

## 1 Eigenvalue, Eigenvector and Diagonalizability

Just like the space  $\mathbb{R}^n$ , we also define the space  $\mathbb{C}^n$ . Indeed,

$$\mathbb{C}^n = \{[x_1, x_2, \dots, x_n]^t : x_1, x_2, \dots, x_n \in \mathbb{C}\}.$$

The definitions of vector addition and scalar multiplication *etc.*, and most of the results that we have studied so far in case of  $\mathbb{R}^n$ , can also be accomplished for the space  $\mathbb{C}^n$ , in a similar manner.

**Definition 1.1.** Let  $A$  be an  $n \times n$  matrix. A complex number  $\lambda$  is called an **eigenvalue** of  $A$  if there is a vector  $\mathbf{x} \in \mathbb{C}^n, \mathbf{x} \neq \mathbf{0}$  such that  $A\mathbf{x} = \lambda\mathbf{x}$ . Such a vector  $\mathbf{x}$  is called an **eigenvector** of  $A$  corresponding to  $\lambda$ .

**Example 1.1.** The numbers  $4, -2$  are eigenvalues of  $A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$  with corresponding eigenvectors  $[1, 1]^t$  and  $[1, -1]^t$ , respectively.

**Definition 1.2.** Let  $\lambda$  be an eigenvalue of a matrix  $A$ . Then the collection of all eigenvectors of  $A$  corresponding to  $\lambda$ , together with the zero vector, is called the **eigenspace** of  $\lambda$ , and is denoted by  $E_\lambda$ .

**Result 1.1.** Let  $A$  be an  $n \times n$  matrix and let  $\lambda$  be an eigenvalue of  $A$ . Then

- $\lambda$  is an eigenvalue of  $A$  iff  $\det(A - \lambda I) = 0$ .
- $0$  is an eigenvalue of  $A$  iff  $A$  is not invertible.
- $E_\lambda = \text{null}(A - \lambda I)$ , that is,  $E_\lambda$  is a subspace of  $\mathbb{C}^n$ .
- Let  $\mathbf{v}_1, \dots, \mathbf{v}_k$  be eigenvectors of  $A$  corresponding to  $\lambda$  and  $\mathbf{v} = \alpha_1\mathbf{v}_1 + \dots + \alpha_k\mathbf{v}_k \neq \mathbf{0}$ . Then  $\mathbf{v}$  is also an eigenvector of  $A$  corresponding to  $\lambda$ .
- Eigenvalues of a **triangular** matrix are its **diagonal** entries.
- Eigenvalues of  $\left[ \begin{array}{c|c} A_p & C \\ \hline O & B_q \end{array} \right]$  are the eigenvalues of  $A$  and  $B$ .

**Definition 1.3.** Let  $A$  be an  $n \times n$  matrix. Then

- $P_A(x) = \det(A - xI)$  is called **characteristic polynomial** of  $A$ .
- $P_A(x) = 0$  is called **characteristic equation** of  $A$ .

**Example 1.2.** Find the eigenvalues and the corresponding eigenspaces of the following matrices:

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 2 & -5 & 4 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} -1 & 0 & 1 \\ 3 & 0 & -3 \\ 1 & 0 & -1 \end{bmatrix}.$$

**Result 1.2 (The Fundamental Theorem of Invertible Matrices: Version II).** Let  $A$  be an  $n \times n$  matrix. Then the following statements are equivalent.

1.  $A$  is invertible.
2.  $A^t$  is invertible.
3.  $A\mathbf{x} = \mathbf{b}$  has a solution for every  $\mathbf{b}$  in  $\mathbb{R}^n$ .
4.  $A\mathbf{x} = \mathbf{b}$  has a unique solution for every  $\mathbf{b}$  in  $\mathbb{R}^n$ .
5.  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
6. The reduced row echelon form of  $A$  is  $I_n$ .

7. The rows of  $A$  are linearly independent.
8. The columns of  $A$  are linearly independent.
9.  $\text{rank}(A) = n$ .
10.  $A$  is a product of elementary matrices.
11.  $\text{nullity}(A) = 0$ .
12. The column vectors of  $A$  span  $\mathbb{R}^n$ .
13. The column vectors of  $A$  form a basis for  $\mathbb{R}^n$ .
14. The row vectors of  $A$  span  $\mathbb{R}^n$ .
15. The row vectors of  $A$  form a basis for  $\mathbb{R}^n$ .
16.  $\det A \neq 0$ .
17.  $0$  is **not** an eigenvalue of  $A$ .

**Result 1.3.** Let  $A$  be a matrix with eigenvalue  $\lambda$  and corresponding eigenvector  $\mathbf{x}$ .

1. For any positive integer  $n$ ,  $\lambda^n$  is an eigenvalue of  $A^n$  with corresponding eigenvector  $\mathbf{x}$ .
2. If  $A$  is invertible, then  $\frac{1}{\lambda}$  is an eigenvalue of  $A^{-1}$  with corresponding eigenvector  $\mathbf{x}$ .
3. If  $A$  is invertible then for any integer  $n$ ,  $\lambda^n$  is an eigenvalue of  $A^n$  with corresponding eigenvector  $\mathbf{x}$ .

