Linear Algebra

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1 Vector Spaces

1.1 Basics

Definition 1.1.1. We say that V is a vector space over a field \mathbf{F} if and only if for all $u, v, w \in V$ and all $a, b \in \mathbf{F}$

- 1. v + u = u + v;
- 2. (v + u) + w = v + (u + w);
- 3. There is an element $0 \in V$ such that v + 0 = v;
- 4. There is an element x such that v + x = 0;
- 5. 1v = v;
- 6. a(v + w) = av + aw and (a + b)v = av + bv.

Proposition 1.1.1. Every vector field has an unique additive identity.

Proof. Let 0 and 0' be additive identities in V. Then 0 = 0 + 0' = 0' + 0 = 0'. \square

Proposition 1.1.2. For all $v \in V$ there is an unique w such that v + w = 0.

Proof. Choose some arbitrary $v \in V$ and assume that v + w = v + w' = 0. Then w = (v + w') + w = (v + w) + w' = w'.

Definition 1.1.2. For every $v \in V$ we define -v as the unique $w \in V$ such that v + w = 0. We also define v - u as v + (-u).

Proposition 1.1.3. -(-v) = v

Proof. By definition we have v + (-v) = 0, so -v + v = 0, therefore v is the additive inverse of -v, thus -(-v) = v.

Definition 1.1.3. A subset U of V is called a subspace of V if and only if U is also a vector space.

Proposition 1.1.4. A subset U of V is a subspace of V if and only if U is

- 1. Closed under vector addition;
- 2. Closed under scalar multiplication;
- 3. Contains the additive identity of V.

2 Finite-Dimensional Vector Spaces

2.1 Span and linear independence

Definition 2.1.1. A list of vectors v_1, \ldots, v_m in the vector space V is linearly independent if and only if the following holds: $0 = a_1v_1 + a_2v_2 + \cdots + a_mv_m \implies a_1 = a_2 = \cdots = a_m = 0$. The empty list is also linearly independent. A list that is not linearly independent is called linearly dependent.

Definition 2.1.2. If v_1, \ldots, v_m is a list of vectors in V, we define

$$\mathrm{span}(v_1,\ldots,v_m) := \{a_1v_1 + \cdots + a_mv_m : a_1,a_2,\ldots,a_m \in F\}.$$

The span of the empty list is $\{0\}$. If the span of a list is V the list is called spanning.

Proposition 2.1.1. If a list is linearly independent then every vector in its span can be written as a linear combination of the vectors in the list in exactly one way.

Proof. We prove by contrapositive, so let v_1, \ldots, v_m be a list of vectors and assume that

$$v = a_1 v_1 + \dots + a_m v_m$$
$$v = b_1 v_1 + \dots + b_m v_m$$

where $a_j \neq b_j$ for some $j \in 1, ..., m$. Subtracting the equations we get

$$0 = (a_1 - b_1)v_1 + \dots + (a_j - b_j)v_j + \dots + (a_m - b_m)v_m.$$

Since $a_j \neq b_j$ we have $a_j - b_j \neq 0$, so not all the coefficients are equal to zero. Therefore the list is linearly dependent.

Corollary 2.1.0.1. A list is linearly independent if and only if none of its vectors are a linear combination of the other.

Definition 2.1.3. A vector space is called finite dimensional if and only if it contains a spanning list.

Lemma 2.1.1 (Linear Dependence Lemma). Let v_1, \ldots, v_m be a linearly dependent list of vectors. Then there is some $j \in 1, \ldots, m$ such that

- 1. $v_j \in \text{span}(v_1, \dots, v_{j-1})$
- 2. The list v_1, \ldots, v_m with the jth term removed preserves its original span.

Proof. Since the list is linearly dependent we have $a_1v_1 + \cdots + a_mv_m = 0$ where not all of the coefficients are zero. Let j be the largest number where $a_j \neq 0$. Then

$$v_j = \frac{-a_1}{a_j}v_1 + \dots + \frac{-a_{j-1}}{a_j}v_m$$

so $v_j \in \operatorname{span}(v_1, \dots, v_{j-1})$. The second item also follows easily from the previous equation.

Lemma 2.1.2. In a finite-dimensional vector space, every spanning list is at least as big as every linearly independent list.

Proof. Let u_1, \ldots, u_n be a linearly independent list and w_1, \ldots, w_m be spanning. We need to show that $n \leq m$. We do so in a multi-step process, as follows:

Step 1: Since w_1, \ldots, w_m spans V, the list u_1, w_1, \ldots, w_m is linearly dependent. By the Linear Dependence Lemma, we can remove one of the items in this list in such a way that it still spans V. That item cannot be u_1 , otherwise u_1 would have to be zero, making u_1, \ldots, u_n linearly dependent. Now we have a list u_1, w_1, \ldots, w_m with one of the w's removed, thus the list still has length m.

Step j: We add u_j to the list so that it now looks like $u_1, \ldots, u_j, w_1, \ldots, w_m$ and has length n+1. By the linear dependence lemma we can remove one of the vectors from the list while keeping it spanning. This vector cannot be one of the u's, since that would again imply that u_1, \ldots, u_n is linearly dependent. Thus another w is removed.

After step n the list looks like u_1, \ldots, u_j and the process stops. At every step one of the w's could be removed, so there were at least as many of them as there were u's.

Definition 2.1.4. A vector space is finite-dimensional if and only if there is a list of vectors in the space that spans it.

