# Bases and Dimension

Textbook: Section 1.6

Warning: These notes contain probably more content than we can cover in one week. Don't panic, I included some extra material for easy access and reference for you.

#### Definition 1.6.1

A family of vectors  $\mathcal{B}$  in a vector space V is called a basis of V if

- 1.  $\mathcal{B}$  spans V
- 2.  $\mathcal{B}$  is linearly independent

### Examples

- 1.  $\{\vec{e}_1, \dots, \vec{e}_n\}$  is a basis of  $\mathbb{R}^n$ .
- 2. We have seen last week that  $\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\}$  is a basis of  $\mathbb{R}^2$
- 3. Which families of polynomials of the ones we have seen before are bases of  $\mathcal{P}_2(\mathbb{R})$ ? Can you write down a basis of  $\mathcal{P}_2(\mathbb{R})$  that does not contain any monomials?

A vector space does not have just one unique basis as we can easily verify.

#### Discussion

Given the vector 
$$\begin{pmatrix} 5 \\ 3 \end{pmatrix} \in \mathbb{R}^2$$
, write down all linear combinations of  $\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\}$  equal to  $\begin{pmatrix} 5 \\ 3 \end{pmatrix} \in \mathbb{R}^2$ .

# Theorem 1.6.3

A family of vectors  $\mathcal{B}$  is a basis of V if and only if every vector  $\vec{v} \in V$  can be written uniquely as a linear combination of vectors in  $\mathcal{B}$ .

Proof.

# **Theorem** (Extend & Reduce pt. II)

Let V be a vector space with a finite spanning set.

- 1. Every linearly independent family S in V may be enlarged to a basis  $\mathcal{B}$  containing S. (Theorem 1.6.6)
- 2. Every family S that spans V may be reduced to a basis  $\mathcal B$  contained in S.

*Proof.* This follows from the *Extend* and *Reduce* theorems from last week

# Remark

- 1. Why do we need a finite spanning set for V? Some vector spaces, such as  $\mathcal{P}(\mathbb{R})$  can not be spanned by finitely many polynomials. Can you show why?
- 2. Observe that a basis hits a sweet spot as it is not too large to contain redundant vectors, but also not too small to not span the vector space.

# Corollary

Every finitely spanned vector space has a basis.

### Discussion

Consider the vectors p(x) = 1 + x and  $q(x) = 1 + x + x^2$  in  $\mathcal{P}_2(\mathbb{R})$ . Find a third vector r(x) such that the family  $\{p, q, r\}$  is a basis for  $\mathcal{P}_2(\mathbb{R})$ .

Even though a basis is not unique to a vector space, we would like to extract an invariant, a label, something that characterizes the vectors space. This invariant is motivated by the Corollary following below.

### Theorem 1.6.10

If V is spanned by a family S with m elements, then no linearly independent family R in V can have more than m elements.

Proof.

# Corollary 1.6.11

Any two bases  $\mathcal B$  and  $\mathcal B'$  of V have the same number of elements

# **Definitions**

- 1. If a vector space V has a finite basis, we say that V is finite dimensional.
- 2. For a finite dimensional vector space V, the dimension of V

$$\dim(V)$$

is the number of elements of a basis of V.

#### Discussion

What is the dimension of

$$\begin{aligned} \dim(\mathbb{R}) &= \\ \dim(\mathcal{P}_n(\mathbb{R})) &= \\ \dim(\mathrm{Mat}_2(\mathbb{R})) &= \\ \dim(\mathrm{Mat}_2^{\mathrm{sym}}(\mathbb{R})) &= \\ \dim(\mathrm{Mat}_2^{\mathrm{anti}}(\mathbb{R})) &= \end{aligned}$$

Remember that a matrix A is symmetric if  $A^T = A$  and antisymmetric if  $A^T = -A$ .

# Discussion

- Can you argue that if  $U \subseteq V$  is a subspace then  $\dim(U) \leq \dim(V)$ ?
- Is it on the contrary true that for every subspace  $U \subseteq V \dim(U) = \dim(V)$  implies U = V?

The following result is very useful when we are looking for a basis of a vector space that we already know the dimension of.

### **Proposition** (Corollary of last discussion)

Let V be a vector space of dimension n with a family S containing n vectors, then the following are equivalent.

- (a) S is a basis of V
- (b) S is linearly independent
- (c) S spans V

Proof.

