

Eigenvectors and Eigenvalues

CE282: Linear Algebra

Computer Engineering Department Sharif University of Technology

Hamid R. Rabiee

Maryam Ramezani

Review



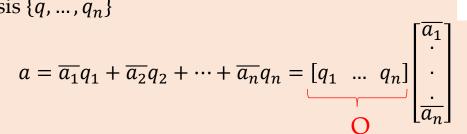
Note

 \square n-vector a based on basis $\{e_1, ..., e_n\}$

$$a = a_1 e_1 + a_2 e_2 + \dots + a_n e_n$$

 \square n-vector a based on new basis $\{q, ..., q_n\}$

$$a = \overline{a_1}q_1 + \overline{a_2}q_2 + \dots + \overline{a_n}q_n = \begin{bmatrix} q_1 & \dots & q_n \end{bmatrix} \begin{bmatrix} a_1 \\ \vdots \\ \overline{a_n} \end{bmatrix}$$



- ☐ Matrix Q is invertible.
- ☐ Any invertible matrix is a basic matrix.

Review



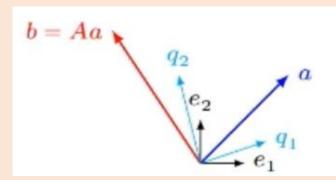
Note

☐ A square matrix for a linear transform

$$A: n \times n$$
 $A: \mathbb{R}^n \to \mathbb{R}^n \Rightarrow Aa = b$ $a. b \in \mathbb{R}^n$

$$\begin{vmatrix}
a = Q\bar{a} \\
b = Q\bar{b}
\end{vmatrix} \Rightarrow AQ\bar{a} = Q\bar{b} \Rightarrow Q^{-1}AQ\bar{a} = \bar{b} \Rightarrow \bar{A}\bar{a} = b$$

$$\bar{A}$$



- \Box Linear transform in new basis $\bar{A} = Q^{-1}AQ$
- \Box \bar{A} is the standard matrix of linear transform in new basis.
- ☐ Similarity Transformation

Similar Matrices



Note

☐ Two n-by-n matrices A and B are called **similar** if there exists an **invertible n-by-n matrix** *Q* such that

$$A = Q^{-1}BQ$$

- \square A and B are similar if QA = BQ
- $\square A = Q^{-1}BQ \rightarrow B = QAQ^{-1}$
- ☐ Same determinant
- ☐ Inverse of A and B are similar (if exists)

Similarity Transformation



☐ We can use similarity transformation for changing the standard matrix of linear transformation

Example

$$A = \begin{bmatrix} 2 & -1 & -1 \\ 1 & 4 & 1 \\ -1 & -1 & 2 \end{bmatrix} \quad Q = \begin{bmatrix} -1 & 1 & 0 \\ 1 & -1 & 1 \\ 0 & 1 & -1 \end{bmatrix} \quad \bar{A} = Q^{-1}AQ = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

Think!



☐ Why trace is a similarity invariant?

☐ Why rank is a similarity invariant?

Motivation

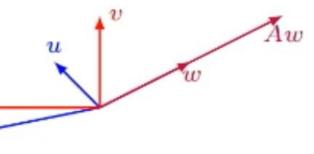


$$\Box A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix}$$

$$u = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow Au = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -5 \\ 1 \end{bmatrix}$$

$$v = \begin{bmatrix} 0 \\ 2 \end{bmatrix} \Rightarrow Av = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 2 \end{bmatrix} = \begin{bmatrix} -4 \\ 0 \end{bmatrix}$$

$$w = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \Rightarrow Aw = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix}$$



Definition



Definition

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

☐ An eigenvector must be nonzero, by definition, but an eigenvalue may be zero.

Example

$$A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix}, v = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \lambda = 2$$

Show that 7 is an eigenvalue of matrix A, and find the corresponding eigenvectors.

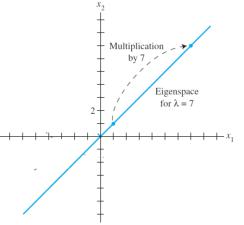
$$A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$$

Eigenspace



Note

has a nontrivial solution. The set of all solutions of (3) is just the null space of the matrix $A - \lambda I$. So this set is the *subspace* of \mathbb{R}^n and is called the **eigenspace** of A corresponding to λ . The eigenspace consists of the zero vector and all the eigenvectors corresponding to λ .



Hamid R. Rabiee & Maryam Ramezani

Characteristic Equation



Note

- $\square Av = \lambda v \Rightarrow Av \lambda vI = 0 \Rightarrow (A \lambda I)v = 0 \quad v \neq 0$
- \Box Characteristic equation $|A \lambda I| = 0$
- □ Characteristic polynomial $|A \lambda I|$ $\Delta_A(\lambda), \Delta(\lambda)$
 - \square Matrix $n \times n$ has eigenvalue

Characteristic Equation



Example

The characteristic polynomial of a 6 × 6 matrix is λ^6 - $4\lambda^5$ - $12\lambda^4$. Find the eigenvalues and their multiplications.

$$A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \quad A = \begin{bmatrix} 2 & 1 \\ -1 & 0 \end{bmatrix} \quad A = \begin{bmatrix} -2 & 1 \\ -2 & 0 \end{bmatrix}$$

Matrix Spectrum



Definition

Set of all eigenvalues of matrix is $\sigma(A)$.

