

MATH 423

Linear Algebra II

**Lecture 33:**

**Diagonalization of normal operators.**

## Adjoint operator and adjoint matrix

Given a linear operator  $L$  on an inner product space  $V$ , the **adjoint** of  $L$  is a transformation  $L^* : V \rightarrow V$  satisfying  $\langle L(\mathbf{x}), \mathbf{y} \rangle = \langle \mathbf{x}, L^*(\mathbf{y}) \rangle$  for all  $\mathbf{x}, \mathbf{y} \in V$ .

**Theorem 1** If  $V$  is finite-dimensional, then the adjoint operator  $L^*$  always exists.

Given a matrix  $A$  with complex entries, its **adjoint matrix** is  $A^* = \overline{A^t}$ .

**Theorem 2** If  $A$  is the matrix of a linear operator  $L$  relative to an orthonormal basis  $\beta$ , then the matrix of  $L^*$  relative to the same basis is  $A^*$ .

Let  $L : V \rightarrow V$  be a linear operator on an inner product space  $V$ . Recall that  $\mathcal{N}(L)$  denotes the **null-space** of  $L$  and  $\mathcal{R}(L)$  denotes the **range** of  $L$ :

$$\mathcal{N}(L) = \{\mathbf{x} \in V \mid L(\mathbf{x}) = \mathbf{0}\}, \quad \mathcal{R}(L) = \{L(\mathbf{y}) \mid \mathbf{y} \in V\}.$$

**Theorem** If the adjoint operator  $L^*$  exists, then  $\mathcal{N}(L) = \mathcal{R}(L^*)^\perp$  as well as  $\mathcal{N}(L^*) = \mathcal{R}(L)^\perp$ .

*Proof:*  $\mathbf{x} \in \mathcal{N}(L) \iff L(\mathbf{x}) = \mathbf{0} \iff \langle L(\mathbf{x}), \mathbf{y} \rangle = 0$  for all  $\mathbf{y} \in V \iff \langle \mathbf{x}, L^*(\mathbf{y}) \rangle = 0$  for all  $\mathbf{y} \in V \iff \mathbf{x} \perp \mathcal{R}(L^*) \iff \mathbf{x} \in \mathcal{R}(L^*)^\perp$ .

The second equality follows in the same way since  $(L^*)^* = L$ .

*Example.*  $V = \mathbb{R}^n$  with the dot product,  $L(\mathbf{x}) = A\mathbf{x}$ , where  $A \in \mathcal{M}_{n,n}(\mathbb{R})$  and vectors are regarded as column vectors.

We have  $L^*(\mathbf{x}) = A^*\mathbf{x}$  for all  $\mathbf{x} \in \mathbb{R}^n$ . The range of  $L^*$  is the column space of the matrix  $A^* = A^t$ , which is the row space of  $A$ . Therefore  $\mathcal{N}(L) = \{\text{null-space of the matrix } A\}$  is the orthogonal complement of the row space of  $A$ .

## Normal operators

*Definition.* A linear operator  $L$  on an inner product space  $V$  is called **normal** if it commutes with its adjoint. That is, if the adjoint operator  $L^*$  exists and  $L \circ L^* = L^* \circ L$ .

There are several special classes of normal operators important for applications.

The operator  $L$  is **self-adjoint** if  $L^* = L$ .

Equivalently,  $\langle L(\mathbf{x}), \mathbf{y} \rangle = \langle \mathbf{x}, L(\mathbf{y}) \rangle$  for all  $\mathbf{x}, \mathbf{y} \in V$ .

The operator  $L$  is **anti-selfadjoint** if  $L^* = -L$ .

The operator  $L$  is **unitary** if  $L^* = L^{-1}$ .

## Normal matrices

*Definition.* A square matrix  $A$  with real or complex entries is **normal** if  $AA^* = A^*A$ .