### Discussion

is the family of polynomials  $\{1+2x-x^2,1+x+x^2\}$  is a basis for the subspace  $W=\{a+bx+cx^2\in\mathcal{P}_2(\mathbb{R})|-3a+2b+c=0\}$ ?

Since subspaces are vector spaces too, all constructions above apply also to subspaces.



Let U and W be subspaces in V, can the sum  $\dim(U) + \dim(W)$  be greater than  $\dim(V)$ ? Argue why it is true or find a counterexample.

# Example (continued)

Consider again the subspaces of symmetric and anti symmetric  $2 \times 2$  matrices,  $\operatorname{Mat_2^{sym}}(\mathbb{R})$  and  $\operatorname{Mat_2^{anti}}(\mathbb{R})$  respectively, in the vector space of all  $2 \times 2$  matrices  $\operatorname{Mat_2}(\mathbb{R})$ . Compute the intersection  $\operatorname{Mat_2^{sym}}(\mathbb{R}) \cap \operatorname{Mat_2^{anti}}(\mathbb{R})$ .

This example can be abstractly formalized as

### Theorem 1.6.18

Let U and W be finite dimensional subspaces of a vector space V, then

$$\dim(U+W) = \dim(U) + \dim(W) - \dim(U \cap W)$$

*Proof.* Read the proof in the textbook (voluntarily).

### Discussion

Let  $S = {\vec{v}_1, \vec{v}_2, \vec{v}_3}$  be a spanning set of V.

- 1. What are the possible dimensions of V?
- 2. Suppose  $\{v_1, v_3\}$  is linearly independent, what are the possible dimensions of V?
- 3. Can you create a basis for V from S?

With the results from this section, we can immediately obtain the following.

# Corollary & Discussion

Let V be an n-dimensional vector space and S a linearly independent family with l elements, then

- 1.  $l \leq n$
- 2. if l = n, S is a basis of V.

### Discussion

Let  $\vec{v}_1, \vec{v}_2, \vec{v}_3, \vec{v}_4$  be vectors in a vector space V and define the subspace  $U = \text{span}\{\vec{v}_1, \vec{v}_2, \vec{v}_3, \vec{v}_4\}$  Suppose

$$\vec{v}_3 = \vec{v}_1 - \vec{v}_2$$

$$\vec{v}_4 = 2\vec{v}_1 + 3\vec{v}_2 - \vec{v}_3$$

- 1. What are the possible dimensons of U?
- 2. Suppose dim(U) = 2. Does this imply that  $\{\vec{v}_3, \vec{v}_4\}$  is linealry independent?

Textbook: Section 2.1

# **Linear Transformatoins**

#### Definition 2.1.1

A function  $V \xrightarrow{T} W$  between vector spaces  $(V, +_V, \bullet_V)$  and  $(W, +_W, \bullet_W)$  is called *linear* if

- 1.  $T(\vec{u} +_V \vec{v}) = T(\vec{v}) +_W T(\vec{v})$  for all  $\vec{u}, \vec{v} \in V$
- 2.  $T(\alpha \bullet_V \vec{u}) = \alpha \bullet_W T(\vec{v})$  for all  $\vec{u} \in V$  and  $\alpha \in \mathbb{R}$ .

#### Remark

- We usually use the expression linear transformation or short transformation and not linear function.
- Notice that the operations + and  $\cdot$  are exactly what distinguishes vector spaces from sets. The requirements for a function to be linear guarantee that it mediates between the operations on the domain and the target.

### Examples

1. Consider the function  $x \mapsto e^x$ , are there vector space structures on  $\mathbb{R}$  such that this defines a linear transformation?

#### Discussion

Given a transformation  $V \xrightarrow{T} W$ , is it true that

$$T\left(\sum_{i=1}^{k} \alpha_i \vec{v}_i\right) = \sum_{i=1}^{k} \alpha_i T(\vec{v}_i) \qquad ?$$

### Discussion

Which of the following functions are linear transformations?

