

# Module 3: Linear Algebra I

## Operational math bootcamp



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# Outline

- Vector spaces and subspaces
- Linear combinations and bases
- Linear transformations

## Definition

We call  $V$  a **vector space** if the following hold:

- (A) *Commutativity in addition:*  $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$  for all  $\mathbf{u}, \mathbf{v} \in V$
- (B) *Associativity in addition:*  $\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$  for all  $\mathbf{u}, \mathbf{v}, \mathbf{w} \in V$
- (C) *Existence of a neutral element, addition:* There exists a vector  $\mathbf{0}$  such that for any  $\mathbf{v} \in V$ ,  $\mathbf{0} + \mathbf{v} = \mathbf{v}$
- (D) *Additive inverse:* For every  $\mathbf{v} \in V$ , there exists another vector, which we denote  $-\mathbf{v}$ , such that  $\mathbf{v} + (-\mathbf{v}) = \mathbf{0}$ .
- (E) *Existence of a neutral element, multiplication:* For any  $\mathbf{v} \in V$ ,  $1 \times \mathbf{v} = \mathbf{v}$
- (F) *Associativity in multiplication:* Let  $\alpha, \beta \in \mathbb{F}$ . For any  $\mathbf{v} \in V$ ,  $(\alpha\beta)\mathbf{v} = \alpha(\beta\mathbf{v})$
- (G) Let  $\alpha \in \mathbb{F}, \mathbf{u}, \mathbf{v} \in V$ .  $\alpha(\mathbf{u} + \mathbf{v}) = \alpha\mathbf{u} + \alpha\mathbf{v}$ .
- (H) Let  $\alpha, \beta \in \mathbb{F}, \mathbf{v} \in V$ .  $(\alpha + \beta)\mathbf{v} = \alpha\mathbf{v} + \beta\mathbf{v}$ .

## Definition

A subset  $U$  of  $V$  is called a **subspace** of  $V$  if  $U$  is also a vector space (using the same addition and scalar multiplication as on  $V$ ).

## Proposition

A subset  $U$  of  $V$  is a subspace of  $V$  if and only if  $U$  satisfies the following three conditions:

- ①  $\mathbf{0} \in U$
- ② Closed under addition:  $u, w \in U$  implies  $\mathbf{u} + \mathbf{v} \in U$
- ③ Closed under scalar multiplication:  $\alpha \in \mathbb{F}$  and  $u \in U$  implies  $\alpha \mathbf{u} \in U$

## Proof.

$\Rightarrow$  If  $U$  is a subspace of  $V$ , then  $U$  satisfies these 3 properties by the definition of a vector space.

$\Leftarrow$  Suppose  $U$  satisfies the given 3 conditions.

Then for any  $\mathbf{v} \in U$ , there must exist  $-\mathbf{v} \in U$  by property 3, since  $-\mathbf{v} = (-1) \times \mathbf{v}$  (exercise). Property 1 assures property C. Properties 2 and 3, and the fact that  $U \subset V$ , assure the remaining properties hold. □

# Linear combinations

## Definition

A linear combination of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$  of vectors in  $V$  is a vector of the form

$$\alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n = \sum_{k=1}^n \alpha_k \mathbf{v}_k$$

where  $\alpha_1, \dots, \alpha_m \in \mathbb{F}$ .

# Span

## Definition

The set of all linear combinations of a list of vectors  $v_1, \dots, v_m$  in  $V$  is called the **span** of  $v_1, \dots, v_m$ , denoted  $\text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ . In other words,

$$\text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\} = \{\alpha_1 \mathbf{v}_1 + \dots + \alpha_m \mathbf{v}_n : \alpha_1, \dots, \alpha_n \in \mathbb{F}\}$$

The span of the empty list is defined to be  $\{\mathbf{0}\}$ .

We say a vector space is *finite dimensional* if it can be spanned by a finite list of vectors; otherwise it is *infinite dimensional*.

# Linear independence

## Definition

A list of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_n \in \mathbb{F}^n$  is said to be *linearly independent* if

$$0 = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

where the  $\alpha_i$ ,  $i = 1, \dots, n$  are scalars, admits only the solution  $\alpha_1 = \dots = \alpha_n = 0$ .

Otherwise we say the vectors are *linearly dependent*.



# Basis

## Definition

A system of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$  is called a basis (for the vector space  $V$ ) if any vector  $\mathbf{v} \in V$  admits a unique representation as a linear combination

$$\mathbf{v} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n = \sum_{k=1}^n \alpha_k \mathbf{v}_k$$

In undergrad, you likely thought about this as: the equation  $\mathbf{v} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$ , where the  $\alpha_i$  are unknown, has a unique solution.

## Claim

All bases of a vector space  $V$  have the same length.

Proof.

## Definition

The *dimension* of a vector space  $V$ , denoted  $\dim V$ , is the length of any basis of  $V$ .



# Bases

Example of bases:

For  $\mathbb{R}^n$ :  $e_1 = (1, 0, \dots, 0)$ ,  $e_2 = (0, 1, 0, \dots, 0)$ ,  $\dots$ ,  $e_n = (0, \dots, 0, 1)$

For  $\mathbb{P}^n$ :  $1, x, x^2, \dots, x^n$

## Definition

The linear combination  $\alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$  is called trivial if  $\alpha_k = 0$  for every  $k$ .

## proposition

A system of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$  is a basis if and only if it is linearly independent and complete (generating).

# Linear transformations

## Definition

A **transformation**  $T$  from domain  $X$  to codomain  $Y$  is a rule that assigns an output  $y = T(x) \in Y$  to each input  $x \in X$

## Definition

A transformation from a vector space  $U$  to a vector space  $V$  is **linear** if

$$T(\alpha \mathbf{u} + \beta \mathbf{v}) = \alpha T(\mathbf{u}) + \beta T(\mathbf{v}) \quad \text{for any } \mathbf{u}, \mathbf{v} \in V, \alpha, \beta \in \mathbb{F}$$

# Examples

- Differentiation
- Integration
- Rotation of vectors
- Reflection of vectors



# Linear transformations

## Definition

Let  $T : U \rightarrow V$  be a linear transformation. We define the following important subspaces:

- *Kernel or Null space:*

$$\text{Ker } T = \{\mathbf{u} \in U : T\mathbf{u} = \mathbf{0}\}$$

- *Range*

$$\text{Range } T = \{\mathbf{v} \in V : \exists \mathbf{u} \in U \text{ such that } \mathbf{v} = T\mathbf{u}\}$$

The dimensions of these spaces are often called the following:

- *Nullity*

$$\text{Nullity}(T) = \dim(\text{Ker}(T))$$

- *Rank*

$$\text{Rank}(T) = \dim(\text{Range}(T))$$

# Linear transformations

## Rank Theorem

For a matrix  $A$  or equivalently a linear transformation  $A : \mathbb{F}^n \rightarrow \mathbb{F}^m$ :

$$\text{Rank } A = \text{Rank } A^T$$

## Rank Nullity Theorem

Let  $T : U \rightarrow V$  be a linear transformation, where  $U$  and  $V$  are finite-dimensional vector spaces. Then

$$\text{Rank } T + \text{Nullity } T = \dim U.$$

# Exercises

- ① Let  $U$  and  $V$  be finite-dimensional vector spaces of the same dimension and let  $T : U \rightarrow V$  be a linear map. Prove that the following are equivalent:
- $T$  is bijective
  - $T$  is injective
  - $T$  is surjective



# References

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