

# Some Properties

Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$

$$\mathbf{A}\mathbf{v} = \lambda\mathbf{v}$$

$\Updownarrow$  Hermitian transpose

$$\mathbf{v}^H \mathbf{A}^H = \lambda^* \mathbf{v}^H \quad * \text{ conjugate}$$

$\Updownarrow$  transpose,  $\mathbf{A}^H = \mathbf{A}^T$

$$\mathbf{A}\mathbf{v}^* = \lambda^* \mathbf{v}^*$$

$$\mathbf{w}\mathbf{A} = \lambda\mathbf{w}$$

$\Updownarrow$

$$\mathbf{A}^H \mathbf{w}^H = \lambda^* \mathbf{w}^H$$

$\Updownarrow$

$$\mathbf{A}^T \mathbf{w}^H = \lambda^* \mathbf{w}^H$$

- $\mathbf{v}^*$  is an eigenvector associated with eigenvalue  $\lambda^*$
- Complex eigenvalues appear in conjugate pairs

- $\mathbf{w}^H$  is an eigenvector associated with eigenvalue  $\lambda^*$  of  $\mathbf{A}^T$
- $\mathbf{w}^T$  is an eigenvector associated with eigenvalue  $\lambda$  of  $\mathbf{A}^T$

- $\mathbf{A}$  and  $\mathbf{A}^T$  have the same set of eigenvalues because  $\det(\lambda\mathbf{I} - \mathbf{A}) = \det(\lambda\mathbf{I} - \mathbf{A})^T = \det(\lambda\mathbf{I} - \mathbf{A}^T)$
- The set of eigenvalues corresponding to (right) eigenvectors is the set of eigenvalues corresponding to left eigenvectors

## Some Properties (cont'd)

**Fact:** The eigenvalues of any triangular matrix are its diagonal entries

$$A = \begin{bmatrix} a_{11} & & * \\ & \ddots & \\ 0 & & a_{nn} \end{bmatrix}, \quad \lambda I - A = \begin{bmatrix} \lambda - a_{11} & & * \\ & \ddots & \\ 0 & & \lambda - a_{nn} \end{bmatrix} \quad \det(\lambda I - A) = \prod_{i=1}^n (\lambda - a_{ii})$$

**Fact:**  $A \in \mathbb{R}^{n \times n}$  is nonsingular if and only if all its eigenvalues are nonzero

$$\begin{aligned} A \text{ has an eigenvalue at } 0 &\Leftrightarrow 0 = \det(0 \cdot I - A) = (-1)^n \det(A) \\ &\Leftrightarrow \det(A) = 0 \Leftrightarrow A \text{ singular} \end{aligned}$$

**Fact:** Suppose  $(\mathbf{v}, \lambda)$  is an eigen-pair of  $\mathbf{A}$ , then  $(\mathbf{v}, \lambda^k)$  is an eigen-pair of  $\mathbf{A}^k$  for any positive integer  $k$

$$\begin{aligned} A^k \mathbf{v} &= A^{k-1} (A \mathbf{v}) = \lambda A^{k-1} \mathbf{v} = \lambda A^{k-2} (A \mathbf{v}) \\ &\quad \lambda \mathbf{v} = \dots = \lambda^{k-1} (A \mathbf{v}) = \lambda^k \mathbf{v} \end{aligned}$$

## Repeated Eigenvalues

- Let  $\lambda_1, \dots, \lambda_n$  be the  $n$  eigenvalues of  $\mathbf{A} \in \mathbb{R}^{n \times n}$
- WLOG, order  $\lambda_1, \dots, \lambda_n$  so that  $\{\lambda_1, \dots, \lambda_k\}$ ,  $k \leq n$  is the set of all **distinct** eigenvalues of  $\mathbf{A}$ :  $\lambda_i \neq \lambda_j$  for all  $i, j \in \{1, \dots, k\}$ ,  $i \neq j$  and  $\lambda_i \in \{\lambda_1, \dots, \lambda_k\}$  for all  $i \in \{1, \dots, n\}$
- Define the **algebraic multiplicity** of eigenvalue  $\lambda_i$  as the multiplicity of  $\lambda_i$  as root of  $p(\lambda)$ , denoted by  $\mu_i$
- Every  $\lambda_i$  may have multiple eigenvectors (scaling not counted)
- If  $\dim \mathcal{N}(\lambda_i \mathbf{I} - \mathbf{A}) = r$ , we can find  $r$  linearly independent  $\mathbf{v}_i$ 's
- Define the **geometric multiplicity** of eigenvalue  $\lambda_i$  as the maximum number of linearly independent eigenvectors associated with  $\lambda_i$ , denoted by  $\gamma_i$ 
  - $\gamma_i = \dim \mathcal{N}(\lambda_i \mathbf{I} - \mathbf{A}) = n - \text{rank}(\lambda_i \mathbf{I} - \mathbf{A})$