1. 
$$\mathbb{R}^2 \xrightarrow{T} \mathbb{R}$$
 defined by  $T\left(\begin{pmatrix} x \\ y \end{pmatrix}\right) = \max\{x, y\}$ 

2. The derivative 
$$\mathcal{P}_n(\mathbb{R}) \xrightarrow{\frac{d}{dx}} \mathcal{P}_n(\mathbb{R})$$

$$f(x) \mapsto \frac{df}{dx}(x)$$

- 3. The transposition of a matrix  $A \mapsto A^T$  as a function  $\operatorname{Mat}_n(\mathbb{R}) \xrightarrow{T} \operatorname{Mat}_n(\mathbb{R})$
- 4. The function  $\operatorname{Mat}_n(\mathbb{R}) \xrightarrow{RREF} \operatorname{Mat}_n(\mathbb{R})$  that computes the reduced row echolon form of a matrix  $A \mapsto \operatorname{RREF}(A)$ .
- 5. The evaluation function  $\mathcal{P}_n(\mathbb{R}) \xrightarrow{\operatorname{ev}_7} \mathbb{R}$  evaluating a polynomial at the value 7.

$$ev_7(p) = p(7)$$

### Proposition

For any linear transformation  $V \xrightarrow{T} W$  we have

1. 
$$T(\vec{0}_V) = \vec{0}_W$$

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

Is the converse also true?

# Definition

- 1. A transformation that is bijective is called an isomorphism. If there exists and isomorphism between two vector spaces V and W, they are called isomorphic.
- 2. A transformation  $V \xrightarrow{T} V$  for which domain and codomain agree is called an *endomorphism*.
- 3. Given two vector space V and W, the set of all linear transformations from V to W is denoted

$$\mathcal{L}(V, W)$$

or just 
$$\mathcal{L}(V)$$
 if  $V = W$ .

#### Exercise

Show that for any vector spaces V and W the set  $\mathcal{L}(V,W)$  is a vector space. Hint: Compare this to the vector space  $\mathcal{F}(\mathbb{R},\mathbb{R})$  of functions from  $\mathbb{R}$  to  $\mathbb{R}$ . Recall that a basis in a vector space enhales us to write every vector uniquely as a linear combination. Since a linear combination is determined by its coefficients, all we need to remember for a fixed basis is the n-tuple of coefficients. To formalize this, we define...

#### Definition

Let V be a vector space with a basis  $\alpha = \{\vec{b}_1, \dots, \vec{b}_n\}$  and for  $\vec{v} \in V$ 

$$\vec{v} = v_1 \cdot \vec{b}_1 + \dots + v_n \cdot \vec{b}_n$$

the unique representation in this basis. We call  $v_1, \ldots, v_n$  the coordinates of  $\vec{v}$  in the basis  $\alpha$  and  $[\vec{v}]_{\alpha} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$ 

the *coordinate vector* of  $\vec{v}$  in the basis  $\mathcal{B}$ .

### Discussion

Consider the basis  $\alpha = \{1, 1+x, 1+x+x^2\}$  of  $\mathcal{P}_2(\mathbb{R})$ .

- 1. Determine  $[1]_{\alpha}$ ,  $[x]_{\alpha}$ ,  $[x^2]_{\alpha}$  and  $[1+x]_{\alpha}$ .
- 2. Determine  $[a+bx+cx^2]_{\alpha}$  for arbitrary scalars  $a,b,c\in\mathbb{R}$  and show that

$$[a + bx + cx^{2}]_{\alpha} = a[1]_{\alpha} + b[x]_{\alpha} + c[x^{2}]_{\alpha}$$

#### Theorem

Given an *n*-dimensional vector space V with a basis  $\mathcal{B}$ . The assignment of its coordinate vector  $[\vec{v}]_{\alpha}$  to every vector  $\vec{v} \in V$  is a linear transformation

$$V \xrightarrow{\gamma^{\alpha}} \mathbb{R}^n$$

Even more is true,  $\gamma^{\alpha}$  is an isomorphism from V to  $\mathbb{R}^{n}$ .

Proof.

#### Intuition

Isomorphic vector spaces V and W can be regarde as practically the same vector space, because we can use the isomorphism to go back and forth between elements from V to W bijectively. Moreover, an isomorphism identifies the operations on V and W.

The above theorem says that every vector space of dimension n is isomorphic to  $\mathbb{R}^n$  and that the choice of a basis (there are many) tells us how to uniquely identify vectors  $\vec{v} \in V$  with n-tuples  $(v_1, \ldots, v_n) \in \mathbb{R}^n$ .

That is to say  $\mathbb{R}^n$  is the prototypical *n*-dimensional vector space.

# Discussion

Suppose  $\mathcal{P}_2(\mathbb{R}) \xrightarrow{T} \mathbb{R}$  is given by

$$T(1+x^2) = 5$$

$$T(x - x^2) = 3$$

$$T(1) = 1$$

What is T(x)?

# Proposition

A transformation  $V \xrightarrow{T} W$  is uniquely determined by its values on elements of a basis on V.

Proof.