

# Chapter 5 Eigenvalues, Eigenvectors, and Diagonalization

# 5.1 Eigenvalues and Eigenvectors

**Definitions** Let  $T$  be a linear operator on  $R^n$ . A *nonzero* vector  $v$  in  $R^n$  is called an **eigenvector** of  $T$  if  $T(v)$  is a multiple of  $v$ ; that is,  $T(v)=\lambda v$  for some scalar  $\lambda$ . The scalar  $\lambda$  is called the **eigenvalue** of  $T$  that corresponds to  $v$ .

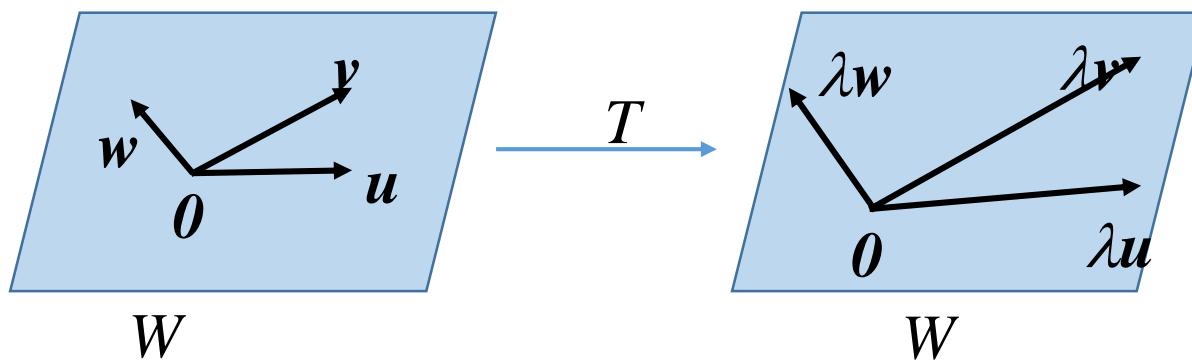
**Definitions** Let  $A$  be an  $n \times n$  matrix. A *nonzero* vector  $v$  in  $R^n$  is called an **eigenvector** of  $A$  if  $Av=\lambda v$  for some scalar  $\lambda$ . The scalar  $\lambda$  is called the **eigenvalue** of  $A$  that corresponds to  $v$ .

The eigenvectors and corresponding eigenvalues of a linear operator are the same as those of its standard matrix.

- An eigenvector  $v$  of a matrix  $A$  is associated with exactly one eigenvalue.
- In contrast, if  $v$  is an eigenvector of  $A$  corresponding to eigenvalue  $\lambda$ , then every nonzero multiple of  $v$  is also an eigenvector of  $A$  corresponding to  $\lambda$ .

Let  $A$  be an  $n \times n$  matrix with eigenvalue  $\lambda$ . The eigenvectors of  $A$  corresponding to  $\lambda$  are the nonzero solutions of  $(A - \lambda I_n)x = 0$ .

- The set of solutions of  $(A - \lambda I_n)x = 0$  is called the **eigenspace of A corresponding to the eigenvalue  $\lambda$** .
  - ✓ It is just the null space of  $(A - \lambda I_n)$ , **the kernel of  $(A - \lambda I_n)$** , a subspace of  $R^n$ .
- If  $\lambda$  is an eigenvalue of a linear operator  $T$  on  $R^n$ , the set of vectors  $v$  in  $R^n$  such  $T(v) = \lambda v$  is called the **eigenspace of  $T$  corresponding to  $\lambda$** .



$W$  is the eigenspace of  $T$  corresponding to eigenvalue  $\lambda$ .

## 5.2 The characteristic polynomial

- If  $\lambda$  is an eigenvalue of  $A$ , there must be a nonzero vector  $v$  in  $R^n$  such that  $Av=\lambda v$ .
  - ✓ For an  $n \times n$  matrix  $A$  and the homogeneous system of linear equations  $(A - \lambda I_n)x = 0$  to have nonzero solutions, the rank of  $A - \lambda I_n$  must be less than  $n$ . By the Invertible Matrix Theorem,  $A - \lambda I_n$  is then not invertible, so its determinant must be zero.

