# 4.5 The Dimension of a Vector Space

# Theorem 4.10

If a vector space V has a basis  $\mathcal{B} = \{b_1, ..., b_n\}$ , then any set in V containing more than n vectors must be linearly dependent.

## Proof

Let  $\{u1, ..., u_p\}$  be a set in V with more than n vectors. The coordinate vectors  $[u_1]_{\mathcal{B}}, ..., [u_p]_{\mathcal{B}}$  form a linearly dependent set in  $\mathbb{R}^n$ , because there are more vectors (p) than entries (n) in each vector. So there exists scalars  $c_1, ..., c_p$ , not all zero such that

$$c_1 [u_1]_{\mathcal{B}} + \dots + c_p [u_p]_{\mathcal{B}} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

Since the coordinate mapping is a linear transformation,

$$\begin{bmatrix} c_1 u_1 + \dots + c_p u_p \end{bmatrix}_{\mathcal{B}} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

The zero vector displays the n weights needed to build the vectors  $c_1u_1 + ... + c_pu_p$  from the basis vectors in  $\mathcal{B}$ . That is,  $c_1u_1 + ... + c_pu_p = 0b_1 + ... + 0b_n = 0$ . Then since  $c_i$  are not all zero,  $\{u_1, ..., u_p\}$  is linearly dependent.

Meaning that Theorem 4.10 implies that if a vector space V has a basis  $\mathcal{B} = \{b_1, ..., b_n\}$ , then each linearly independent set in V has no more than n vectors.

#### Theorem 4.11

If a vector space V has a basis of n vectors, then every basis of V must consistent of exactly n vectors.

## **Definition**

If a vector space V is spanned by a finite set, then V is said to be finite-dimensional, and the dimension of V, written as dim V, is the number of vectors in a basis for V. The dimension of the zero vector space  $\{0\}$  is defined to be zero. If V is not spanned by a finite set, then V is said to be infinite-dimensional.

#### $\mathbf{E}\mathbf{x}$ 1

The standard basis for  $\mathbb{R}^n$  contains n vectors, so dim  $\mathbb{R}^n = n$ . The standard polynomial basis  $\{1, t, t^2\}$  shows that dim  $\mathbb{P}_2 = 3$ . In general, dim  $\mathbb{P}_n = n + 1$ . The space  $\mathbb{P}$  of all polynomials is infinite-dimensional.

# $\mathbf{Ex} \ \mathbf{2}$

Let  $H = \text{Span } \{v_1, v_2\}$ , where  $v_1 = \begin{bmatrix} 3 \\ 6 \\ 2 \end{bmatrix}$  and  $v_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ . A basis for H is  $\{v_1, v_2\}$ , since  $v_1 \& v_2$  are not multiples and hence are linearly independent. Thus dim H = 2.

## Ex 3

Find the dimension of the subspace

$$H = \left\{ \begin{bmatrix} a - 3b + 6c \\ 5a + 4d \\ b - 2c - d \\ 5d \end{bmatrix} : a, b, c, d \in \mathbb{R} \right\}$$

H is the set of all linear combinations of the vectors

$$v_1 = \begin{bmatrix} 1 \\ 5 \\ 0 \\ 0 \end{bmatrix}, \qquad v_2 = \begin{bmatrix} -3 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \qquad v_3 = \begin{bmatrix} 6 \\ 0 \\ -2 \\ 0 \end{bmatrix}, \qquad v_4 = \begin{bmatrix} 0 \\ 4 \\ -1 \\ 5 \end{bmatrix}$$

We can see that  $v_3$  is a multiple of  $v_2$ . So by the Spanning Set Theorem,  $v_3$  can be discarded and we would stil have a set that spans H. The 3 other vectors are linearly independent so  $\{v_1, v_2, v_4\}$  is linearly independent and hence it is a basis for H. Thus

$$\dim H = 3$$

#### Subspaces of a Finite-Dimensional Space

The next theorem is a natural counterpart to the Spanning Set Theorem.

# Theorem 4.12

Let H be a subspace of a finite-dimensional vector space V. Any linearly independent set in H can be expanded, if necessary, to a basis for H. Also, H is finite-dimensional and

$$\dim H \leq \dim V$$

# Theorem 4.13 The Basis Theorem

Let V be a p-dimensional vector space,  $b \ge 1$ . Any linearly independent set of exactly p elements in V is automatically a basis for V. Any set of exactly p elements that spans V is automatically a basis for V.

# The Dimensions of Nul A, Col A, & Row A

Since the dimensions of the null space and column space of an  $m \times n$  matrix are referred to frequently, they have specific names:

#### Definition

The rank of an  $m \times n$  matrix A is the dimension of the column space and the nullity of A is the dimension of the null space.

The rank of an  $m \times n$  matrix A is the number of pivot columns and the nullity of A is the number of free variables. Since the dimension of the row space is the number of pivot rows, it is also equal to the rank of A.

# Theorem 4.14 The Rank Theorem

The dimensions of the column space and the null space of an  $m \times n$  matrix A satisfy the equation

 $\operatorname{rank} A + \operatorname{nullity} A = \operatorname{number} \operatorname{of} \operatorname{columns} \operatorname{in} A$ 

### $\mathbf{Ex} \ \mathbf{5}$

Find the nullity and rank of

$$A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}$$

$$Ax = 0$$
, after row reduction 
$$\begin{bmatrix} 1 & -2 & 2 & 3 & -1 & 0 \\ 0 & 01 & 2 & -2 & 0 \\ 0 & 00 & 0 & 0 & 0 \end{bmatrix}$$

There are three free variables, ll.run(2133)  $.x_2, x_4, \& x_5$ . Hence the nullity of A is 3. Also the rank of A is 2 because A has two pivot columns.

## **Ex** 6

a) If A is a  $7 \times 9$  matrix with nullity 2, what is the rank of a?

Since A has 9 columns, rank A + 2 = 9, rank A = 7.

b) Could a  $6 \times 9$  matrix have nullity 2?

The columns of the matrix are vectors in  $\mathbb{R}^6$ , thus the dimension of the columns cannot exceed 6, that is the rank cannot exceed 6. So 6 + nullity A = 9 to be true, the nullity has to be 3.

# The Invertible Matrix Theorem (continued)

Let A be an  $n \times n$  matrix. Then the following statements are each equivalent to the statement that A is an invertible matrix.

m) The columns of A form a basis of  $\mathbb{R}^n$ . n) Col  $A = \mathbb{R}^n$  o) rank A = n p) nullity A = 0 q) Nul  $A = \{0\}$ 

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