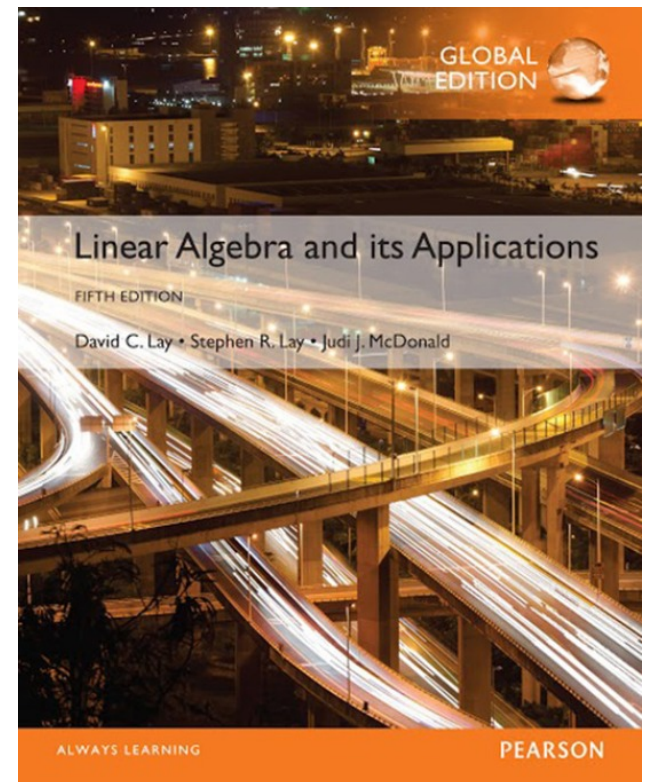


# 2

# Matrix Algebra

## 2.8

## SUBSPACES OF $\mathbb{R}^n$

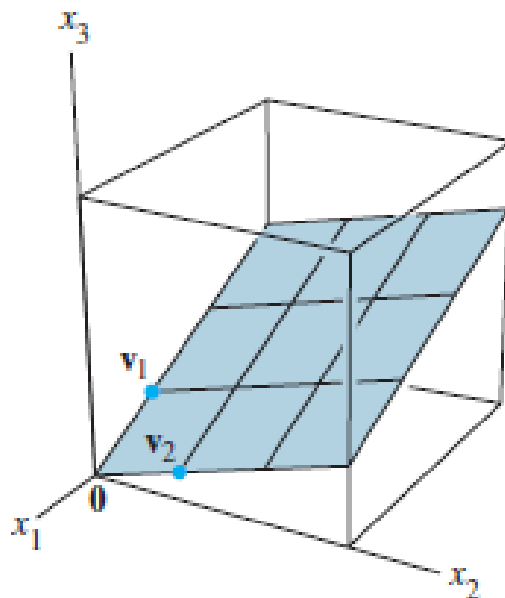


# SUBSPACES OF $\mathbb{R}^n$

- **Definition:** A subspace of  $\mathbb{R}^n$  is any set  $H$  in  $\mathbb{R}^n$  that has three properties:
  - a) The zero vector is in  $H$ .
  - b) For each  $\mathbf{u}$  and  $\mathbf{v}$  in  $H$ , the sum  $\mathbf{u} + \mathbf{v}$  is in  $H$ .
  - c) For each  $\mathbf{u}$  in  $H$  and each scalar  $c$ , the vector  $c\mathbf{u}$  is in  $H$ .

# SUBSPACES OF $\mathbb{R}^n$

- A plane through the origin is the standard way to visualize the subspace in Example 1 on the next slide. See Fig. 1 below:



**FIGURE 1**

**Span  $\{v_1, v_2\}$  as a plane through the origin.**

# SUBSPACES OF $\mathbb{R}^n$

- **Example 1** If  $v_1$  and  $v_2$  are in  $\mathbb{R}^n$  and  $H = \text{Span}\{v_1, v_2\}$ , then  $H$  is a subspace of  $\mathbb{R}^n$ . To verify this statement, note that the zero vector is in  $H$  (because  $0v_1 + 0v_2$  is a linear combination of  $v_1$  and  $v_2$ ).
- Now take two arbitrary vectors in  $H$ , say,  
$$u = s_1v_1 + s_2v_2 \quad \text{and} \quad v = t_1v_1 + t_2v_2$$
- Then  
$$u + v = (s_1 + t_1)v_1 + (s_2 + t_2)v_2$$
- which shows that  $u + v$  is a linear combination of  $v_1$  and  $v_2$  and hence is in  $H$ . Also, for any scalar  $c$ , the vector  $cu$  is in  $H$ , because  $cu = c(s_1v_1 + s_2v_2) = cs_1(v_1) + cs_2(v_2)$ .

# COLUMN SPACE AND NULL SPACE OF A MATRIX

- **Definition:** The **column space** of a matrix  $A$  is the set  $\text{Col } A$  of all linear combinations of the columns of  $A$ .
- If  $A = [a_1 \dots a_n]$  with the columns of  $\mathbb{R}^m$ , then  $\text{Col } A$  is the same as  $\text{Span}\{a_1 \dots a_n\}$ . Example 4 shows that the column space of an  $m \times n$  matrix is a subspace of  $\mathbb{R}^m$ .
- **Example 4** Let  $A = \begin{bmatrix} 1 & -3 & -4 \\ -4 & 6 & -2 \\ -3 & 7 & 6 \end{bmatrix}$  and  $b = \begin{bmatrix} 3 \\ 3 \\ -4 \end{bmatrix}$ .  
Determine whether  $b$  is in the column space of  $A$ .

