

## 1 Eigenvalues & Eigenvectors

**Definition 1.1 (Eigenvalue & Eigenvector)** Let  $A \in \mathbb{C}^{n \times n}$ . Then  $\lambda \in \mathbb{C}$  is said to be an eigenvalue of  $A$  if  $\exists v \neq 0$  such that

$$Av = \lambda v \iff (A - \lambda I)v = 0v.$$

Such a  $v$  is called eigenvector of  $A$  with eigenvalue  $\lambda$ .

**Exercise 1.2**  $\lambda$  is an eigenvalue of  $A$  if and only if  $\det(\lambda I - A) = 0$

**Exercise 1.3** Let  $U_\lambda = \{v : Av = \lambda v\}$ . Prove:  $U_\lambda$  is a subspace of  $\mathbb{C}^n$ .

**Definition 1.4 (Characteristic Polynomial)**  $f_A(t) = \det(tI - A)$  is called the characteristic polynomial of  $A$ .

From the above, we know that  $\lambda$  is an eigenvalue of  $A$  iff  $\lambda$  is a root of the characteristic polynomial  $f_A(t)$ .

**Example 1.5** Consider  $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ . The characteristic polynomial is  $f_A(t) = \det(tI - A) = (t - 1)^2$ . The only eigenvalue is  $\lambda = 1$ , and  $U_\lambda = \mathbb{C}^2$ , since  $Av = Iv = v$  for all  $v \in \mathbb{C}^2$ .

**Example 1.6** Consider  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ . The characteristic polynomial is still  $f_A(t) = \det(tI - A) = (t - 1)^2$  and the only eigenvalue is  $\lambda = 1$ . However,  $U_\lambda$  is now only the one-dimensional space

$$U_\lambda = \left\{ \alpha \cdot \begin{pmatrix} 1 \\ 0 \end{pmatrix} \mid \alpha \in \mathbb{C} \right\}.$$

**Exercise 1.7** Calculate the eigenvalues of  $A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ .

Note that the eigenvalue of a real matrix may be complex valued. Also, any polynomial  $p(t)$  of degree  $n$  over  $\mathbb{C}$  can be factored as  $c(t - \lambda_1)\dots(t - \lambda_n)$  for  $c, \lambda_1, \dots, \lambda_n \in \mathbb{C}$ . Thus, the characteristic polynomial of a matrix always has  $n$  (not necessarily distinct) complex roots.

**Definition 1.8 (Algebraic Multiplicity)** The algebraic multiplicity of an eigenvalue  $\lambda$  is the number of times  $t - \lambda$  appears as a factor in the characteristic polynomial  $f_A(t)$ .

**Definition 1.9 (Geometric Multiplicity)** The geometric multiplicity of an eigenvalue  $\lambda$  is  $\dim(U_\lambda)$ .

**Exercise 1.10** Algebraic multiplicity  $\geq$  geometric multiplicity.

Note that the matrix  $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  gives an example where the algebraic multiplicity is strictly greater than the geometric multiplicity of an eigenvalue.

**Exercise 1.11** Let  $Av = \lambda v$  for  $v \in \mathbb{C}^n, A \in \mathbb{R}^{n \times n}, \lambda \in \mathbb{R}$ . Then  $\operatorname{Re}(v), \operatorname{Im}(v)$  are also eigenvectors with eigenvalue  $\lambda$ .

Thus, if a real matrix  $A$  has an eigenvector with a real eigenvalue  $\lambda \in \mathbb{R}$ , then it also has a real eigenvector with the same eigenvalue.

**Example 1.12** Calculate the eigenvalues, eigenvectors, and the characteristic polynomial for the rotation matrix,  $R_\theta = \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix}$ .

$f_{R_\theta}(t) = (t - \cos\theta)^2 + (\sin\theta)^2 = 0$  gives  $t = \cos\theta \pm i\sin\theta = e^{\pm i\theta}$ . Let  $\lambda_1 = \cos\theta + i\sin\theta$ ,  $\lambda_2 = \cos\theta - i\sin\theta$ . Then  $U_{\lambda_1} = \{\alpha \begin{pmatrix} i \\ 1 \end{pmatrix} : \alpha \in \mathbb{C}\}$ ,  $U_{\lambda_2} = \{\alpha \begin{pmatrix} 1 \\ i \end{pmatrix} : \alpha \in \mathbb{C}\}$ . Note that the eigenvectors do not depend on  $\theta$ .

**Exercise 1.13** If  $v_1, \dots, v_n$  are eigenvectors with distinct eigenvalues  $\lambda_1, \dots, \lambda_n$ , then  $v_1, \dots, v_n$  are linearly independent.