Theorem

The eigenvalues of a triangular (upper/lower/diagonal) matrix are the entries on its main diagonal.

- ☐ Proof?
- $\Box 0 \in \sigma(A) \Leftrightarrow |A| = 0$
- \square *A* is invertible if and only if ...
- \square 0 is an eigenvalue of *A* if and only if *A* is not invertible.

Similar Matrices



- □Similar matrices have equal characteristic equations.
 - □vice versa?

Example

$$\square A = \begin{bmatrix} 2 & -1 & -1 \\ 1 & 4 & 1 \\ -1 & -1 & 2 \end{bmatrix}, A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

$$\square \begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

Eigenvectors Linear Independence



The Invertible Matrix Theorem

If v_1, \dots, v_n are eigenvectors that correspond to distinct eigenvalues $\lambda_1, \dots, \lambda_n$ of an $n \times n$ matrix A, then the set $\{\lambda_1, \dots, \lambda_n\}$ is linearly independent.

- □One way to prove the statement "If P then Q" is to show that P and the negation of Q leads to a contradiction
- □Distinct eigenvalues -> eigenvectors are LI
- □Duplicate eigenvalues -> ???
 - **□**Example

Some notes



The Invertible Matrix Theorem

Let *A* be an $n \times n$ matrix. Then *A* is invertible if and only if:

- \square The number 0 is not an eigenvalue of A.
- \Box The determinant of *A* is not zero.

Warnings

1. The matrices

$$\begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}$$
 and
$$\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

are not similar even though they have the same eigenvalues.

2. Similarity is not the same as row equivalence. (If A is row equivalent to B, then B = EA for some invertible matrix E.) Row operations on a matrix usually change its eigenvalues.

Example



Example

Find eigenvalues and eigenvectors?

$$A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix}$$

$$|A - \lambda I| = \begin{vmatrix} 3 - \lambda & -2 \\ 1 & -\lambda \end{vmatrix} = \lambda^3 - 3\lambda + 2 = 0 \implies \begin{cases} \lambda_1 = 1 \\ \lambda_2 = 2 \end{cases}$$

$$\Rightarrow \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$
$$\Rightarrow \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$$

Diagonalization



Definition

- \square With similarity transformation Q, matrix A changed to a diagonal matrix $diag(\lambda_1,\lambda_2)$
- ☐ Matrix *A* has n linear independent eigenvectors

$$\square \ [Aq_1 \ Aq_2 \ \cdots \ Aq_n] = [q_1 \ q_2 \cdots \ q_n] \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

$$\square$$
 $A[q_1 \ q_2 \ \cdots \ q_n] = Q\Lambda \Longrightarrow AQ = Q\Lambda$

$$\square \Lambda = Q^{-1}AQ^T$$

$$\Box A = Q\Lambda Q^{-1}$$

Diagonalizable



Definition

A matrix A is said to be diagonalizable if A is similar to a diagonal matrix, that is, if $A = PDP^{-1}$ for some invertible matrix P and some diagonal matrix D.

Theorem

An $n \times n$ matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

- \square The columns of *P* is called an eigenvector basis of \mathbb{R}^n .
- \square An $n \times n$ matrix with n distinct eigenvalues is diagonalizable.

Symmetric Matrix



Theorem

If *A* is symmetric, then any two eigenvectors from different eigenspace are orthogonal.

$$Av_1 = \lambda_1 v_1 Av_2 = \lambda_2 v_2 \lambda_1 \neq \lambda_2$$
 $\Rightarrow v_1^T v_2 = 0$

Symmetric Matrix



Important

- ☐ Eigenvalues of a real symmetric matrix are real.
- \square If *A* is diagonalizable by an orthogonal matrix, then *A* is a symmetric matrix.
- ☐ A symmetric matrix is always diagonalizable.
- ☐ A similar transform that diagonalized the symmetric matrix is orthogonal.

$$\square Q^T Q = I$$

$$A = Q\Lambda Q^T,$$

$$\Lambda = \begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix}, \, \lambda_i \in \mathbb{R}$$

Orthogonally Diagonalizable



Theorem

An $n \times n$ matrix A is orthogonally diagonalizable if and only if A is a symmetric matrix.

$$A = A^T \Rightarrow A = Q\Lambda Q^T, \Lambda = diag\{\lambda_1, \dots, \lambda_n\}$$

$$A = A^T \Leftarrow A = Q\Lambda Q^T, \Lambda = diag\{\lambda_1, \cdots, \lambda_n\}$$
$$A^T = (Q\Lambda Q^T)^T = Q\Lambda^T Q^T = Q\Lambda Q^T = A$$

Spectral Theorem



The Spectral Theorem for Symmetric Matrices

An $n \times n$ symmetric matrix A has the following properties:

- *a.* A has *n* real eigenvalues, counting multiplicities.
- b. The dimension of the eigenspace for each eigenvalue λ equals the multiplicity of λ as a root of the characteristic equation.
- c. The eigenspaces are mutually orthogonal, in the sense that eigenvectors corresponding to different eigenvalues are orthogonal.
- d. A is orthogonally diagonalizable.

Gram Matrix



- ☐ Eigenvalues are real.
- ☐ Eigenvalues are nonnegative.