The eigenvalues of a square matrix  $A$  are the values of  $t$  that satisfy

$$\det(A - tI_n) = 0.$$

- The equation  $\det(A - \lambda I_n) = 0$  is called the **characteristic equation** of  $A$ .
- $\det(A - \lambda I_n)$  is called the **characteristic polynomial** of  $A$ , a polynomial of degree  $n$ .

Let  $R$  be the reduced row echelon form of an  $n \times n$  matrix  $A$ .

- The characteristic polynomial of  $A$  is not usually equal to the characteristic polynomial of  $R$ !
- In general, the eigenvalues of  $A$  and  $R$  are not the same!
- The eigenvectors of  $A$  and  $R$  are not usually the same!

The eigenvalues of an upper triangular or low triangular matrix are its diagonal entries.

For a linear operator  $T$ :

- The characteristic equation of the standard matrix of  $T$  is called the **characteristic equation** of  $T$ .
- The characteristic polynomial of the standard matrix of  $T$  is called the **characteristic polynomial** of  $T$ .
- The characteristic polynomial of a linear operator  $T$  on  $R^n$  is a polynomial of degree  $n$ , whose roots are the eigenvalues of  $T$ .

## The multiplicity of an eigenvalue:

If  $\lambda$  is an eigenvalue of an  $n \times n$  matrix  $M$ , then the largest positive integer  $k$  such that  $(t - \lambda)^k$  is a factor of the characteristic polynomial of  $M$  is called the **multiplicity** of  $\lambda$ .

- For matrix  $A = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 1 \end{bmatrix}$ ,  $\det(A - tI_3) = -(t + 1)^2(t - 3)$ .
- For matrix  $B = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 1 \end{bmatrix}$ ,  $\det(B - tI_3) = -(t + 1)(t - 3)^2$ .

### Theorem 5.1

Let  $\lambda$  be an eigenvalue of a matrix  $A$ . The dimension of the eigenspace of  $A$  corresponding to  $\lambda$  is less than or equal to the multiplicity of  $\lambda$ .

*(Geometric multiplicity  $\leq$  Algebraic multiplicity)*

# The eigenvalue of similar matrices

Matrices  $A$  and  $B$  are similar if there exists an invertible matrix  $P$  such that  $B=P^{-1}AP$ .

By Theorem 3.4 we have

$$\begin{aligned}\det(B - tI_n) &= \det(P^{-1}AP - tP^{-1}I_nP) \\ &= \det(P^{-1}(A - tI_n)P) \\ &= (\det P^{-1}) [\det(A - tI_n)] (\det P) \\ &= \det(A - tI_n)\end{aligned}$$

Similar matrices have the same characteristic polynomial and hence have the same eigenvalues and multiplicities. In addition, their eigenspaces corresponding to the same eigenvalue have the same dimension.

Complex eigenvalues ( $R \Rightarrow C$ ;  $R^n \Rightarrow C^n$ ; Self-study)

## 5.3 Diagonalization of Matrices

**Definition** An  $n \times n$  matrix  $A$  is called diagonalizable if  $A = PDP^{-1}$  for some diagonal  $n \times n$  matrix  $D$  and some invertible  $n \times n$  matrix  $P$ .

- Because  $A = PDP^{-1}$  can be written as  $D = P^{-1}AP$ , a diagonalizable matrix  $A$  is *similar* to a diagonal matrix  $D$ .
  - ✓ The eigenvalues of  $A$  are the diagonal entries of  $D$ .
- Every diagonal matrix is diagonalizable.
- Not every matrix is diagonalizable.

## Theorem 5.2

An  $n \times n$  matrix  $A$  is diagonalizable iff there is a basis for  $\mathbb{R}^n$  consisting of eigenvectors of  $A$ .

Furthermore,  $A = PDP^{-1}$ , where  $D$  is a diagonal matrix and  $P$  is an invertible matrix iff the columns of  $P$  are a basis for  $\mathbb{R}^n$  consisting of eigenvectors of  $A$  and the diagonal entries of  $D$  are the eigenvalues corresponding to the respective columns of  $P$ .

