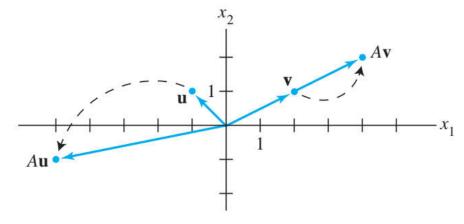
# 5.1 EIGEN VALUES AND EIGEN VECTORS

Although a transformation  $\mathbf{x} \mapsto A\mathbf{x}$  may move vectors in a variety of directions, it often happens that there are special vectors on which the action of A is quite simple.

**EXAMPLE 1** Let 
$$A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix}$$
,  $\mathbf{u} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ , and  $\mathbf{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ .

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**FIGURE 1** Effects of multiplication by A.

An **eigenvector** of an  $n \times n$  matrix A is a nonzero vector  $\mathbf{x}$  such that  $A\mathbf{x} = \lambda \mathbf{x}$  for some scalar  $\lambda$ . A scalar  $\lambda$  is called an **eigenvalue** of A if there is a nontrivial solution  $\mathbf{x}$  of  $A\mathbf{x} = \lambda \mathbf{x}$ ; such an  $\mathbf{x}$  is called an *eigenvector corresponding to*  $\lambda$ .<sup>1</sup>

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It is easy to determine if a given vector is an eigenvector of a matrix. It is also easy to decide if a specified scalar is an eigenvalue.

**EXAMPLE 2** Let  $A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$ ,  $\mathbf{u} = \begin{bmatrix} 6 \\ -5 \end{bmatrix}$ , and  $\mathbf{v} = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$ . Are  $\mathbf{u}$  and  $\mathbf{v}$  eigen-

vectors of A?

**EXAMPLE 2** Let  $A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$ ,  $\mathbf{u} = \begin{bmatrix} 6 \\ -5 \end{bmatrix}$ , and  $\mathbf{v} = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$ . Are  $\mathbf{u}$  and  $\mathbf{v}$  eigenvectors of A?

$$A\mathbf{u} = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} \begin{bmatrix} 6 \\ -5 \end{bmatrix} = \begin{bmatrix} -24 \\ 20 \end{bmatrix} = -4 \begin{bmatrix} 6 \\ -5 \end{bmatrix} = -4\mathbf{u}$$
$$A\mathbf{v} = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} \begin{bmatrix} 3 \\ -2 \end{bmatrix} = \begin{bmatrix} -9 \\ 11 \end{bmatrix} \neq \lambda \begin{bmatrix} 3 \\ -2 \end{bmatrix}$$

**EXAMPLE 3** Show that 7 is an eigenvalue of matrix A in Example 2, and find the corresponding eigenvectors.

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$$A\mathbf{x} = 7\mathbf{x} \qquad A\mathbf{x} - 7\mathbf{x} = \mathbf{0} \qquad (A - 7I)\mathbf{x} = \mathbf{0}$$

$$A - 7I = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} - \begin{bmatrix} 7 & 0 \\ 0 & 7 \end{bmatrix} = \begin{bmatrix} -6 & 6 \\ 5 & -5 \end{bmatrix}$$

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$$\begin{bmatrix} -6 & 6 & 0 \\ 5 & -5 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \qquad x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

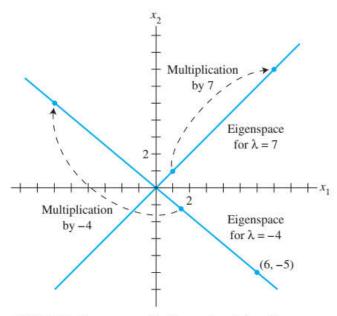
Thus  $\lambda$  is an eigenvalue of an  $n \times n$  matrix A if and only if the equation

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Thus  $\lambda$  is an eigenvalue of an  $n \times n$  matrix A if and only if the equation

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has a nontrivial solution. The set of *all* solutions of (3) is just the null space of the matrix  $A - \lambda I$ . So this set is a *subspace* of  $\mathbb{R}^n$  and is called the **eigenspace** of A corresponding to  $\lambda$ . The eigenspace consists of the zero vector and all the eigenvectors corresponding to  $\lambda$ .



**FIGURE 2** Eigenspaces for  $\lambda = -4$  and  $\lambda = 7$ .

**EXAMPLE 4** Let  $A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix}$ . An eigenvalue of A is 2. Find a basis for

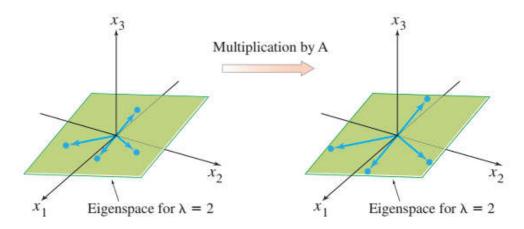
the corresponding eigenspace.