**Theorem** Let  $L$  be a linear operator on a finite-dimensional inner product space. Suppose  $A$  is the matrix of  $L$  relative to an orthonormal basis. Then the operator  $L$  is normal if and only if the matrix  $A$  is normal.

Special classes of normal operators give rise to special classes of normal matrices.

A matrix  $A \in \mathcal{M}_{n,n}(\mathbb{C})$  is **Hermitian** if  $A^* = A$ , **skew-Hermitian** if  $A^* = -A$ , and **unitary** if  $A^* = A^{-1}$ .

A square matrix  $B$  with real entries is **symmetric** if  $B^t = B$ , **skew-symmetric** if  $B^t = -B$ , and **orthogonal** if  $B^t = B^{-1}$ .

## Properties of normal operators

**Theorem** Suppose  $L$  is a normal operator on an inner product space  $V$ . Then

- (i)  $\|L(\mathbf{x})\| = \|L^*(\mathbf{x})\|$  for all  $\mathbf{x} \in V$ ;
- (ii)  $\mathcal{N}(L) = \mathcal{N}(L^*)$ ;
- (iii) an operator given by  $\mathbf{x} \mapsto L(\mathbf{x}) - \lambda\mathbf{x}$  is normal for any scalar  $\lambda$ ;
- (iv) the operators  $L$  and  $L^*$  share eigenvectors; namely, if  $L(\mathbf{v}) = \lambda\mathbf{v}$  then  $L^*(\mathbf{v}) = \bar{\lambda}\mathbf{v}$ ;
- (v) eigenvectors of  $L$  belonging to distinct eigenvalues are orthogonal;
- (vi) if a subspace  $V_0 \subset V$  is invariant under  $L$ , then the orthogonal complement  $V_0^\perp$  is also invariant under  $L$ .

## Properties of normal operators

- $\|L(\mathbf{x})\| = \|L^*(\mathbf{x})\|$  for all  $\mathbf{x} \in V$ .

$$\begin{aligned}\|L(\mathbf{x})\|^2 &= \langle L(\mathbf{x}), L(\mathbf{x}) \rangle = \langle \mathbf{x}, L^*(L(\mathbf{x})) \rangle = \langle \mathbf{x}, L(L^*(\mathbf{x})) \rangle \\ &= \overline{\langle L(L^*(\mathbf{x})), \mathbf{x} \rangle} = \overline{\langle L^*(\mathbf{x}), L^*(\mathbf{x}) \rangle} = \langle L^*(\mathbf{x}), L^*(\mathbf{x}) \rangle = \|L(\mathbf{x})\|^2.\end{aligned}$$

- If  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are eigenvectors of  $L$  belonging to distinct eigenvalues  $\lambda_1$  and  $\lambda_2$ , then  $\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = 0$ .

We have  $L(\mathbf{v}_1) = \lambda_1 \mathbf{v}_1$  and  $L^*(\mathbf{v}_2) = \overline{\lambda_2} \mathbf{v}_2$ . Then

$$\begin{aligned}\lambda_1 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle &= \langle \lambda_1 \mathbf{v}_1, \mathbf{v}_2 \rangle = \langle L(\mathbf{v}_1), \mathbf{v}_2 \rangle = \langle \mathbf{v}_1, L^*(\mathbf{v}_2) \rangle \\ &= \langle \mathbf{v}_1, \overline{\lambda_2} \mathbf{v}_2 \rangle = \lambda_2 \langle \mathbf{v}_1, \mathbf{v}_2 \rangle.\end{aligned}$$

It follows that  $(\lambda_1 - \lambda_2) \langle \mathbf{v}_1, \mathbf{v}_2 \rangle = 0$ . Since  $\lambda_1 \neq \lambda_2$ , we obtain  $\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = 0$ .

## Diagonalization of normal operators

**Theorem** A linear operator  $L$  on a finite-dimensional inner product space  $V$  is normal if and only if there exists an orthonormal basis for  $V$  consisting of eigenvectors of  $L$ .

