

## Bases and Dimension

**Textbook:** Section 1.6

### 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 of families of polynomials we have seen before is a basis of  $\mathcal{P}_2(\mathbb{R})$ ? Can you write down a basis of  $\mathcal{P}_2(\mathbb{R})$  that does not contain any monomials?

### 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.* ■

### Remark

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

### Theorem 1.6.6

Let  $V$  be a vector space with a finite spanning set. For every linearly independent family  $S$  in  $V$ , there is a basis  $\mathcal{B}$  containing  $S$ .

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?

Observe that a basis hits a sweet spot of a family that is not too large as that it would contain redundant vectors, but also not too small of a family that it couldn't span the vector space.

The *Extend* and *Reduce* theorems from last week give us the following:

A family of vectors that is linearly independent, but not a spanning set can be enlarged to a basis, a family that spans  $V$ , but is linearly dependent, can be cut down to form a basis.

Even though a basis is not unique to a vector space, we would like to extract an invariant, a label, something that characterizes the vector 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

$$\dim(\mathbb{R}) =$$

$$\dim(\mathcal{P}_n(\mathbb{R})) =$$

$$\dim(\text{Mat}_2(\mathbb{R})) =$$

$$\dim(\text{Mat}_2^{\text{sym}}(\mathbb{R})) =$$

$$\dim(\text{Mat}_2^{\text{anti}}(\mathbb{R})) =$$

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 space that we already know the dimension of.

**Proposition**

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  $V$

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

**Discussion**

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

**Example** (continued)

Consider again the subspaces of symmetric and anti symmetric  $2 \times 2$  matrices,  $\text{Mat}_2^{\text{sym}}(\mathbb{R})$  and  $\text{Mat}_2^{\text{anti}}(\mathbb{R})$  respectively, in the vector space of all  $2 \times 2$  matrices  $\text{Mat}_2(\mathbb{R})$ . Compute the intersection  $\text{Mat}_2^{\text{sym}}(\mathbb{R}) \cap \text{Mat}_2^{\text{anti}}(\mathbb{R})$ .

As a final result, we will state without proof

**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). ■

**Corollary**

Every finite dimensional vector space has basis.

**Textbook:** Section 2.1

## Linear Transformations

### Definition 2.1.1

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

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

### Discussion

Given a transformation  $T: V \rightarrow W$ , is it true that

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

### Discussion

Which of the following functions are linear transformations?

1. The derivative of a polynomial  $\mathcal{P}_n(\mathbb{R}) \rightarrow \mathcal{P}_n(\mathbb{R})$   $p(x) \mapsto \frac{d}{dx}p(x)$

### Proposition

For any linear transformation  $T: V \rightarrow W$ ,  $T(\vec{0}_V) = \vec{0}_W$ .

### Proposition

A transformation  $T: V \rightarrow W$  is uniquely specified by its values on elements of a basis on  $V$ .

Recall that a basis in a vector space enables 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  $\mathcal{B} = \{\vec{b}_1, \dots, \vec{b}_n\}$  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  $[\vec{v}]_{\mathcal{B}} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$  the *coordinate vector* of  $\vec{v}$  in the basis  $\mathcal{B}$ .

**Proposition**

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

$$V \xrightarrow{\gamma^{\mathcal{B}}} \mathbb{R}^n$$