$\uparrow$   
 $\mathcal{N}(\cdot)$

$\dim \mathcal{N}(\cdot)$   
nullity

## Repeated Eigenvalues (cont'd)

**Fact:** For every eigenvalue  $\lambda_i$  of  $\mathbf{A}$ ,  $\mu_i \geq \gamma_i$  21

Proof: Let  $q_1, \dots, q_{r_i}$  be an orthonormal basis of  $\mathcal{N}(\lambda_i I - A)$  and let  $q_{r_i+1}, \dots, q_n$  be s.t.  $Q = \begin{bmatrix} q_1 & \dots & q_{r_i} & q_{r_i+1} & \dots & q_n \end{bmatrix}$  is unitary

$$Q^H(AQ) = \begin{bmatrix} Q_1^H \\ Q_2^H \end{bmatrix} \begin{bmatrix} AQ_1 & AQ_2 \end{bmatrix} = \begin{bmatrix} Q_1^H A Q_1 & Q_1^H A Q_2 \\ Q_2^H A Q_1 & Q_2^H A Q_2 \end{bmatrix}$$

For  $i=1, \dots, r_i$ ,  $Aq_i = \lambda_i q_i \Rightarrow AQ_1 = \lambda_i Q_1$  (\*)

Also,  $Q_1^H Q_1 = I$ ,  $Q_2^H Q_2 = I$

$$(*) = \begin{bmatrix} \lambda_i I_{r_i} & Q_1^H A Q_2 \\ 0 & Q_2^H A Q_2 \end{bmatrix}$$

$$\lambda I - Q^H A Q = \begin{bmatrix} (\lambda - \lambda_i) I_{r_i} & -Q_1^H A Q_2 \\ 0 & \lambda I - Q_2^H A Q_2 \end{bmatrix}$$

$$\det(\lambda I - A) = \det(Q^H(\lambda I - A)Q) = \det(\lambda I - Q^H A Q)$$

$$\stackrel{(*)}{=} \det((\lambda - \lambda_i) I_{r_i}) \det(\lambda I_{n-r_i} - Q_2^H A Q_2) = (\lambda - \lambda_i)^{r_i} \cdot \underbrace{\det(\dots)}_{\text{a polynomial of } \lambda}$$

$\Rightarrow \det(\lambda I - A)$  has at least  $r_i$  roots at  $\lambda_i$

## Repeated Eigenvalues (cont'd)

upper triangular

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

$$\det(\lambda I - A) = (\lambda - 3)(\lambda - 2)^2$$

$$\lambda_1 = 3 \quad \mu_1 = 1 \quad \delta_1 = 1$$

$$\lambda_2 = 2 \quad \mu_2 = 2 \quad \delta_2 = ?$$

$$Av = 2v \Leftrightarrow \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} = \begin{bmatrix} 2v_1 \\ 2v_2 \\ 2v_3 \end{bmatrix} \Leftrightarrow \begin{cases} 3v_1 = 2v_1 \\ 2v_2 + v_3 = 2v_2 \\ 2v_3 = 2v_3 \end{cases}$$

$$\Leftrightarrow v = \begin{bmatrix} 0 \\ * \\ 0 \end{bmatrix}$$

eigenspace  $\text{span} \left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\} \Rightarrow \delta_2 = 1$

or

$$2I - A = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{rank } 2 \Rightarrow \delta_2 = 3 - 2 = 1$$

## Repeated Eigenvalues (cont'd)

Example:  $A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$

$$\lambda_1 = 3$$

$$n_1 = 1$$

$$d_1 = 1$$

$$\lambda_2 = 2$$

$$n_2 = 2$$

$$d_2 = ?$$

$$Av = 2v \Leftrightarrow \begin{cases} 3v_1 = 2v_1 \\ 2v_2 = 2v_2 \\ 2v_3 = 2v_3 \end{cases} \Leftrightarrow v = \begin{bmatrix} 0 \\ * \\ * \end{bmatrix}$$

eigenspace  $\text{Span} \left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\} \Rightarrow d_2 = 2$

or

$$2I - A = \begin{bmatrix} -1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ rank } 1 \Rightarrow d_2 = 3 - 1 = 2$$

# Similarity Transformation

- Let  $\mathbf{Q} = [\mathbf{q}_1 \cdots \mathbf{q}_n] \in \mathbb{R}^{n \times n}$  (or  $\mathbb{C}^{n \times n}$ ) be a nonsingular matrix
  - The columns of  $\mathbf{Q}$  form a basis of  $\mathbb{R}^n$  (or  $\mathbb{C}^n$ )
- Let  $\mathbf{A} \in \mathbb{R}^{n \times n}$  (or  $\mathbb{C}^{n \times n}$ ). We call  $\tilde{\mathbf{A}} = \mathbf{Q}^{-1}\mathbf{A}\mathbf{Q}$  a **similarity transformation**
- $\mathbf{A}$  and  $\tilde{\mathbf{A}}$  are said to be **similar**
- Similar matrices represent the same linear map under two (possibly) different bases, with  $\mathbf{Q}$  being the change of basis matrix
- **Interpretation:** Consider a linear system  $\mathbf{A}\mathbf{x} = \mathbf{y}$  and let  $\mathbf{x} = \mathbf{Q}\tilde{\mathbf{x}}$ ,  $\mathbf{y} = \mathbf{Q}\tilde{\mathbf{y}}$

