Linear algebra

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Contents

1	Vector spaces			
	1.1	Linear maps	3	
	1.2	Subspaces, span, and basis	3	
\mathbf{R}_{0}	efere	nces	5	

1 Vector spaces

Until the 19th century, linear algebra was introduced through systems of linear equations and matrices. In modern mathematics, the presentation through vector spaces is generally preferred, since it is more synthetic, more general (not limited to the finite-dimensional case), and conceptually simpler, although more abstract.

A vector space over a field F (often the field of the real numbers) is a set V equipped with two binary operations satisfying the following axioms. Elements of V are called vectors, and elements of F are called scalars. The first operation, vector addition, takes any two vectors v and w and outputs a third vector $\mathbf{v} + \mathbf{w}$. The second operation, scalar multiplication, takes any scalar a and any vector v and outputs a new vector av. The axioms that addition and scalar multiplication must satisfy are the following. (In the list below, u, v and w are arbitrary elements of V, and a and b are arbitrary scalars in the field F.)[2]

Axiom	Signification
Associativity of addition	u + (v + w) = (u + v) + w
Commutativity of addition	u + v = v + u
Identity element of addition	There exists an element 0 in V,
	called the zero vector (or simply zero),
	such that $v + 0 = v$ for all v in V.

The first four axioms mean that V is an abelian group under addition.

An element of a specific vector space may have various nature; for example, it could be a sequence, a function, a polynomial or a matrix. Linear algebra is concerned with those properties of such objects that are common to all vector spaces.

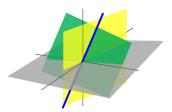


Figure 1: In three-dimensional Euclidean space, these three planes represent solutions of linear equations, and their intersection represents the set of common solutions: in this case, a unique point. The blue line is the common solution to two of these equations.

1.1 Linear maps

Linear maps are mappings between vector spaces that preserve the vectorspace structure. Given two vector spaces V and W over a field F, a linear map (also called, in some contexts, linear transformation or linear mapping) is a map

$$T:V\to W$$

that is compatible with addition and scalar multiplication, that is

$$T(u+v) = T(u) + T(v), \quad T(av) = aT(v)$$

for any vectors u,v in V and scalar a in F.

This implies that for any vectors u, v in V and scalars a, b in F, one has

$$T(au + bv) = T(au) + T(bv) = aT(u) + bT(v)$$

When V=W are the same vector space, a linear map $T:V\to V$ is also known as a linear operator on V.

A bijective linear map between two vector spaces (that is, every vector from the second space is associated with exactly one in the first) is an isomorphism. Because an isomorphism preserves linear structure, two isomorphic vector spaces are "essentially the same" from the linear algebra point of view, in the sense that they cannot be distinguished by using vector space properties. An essential question in linear algebra is testing whether a linear map is an isomorphism or not, and, if it is not an isomorphism, finding its range (or image) and the set of elements that are mapped to the zero vector, called the kernel of the map. All these questions can be solved by using Gaussian elimination or some variant of this algorithm.

1.2 Subspaces, span, and basis

The study of those subsets of vector spaces that are in themselves vector spaces under the induced operations is fundamental, similarly as for many mathematical structures. These subsets are called linear subspaces. More precisely, a linear subspace of a vector space V over a field F is a subset W of V such that u+v and au are in W, for every u,v in W, and every a in F. (These conditions suffice for implying that W is a vector space.)

For example, given a linear map $T:V\to W$, the image T(V) of V, and the inverse image $T^{-1}(0)$ of 0 (called kernel or null space), are linear subspaces of W and V, respectively. Another important way of forming a

subspace is to consider linear combinations of a set S of vectors: the set of all sums

$$a_1v_1 + a_2v_2 + \cdots + a_kv_k$$

where $v_1 + v_2 + \cdots + v_k$ are in S, and $a_1 + a_2 + \cdots + a_k$, ak are in F form a linear subspace called the span of S. The span of S is also the intersection of all linear subspaces containing S. In other words, it is the (smallest for the inclusion relation) linear subspace containing S.

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A set of vectors that spans a vector space is called a spanning set or generating set. If a spanning set S is linearly dependent (that is not linearly independent), then some element w of S is in the span of the other elements of S, and the span would remain the same if one remove w from S. One may continue to remove elements of S until getting a linearly independent spanning set. Such a linearly independent set that spans a vector space V is called a basis of V. The importance of bases lies in the fact that there are together minimal generating sets and maximal independent sets. More precisely, if S is a linearly independent set, and T is a spanning set such that $S \subseteq T$, then there is a basis S such that $S \subseteq T$.

Any two bases of a vector space V have the same cardinality, which is called the dimension of V; this is the dimension theorem for vector spaces. Moreover, two vector spaces over the same field F are isomorphic if and only if they have the same dimensio.[3]

If any basis of V (and therefore every basis) has a finite number of elements, V is a finite-dimensional vector space. If U is a subspace of V, then $dimU \leq dimV$. In the case where V is finite-dimensional, the equality of the dimensions implies U = V.

If U_1 and U_2 are subspaces of V, then

$$\dim(U_1 + U_2) = \dim U_1 + \dim U_2 - \dim(U_1 \cap U_2),$$

where $U_1 + U_2$ denotes the span of $U_1 \cup U_2$.[4]

References

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- [3] Roman (2005, ch. 1, p. 27)
- [4] Axler (2004, p. 55)