T)$  is a linearly independent set. If  $\text{span}(S) \neq \text{span}(T)$ , then there exists  $v \in T \setminus S$  such that  $S \cup \{v\}$  is linearly independent.*

## Proof.

Since  $\text{span}(S) \neq \text{span}(T)$ , there exists  $v \in T \setminus \text{span}(S)$ .  
Suppose that

$$\alpha v + \alpha_1 v_1 + \dots + \alpha_n v_n = 0$$

for some  $v_1, \dots, v_n \in S$ . If  $\alpha \neq 0$ , then

$$v = -\alpha^{-1}(\alpha_1 v_1 + \dots + \alpha_n v_n) \in \text{span}(S), \text{ a contradiction.}$$

Hence  $\alpha = 0$ , and since  $S$  is linearly independent,  $\alpha_1 = \dots = \alpha_n = 0$ . Therefore,  $S \cup \{v\}$  is linearly independent. □

# Extension to basis

## Corollary (Extension lemma)

*If a vector space  $\mathbf{V}$  has a finite basis  $B$ , then every independent set is a subset of some basis.*

## Proof.

Let  $S$  be an independent set. Keep adding elements of  $B$  to  $S$  by the Transfer lemma until  $S$  becomes generating. □

# Exchange lemma

## Corollary (Exchange lemma)

Let  $S \subset \text{span}(T)$  be a linearly independent set. For every  $s \in S \setminus T$  there exists  $t \in T \setminus S$  such that

$$(S \setminus \{s\}) \cup \{t\}$$

is linearly independent.

## Proof.

Since  $S$  is linearly independent,

$$\text{span}(S \setminus \{s\}) \subsetneq \text{span}(S) \subseteq \text{span}(T).$$

By the Transfer Lemma, there exists  $t \in T \setminus S$  such that  $(S \setminus \{s\}) \cup \{t\}$  is linearly independent. □

# Generating and independent sets

## Lemma (Generating-independent inequality)

*Suppose that  $S$  and  $T$  are sets of vectors, where  $T$  is finite. If  $S \subseteq \text{span}(T)$  is linearly independent, then  $S$  is finite and  $|S| \leq |T|$ .*

## Proof.

- Using the Exchange lemma, replace elements of  $S \setminus T$  by elements of  $T \setminus S$  in  $S$  as long as possible (at most  $|T| \times$ ).
- In the end,  $S \subseteq T$ , and thus  $|S| \leq |T|$ .



# Sizes of bases

## Lemma

*If a vector space  $\mathbf{V}$  has a finite basis, then all its bases are finite and have the same size.*

## Proof.

Let  $B_1$  and  $B_2$  be two bases of  $\mathbf{V}$ , where  $B_1$  is finite.

- $\text{span}(B_1) = \mathbf{V}$  and  $B_2 \subseteq \mathbf{V}$  is linearly independent.
- By the Generating-independent inequality,  $|B_2| \leq |B_1|$ .
- Symmetrically,  $|B_1| \leq |B_2|$ .



## Definition

The **dimension**  $\dim(\mathbf{V})$  of a vector space is the size of its basis.

# Examples

- $\mathbf{R}^n$  has dimension  $n$ .
- $\mathbf{R}^{n \times m}$  has dimension  $nm$ .
- Complex numbers as a vector space over  $\mathbf{R}$  have dimension 2.
- The space of polynomials has infinite dimension.
- The space of polynomials of degree at most  $n$  has dimension  $n + 1$ .
- The trivial space  $\{0\}$  has dimension 0.

# Dimension, independent and generating sets

## Lemma

Let  $\mathbf{V}$  be a vector space of a finite dimension  $n$ .

- Every independent set in  $\mathbf{V}$  has size at most  $n$ , and *all independent sets of size  $n$  are bases*.
- Every set that generates  $\mathbf{V}$  has size at least  $n$ , and *all generating sets of size  $n$  are bases*.

## Proof.

Let  $S$  be independent,  $G$  generating.

- $|S| \leq n \leq |G|$  by the Generating-independent inequality.
- *If  $|S| = n$ , then no proper superset of  $S$  is independent.*
- *By the Transfer lemma,  $\text{span}(S) = \mathbf{V}$ .*



# Dimension, independent and generating sets

## Lemma

Let  $\mathbf{V}$  be a vector space of a finite dimension  $n$ .

- Every independent set in  $\mathbf{V}$  has size at most  $n$ , and *all independent sets of size  $n$  are bases*.
- Every set that generates  $\mathbf{V}$  has size at least  $n$ , and *all generating sets of size  $n$  are bases*.