2.2 Bases

Definition 2.2.1. A list of vectors in V is a basis for V if and only if it is linearly independent and spans V.

Proposition 2.2.1. A list of vectors in V is a basis if and only if every vector in V can be written as a linearly combination of the vectors in the list in exactly one way.

Proposition 2.2.2. Every spanning list can be reduced to a basis.

Proof. Let v_1, \ldots, v_m be a spanning list. If it is linearly independent then we are done, so assume it is not. We can apply the Linearly Dependence lemma repeatedly to remove v_i 's until the remaining list is linearly independent while keeping it spanning.

Proposition 2.2.3. Every finite dimensional vector space has a basis.

Proof. By definition, the finite dimensional vector space V has a spanning list. This list can then be reduced to a basis.

Proposition 2.2.4. Every linearly independent list in a finite dimensional vector space can be extended to a basis.

Proof. Consider the linearly independent list v_1, \ldots, v_m in the finite-dimensional vector space V. Append to it the spanning list u_1, \ldots, u_n so that we now have the list $v_1, \ldots, v_m, u_1, \ldots, u_n$ and reduce this list to a basis. None of the v's is removed in the process, as is guaranteed by the Linear Dependence Lemma. \square

2.3 Dimension

Lemma 2.3.1. Every basis of a finite-dimensional vector space has the same length.

Proof. Let B_1 and B_2 be bases for V. Since B_1 is linearly independent and B_2 is spanning, the length of B_1 is less than or equal to the length of B_2 . The same argument applies in the other direction, therefore B_1 and B_2 have the same length.

Definition 2.3.1. Let V be a finite-dimensional vector space. We define the dimension $\dim(V)$ of V as the length of any basis of V.

Proposition 2.3.1. If v_1, \ldots, v_m is linearly independent and $\dim(V) = m$ then v_1, \ldots, v_m is a basis for V.

Proof. Since v_1, \ldots, v_m is linearly independent and V is finite dimensional the list can be extended to a basis. However the resulting list has to have length m, so the extension is the trivial one and the list is left unchanged, so the original list is a basis.

Proposition 2.3.2. If V is finite-dimensional and U is a subspace of V then U is finite-dimensional and $\dim(U) \leq \dim(V)$.

Proof. If $U = \{0\}$ the result follows trivially, so assume otherwise. At step 1, choose some non-zero vector $u_1 \in U$. The list (u_1) is clearly linearly independent. At step j, if $\operatorname{span}(u_1, \ldots, u_{j-1}) \neq U$ then add some $u_j \in U \setminus \operatorname{span}(u_1, \ldots, u_{j-1})$ to the end of the list.

Since the list being generated is linearly independent in V, its length has to be less than or equal to dim V, so the process eventually terminates, thus the linearly independent list created spans all of U, and is therefore a basis. Clearly its length is less than dim(V), as wanted.

Proposition 2.3.3. If V is finite-dimensional, U is a subspace of V and $\dim U = \dim V$ then U = V.

Proof. Let $u = u_1, \ldots, u_n$ be a basis for U. This is a linearly dependent list with length dim $U = \dim V$, therefore it is a basis for V. Thus given any $v \in V$ we have $v \in \operatorname{span}(u)$, thus v is a linear combination of vectors in U, therefore $v \in U$. Since $U \subset V$ and $V \subset U$, we have U = V.

Proposition 2.3.4. $\dim(V) = 0$ if and only if $V = \{0\}$.

Exercise 2.3.1. Proven in Proposition 2.3.3.

Exercise 2.3.2. Let U be a subspace of \mathbf{R}^2 . By Proposition 2.3.2, dim $U \leq \dim \mathbf{R}^2 = 2$, so dim $U \in \{0,1,2\}$. if dim U = 2 then $U = \mathbf{R}^2$ by Proposition 2.3.3. If dim U = 0 then $U = \{0\}$. If dim U = 1 then choose any non-zero $(a,b) \in U$. Then the list containing just (a,b) is linearly independent and has length 1, so it is a basis for U, therefore U is a line through the origin.

3 Linear Maps

3.1 The Vector Space of Linear Maps

Definition 3.1.1. A function $T: V \to W$ is a linear map if and only if

- 1. T(u+v) = T(u) + T(v) for all $u, v \in V$;
- 2. $T(\lambda v) = \lambda T(v)$ for all $v \in V$ and all $\lambda \in \mathbf{F}$.

Definition 3.1.2. $\mathcal{L}(V, W)$ is the set of all linear maps from V to W.

3.2 Null Spaces and Ranges

Definition 3.2.1. Let $T:V\to W$ be a linear map. We define $\operatorname{null}(T):=\{v\in V:T(v)=0\}.$

Proposition 3.2.1. A linear map $T: V \to W$ is injective if and only if $null(T) = \{0\}$.

Proof. First assume that T is injective. Notice that T(0) = T(0+0) = T(0) + T(0), therefore T(0) = 0. By the injectivity of T, no other input maps to T.

For the converse direction assume $\operatorname{null}(T) = \{0\}$ and that T(u) = T(v). Then T(u) - T(v) = 0 = T(u - v), thus u - v = 0 and u = v, so T is injective. \square

Theorem 3.2.1 (Fundamental Theorem of Linear Maps). Let V, W be finite-dimensional vector spaces and $T: V \to W$ be a linear map. Then $\dim V = \dim \operatorname{null} T + \dim \operatorname{range} T$.