

# **MTHE 212 Notes**

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# Chapter 1

## Week 1

### 1.1 Lecture 1

#### 1.1.1 Introduction to Vector Spaces

In mathematics, we often study families of objects and the rules governing their interactions. In this course, we focus on **linear structures**, specifically **Vector Spaces**.

A vector space essentially consists of a collection of elements (vectors) and two operations:

1. **Addition:** Combines two elements to produce a third.
2. **Scalar Multiplication:** Scales an element by a number (scalar).

**Example.**

Consider  $\mathbb{R}^n$  (the set of column vectors with  $n$  real entries). Let  $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} \in \mathbb{R}^n$ .

The standard operations are defined component-wise:

- Addition:  $x + y = \begin{pmatrix} x_1 + y_1 \\ \vdots \\ x_n + y_n \end{pmatrix}$ .
- Scalar Multiplication: For  $\alpha \in \mathbb{R}$ ,  $\alpha x = \begin{pmatrix} \alpha x_1 \\ \vdots \\ \alpha x_n \end{pmatrix}$ .

The operations in  $\mathbb{R}^n$  satisfy 8 key properties:

1. Commutativity of addition:  $x + y = y + x$ .
2. Associativity of addition:  $(x + y) + z = x + (y + z)$ .
3. Additive identity: There exists  $0 \in \mathbb{R}^n$  such that  $x + 0 = x$ .

4. Additive inverse: For every  $x$ , there exists  $-x$  such that  $x + (-x) = 0$ .
5. Associativity of scalar multiplication:  $\alpha(\beta x) = (\alpha\beta)x$ .
6. Multiplicative identity:  $1 \cdot x = x$ .
7. Distributivity over vector sums:  $\alpha(x + y) = \alpha x + \alpha y$ .
8. Distributivity over scalar sums:  $(\alpha + \beta)x = \alpha x + \beta x$ .

### 1.1.2 Definition of a Vector Space

We abstract these properties to define a vector space generally.

#### Definition 1.1.1: Vector Space

A **vector space** is a set  $V$  equipped with two operations:

- Addition:  $+: V \times V \rightarrow V$
- Scalar Multiplication:  $\cdot: F \times V \rightarrow V$  (where  $F$  is a field of scalars, usually  $\mathbb{R}$  or  $\mathbb{C}$ )

satisfying the following axioms for all  $u, v, w \in V$  and  $\alpha, \beta \in F$ :

1.  $u + v = v + u$  (Commutativity)
2.  $(u + v) + w = u + (v + w)$  (Associativity)
3.  $\exists 0 \in V$  such that  $u + 0 = u$  (Additive Identity)
4.  $\forall u \in V, \exists -u \in V$  such that  $u + (-u) = 0$  (Additive Inverse)
5.  $1 \cdot u = u$  (Multiplicative Identity)
6.  $\alpha(\beta u) = (\alpha\beta)u$  (Associativity of Scalar Mult.)
7.  $\alpha(u + v) = \alpha u + \alpha v$  (Distributivity)
8.  $(\alpha + \beta)u = \alpha u + \beta u$  (Distributivity)

The scalars come from a field  $F$ .

- If  $F = \mathbb{R}$ , we call  $V$  a **real vector space**.
- If  $F = \mathbb{C}$ , we call  $V$  a **complex vector space**.

### 1.1.3 Examples and Non-Examples

**Example.**

Let  $V = \{(x, 0) \in \mathbb{R}^2 \mid x \in \mathbb{R}\}$ . This is the x-axis in  $\mathbb{R}^2$ . With standard component-wise operations,  $V$  is closed under addition and scalar multiplication, and inherits all axioms from  $\mathbb{R}^2$ . Thus,  $V$  is a vector space.

**Example.**

Let  $V = \{(x, 1) \in \mathbb{R}^2 \mid x \in \mathbb{R}\}$ . With standard operations:

- Addition:  $(x, 1) + (y, 1) = (x + y, 2) \notin V$ .
- Scalar Multiplication:  $\alpha(x, 1) = (\alpha x, \alpha) \notin V$  (unless  $\alpha = 1$ ).

Since  $V$  is not closed under these operations, it is **not** a vector space with standard operations. However, we could define *different* operations to make it a vector space (e.g., define addition such that the second component remains 1).

**Example.**

Consider the set of functions from a set  $S$  to a field  $F$ , denoted  $F^S = \{f : S \rightarrow F\}$ . Operations are defined pointwise:

- $(f + g)(x) = f(x) + g(x)$  for all  $x \in S$ .
- $(\alpha f)(x) = \alpha f(x)$  for all  $x \in S$ .