## Proof.

Let  $S$  be independent,  $G$  generating.

- $|S| \leq n \leq |G|$  by the Generating-independent inequality.
- If  $|G| = n$ , then no proper subset of  $G$  is generating.
- Hence,  $\text{span}(A) \neq \text{span}(G)$  for every  $A \subsetneq G$ .
- Implies that  $G$  is linearly independent.





# Dimension and subspaces

## Lemma

*Suppose that  $\mathbf{V}$  has finite dimension, and  $\mathbf{U} \subseteq \mathbf{V}$ .*

- $\dim(\mathbf{U}) \leq \dim(\mathbf{V})$
- *If  $\dim(\mathbf{U}) = \dim(\mathbf{V})$ , then  $\mathbf{U} = \mathbf{V}$ .*

## Proof.

- Let  $B_U$  be a basis of  $\mathbf{U}$ .
- By the Extension lemma, we have a basis  $B_V \supseteq B_U$  of  $\mathbf{V}$ .
- $\dim(\mathbf{U}) = |B_U| \leq |B_V| = \dim(\mathbf{V})$
- If  $\dim(\mathbf{U}) = \dim(\mathbf{V})$ , then  $B_U = B_V$  and

$$\mathbf{U} = \text{span}(B_U) = \text{span}(B_V) = \mathbf{V}.$$



# Example: Dimension and subspaces

Subspaces of  $\mathbf{R}^3$ :

- Dimension 3:  $\mathbf{R}^3$
- Dimension 2: spans of 2 independent vectors = planes containing  $(0, 0, 0)$ .
- Dimension 1: spans of vectors = lines containing  $(0, 0, 0)$ .
- Dimension 0:  $\{(0, 0, 0)\}$

## Example: bases of polynomials

$\mathcal{P}_n$  has dimension  $n + 1$

- Basis  $1, x, x^2, \dots, x^n$ .

# Lagrange polynomials

Let  $a_0, \dots, a_n \in \mathbf{R}$  be pairwise distinct.

- For  $k = 0, \dots, n$ , let

$$p_k(x) = \frac{(x - a_0) \cdots (x - a_{k-1})(x - a_{k+1}) \cdots (x - a_n)}{(a_k - a_0) \cdots (a_k - a_{k-1})(a_k - a_{k+1}) \cdots (a_k - a_n)}.$$

- We have  $p_k(a_i) = \begin{cases} 1 & \text{if } i = k \\ 0 & \text{if } i \neq k \end{cases}$
- The set  $B = \{p_0, \dots, p_n\}$  is another basis of  $\mathcal{P}_n$ .
  - $|B| = \dim(\mathcal{P}_n)$
  - $B$  is linearly independent:

$$(\alpha_0 p_0 + \dots + \alpha_n p_n)(a_i) = \alpha_0 p_0(a_i) + \dots + \alpha_n p_n(a_i) = \alpha_i,$$

hence if  $\alpha_0 p_0 + \dots + \alpha_n p_n = \mathbf{o}$ , then  $\alpha_i = \mathbf{o}(a_i) = 0$  for  $i = 0, \dots, n$ .

# Polynomial interpolation

## Corollary (Polynomial interpolation lemma)

*A polynomial  $p$  of degree at most  $n$  is uniquely determined by its values in  $n + 1$  distinct points.*

## Proof.

- Since  $B$  generates  $\mathcal{P}_n$ , there exist  $\alpha_0, \dots, \alpha_n \in \mathbf{R}$  such that

$$p = \alpha_0 p_0 + \dots + \alpha_n p_n.$$

- For  $i = 0, \dots, n$ ,

$$p(a_i) = \alpha_0 p_0(a_i) + \dots + \alpha_n p_n(a_i) = \alpha_i.$$

- Therefore,

$$p = p(a_0)p_0 + \dots + p(a_n)p_n$$

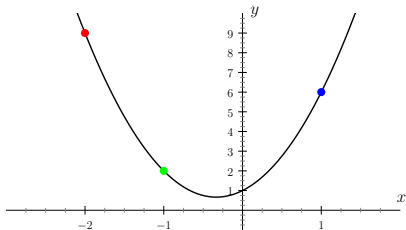
is uniquely determined by the values of  $p$  in  $a_0, \dots, a_n$ .