**Definition 1.14 (Similar Matrices)**  $A$  and  $B$  are similar,  $A \sim B$ , if  $\exists S \in \mathbb{C}^{n \times n}$  such that  $A = S^{-1}BS$ .

**Definition 1.15 (Diagonalizable Matrices)**  $A$  is diagonalizable if  $A = S^{-1}DS$  where  $D$  is a diagonal matrix. ( $A$  is similar to a diagonal matrix).

**Exercise 1.16** Prove that similarity between matrices is an equivalence relation.

**Exercise 1.17** Prove: If  $A \sim B$ , then  $f_A(t) = f_B(t)$ . Thus, if  $A \sim B$ , then they have the same eigenvalues and each eigenvalue has the same algebraic multiplicity for both  $A$  and  $B$ .

**Exercise 1.18** Let  $A \sim B$ . Then for each  $\lambda$  which is an eigenvalue of  $A$  (and hence also of  $B$ ), show that  $U_\lambda^{(A)}$  is isomorphic to  $U_\lambda^{(B)}$ . Thus, each eigenvalue also has the same geometric multiplicity for both  $A$  and  $B$ .

This follows by noting that  $S^{-1}BSv = \lambda v \Rightarrow BSv = \lambda Sv$ . Hence,  $v \mapsto Sv$  is a bijective linear map from  $U_\lambda^{(A)}$  to  $U_\lambda^{(B)}$ .

## 2 Inner Products and Unitary Matrices

**Definition 2.1 (Inner Product)** Let  $u, v \in \mathbb{C}^n$ . Then the Hermitian inner product of  $u$  and  $v$  is defined as

$$\langle u, v \rangle = \sum_{i=1}^n \overline{u_i} v_i,$$

where  $\overline{u_i}$  denotes the conjugate of  $u_i$ . Note that this is the same as the usual dot-product if  $u, v \in \mathbb{R}^n$ .

Note that the quantity  $\langle u, u \rangle = \sum_i |u_i|^2$  is always non-negative and is zero only when  $u = 0$ . We define  $\|u\| = \sqrt{\langle u, u \rangle}$ , which extends the notion of *length* of a vector, to vectors in  $\mathbb{C}^n$ .

**Definition 2.2 (Orthogonal Vectors)** Two vectors  $u, v$  are said to be orthogonal if  $\langle u, v \rangle = 0$ .

**Definition 2.3 (Orthonormal Basis)**  $\{v_1, \dots, v_k\}$  is an orthonormal basis for  $V$  if it is a basis such that

$$\forall i, j \quad \langle v_i, v_j \rangle = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}.$$

**Example 2.4** For the rotation matrix  $R_\theta$ , the eigenvectors  $\begin{pmatrix} 1 \\ i \end{pmatrix}$  and  $\begin{pmatrix} 1 \\ -i \end{pmatrix}$  are orthogonal. After scaling, the vectors to  $\frac{1}{\sqrt{2}} \cdot \begin{pmatrix} 1 \\ i \end{pmatrix}$  and  $\frac{1}{\sqrt{2}} \cdot \begin{pmatrix} 1 \\ -i \end{pmatrix}$ , we obtain an orthonormal basis of  $\mathbb{C}^2$  consisting of eigenvectors of  $R_\theta$ .

**Definition 2.5 (Unitary Matrix)** A matrix  $U$  is called a unitary matrix if the columns of  $U$  form an orthonormal basis of  $\mathbb{C}^n$ .

**Definition 2.6 (Adjoint of a matrix)** For a matrix  $A \in M_n(\mathbb{C})$ , its adjoint, denoted as  $A^*$  is the matrix defined as  $(A^*)_{ij} = \overline{a_{ji}}$  i.e.,  $A^* = \overline{A}^T$ .

Note that it follows from the fact that  $(AB)^T = B^T A^T$  that we have  $(AB)^* = B^* A^*$ .

**Exercise 2.7**  $U$  is unitary if and only if  $U^*U = UU^* = I$ .

Note that  $(U^*U)_{ij} = \langle U^{(i)}, U^{(j)} \rangle$ , where  $U^{(i)}$  denotes the  $i$ th column of  $U$ . Hence  $U^*U = I$  if and only if the columns form an orthonormal basis. Also, we have that

$$U^*U = I \Leftrightarrow UU^* = I^* = I.$$

**Exercise 2.8** Show that if  $U_1, U_2$  are unitary, then so is  $U_1 U_2$ .