

Chapter 5 Eigenvalues, Eigenvectors, and Diagonalization

5.1 Eigenvalues and Eigenvectors

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

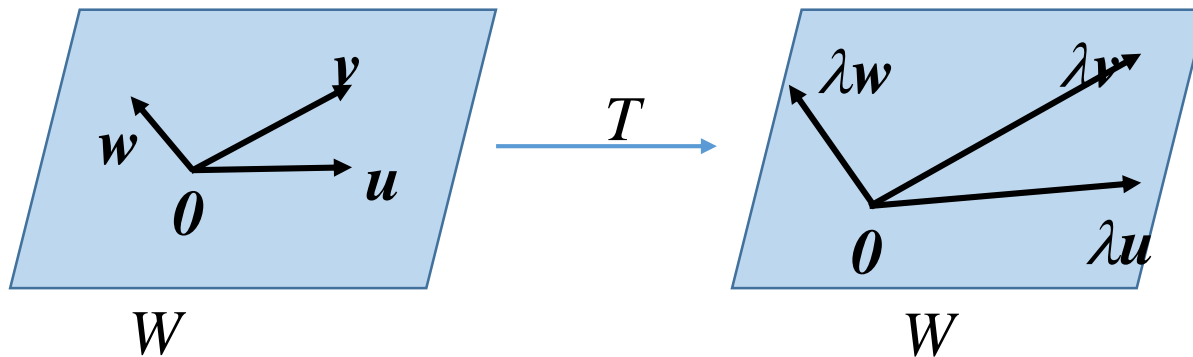
Definitions Let A be an $n \times n$ matrix. A *nonzero* vector \mathbf{v} in R^n is called an **eigenvector** of A if $A\mathbf{v} = \lambda \mathbf{v}$ for some scalar λ . The scalar λ is called the **eigenvalue** of A that corresponds to \mathbf{v} .

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

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

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

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



W is the eigenspace of T corresponding to eigenvalue λ .

5.2 The characteristic polynomial

- If λ is an eigenvalue of A , there must be a nonzero vector \mathbf{v} in \mathbb{R}^n such that $A\mathbf{v}=\lambda\mathbf{v}$.
 - ✓ For an $n \times n$ matrix A and the homogeneous system of linear equations $(A-\lambda I_n)\mathbf{x}=0$ 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 λ 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 λ .

- 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 λ be an eigenvalue of a matrix A . The dimension of the eigenspace of A corresponding to λ is less than or equal to the multiplicity of λ .

(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 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 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 \mathbb{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 λ of A , the dimension of the corresponding eigenspace, which is $n - \text{rank}(A - \lambda I_n)$, is equal to the multiplicity of λ .

(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 λ_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

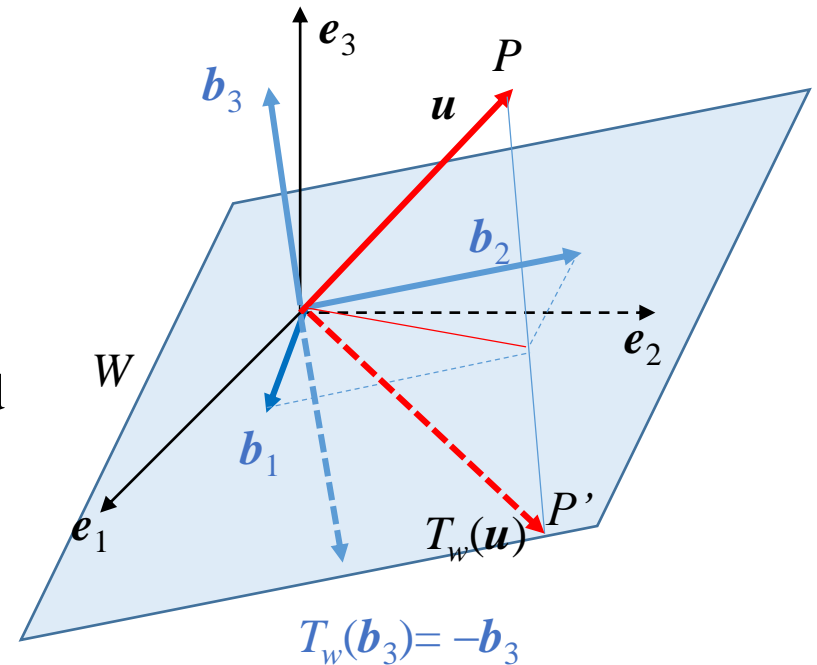
$[T]_{\mathcal{B}} = [[T(\mathbf{v}_1)]_{\mathcal{B}} \quad [T(\mathbf{v}_2)]_{\mathcal{B}} \quad \dots \quad [T(\mathbf{v}_n)]_{\mathcal{B}}] = [\lambda_1 \mathbf{e}_1 \quad \lambda_2 \mathbf{e}_2 \quad \dots \quad \lambda_n \mathbf{e}_n]$
is a diagonal matrix.

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 \mathcal{R}^3 , a plane containing the origin.
- $T_w: \mathcal{R}^3 \rightarrow \mathcal{R}^3$, defined as follows: For a vector \mathbf{u} in \mathcal{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 \mathcal{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

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

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

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

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

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