**EXAMPLE 4** Let  $A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix}$ . An eigenvalue of A is 2. Find a basis for

the corresponding eigenspace.

$$A - 2I = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix} - \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} 2 & -1 & 6 \\ 2 & -1 & 6 \\ 2 & -1 & 6 \end{bmatrix}$$

$$\begin{bmatrix} 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \end{bmatrix} \sim \begin{bmatrix} 2 & -1 & 6 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \qquad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_2 \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix}, \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \right\}$$



# THEOREM 1

The eigenvalues of a triangular matrix are the entries on its main diagonal.

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$$A - \lambda I = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{bmatrix} - \begin{bmatrix} \lambda & 0 & 0 \\ 0 & \lambda & 0 \\ 0 & 0 & \lambda \end{bmatrix}$$
$$= \begin{bmatrix} a_{11} - \lambda & a_{12} & a_{13} \\ 0 & a_{22} - \lambda & a_{23} \\ 0 & 0 & a_{33} - \lambda \end{bmatrix}$$

Lower triangular??

**EXAMPLE 5** Let 
$$A = \begin{bmatrix} 3 & 6 & -8 \\ 0 & 0 & 6 \\ 0 & 0 & 2 \end{bmatrix}$$
 and  $B = \begin{bmatrix} 4 & 0 & 0 \\ -2 & 1 & 0 \\ 5 & 3 & 4 \end{bmatrix}$ . The eigenval-

ues of A are 3, 0, and 2. The eigenvalues of B are 4 and 1.

What does it mean for a matrix A to have an eigenvalue of 0, such as in Example 5?

**EXAMPLE 5** Let 
$$A = \begin{bmatrix} 3 & 6 & -8 \\ 0 & 0 & 6 \\ 0 & 0 & 2 \end{bmatrix}$$
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ues of A are 3, 0, and 2. The eigenvalues of B are 4 and 1.

What does it mean for a matrix A to have an eigenvalue of 0, such as in Example 5?

$$A\mathbf{x} = 0\mathbf{x}$$

Thus 0 is an eigenvalue of A if and only if A is not invertible.

# THEOREM 2

If  $\mathbf{v}_1, \dots, \mathbf{v}_r$  are eigenvectors that correspond to distinct eigenvalues  $\lambda_1, \dots, \lambda_r$  of an  $n \times n$  matrix A, then the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$  is linearly independent.

# THEOREM 2

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Suppose  $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$  is linearly dependent.

$$c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p = \mathbf{v}_{p+1}$$

$$c_1A\mathbf{v}_1 + \dots + c_pA\mathbf{v}_p = A\mathbf{v}_{p+1}$$

$$c_1\lambda_1\mathbf{v}_1 + \dots + c_p\lambda_p\mathbf{v}_p = \lambda_{p+1}\mathbf{v}_{p+1}$$

$$c_1(\lambda_1 - \lambda_{p+1})\mathbf{v}_1 + \dots + c_p(\lambda_p - \lambda_{p+1})\mathbf{v}_p = \mathbf{0}$$

$$\mathbf{x}_{k+1} = A\mathbf{x}_k \quad (k = 0, 1, 2, \ldots)$$

If A is an  $n \times n$  matrix, then (8) is a *recursive* description of a sequence  $\{\mathbf{x}_k\}$  in  $\mathbb{R}^n$ .

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If A is an  $n \times n$  matrix, then (8) is a *recursive* description of a sequence  $\{\mathbf{x}_k\}$  in  $\mathbb{R}^n$ .

The simplest way to build a solution of (8) is to take an eigenvector  $\mathbf{x}_0$  and its corresponding eigenvalue  $\lambda$  and let

$$\mathbf{x}_k = \lambda^k \mathbf{x}_0 \quad (k = 1, 2, \ldots) \tag{9}$$

This sequence is a solution because

$$A\mathbf{x}_k = A(\lambda^k \mathbf{x}_0) = \lambda^k (A\mathbf{x}_0) = \lambda^k (\lambda \mathbf{x}_0) = \lambda^{k+1} \mathbf{x}_0 = \mathbf{x}_{k+1}$$

**2.** If **x** is an eigenvector of A corresponding to  $\lambda$ , what is  $A^3$ **x**?

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$$A\mathbf{x} = \lambda \mathbf{x}$$

$$A^2$$
**x** =  $A(\lambda$ **x**) =  $\lambda A$ **x** =  $\lambda^2$ **x**

$$A^3$$
**x** =  $A(A^2$ **x**) =  $A(\lambda^2$ **x**) =  $\lambda^2 A$ **x** =  $\lambda^3$ **x**

$$A^k \mathbf{x} = \lambda^k \mathbf{x}$$

If A is an  $n \times n$  matrix and  $\lambda$  is an eigenvalue of A, show that  $2\lambda$  is an eigenvalue of 2A.

