# Chapter 4 Vector Spaces

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# **Vector Spaces and Subspaces**

A vector space is a nonempty set V of objects, called vectors, on which are defined two operations, called addition and mulitplication by scalars (real numbers), subject to the ten rules listed below. The rules must hold for all vectors  $\mathbf{u}$ ,  $\mathbf{v}$ , and  $\mathbf{w}$  in V and for all scalars c and d.

- 1. The sum of  $\mathbf{u}$  and  $\mathbf{v}$ , denoted by  $\mathbf{u} + \mathbf{v}$ , is in V.
- $2. \ \mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}.$
- 3.  $(\mathbf{u} + \mathbf{v}) + \mathbf{w} = \mathbf{u} + (\mathbf{v} + \mathbf{w})$ .
- 4. There is a **zero** vector **0** in V such that  $\mathbf{u} + \mathbf{0} = \mathbf{u}$ .
- 5. For each **u** in V, there is a vector  $-\mathbf{u}$  in V such that  $\mathbf{u} + (-\mathbf{u}) = \mathbf{0}$ .
- 6. The scalar multiple of  $\mathbf{u}$  by c, denoted by  $c\mathbf{u}$ , is in V.
- 7.  $c(\mathbf{u} + \mathbf{v}) = c\mathbf{u} + c\mathbf{v}$ .
- 8.  $(c+d)\mathbf{u} = c\mathbf{u} + d\mathbf{u}$ .
- 9. c(d**u**) = (cd)**u**.
- 10.  $1\mathbf{u} = \mathbf{u}$ .

A subspace of a vector space V is a subset H of V that has three properties:

- 1. The zero vector of V is in H.
- 2. H is closed under vector addition. That is, for each  $\mathbf{u}$  and  $\mathbf{v}$  in H, the sum  $\mathbf{u} + \mathbf{v}$  is in H.
- 3. H is closed under multiplication by scalars. That is, for each  $\mathbf{u}$  in H and each scalar c, the vector  $c\mathbf{u}$  is in H.

#### Theorem 1

If  $\mathbf{v}_1, \dots, \mathbf{v}_p$  are in a vector space V, then  $\mathrm{Span}\{\mathbf{v}_1, \dots, \mathbf{v}_p\}$  is a subspace of V.

### **Key Points**

1. If **v** is in  $\mathbb{R}^3$ ,  $H = \text{Span}\{\mathbf{v}\}$  is a subspace of  $\mathbb{R}^3$ .

# Null Spaces, Column Spaces, Row Spaces, and Linear Transformations

## **Null Spaces**

The **null space** of an  $m \times n$  matrix A, written as Nul A, is the set of all solutions of the homogeneous equation  $A\mathbf{x} = \mathbf{0}$ . In set notation,

Nul 
$$A = \{ \mathbf{x} : \mathbf{x} \text{ is in } \mathbb{R}^n \text{ and } A\mathbf{x} = \mathbf{0} \}$$

To find the vectors that span the null space:

- 1. Row reduce the augmented matrix  $\begin{bmatrix} A & 0 \end{bmatrix}$  to reduced echelon form.
- 2. Write the solution in terms of the free variables.
- 3. The column vectors that are multiplied by the free variables in the previous step form the spanning set for Nul A.

#### Theorem 2

The null space of an  $m \times n$  matrix A is a subspace of  $\mathbb{R}^n$ . Equivalently, the set of all solutions to a system  $A\mathbf{x} = \mathbf{0}$  of m homogeneous linear equations in n unknowns is a subspace of  $\mathbb{R}^n$ .

## Column Spaces

The **column space** of an  $m \times n$  matrix A, written as Col A, is the set of all linear combinations of the columns of A. If  $A = \begin{bmatrix} a_1 & \cdots & a_n \end{bmatrix}$ , then

$$Col A = Span\{a_1, \dots, a_n\}$$

To find the vectors that span the column space:

- 1. Determine the pivot columns in the matrix.
- 2. The pivot columns in the original matrix form the spanning set for Col A.

### Theorem 3

The column space of an  $m \times n$  matrix A is a subspace of  $\mathbb{R}^m$ .

## **Row Spaces**

The **row space** of an  $m \times n$  matrix A, written as Row A, is the set of all linear

combinations of the rows of 
$$A$$
. If  $A = \begin{bmatrix} r_1 \\ \cdots \\ r_n \end{bmatrix}$ , then

Row 
$$A = \operatorname{Span}\{r_1, \dots, r_n\}$$

To find the vectors that span the row space:

- 1. Row reduce the matrix to echelon form.
- 2. The pivot rows in the resulting matrix form the the spanning set for Row A.

### **Linear Transformations**

A linear transformation T from a vector space V into a vector space W is a rule that assigns to each vector  $\mathbf{x}$  in V a unique vector  $T(\mathbf{x})$  in W, such that

(i)  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$ 

for all  $\mathbf{u}, \mathbf{v}$  in V

(ii)  $T(c\mathbf{u}) = cT(\mathbf{u})$ 

for all **u** in *Vtextandallscalarsc*.

# Linear Independent Sets; Bases

#### Theorem 4

An indexed set  $\{v_1, \ldots, v_p\}$  of two or more vectors, with  $v_1 \neq 0$ , is linear dependent if and only if some  $v_j$  (with j > 1) is a linear combination of the preceding vectors,  $v_1, \ldots, v_{j-1}$ .

### **Bases**

Let H be a subspace of a vector space V. A set of vectors  $\mathcal{B}$  in V is a **basis** for H if

- (i)  $\mathcal{B}$  is a linearly independent set, and
- (ii) the subspace spanned by  $\mathcal{B}$  coincides with H; that is,

$$H = \operatorname{Span} \mathcal{B}$$

### Theorem 5 — The Spanning Set Theorem

Let  $S = \{v_1, \dots, v_p\}$  be a set in a vector space V, and let  $H = \text{Span}\{v_1, \dots, v_p\}$ .

- 1. If one of the vectors in S,  $v_k$ , is a linear combination of the remaining vectors in S, then the set formed from S by removing  $v_k$  still spans H.
- 2. If  $H \neq 0$ , some subset of S is a basis for H.

#### Theorem 6

The pivot columns of a matrix A form a basis for Col A.

### Theorem 7

If two matrices A and B are row equivalent, then their row spaces are the same. If B is in echelon form, the nonzero rows of B form a basis for the row space of A as well as for that of B.

# Coordinate Systems

### Theorem 8 — The Unique Representation Theorem

Let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis for a vector space V. Then for each  $\mathbf{x}$  in V, there exists a unique set of scalars  $c_1, \dots, c_n$  such that

$$\mathbf{x} = c_1 \mathbf{b}_1 + \dots + c_n \mathbf{b}_n$$

# Coordinates

Suppose  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  is a basis for a vector space V and  $\mathbf{x}$  is in V. The coordinates of  $\mathbf{x}$  relative to the basis  $\mathcal{B}$  (or the  $\mathcal{B}$ -coordinates of  $\mathbf{x}$ ) are the weights  $c_1, \dots, c_n$  such that  $\mathbf{x} = c_1 \mathbf{b}_1 + \dots + c_n \mathbf{b}_n$ .

#### Theorem 9

Let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis for a vector space V. Then the coordinate mapping  $\mathbf{x} \mapsto [\mathbf{x}]_{\mathcal{B}}$  is a one-to-one linear transformation from V onto  $\mathbb{R}^n$ .

# Extra Notes

 $1. \sin x \cos x = \frac{1}{2} \sin 2x$