**Result 1.4.** Let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  be eigenvectors of a matrix  $A$  with corresponding eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_m$ , respectively. Let  $\mathbf{x} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_m\mathbf{v}_m$ . Then for any positive integer  $k$ ,

$$A^k \mathbf{x} = c_1 \lambda_1^k \mathbf{v}_1 + c_2 \lambda_2^k \mathbf{v}_2 + \dots + c_m \lambda_m^k \mathbf{v}_m.$$

**Result 1.5.** Let  $\lambda_1, \lambda_2, \dots, \lambda_m$  be distinct eigenvalues of a matrix  $A$  with corresponding eigenvectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$ , respectively. Then the set  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$  is linearly independent.

**Similar Matrices:** Let  $A$  and  $B$  be two  $n \times n$  matrices. Then  $A$  is said to be **similar** to  $B$  if there is an  $n \times n$  invertible matrix  $T$  such that  $T^{-1}AT = B$ .

- If  $A$  is similar to  $B$ , we write  $A \approx B$ .
- If  $A \approx B$ , we can equivalently write that  $A = TBT^{-1}$  or  $AT = TB$ .

**Example 1.3.** Let  $A = \begin{bmatrix} 1 & 2 \\ 0 & -1 \end{bmatrix}$ ,  $B = \begin{bmatrix} 1 & 0 \\ -2 & -1 \end{bmatrix}$  and  $T = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$ . Then  $A \approx B$  since  $AT = TB$ .

**Result 1.6.** Let  $A, B$  and  $C$  be  $n \times n$  matrices. Then

1.  $A \approx A$ .
2. If  $A \approx B$  then  $B \approx A$ .
3. If  $A \approx B$  and  $B \approx C$  then  $A \approx C$ .

**Result 1.7.** Let  $A$  and  $B$  be two matrices such that  $A \approx B$ . Then

1.  $\det A = \det B$ .
2.  $A$  is invertible iff  $B$  is invertible.
3.  $A$  and  $B$  have the same rank.
4.  $A$  and  $B$  have the same characteristic polynomial.
5.  $A$  and  $B$  have the same set of eigenvalues.

6.  $\lambda$  is an eigenvalue of  $B$  with corresponding eigenvector  $\mathbf{v}$  **iff**  $\lambda$  is an eigenvalue of  $A$  with corresponding eigenvector  $T\mathbf{v}$ .

7. The  $\dim(E_\lambda)$  for  $A$  is same as  $\dim(E_\lambda)$  for  $B$ .

**Diagonalizable Matrix:** A matrix  $A$  is said to be **diagonalizable** if there is a diagonal matrix  $D$  such that  $A \approx D$ , that is, if there is an invertible matrix  $T$  and a diagonal matrix  $D$  such that  $AT = TD$ .

**Example 1.4.** The matrix  $A = \begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix}$  is diagonalizable, since if  $D = \begin{bmatrix} 4 & 0 \\ 0 & -1 \end{bmatrix}$  and  $T = \begin{bmatrix} 1 & 3 \\ 1 & -2 \end{bmatrix}$  then  $AT = TD$ .

**Result 1.8.** Let  $A$  be an  $n \times n$  matrix. Then  $A$  is diagonalizable iff  $A$  has  $n$  linearly independent eigenvectors.

- Let  $A$  be an  $n \times n$  matrix. Then there exists an invertible matrix  $T$  and a diagonal matrix  $D$  satisfying  $T^{-1}AT = D$  **iff** the columns of  $T$  are  $n$  linearly independent eigenvectors of  $A$  and the diagonal entries of  $D$  are the eigenvalues of  $A$  corresponding to the columns (eigenvectors of  $A$ ) of  $T$  in the same order.

**Example 1.5.** Check for the diagonalizability of the following matrices. If they are diagonalizable, find invertible matrices  $T$  that diagonalizes them:

$$\begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 2 & -5 & 4 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} -1 & 0 & 1 \\ 3 & 0 & -3 \\ 1 & 0 & -1 \end{bmatrix}.$$

**Result 1.9.** If  $A$  is an  $n \times n$  matrix with  $n$  distinct eigenvalues then  $A$  is diagonalizable.

**Result 1.10.** Let  $\lambda_1, \lambda_2, \dots, \lambda_k$  be distinct eigenvalues of a matrix  $A$ . If  $\mathcal{B}_i$  is a basis for the eigenspace  $E_{\lambda_i}$ , then  $\mathcal{B} = \mathcal{B}_1 \cup \mathcal{B}_2 \cup \dots \cup \mathcal{B}_k$  is a linearly independent set.

**Definition 1.4.** Let  $\lambda$  be an eigenvalue of a matrix  $A$ .

- The **algebraic multiplicity** of  $\lambda$  is the multiplicity of  $\lambda$  as a root of the characteristic polynomial of  $A$ .
- The **geometric multiplicity** of  $\lambda$  is the dimension of  $E_\lambda$ .

**Result 1.11.** The geometric multiplicity of each eigenvalue of a matrix is less than or equal to its algebraic multiplicity.

**Result 1.12 (The Diagonalization Theorem).** Let  $A$  be an  $n \times n$  matrix whose distinct eigenvalues are  $\lambda_1, \lambda_2, \dots, \lambda_k$ . Then the following statements are equivalent:

1.  $A$  is diagonalizable.
2. The union  $\mathcal{B}$  of the bases of the eigenspaces of  $A$  contains  $n$  vectors.
3. The algebraic multiplicity of each eigenvalue equals its geometric multiplicity.