

## Definition

If  $V$  is a vector space and  $S$  is a subset of  $V$  then  $S$  is a basis for  $V$  if

- 1  $S$  is linearly independent and
- 2  $S$  spans  $V$ .

## Theorem

*If  $S$  is a basis for  $V$  then every vector  $v \in V$  can be expressed in the form*

$$v = k_1 v_1 + k_2 v_2 + \dots + k_n v_n$$

*for distinct  $v_1, \dots, v_n \in S$  in exactly one way.*

## Definition

We say that a vector space is finite-dimensional if it has a finite basis.

## Theorem

*If  $V$  is a vector space with a basis of  $n$  vectors then*

- 1 any subset of  $V$  with more than  $n$  vectors is linearly dependent, and*
- 2 any subset of  $V$  with fewer than  $n$  vectors does not span  $V$ .*

## Corollary

*If  $V$  is a finite-dimensional vector space then all bases for  $V$  have the same size; we call this size the dimension of  $V$ .*

# Subspaces of $\mathbb{R}^n$

- Suppose that  $v_1, \dots, v_k$  are vectors in  $\mathbb{R}^n$ ; the big question is, what is the span of these vectors?
- Consider the matrix  $A$  with rows consisting of  $v_1, \dots, v_k$ ;  $A$  is  $k \times n$ .
- Now let  $B = rref(A)$ , the reduced row echelon form of  $A$ . We claim two things:
- First, the rows of  $B$  still span the same subspace as  $v_1, \dots, v_k$  and
- second, the non-zero rows form a basis for this subspace.
- In general, the subspace spanned by the rows of a matrix is called the row space and if the matrix is  $k \times n$  then the row space is a subspace of  $\mathbb{R}^n$ . The dimension of the row-space of  $A$  is called the rank of  $A$ .

# Example

- $$A = \begin{pmatrix} 1 & 1 & 1 & 2 & 3 & -1 \\ 1 & 1 & 1 & 0 & 3 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 3 & 0 \end{pmatrix}$$

- $$\text{rref}(A) = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 1 & 0 & -1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

# The Plus/Minus Theorem

## Theorem (Plus/Minus Theorem, 4.5.3)

*Suppose that  $S$  is a non-empty subset of a vector space  $V$ .  
Then*

- 1 if  $S$  is linearly independent and  $v \in V$  is not in  $\text{span}(S)$  then  $S \cup \{v\}$  is linearly independent, and*
- 2 if  $v \in S$  can be expressed as a linear combination of vectors from  $S \setminus \{v\}$  then  $\text{span}(S) = \text{span}(S \setminus \{v\})$ .*

## Corollary

*Suppose that  $V$  is an  $n$ -dimensional vector space and  $S$  is a subset of  $V$ .*

- 1 If  $S$  spans  $V$  then  $S$  contains a basis for  $V$ .*
- 2 If  $S$  is linearly independent then  $S$  is contained in a basis for  $V$ .*
- 3 If  $S$  contains exactly  $n$  vectors then  $S$  is a basis for  $V$  iff  $S$  is linearly independent.*

## Corollary

*If  $W$  is a subspace of a finite-dimensional vector space  $V$  then*

- 1  $W$  is finite-dimensional;*
- 2  $\dim(W) \leq \dim(V)$ ; and*
- 3  $\dim(W) = \dim(V)$  iff  $W = V$ .*

# The column-space of a matrix

- We return to the problem of subspaces of  $\mathbb{R}^n$ . The question is: given vectors  $v_1, \dots, v_k$  in  $\mathbb{R}^n$ , how do we find a basis for the subspace  $W$  spanned by these vectors *from among these vectors*.
- Consider a matrix  $A$  formed by placing  $v_1, \dots, v_k$  in the columns;  $A$  is  $n \times k$ ; let  $B = rref(A)$ , the reduced row-echelon form of  $A$ .
- The claim is that the vectors in  $A$  which correspond to the columns of  $B$  with leading 1's form a basis for  $W$ .
- Notice that this means that the dimension of  $W$  is the same as the rank of  $A$ .
- In general, for a matrix  $A$  which is  $k \times n$ , the subspace generated by the columns is called the column space of  $A$ , a subspace of  $\mathbb{R}^k$ , and its dimension is the same as the rank of  $A$ .