# COLUMN SPACE AND NULL SPACE OF A MATRIX

- **Solution:** The vector  $\mathbf{b}$  is a linear combination of the columns of  $A$  if and only if  $\mathbf{b}$  can be written as  $A\mathbf{x}$  for some  $\mathbf{x}$ , that is, if and only if the equation  $A\mathbf{x} = \mathbf{b}$  has a solution.

- Row reducing the augmented matrix  $[A\mathbf{b}]$ ,

$$\begin{bmatrix} 1 & -3 & -4 & 3 \\ -4 & 6 & -2 & 3 \\ -3 & 7 & 6 & -4 \end{bmatrix} \sim \begin{bmatrix} 1 & -3 & -4 & 3 \\ 0 & -6 & -18 & 15 \\ 0 & -2 & -6 & 5 \end{bmatrix} \sim \begin{bmatrix} 1 & -3 & -4 & 3 \\ 0 & -6 & -18 & 15 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

- We conclude that  $A\mathbf{x} = \mathbf{b}$  is consistent and  $\mathbf{b}$  is in  $\text{Col } A$ .

# COLUMN SPACE AND NULL SPACE OF A MATRIX

- **Definition:** The **null space** of a matrix  $A$  is the set  $\text{Nul}A$  of all solutions of the homogenous equation  $Ax = 0$ .
- **Theorem 12:** The null space of an  $m \times n$  matrix  $A$  is a subspace of  $\mathbb{R}^n$ . Equivalently, the set of all solutions of a system  $Ax = 0$  of  $m$  homogenous linear equations in  $n$  unknowns is a subspace of  $\mathbb{R}^n$ .
- **Proof:** The zero vector is in  $\text{Nul}A$  (because  $A0 = 0$ ). To show that  $\text{Nul}A$  satisfies that other two properties required for a subspace, take any  $u$  and  $v$  in  $\text{Nul}A$ .

# COLUMN SPACE AND NULL SPACE OF A MATRIX

- That is, suppose  $A\mathbf{u} = 0$  and  $A\mathbf{v} = 0$ . Then, by a property of matrix multiplication,

$$A(u + v) = Au + Av = 0 + 0 = 0$$

- Thus  $\mathbf{u} + \mathbf{v}$  satisfies  $A(\mathbf{u} + \mathbf{v}) = 0$ , and so  $\mathbf{u} + \mathbf{v}$  is in  $\text{Nul}A$ . Also, for any scalar  $c$ ,  $A(c\mathbf{u}) = c(A\mathbf{u}) = c(0) = 0$ , which shows that  $c\mathbf{u}$  is in  $\text{Nul}A$ .



# BASIS FOR A SUBSPACE

- **Definition:** A **basis** for a subspace  $H$  of  $\mathbb{R}^n$  is a linearly independent set in  $H$  that spans  $H$ .
- **Example 5** The columns of an invertible  $n \times n$  matrix form a basis for all of because they are linearly independent and span  $\mathbb{R}^n$ , by the Invertible Matrix Theorem.

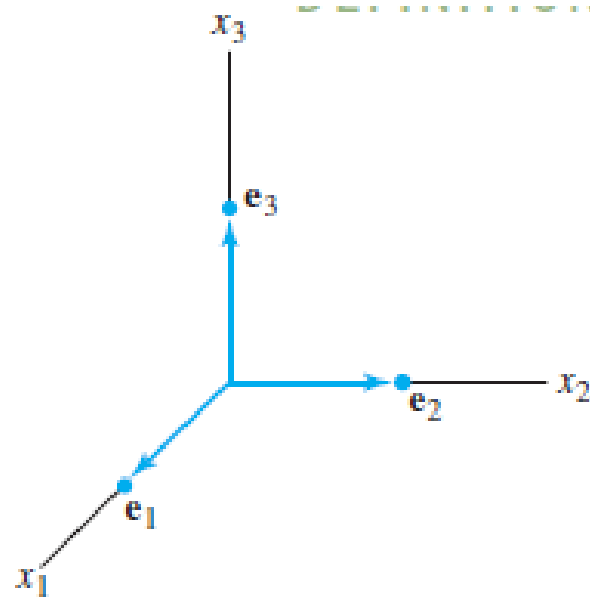
# BASIS FOR A SUBSPACE

- One such matrix is the  $n \times n$  identity matrix. Its columns are denoted by  $e_1, \dots, e_n$ :

$$e_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad e_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \quad \dots, \quad e_n = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix},$$

- The set  $\{e_1, \dots, e_n\}$  is called the **standard basis** for  $\mathbb{R}^n$ . See Fig. 3 on the next slide.

# BASIS FOR A SUBSPACE



**FIGURE 3**

The standard basis for  $\mathbb{R}^3$ .

- **Theorem 13:** The pivot columns of a matrix  $A$  form a basis for the column space of  $A$ .