

# MA2102: LINEAR ALGEBRA

## Lecture 6: Linear Independence

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As observed in previous examples, we may eliminate an element from a linearly dependent set without changing the span.

**Observation** *Let  $S$  be a linearly dependent subset of a vector space  $V$ . Then there exists  $v \in S$  such that  $\text{span}(S) = \text{span}(S - \{v\})$ .*

**Proof.**

If  $0 \in S$ , then  $\text{span}(S) = \text{span}(S - \{0\})$ . Now suppose that  $0 \notin S$ . As  $S$  is linearly dependent, there exists  $v_1, \dots, v_k \in S$  and scalars  $c_i$ 's, not all zero, such that

$$c_1 v_1 + \dots + c_k v_k = 0.$$

By rearranging the indices, we may assume that  $c_1 \neq 0$ . Then

$$v_1 = -\frac{c_2}{c_1} v_2 - \dots - \frac{c_k}{c_1} v_k.$$

Now it follows that  $\text{span}(S) = \text{span}(S - \{v_1\})$ . □

The observation implies that if  $S$  is a set such that no proper subset of it spans  $\text{span}(S)$ , then  $S$  is linearly independent. Note that a subset  $S$  is linearly independent if and only if there is no non-trivial linear combination of elements in  $S$  that equal  $\mathbf{0}$ .

Show that if  $S$  is a linearly independent set, then any proper subset  $T$  of  $S$  will span a strictly smaller subspace of  $\text{span}(S)$ , i.e.,  $\text{span}(T) \subset \text{span}(S)$ .

**Convention** Recall that we had defined  $\text{span}(\emptyset) = \{\mathbf{0}\}$ .

(i) The empty set  $\emptyset$  is linearly independent, as linearly dependent sets are necessarily non-empty.

(ii) If we put  $S_1 = \emptyset$  in

$$\text{span}(S_1 \cup S_2) = \text{span}(S_1) + \text{span}(S_2)$$

which was proved for non-empty sets, then we get

$$\text{span}(S_2) = \text{span}(\emptyset) + \text{span}(S_2)$$

implying  $\text{span}(\emptyset) = \{0\}$ .

(iii) If we want  $\text{span}(S)$  to be a subspace, then  $0 \in \text{span}(S)$ .

**Theorem** *Let  $S$  be a linearly independent subset of a vector space  $V$ . Given  $v \in V - S$ , the set  $S \cup \{v\}$  is linearly independent if and only if  $v \notin \text{span}(S)$ .*

**Proof.**

Assume that  $S \cup \{v\}$  is linearly independent. If  $v \in \text{span}(S)$ , then write

$$v = a_1 v_1 + \cdots + a_k v_k$$

with scalars  $a_i$  and  $v_i \in S$ . Thus,

$$a_1 v_1 + \cdots + a_k v_k + (-1)v = 0$$

expresses zero as a non-trivial linear combination of elements in  $S \cup \{v\}$ , contradicting linear independence of  $S \cup \{v\}$ . This implies that  $v \notin \text{span}(S)$ .

Assume that  $v \notin \text{span}(S)$ . If  $S \cup \{v\}$  is linearly dependent, then

$$a_1 v_1 + \cdots + a_k v_k + \lambda v = 0$$

for  $v_i \in S$  and the scalars  $a_i, \lambda$  cannot all be zero. If  $\lambda = 0$ , then we can represent zero as a non-trivial linear combination of elements in  $S$ , contradicting linear independence of  $S$ . If  $\lambda \neq 0$ , then rewrite

$$v = -\frac{a_1}{\lambda} v_1 - \frac{a_2}{\lambda} v_2 - \cdots - \frac{a_k}{\lambda} v_k$$

whence  $v \in \text{span}(S)$ , a contradiction. □

We may either start with a set that spans  $V$  and keep removing elements from it until it fails to span  $V$ , or we may start with a linearly independent set and keep adding elements to it, increasing the span until it spans  $V$ . Both these process lead to the following notion.

**Definition [Basis]** A basis  $\beta$  of a vector space  $V$  is a subset such that  $\beta$  is linearly independent and spans  $V$ .

**Examples** (1) The vector space  $\{0\}$  has  $\emptyset$  as a basis.

(2) The vector space  $\mathbb{R}$  (over  $\mathbb{R}$ ) has any non-zero real number as a basis.

(3) The vector space  $\mathbb{R}^n$  (over  $\mathbb{R}$ ) has  $\{e_1, \dots, e_n\}$  as a basis. It is called the *standard basis*.

Show that  $\{e_1, e_1 + e_2, e_1 + e_2 + e_3, \dots, e_1 + \dots + e_n\}$  is a basis for  $\mathbb{R}^n$ .

(4) The vector space  $M_2(\mathbb{R})$  (over  $\mathbb{R}$ ) has

$$\left\{ \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}, \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \right\}$$

as a basis (**exercise**).

(5) The vector space  $P(\mathbb{R})$  (over  $\mathbb{R}$ ) has  $\beta = \{1, x, x^2, \dots\}$  as a basis. Any polynomial is in the span of  $\beta$ . If  $\beta$  is linearly dependent, then for some  $x^{k_1}, \dots, x^{k_n}$  we must have

$$a_1 x^{k_1} + \dots + a_n x^{k_n} = 0$$

where 0 stands for the zero polynomial. Any real number is the root of the zero polynomial while if at least some of the  $a_i$ 's are non-zero, then the left hand side is a non-zero polynomial, which has finitely many roots, a contradiction. Thus,  $\beta$  is linearly independent.

**Lemma** Let  $\beta = \{v_1, \dots, v_n\}$  be a basis for  $V$ . Then every vector  $v \in V$  can be expressed uniquely as  $v = a_1v_1 + \dots + a_nv_n$  for scalars  $a_i$ .

**Proof.**

As a basis spans  $V$ , every  $v$  can be written as claimed. If  $v = c_1v_1 + \dots + c_nv_n$  is another expression, then

$$a_1v_1 + \dots + a_nv_n = v = c_1v_1 + \dots + c_nv_n.$$

Therefore,  $(a_1 - c_1)v_1 + \dots + (a_n - c_n)v_n = \mathbf{0}$  coupled with linear independence of  $\beta$ , implies that  $a_j = c_j$ . □

**Proposition** Let  $V$  be a vector space that admits a finite subset  $S$  that spans  $V$ . Then there exists a subset  $\beta \subseteq S$  which is a basis for  $V$ .



**Remark** The space  $P(\mathbb{R})$  does not have a finite spanning set. If it did, let  $S = \{p_1, \dots, p_k\}$  be such a set. Let  $n_k = \deg(p_k)$  and  $N = \max\{n_1, \dots, n_k\}$ . Then any linear combination of  $p_j$ 's will result in a polynomial of degree at most  $N$ . Thus,  $x^{N+1}$  will not be in the span.