$$\mathbf{A}\mathbf{x} = \mathbf{y} \Leftrightarrow \mathbf{A}\mathbf{Q}\tilde{\mathbf{x}} = \mathbf{Q}\tilde{\mathbf{y}} \Leftrightarrow \mathbf{Q}^{-1}\mathbf{A}\mathbf{Q}\tilde{\mathbf{x}} = \tilde{\mathbf{y}} \Leftrightarrow \tilde{\mathbf{A}}\tilde{\mathbf{x}} = \tilde{\mathbf{y}}$$

$\mathbf{x} = \tilde{x}_1 \mathbf{q}_1 + \cdots + \tilde{x}_n \mathbf{q}_n$   
 $= \tilde{x}_1 \mathbf{e}_1 + \cdots + \tilde{x}_n \mathbf{e}_n$

## Similarity Transformation (cont'd)

- Every **square** matrix is similar to itself  $I^{-1} A I = A$
- If **A** is similar to **B** and **B** is similar to **C**, then **A** is similar to **C**

$\exists$  nonsingular  $Q, P$  s.t.  $B = Q^{-1} A Q, C = P^{-1} B P$

$$C = P^{-1} (Q^{-1} A Q) P = (QP)^{-1} A (QP), \quad QP \text{ nonsingular}$$

- If  $A, B$  are invertible and similar, then  $A^{-1}$  and  $B^{-1}$  are also similar

$$B = Q^{-1} A Q \Leftrightarrow B^{-1} = (Q^{-1} A Q)^{-1} = Q^{-1} A^{-1} Q$$

- Similar matrices have the same characteristic polynomial, determinant, rank, nullity, trace, eigenvalues, algebraic multiplicity, geometric multiplicity, etc.

$$\tilde{A} = Q^{-1} A Q$$

$$\begin{aligned} \det(\lambda I - \tilde{A}) &= \det(\lambda I - Q^{-1} A Q) = \det(Q^{-1} (\lambda I - A) Q) \\ &= \underbrace{\det(Q^{-1}) \det(\lambda I - A) \det(Q)}_{= \det(\lambda I - A)} \end{aligned}$$



# Eigendecomposition

A matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$  (or  $\mathbb{C}^{n \times n}$ ) is said to be **diagonalizable**, or admit an **eigendecomposition**, if there exists a nonsingular  $\mathbf{V} \in \mathbb{C}^{n \times n}$  s.t.

$$\mathbf{A} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^{-1}$$

where  $\mathbf{\Lambda} = \text{Diag}(\lambda_1, \dots, \lambda_n)$ , or,  $\mathbf{A}$  is similar to a diagonal matrix

- In this definition, we didn't say that  $(\mathbf{v}_i, \lambda_i)$  is an eigen-pair of  $\mathbf{A}$ , but it indeed has to be

$$\begin{aligned}\mathbf{A} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^{-1} &\iff \mathbf{A}\mathbf{V} = \mathbf{V}\mathbf{\Lambda}, \mathbf{V} \text{ nonsingular} \\ &\iff \mathbf{A}\mathbf{v}_i = \lambda_i\mathbf{v}_i, \quad i = 1, \dots, n, \\ &\quad \mathbf{v}_1, \dots, \mathbf{v}_n \text{ linearly independent}\end{aligned}$$

- The key lies in finding  $n$  linearly independent eigenvectors to form  $\mathbf{V}$

## Eigendecomposition (cont'd)

$$A = V \Lambda V^{-1}$$

$A$  and  $\Lambda$  similar

**Facts:** Suppose  $\mathbf{A}$  admits an eigendecomposition

$$1. \det(\mathbf{A}) = \prod_{i=1}^n \lambda_i$$

including repeated eigenvalues

$$\det(A) = \det(\Lambda) = \prod_{i=1}^n \lambda_i$$

$$2. \operatorname{tr}(\mathbf{A}) = \sum_{i=1}^n a_{ii} = \sum_{i=1}^n \lambda_i$$

$$\operatorname{rank}(\mathbf{A}) = \operatorname{rank}(\Lambda)$$

3.  $\operatorname{rank}(\mathbf{A})$  = number of nonzero eigenvalues of  $\mathbf{A}$

4. Suppose  $\mathbf{A}$  is also nonsingular. Then,  $\mathbf{A}^{-1} = \mathbf{V} \Lambda^{-1} \mathbf{V}^{-1}$

$$\Lambda^{-1} = \begin{bmatrix} \frac{1}{\lambda_1} & & \\ & \ddots & \\ & & \frac{1}{\lambda_n} \end{bmatrix}$$

**Note:** Facts 1–2 are indeed true for any  $\mathbf{A}$ ; Facts 3–4 may not hold when  $\mathbf{A}$  does not admit an eigendecomposition