

## Review of Eigenvalues & Eigenvectors

**Definition 1.** An **eigenvector** of an  $n \times n$  matrix  $\mathbf{A}$  is a nonzero vector  $\mathbf{v}$  for some scalar  $\lambda$ . A scalar  $\lambda$  is called an **eigenvalue** of  $\mathbf{A}$  if there is a nonzero (nontrivial) solution  $\mathbf{v}$  of  $\mathbf{A}\mathbf{v} = \lambda \mathbf{v}$ ; such a  $\mathbf{v}$  is called an *eigenvector corresponding to  $\lambda$* .

*Remark 1.* From Definition-1,  $\lambda$  is an eigenvalue of an  $n \times n$  matrix  $\mathbf{A}$  if and only if the equation

$$(\mathbf{A} - \lambda \mathbf{I}) \mathbf{v} = \mathbf{0} \quad (1)$$

has a nontrivial solution ( $\mathbf{v} \neq \mathbf{0}$ ). Since  $\mathbf{v}$  is a nonzero vector, the matrix  $(\mathbf{A} - \lambda \mathbf{I})$  must be singular (*non-invertible*). A matrix is singular if and only if its determinant is zero, thus, the eigenvalues of  $\mathbf{A}$  are the scalars ( $\lambda$ ) for which

$$\det(\mathbf{A} - \lambda \mathbf{I}) = 0. \quad (2)$$

The determinant in (2) is a polynomial of degree  $n$  called *characteristic polynomial* of  $\mathbf{A}$ . The eigenvalues of  $\mathbf{A}$  are the roots of the *characteristic equation*. Since a polynomial of degree  $n$  has exactly  $n$  roots, a matrix of order  $n$  has  $n$  eigenvalues.

What does it mean for a matrix  $\mathbf{A}$  to have an eigenvalue of 0?

**Ans.** This happens if and only if the equation

$$\mathbf{A} \mathbf{v} = \mathbf{0} \mathbf{v} \quad (3)$$

has a nontrivial solution (according to the Definition 1). But (3) is equivalent to  $\mathbf{A} \mathbf{v} = \mathbf{0}$ , which has a nontrivial solution if only if  $\mathbf{A}$  is not *invertible*.

*Remark 2.* Let the matrix  $\mathbf{A}$  of order  $n$  have  $n$  distinct eigenvalues as

$$\mathbf{V} = \mathbf{v}_1, \dots, \mathbf{v}_n \quad \text{and} \quad \mathbf{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_n),$$

and hence  $\mathbf{A}$  has a complete system of eigenpairs  $(\lambda_i, \mathbf{v}_i), i = 1, \dots, n$ . The the individual relations  $\mathbf{A}\mathbf{v}_i = \lambda_i \mathbf{v}_i$  can be combined in the matrix equation

$$\mathbf{A} \mathbf{V} = \mathbf{V} \mathbf{\Lambda}. \quad (4)$$

Because the eigenvectors  $\mathbf{v}_i$  are linearly independent, the matrix  $\mathbf{V}$  is nonsingular. Hence,  $\mathbf{A}$  is diagonalizable and we may write

$$\mathbf{V}^{-1} \mathbf{A} \mathbf{V} = \mathbf{\Lambda}. \quad (5)$$

**Proposition 1.** Let a  $n \times n$  matrix  $\mathbf{A}$  be diagonalizeable, it is

$$\begin{aligned}
 \mathbf{A}^k &= (\mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1})^k \\
 &= (\mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1}) \cdots (\mathbf{V} \mathbf{\Lambda} \mathbf{V}^{-1}) \\
 &= \mathbf{V} \mathbf{\Lambda}^k \mathbf{V}^{-1} \\
 &= \mathbf{V} \text{diag}(\lambda_1^k, \dots, \lambda_n^k) \mathbf{V}^{-1}.
 \end{aligned} \tag{6}$$

For the case of  $k < 0$ , (6) is hold only if  $\mathbf{A}$  is invertible, i.e. all  $\lambda_i \neq 0$ .

**Definition 2.** A matrix  $\mathbf{B}_{n \times n}$  is said to be similar to a matrix  $\mathbf{A}_{n \times n}$  if there exists a nonsingular square matrix  $\mathbf{T}_{n \times n}$  such that  $\mathbf{B} = \mathbf{T}^{-1} \mathbf{A} \mathbf{T}$ .

The transformation  $\mathbf{A} \rightarrow \mathbf{T}^{-1} \mathbf{A} \mathbf{T}$  is called a *similarity transformation* by the *similarity matrix*  $\mathbf{T}$ . The relation “ $\mathbf{B}$  is similar to  $\mathbf{A}$ ” is sometimes abbreviated  $\mathbf{B} \sim \mathbf{A}$ .

*Note 1.* Defining  $\mathbf{Q} = \mathbf{T}^{-1}$  similarity transformation can be equivalently rewritten as  $\mathbf{B} = \mathbf{Q} \mathbf{A} \mathbf{Q}^{-1}$ , where (now)  $\mathbf{Q}$  is the *similarity matrix*.

*Note 2.* Recall the following properties of determinant of square matrices,

- $\det \mathbf{A} \det \mathbf{B} = \det (\mathbf{A} \mathbf{B})$
- $\det (\mathbf{P}^{-1}) \det (\mathbf{P}) = \det (\mathbf{P}^{-1} \mathbf{P}) = \det (\mathbf{I}) = 1$ .