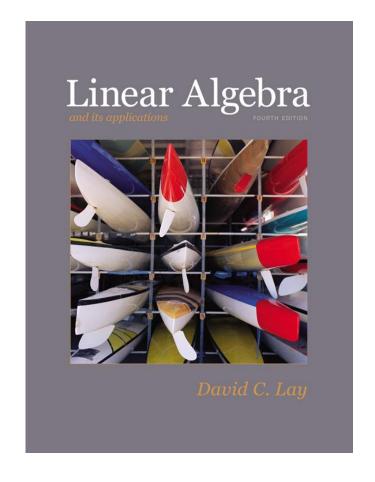
5

# Eigenvalues and Eigenvectors

5.1



- **Definition:** 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*  $\lambda$ .
- $\lambda$  is an eigenvalue of an  $n \times n$  matrix A if and only if the equation

$$(A - \lambda I)x = 0 \quad ----(1)$$

has a nontrivial solution.

The set of *all* solutions of (1) 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  $\lambda$ .
- **Example 1:** Show that 7 is an eigenvalue of matrix

$$A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$$
 and find the corresponding eigenvectors.

• **Solution:** The scalar 7 is an eigenvalue of *A* if and only if the equation

$$Ax = 7x \qquad ----(2)$$

has a nontrivial solution.

• But (2) is equivalent to Ax - 7x = 0, or (A - 7I)x = 0 ----(3)

To solve this homogeneous equation, form the matrix

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

- The columns of A-7I are obviously linearly dependent, so (3) has nontrivial solutions.
- To find the corresponding eigenvectors, use row operations:

$$\begin{bmatrix} -6 & 6 & 0 \\ 5 & -5 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$
• The general solution has the form  $x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ .

- Each vector of this form with  $x_2 \neq 0$  is an eigenvector corresponding to  $\lambda = 7$ .

■ Example 2: 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.

• **Solution:** Form

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

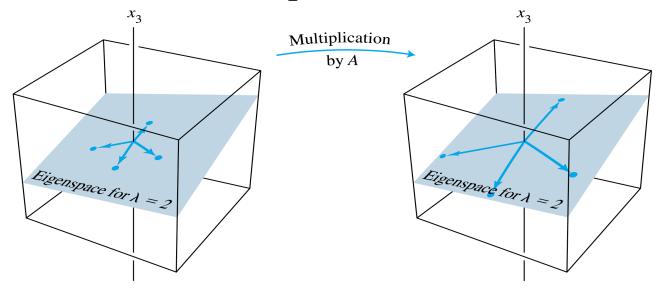
and row reduce the augmented matrix for (A-2I)x = 0.

$$\begin{bmatrix} 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \\ 2 & -1 & 6 & 0 \end{bmatrix} \sim \begin{bmatrix} 2 & -1 & 6 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

- At this point, it is clear that 2 is indeed an eigenvalue of A because the equation (A-2I)x = 0 has free variables.
- The general solution is

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_2 \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix}, x_2 \text{ and } x_3 \text{ free.}$$

• The eigenspace, shown in the following figure, is a two-dimensional subspace of  $\mathbb{R}^3$ .



A acts as a dilation on the eigenspace.

A basis is

 $\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \right\}$ 

- **Theorem 1:** The eigenvalues of a triangular matrix are the entries on its main diagonal.
- **Proof:** For simplicity, consider the  $3 \times 3$  case.
- If A is upper triangular, the  $A \lambda I$  has the form

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

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

- The scalar  $\lambda$  is an eigenvalue of A if and only if the equation  $(A \lambda I)x = 0$  has a nontrivial solution, that is, if and only if the equation has a free variable.
- Because of the zero entries in  $A \lambda I$ , it is easy to see that  $(A \lambda I)x = 0$  has a free variable if and only if at least one of the entries on the diagonal of  $A \lambda I$  is zero.
- This happens if and only if  $\lambda$  equals one of the entries  $a_{11}, a_{22}, a_{33}$  in A.

- **Theorem 2:** If  $\mathbf{v}_1, ..., \mathbf{v}_r$  are eigenvectors that correspond to distinct eigenvalues  $\lambda_1, ..., \lambda_r$  of an  $n \times n$  matrix A, then the set  $\{\mathbf{v}_1, ..., \mathbf{v}_r\}$  is linearly independent.
- **Proof:** Suppose  $\{\mathbf{v}_1, ..., \mathbf{v}_r\}$  is linearly dependent.
- Since  $\mathbf{v}_1$  is nonzero, Theorem 7 in Section 1.7 says that one of the vectors in the set is a linear combination of the preceding vectors.
- Let p be the least index such that  $V_{p+1}$  is a linear combination of the preceding (linearly independent) vectors.

• Then there exist scalars  $c_1, ..., c_p$  such that

$$c_1 V_1 + \dots + c_p V_p = V_{p+1}$$
 ----(4)

 Multiplying both sides of (4) by A and using the fact that

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

$$c_1 \lambda_1 \mathbf{v}_1 + \dots + c_p \lambda_p \mathbf{v}_p = \lambda_{p+1} \mathbf{v}_{p+1} \qquad ----(5)$$

• Multiplying both sides of (4) by  $\lambda_{p+1}$  and subtracting the result from (5), we have

$$c_1(\lambda_1 - \lambda_{p+1})v_1 + \dots + c_p(\lambda_p - \lambda_{p+1})v_p = 0$$
 ----(6)

- Since  $\{\mathbf{v}_1, ..., \mathbf{v}_p\}$  is linearly independent, the weights in (6) are all zero.
- But none of the factors  $\lambda_i \lambda_{p+1}$  are zero, because the eigenvalues are distinct.
- Hence  $c_i = 0$  for i = 1, ..., p.
- But then (4) says that  $V_{p+1} = 0$ , which is impossible.

# EIGENVECTORS AND DIFFERENCE EQUATIONS

- Hence  $\{\mathbf{v}_1, ..., \mathbf{v}_r\}$  cannot be linearly dependent and therefore must be linearly independent.
- If A is an  $n \times n$  matrix, then

$$\mathbf{x}_{k+1} = A\mathbf{x}_k$$
  $(k = 0, 1, 2...)$  ----(7)

is a *recursive* description of a sequence  $\{x_k\}$  in  $\mathbb{R}^n$ .

• A **solution** of (7) is an explicit description of  $\{x_k\}$  whose formula for each  $x_k$  does not depend directly on A or on the preceding terms in the sequence other than the initial term  $\mathbf{x}_0$ .

# **EIGENVECTORS AND DIFFERENCE EQUATIONS**

• The simplest way to build a solution of (7) is to take an eigenvector  $\mathbf{x}_0$  and its corresponding eigenvalue  $\lambda$  and let

$$X_k = \lambda^k X_0 \quad (k = 1, 2, ...)$$
 ----(8)

This sequence is a solution because

$$Ax_{k} = A(\lambda^{k}x_{0}) = \lambda^{k}(Ax_{0}) = \lambda^{k}(\lambda x_{0}) = \lambda^{k+1}x_{0} = x_{k+1}$$