

Eigenvalues and Eigenvectors (review):

Definition 1. An **eigenvector** of an $n \times n$ matrix \mathbf{A} is a nonzero vector \mathbf{v} for some scalar λ . A scalar λ 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 λ* .

Remark 1. From Definition-1, λ 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 (λ) 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?

This happens if and only if the equation

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

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 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

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

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,

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