# 5.2 CHARACTERISTIC EQUATION

**EXAMPLE 1** Find the eigenvalues of  $A = \begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix}$ .

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$$(A - \lambda I)\mathbf{x} = \mathbf{0}$$

$$A - \lambda I = \begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 2 - \lambda & 3 \\ 3 & -6 - \lambda \end{bmatrix}$$

finding all  $\lambda$  such that the matrix  $A - \lambda I$  is *not* invertible determinant is zero

$$det(A - \lambda I) = (2 - \lambda)(-6 - \lambda) - (3)(3)$$
$$= -12 + 6\lambda - 2\lambda + \lambda^2 - 9$$
$$= \lambda^2 + 4\lambda - 21$$
$$= (\lambda - 3)(\lambda + 7)$$

# The Invertible Matrix Theorem (continued)

Let A be an  $n \times n$  matrix. Then A is invertible if and only if:

- s. The number 0 is *not* an eigenvalue of A.
- t. The determinant of A is not zero.

A scalar  $\lambda$  is an eigenvalue of an  $n \times n$  matrix A if and only if  $\lambda$  satisfies the characteristic equation

$$\det(A - \lambda I) = 0$$

**EXAMPLE 3** Find the characteristic equation of

$$A = \begin{bmatrix} 5 & -2 & 6 & -1 \\ 0 & 3 & -8 & 0 \\ 0 & 0 & 5 & 4 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

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$$A = \begin{bmatrix} 5 & -2 & 6 & -1 \\ 0 & 3 & -8 & 0 \\ 0 & 0 & 5 & 4 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

# Characteristic polynomial of A.

$$\det(A - \lambda I) = \det \begin{bmatrix} 5 - \lambda & -2 & 6 & -1 \\ 0 & 3 - \lambda & -8 & 0 \\ 0 & 0 & 5 - \lambda & 4 \\ 0 & 0 & 0 & 1 - \lambda \end{bmatrix}$$

$$= (5 - \lambda)(3 - \lambda)(5 - \lambda)(1 - \lambda)$$

$$\lambda^4 - 14\lambda^3 + 68\lambda^2 - 130\lambda + 75 = 0$$

eigenvalue 5 is said to have multiplicity 2

**EXAMPLE 4** The characteristic polynomial of a  $6 \times 6$  matrix is  $\lambda^6 - 4\lambda^5 - 12\lambda^4$ . Find the eigenvalues and their multiplicities.

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$$\lambda^{6} - 4\lambda^{5} - 12\lambda^{4} = \lambda^{4}(\lambda^{2} - 4\lambda - 12) = \lambda^{4}(\lambda - 6)(\lambda + 2)$$

Because the characteristic equation for an n\* n matrix involves an nth-degree polynomial, the equation has exactly n roots, counting multiplicities, provided complex roots are allowed.

# Similar matrices

If A and B are n\*n matrices, then A is similar to B if there is an invertible matrix P such that

$$P^{-1}AP = B$$
, or, equivalently,  $A = PBP^{-1}$ 

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, or, equivalently,  $A = PBP^{-1}$ 

B is also similar to A, and we say simply that A and B are similar

Changing A into  $P^{-1}AP$  is called a **similarity transformation** 

# Similar matrices

# THEOREM 4

If  $n \times n$  matrices A and B are similar, then they have the same characteristic polynomial and hence the same eigenvalues (with the same multiplicities).

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**PROOF** If 
$$B = P^{-1}AP$$
, then

$$B - \lambda I = P^{-1}AP - \lambda P^{-1}P = P^{-1}(AP - \lambda P) = P^{-1}(A - \lambda I)P$$

Using the multiplicative property (b) in Theorem 3, we compute

$$det(B - \lambda I) = det[P^{-1}(A - \lambda I)P]$$
$$= det(P^{-1}) \cdot det(A - \lambda I) \cdot det(P)$$

**EXAMPLE 5** Let  $A = \begin{bmatrix} .95 & .03 \\ .05 & .97 \end{bmatrix}$ . Analyze the long-term behavior of the dynam-

ical system defined by  $\mathbf{x}_{k+1} = A\mathbf{x}_k$  (k = 0, 1, 2, ...), with  $\mathbf{x}_0 = \begin{bmatrix} .6 \\ .4 \end{bmatrix}$ .