Under these operations,  $F^S$  is a vector space.

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This is where we ended Week 1, Lecture 1

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## 1.2 Lecture 2

### 1.2.1 Properties of Vector Spaces

We proved some basic propositions that follow directly from the axioms.

**Proposition 1.2.1**

For any  $v \in V$  and scalar  $a \in F$ :

1.  $a \cdot 0 = 0$  (where  $0$  is the zero vector).
2.  $(-1) \cdot v = -v$  (the additive inverse of  $v$ ).

**Proof.** 1.  $a \cdot 0 = a \cdot (0 + 0) = a \cdot 0 + a \cdot 0$ . Adding the additive inverse of  $a \cdot 0$  to both sides gives  $0 = a \cdot 0$ .

2. We verify that  $v + (-1)v = 0$ .  $v + (-1)v = 1 \cdot v + (-1)v = (1 + (-1))v = 0 \cdot v = 0$  (using the previous property  $0 \cdot v = 0$  which is similarly proved).  $\square$

### 1.2.2 Subspaces

Often we find vector spaces inside other vector spaces.

**Definition 1.2.2: Subspace**

Let  $V$  be a vector space. A subset  $U \subseteq V$  is a **subspace** if  $U$  is itself a vector space under the same operations as  $V$ , sharing the same zero vector.

To check if a subset is a subspace, we don't need to verify all 8 axioms. We can use the Subspace Test.

**Proposition 1.2.3****Subspace Test**

A subset  $U \subseteq V$  is a subspace if and only if:

1.  $0 \in U$  (Contains the zero vector).
2.  $U$  is closed under addition ( $u, v \in U \implies u + v \in U$ ).
3.  $U$  is closed under scalar multiplication ( $u \in U, \lambda \in F \implies \lambda u \in U$ ).

**Example.**

Let  $U = \{(x_1, x_2, x_3, x_4) \in F^4 \mid 2x_1 - x_2 + x_4 = 0\}$ .

- $0 \in U$  because  $2(0) - 0 + 0 = 0$ .
- Closed under addition: If  $x, y \in U$ , then  $2(x_1 + y_1) - (x_2 + y_2) + (x_4 + y_4) = (2x_1 - x_2 + x_4) + (2y_1 - y_2 + y_4) = 0 + 0 = 0$ . So  $x + y \in U$ .
- Closed under scalar multiplication (exercises).

Thus  $U$  is a subspace of  $F^4$ .

**Example.**

$C(\mathbb{R})$ , the set of continuous real-valued functions, is a subspace of  $\mathbb{R}^{\mathbb{R}}$  (all real functions).  
 $D(\mathbb{R})$ , the set of differentiable functions, is a subspace of  $C(\mathbb{R})$ .

**Proposition 1.2.4**

The intersection of any collection of subspaces is a subspace.

**1.2.3 Sums of Subspaces****Definition 1.2.5: Sum of Subspaces**

Let  $V_1, \dots, V_k$  be subspaces of  $V$ . The sum is defined as:

$$V_1 + \dots + V_k = \{v_1 + \dots + v_k \mid v_i \in V_i\}$$

**Proposition 1.2.6**

The sum  $V_1 + \cdots + V_k$  is a subspace of  $V$ . Moreover, it is the **smallest** subspace containing all  $V_1, \dots, V_k$ .

**1.2.4 Direct Sums**

Sometimes the representation of an element in a sum is not unique. If it is unique, we call it a direct sum.

**Definition 1.2.7: Direct Sum**

The sum  $V_1 + \cdots + V_k$  is a **direct sum**, denoted  $V_1 \oplus \cdots \oplus V_k$ , if every element  $v \in V_1 + \cdots + V_k$  can be written in a **unique** way as  $v = v_1 + \cdots + v_k$  with  $v_i \in V_i$ .

**Proposition 1.2.8**

The sum  $V_1 + \cdots + V_k$  is direct if and only if the zero vector  $0$  has a unique representation:

$$0 = 0 + \cdots + 0$$

**Proposition 1.2.9**

For two subspaces  $U, W$ , the sum  $U + W$  is direct if and only if  $U \cap W = \{0\}$ .

**1.2.5 Span****Definition 1.2.10: Linear Combination**

A linear combination of vectors  $v_1, \dots, v_k$  is a vector of the form  $c_1v_1 + \cdots + c_kv_k$  where  $c_i \in F$ .

**Definition 1.2.11: Span**

The span of  $v_1, \dots, v_k$ , denoted  $\text{span}(v_1, \dots, v_k)$ , is the set of all linear combinations of these vectors.