Day 2: 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 ${\bf 0}$ such that for any ${\bf v} \in V$, ${\bf 0} + {\bf v} = {\bf v}$
- (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} , $\mathbf{v} \in V$. $\alpha(\mathbf{u} + \mathbf{v}) = \alpha \mathbf{u} + \beta \mathbf{v}$.
- (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 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 0** ∈ *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 $\mathbf{v}_1,...,\mathbf{v}_n$ 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 $\alpha_1, ..., \alpha_m \in \mathbb{F}$.



Span

Definition

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

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

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

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

$$0 = \alpha_i \mathbf{v}_i + \ldots + \alpha_n \mathbf{v}_n,$$

where the α_i , $i=1,\ldots,n$ are scalars, admits only the solution $\alpha_1=\cdots=\alpha_n=0$.

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

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

In undergrad, you likely thought about this as: the equation $\mathbf{v} = \alpha_1 \mathbf{v}_1 + \ldots + \alpha_n \mathbf{v}_n$ where the α_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 $\alpha_1 \mathbf{v}_1 + ... + \alpha_n \mathbf{v}_n$ 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

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



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Examples

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



Linear transformations

Definition

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

• Kernel or Null space:

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

Range

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

$$Nullity(T) = dim(Ker(T))$$

Rank

$$Rank(T) = dim(Range(T))$$



Linear transformations

Rank Theorem

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

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

Rank Nullity Theorem

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

Rank T + Nullity T = dim U.



Exercises

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



References

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