**EXAMPLE 5** Let  $A = \begin{bmatrix} .95 & .03 \\ .05 & .97 \end{bmatrix}$ . Analyze the long-term behavior of the dynam-

ical system defined by  $\mathbf{x}_{k+1} = A\mathbf{x}_k$  (k = 0, 1, 2, ...), with  $\mathbf{x}_0 = \begin{bmatrix} .6 \\ .4 \end{bmatrix}$ .

$$0 = \det \begin{bmatrix} .95 - \lambda & .03 \\ .05 & .97 - \lambda \end{bmatrix} = (.95 - \lambda)(.97 - \lambda) - (.03)(.05)$$
$$= \lambda^2 - 1.92\lambda + .92$$

$$\mathbf{v}_1 = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$
 and  $\mathbf{v}_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ 

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$$\mathbf{v}_1 = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$
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$$\mathbf{x}_0 = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2$$

$$\mathbf{x}_0 = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2$$

$$\mathbf{x}_1 = A\mathbf{x}_0 = c_1A\mathbf{v}_1 + c_2A\mathbf{v}_2$$
$$= c_1\mathbf{v}_1 + c_2(.92)\mathbf{v}_2$$

$$\mathbf{x}_2 = A\mathbf{x}_1 = c_1A\mathbf{v}_1 + c_2(.92)A\mathbf{v}_2$$
  
=  $c_1\mathbf{v}_1 + c_2(.92)^2\mathbf{v}_2$ 

$$\mathbf{x}_0 = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 \qquad \mathbf{x}_1 = A \mathbf{x}_0 = c_1 A \mathbf{v}_1 + c_2 A \mathbf{v}_2$$
$$= c_1 \mathbf{v}_1 + c_2 (.92) \mathbf{v}_2$$

$$\mathbf{x}_2 = A\mathbf{x}_1 = c_1A\mathbf{v}_1 + c_2(.92)A\mathbf{v}_2$$
  
=  $c_1\mathbf{v}_1 + c_2(.92)^2\mathbf{v}_2$ 

$$\mathbf{x}_k = c_1 \mathbf{v}_1 + c_2 (.92)^k \mathbf{v}_2 \quad (k = 0, 1, 2, ...)$$

$$\mathbf{x}_k = .125 \begin{bmatrix} 3 \\ 5 \end{bmatrix} + .225 (.92)^k \begin{bmatrix} 1 \\ -1 \end{bmatrix} \qquad \mathbf{x}_k \text{ tends to } \begin{bmatrix} .375 \\ .625 \end{bmatrix} = .125 \mathbf{v}_1$$

Show that if A = QR with Q invertible, then A is similar to  $A_1 = RQ$ .

# DIAGONALIZATION

$$A = PDP^{-1}$$

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$$D = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix}$$
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$$D = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix}$$
,

$$D^2 = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 5^2 & 0 \\ 0 & 3^2 \end{bmatrix}$$

$$D^k = \begin{bmatrix} 5^k & 0 \\ 0 & 3^k \end{bmatrix} \quad \text{for } k \ge 1$$

**EXAMPLE 2** Let  $A = \begin{bmatrix} 7 & 2 \\ -4 & 1 \end{bmatrix}$ . Find a formula for  $A^k$ , given that  $A = PDP^{-1}$ 

where

$$P = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \quad \text{and} \quad D = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix} \qquad P^{-1} = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$$

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$$A^2 = (PDP^{-1})(PDP^{-1}) = PD(P^{-1}P)DP^{-1} = PDDP^{-1}$$

$$= PD^2P^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} 5^2 & 0 \\ 0 & 3^2 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$$

$$A^3 = (PDP^{-1})A^2 = (PDP^{-1})PD^2P^{-1} = PDD^2P^{-1} = PD^3P^{-1}$$

 $A^{k} = PD^{k}P^{-1}$ 

A square matrix A is said to be **diagonalizable** if A is similar to a diagonal matrix, that is, if  $A = PDP^{-1}$  for some invertible matrix P and some diagonal matrix D.

## THEOREM 5

## The Diagonalization Theorem

An  $n \times n$  matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

In fact,  $A = PDP^{-1}$ , with D a diagonal matrix, if and only if the columns of P are n linearly independent eigenvectors of A. In this case, the diagonal entries of D are eigenvalues of A that correspond, respectively, to the eigenvectors in P.

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$$A = PDP^{-1} \qquad AP = PD$$

$$AP = A[\mathbf{v}_1 \ \mathbf{v}_2 \ \cdots \ \mathbf{v}_n] = [A\mathbf{v}_1 \ A\mathbf{v}_2 \ \cdots \ A\mathbf{v}_n]$$

$$PD = P \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} = \begin{bmatrix} \lambda_1 \mathbf{v}_1 & \lambda_2 \mathbf{v}_2 & \cdots & \lambda_n \mathbf{v}_n \end{bmatrix}$$