- Proof ?
- The matrices  $P$ , a modal matrix for  $A$ , and  $D$ , a spectral matrix for  $A$ , such that  $PDP^{-1} = A$  in Theorem 5.2 are not unique.

### Theorem 5.3

A set of eigenvectors of a square matrix that correspond to distinct eigenvalues is linearly independent.

- Proof ?
- Thus, an  $n \times n$  matrix having  $n$  distinct eigenvalues must have  $n$  linearly independent eigenvectors.
- If the bases for distinct eigenspaces are combined, then the resulting set is linearly independent.

Every  $n \times n$  matrix having  $n$  distinct eigenvalues is diagonalizable.

## Algorithm for Matrix Diagonalization

Let  $A$  be a diagonalizable  $n \times n$  matrix. Combining bases for each eigenspace of  $A$  forms a basis  $\mathcal{B}$  for  $R^n$  consisting of eigenvectors of  $A$ . Therefore, if  $P$  is the matrix whose columns are the vectors in  $\mathcal{B}$  and  $D$  is the diagonal matrix whose diagonal entries are eigenvalues of  $A$  corresponding to the respective columns of  $P$ , then  $A = PDP^{-1}$ .

## Test for a Diagonalizable Matrix Whose Characteristic Polynomial is Known

An  $n \times n$  matrix  $A$  is diagonalizable iff both of the following conditions are true:

1. The total number of eigenvalues of  $A$ , when each eigenvalue is counted as often as its multiplicity, is equal to  $n$ .
2. For each eigenvalue  $\lambda$  of  $A$ , the dimension of the corresponding eigenspace, which is  $n - \text{rank}(A - \lambda I_n)$ , is equal to the multiplicity of  $\lambda$ .

(Item 2 states that *Geometric multiplicity = Algebraic multiplicity for each eigenvalue.*)

- By Theorem 5.1, the eigenspace corresponding to an eigenvalue of multiplicity 1 must have dimension 1. Hence condition (2) need be checked only for eigenvalues of multiplicity greater than 1.

## 5.4\* Diagonalization of linear operators

A linear operator on  $R^n$  is defined to be **diagonalizable** if there is a basis for  $R^n$  consisting of eigenvectors of the operator.

- *A linear operator is diagonalizable iff its standard matrix is diagonalizable.*

Let  $T$  be a linear operator on  $R^n$  for which there is a basis  $\mathcal{B}=\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  consisting of eigenvectors of  $T$ .

- Then,  $T(\mathbf{v}_i)=\lambda_i \mathbf{v}_i$  for each  $i$ , where  $\lambda_i$  is the eigenvalue corresponding to  $\mathbf{v}_i$ .
- Thus  $[T(\mathbf{v}_i)]_{\mathcal{B}}=\lambda_i \mathbf{e}_i$  for each  $i$ , and so  $[T]_{\mathcal{B}}=[ [T(\mathbf{v}_1)]_{\mathcal{B}} \ [T(\mathbf{v}_2)]_{\mathcal{B}} \ \dots \ [T(\mathbf{v}_n)]_{\mathcal{B}} ] = [\lambda_1 \mathbf{e}_1 \ \lambda_2 \mathbf{e}_2 \ \dots \ \lambda_n \mathbf{e}_n ]$  is a diagonal matrix.

A linear operator  $T$  on  $R^n$  is diagonalizable iff there is a basis  $\mathcal{B}$  for  $R^n$  such that  $[T]_{\mathcal{B}}$ , the  $\mathcal{B}$ -matrix of  $T$ , is a diagonal matrix. Such a basis  $\mathcal{B}$  must consist of eigenvectors of  $T$ .