*Proof ("if"):* Suppose  $\beta$  is an orthonormal basis consisting of eigenvectors of  $L$ . Then the matrix  $A = [L]_{\beta}$  is diagonal,  $A = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$ . Since  $\beta$  is orthonormal,  $[L^*]_{\beta} = A^* = \text{diag}(\overline{\lambda_1}, \overline{\lambda_2}, \dots, \overline{\lambda_n})$ . Clearly,  $AA^* = A^*A$ . Hence  $L \circ L^* = L^* \circ L$ .

*Idea of the proof ("only if"):* The statement is derived from the following two lemmas.

**Lemma 1 (Schur's Theorem)** There exists an orthonormal basis  $\beta$  for  $V$  such that the matrix  $[L]_{\beta}$  is upper triangular.

**Lemma 2** If a normal matrix is upper triangular, then it is actually diagonal.



## Diagonalization of normal operators

**Theorem** A linear operator  $L$  on a finite-dimensional inner product space  $V$  is normal if and only if there exists an orthonormal basis for  $V$  consisting of eigenvectors of  $L$ .

**Corollary 1** Suppose  $L$  is a normal operator. Then

(i)  $L$  is self-adjoint if and only if all eigenvalues of  $L$  are real ( $\bar{\lambda} = \lambda$ );

(ii)  $L$  is anti-selfadjoint if and only if all eigenvalues of  $L$  are purely imaginary ( $\bar{\lambda} = -\lambda$ );

(iii)  $L$  is unitary if and only if all eigenvalues of  $L$  are of absolute value 1 ( $\bar{\lambda} = \lambda^{-1}$ ).

**Corollary 2** A linear operator  $L$  on a finite-dimensional, real inner product space  $V$  is self-adjoint if and only if there exists an orthonormal basis for  $V$  consisting of eigenvectors of  $L$ .

## Diagonalization of normal matrices

**Theorem (a)**  $A \in \mathcal{M}_{n,n}(\mathbb{C})$  is normal  $\iff$  there exists an orthonormal basis for  $\mathbb{C}^n$  consisting of eigenvectors of  $A$ ;

**(b)**  $A \in \mathcal{M}_{n,n}(\mathbb{R})$  is symmetric  $\iff$  there exists an orthonormal basis for  $\mathbb{R}^n$  consisting of eigenvectors of  $A$ .

*Example.*  $A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 3 & 0 \\ 1 & 0 & 1 \end{pmatrix}.$

- $A$  is symmetric.
- $A$  has three eigenvalues: 0, 2, and 3.
- Associated eigenvectors are  $\mathbf{v}_1 = (-1, 0, 1)$ ,  $\mathbf{v}_2 = (1, 0, 1)$ , and  $\mathbf{v}_3 = (0, 1, 0)$ , respectively.
- Vectors  $\frac{1}{\sqrt{2}}\mathbf{v}_1$ ,  $\frac{1}{\sqrt{2}}\mathbf{v}_2$ ,  $\mathbf{v}_3$  form an orthonormal basis for  $\mathbb{R}^3$ .

*Example.*  $A_\phi = \begin{pmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{pmatrix}.$

- $A_\phi A_\psi = A_{\phi+\psi}$
- $A_\phi^{-1} = A_{-\phi} = A_\phi^t$
- $A_\phi$  is orthogonal
- $\det(A_\phi - \lambda I) = (\cos \phi - \lambda)^2 + \sin^2 \phi.$
- Eigenvalues:  $\lambda_1 = \cos \phi + i \sin \phi = e^{i\phi},$   
 $\lambda_2 = \cos \phi - i \sin \phi = e^{-i\phi}.$
- Associated eigenvectors:  $\mathbf{v}_1 = (1, -i),$   
 $\mathbf{v}_2 = (1, i).$
- Vectors  $\frac{1}{\sqrt{2}}\mathbf{v}_1$  and  $\frac{1}{\sqrt{2}}\mathbf{v}_2$  form an orthonormal basis for  $\mathbb{C}^2.$