# Example

## Problem

Find the equation of a quadratic function through points

$(-2, 9)$ ,  $(-1, 2)$ , and  $(1, 6)$



$$\begin{aligned} & 9 \frac{(x+1)(x-1)}{(-2+1)(-2-1)} + 2 \frac{(x+2)(x-1)}{(-1+2)(-1-1)} + 6 \frac{(x+2)(x+1)}{(1+2)(1+1)} \\ & = 3x^2 + 2x + 1 \end{aligned}$$

# Vandermonde matrix

## Definition

For distinct real numbers  $a_0, \dots, a_n$ ,

$$V(a_0, \dots, a_n) = \begin{pmatrix} 1 & a_0 & a_0^2 & \dots & a_0^n \\ 1 & a_1 & a_1^2 & \dots & a_1^n \\ \dots & \dots & \dots & \dots & \dots \\ 1 & a_n & a_n^2 & \dots & a_n^n \end{pmatrix},$$

is a **Vandermonde matrix**.

For any polynomial  $p(x) = \beta_0 + \beta_1 x + \beta_2 x^2 + \dots + \beta_n x^n$ ,

$$V(a_0, \dots, a_n) \begin{pmatrix} \beta_0 \\ \beta_1 \\ \dots \\ \beta_n \end{pmatrix} = \begin{pmatrix} p(a_0) \\ p(a_1) \\ \dots \\ p(a_n) \end{pmatrix}$$

# Vandermonde matrix and polynomial interpolation

For  $b_0, \dots, b_n$ , if a polynomial

$$p(x) = \beta_0 + \beta_1 x + \beta_2 x^2 + \dots + \beta_n x^n$$

satisfies  $p(a_0) = b_0, p(a_1) = b_1, \dots, p(a_n) = b_n$ , then

$$V^{(a_0, \dots, a_n)} \begin{pmatrix} \beta_0 \\ \beta_1 \\ \dots \\ \beta_n \end{pmatrix} = \begin{pmatrix} b_0 \\ b_1 \\ \dots \\ b_n \end{pmatrix}$$

By the Polynomial interpolation lemma, this system always has a solution,

$$\beta_0 + \beta_1 x + \beta_2 x^2 + \dots + \beta_n x^n = b_0 p_0(x) + b_1 p_1(x) + \dots + b_n p_n(x).$$

## Corollary

*Every Vandermonde matrix is regular.*



# Linear recurrences

## Problem

Describe all infinite sequences  $a_0, a_1, \dots$  that satisfy

$$a_{n+2} = 5a_{n+1} - 6a_n \text{ for every } n \geq 0. \quad (1)$$

Let  $\mathcal{S}$  be the vector space of infinite sequences, and let  $U \subseteq \mathcal{S}$  consist of those satisfying (1). Then  $U$  is a subspace:

- $(0, 0, \dots) \in U$
- If  $A = (\alpha_0, \alpha_1, \dots) \in U$  and  $B = (\beta_0, \beta_1, \dots) \in U$ , and  $\gamma \in \mathbf{R}$ , then

$$\alpha_{n+2} + \beta_{n+2} = 5(\alpha_{n+1} + \beta_{n+1}) - 6(\alpha_n + \beta_n)$$

$$\gamma\alpha_{n+2} = 5\gamma\alpha_{n+1} - 6\gamma\alpha_n,$$

and thus  $A + B, \gamma A \in U$ .

# Linear recurrences

## Problem

*Describe all infinite sequences  $a_0, a_1, \dots$  that satisfy*

$$a_{n+2} = 5a_{n+1} - 6a_n \text{ for every } n \geq 0. \quad (1)$$

The choice of  $a_0$  and  $a_1$  uniquely determines the rest of the sequence. Hence,  $\dim(U) = 2$ . “Standard” basis:

- $a_0 = 0, a_1 = 1 \rightarrow (0, 1, 5, 19, 65, \dots)$
- $a_0 = 1, a_1 = 0 \rightarrow (1, 0, -6, -30, -114, \dots)$

# Linear recurrences

## Problem

Describe all infinite sequences  $a_0, a_1, \dots$  that satisfy

$$a_{n+2} = 5a_{n+1} - 6a_n \text{ for every } n \geq 0. \quad (1)$$

Nicer basis:

- $a_n = 2^n \rightarrow (1, 2, 4, 8, 16, \dots)$
- $a_n = 3^n \rightarrow (1, 3, 9, 27, 81, \dots)$

$$2^{n+2} = 4 \cdot 2^n = 10 \cdot 2^n - 6 \cdot 2^n = 5 \cdot 2^{n+1} - 6 \cdot 2^n$$

$$3^{n+2} = 9 \cdot 3^n = 15 \cdot 3^n - 6 \cdot 3^n = 5 \cdot 3^{n+1} - 6 \cdot 3^n$$

Therefore, for  $\alpha, \beta \in \mathbf{R}$ ,  $a_n = \alpha 2^n + \beta 3^n$  is a solution, and no other solutions exist.