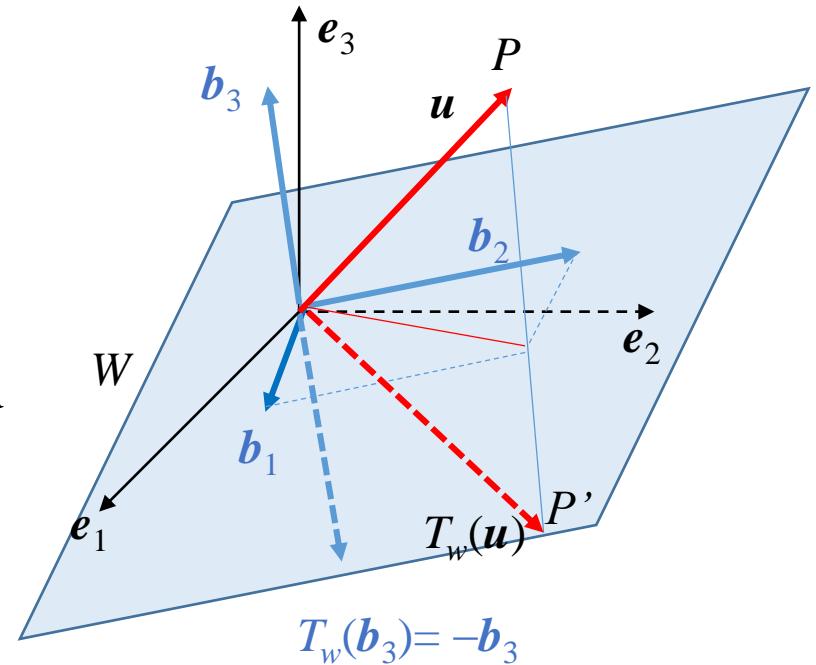
By Theorem 4.12, the  $\mathcal{B}$ -matrix of  $T$  is given by that  $[T]_{\mathcal{B}} = B^{-1}AB$ , where  $B$  is the matrix whose columns are the vectors in  $\mathcal{B}$  and  $A$  is the standard matrix of  $T$ . So, let  $\mathcal{B}$  consist of eigenvectors of  $T$ . Then  $[T]_{\mathcal{B}}$  is the diagonal matrix whose diagonal entries are .....

# A Reflection Operator

- Let  $W$  be a 2-dimensional subspace of  $\mathbb{R}^3$ , a plane containing the origin.
- $T_w: \mathbb{R}^3 \rightarrow \mathbb{R}^3$ , defined as follows: For a vector  $\mathbf{u}$  in  $\mathbb{R}^3$  with end point  $P$ , drop a perpendicular from  $P$  to  $W$ , and extend this perpendicular an equal distance to the point  $P'$  on the other side of  $W$ . Then  $T_w(\mathbf{u})$  is the vector with endpoint  $P'$ .
- The set  $\mathcal{B} = \{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  is linearly independent and hence a basis for  $\mathbb{R}^3$ .
- $T_w(\mathbf{b}_1) = \mathbf{b}_1$ ,  $T_w(\mathbf{b}_2) = \mathbf{b}_2$ ,  $T_w(\mathbf{b}_3) = -\mathbf{b}_3$ .
  - ✓  $\mathbf{b}_1$ ,  $\mathbf{b}_2$ , and  $\mathbf{b}_3$  are eigenvectors of  $T_w$  with corresponding eigenvalues 1, 1, and  $-1$ , respectively

✓  $[T_w(\mathbf{b}_1)]_{\mathcal{B}} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ ,  $[T_w(\mathbf{b}_2)]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ ,  $[T_w(\mathbf{b}_3)]_{\mathcal{B}} = \begin{bmatrix} 0 \\ 0 \\ -1 \end{bmatrix}$ , and

$$[T_w]_{\mathcal{B}} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix}$$



# 5.5\* Applications of Eigenvalues

Markov chain

- States
- Transition matrix
- Regular (All are *recurrent* states in one class)

## Theorem 5.4

If  $A$  is a *regular*  $n \times n$  transition matrix and  $\mathbf{p}$  is a probability vector in  $R^n$ , then

- a) 1 is an eigenvalue of  $A$ ;
- b) there is a unique probability vector  $\mathbf{v}$  of  $A$  that is also an eigenvector corresponding to eigenvalue 1;
- c) The vectors  $A^m\mathbf{p}$  approach  $\mathbf{v}$  for  $m=1, 2, 3, \dots$ .

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