

## EIGENVALUE PROBLEMS

- Background and review on eigenvalue problems
- Diagonalizable matrices
- The Schur form
- Localization of eigenvalues - Gerschgorin's theorem
- Perturbation analysis, condition numbers..

### *Eigenvalue Problems. Introduction*

Let  $A$  an  $n \times n$  real nonsymmetric matrix. The eigenvalue problem:

$$Ax = \lambda x$$

$\lambda \in \mathbb{C}$  : eigenvalue

$x \in \mathbb{C}^n$  : eigenvector

#### *Types of Problems:*

- Compute a few  $\lambda_i$  's with smallest or largest real parts;
- Compute all  $\lambda_i$ 's in a certain region of  $\mathbb{C}$ ;
- Compute a few of the dominant eigenvalues;
- Compute all  $\lambda_i$ 's.

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GvL 7.1-7.4,7.5.2 – EigenPart1

### *Eigenvalue Problems. Their origins*

- Structural Engineering [ $Ku = \lambda Mu$ ]
- Stability analysis [e.g., electrical networks, mechanical system,..]
- Bifurcation analysis [e.g., in fluid flow]
- Electronic structure calculations [Schrödinger equation..]
- Applications of new era: page rank (of the world-wide web) and many types of dimension reduction (SVD instead of eigenvalues)

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### *Basic definitions and properties*

A complex scalar  $\lambda$  is called an **eigenvalue** of a square matrix  $A$  if there exists a nonzero vector  $u$  in  $\mathbb{C}^n$  such that  $Au = \lambda u$ . The vector  $u$  is called an **eigenvector** of  $A$  associated with  $\lambda$ . The set of all eigenvalues of  $A$  is the '**spectrum**' of  $A$ . Notation:  $\Lambda(A)$ .

- $\lambda$  is an eigenvalue iff the columns of  $A - \lambda I$  are linearly dependent.
- ... equivalent to saying that its rows are linearly dependent. So: there is a nonzero vector  $w$  such that

$$w^H(A - \lambda I) = 0$$

- $w$  is a **left** eigenvector of  $A$  ( $u$ = **right** eigenvector)
- $\lambda$  is an eigenvalue iff  $\det(A - \lambda I) = 0$

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### Basic definitions and properties (cont.)

- An eigenvalue is a root of the **Characteristic polynomial**:

$$p_A(\lambda) = \det(A - \lambda I)$$

- So there are  $n$  eigenvalues (counted with their multiplicities).
- The multiplicity of these eigenvalues as roots of  $p_A$  are called **algebraic multiplicities**.
- The **geometric multiplicity** of an eigenvalue  $\lambda_i$  is the number of linearly independent eigenvectors associated with  $\lambda_i$ .

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- Two matrices  $A$  and  $B$  are **similar** if there exists a nonsingular matrix  $X$  such that

$$A = XBX^{-1}$$

- $Av = \lambda v \iff B(X^{-1}v) = \lambda(X^{-1}v)$   
eigenvalues remain the same, eigenvectors transformed.
- Issue: find  $X$  so that  $B$  has a simple structure

**Definition:**  $A$  is **diagonalizable** if it is similar to a diagonal matrix

- **THEOREM:** A matrix is diagonalizable iff it has  $n$  linearly independent eigenvectors
- ... iff all its eigenvalues are semi-simple
- ... iff its eigenvectors form a basis of  $\mathbb{R}^n$

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- Geometric multiplicity is  $\leq$  algebraic multiplicity.
- An eigenvalue is **simple** if its (algebraic) multiplicity is one.
- It is **semi-simple** if its geometric and algebraic multiplicities are equal.

**Ex1** Consider

$$A = \begin{pmatrix} 1 & 2 & -4 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{pmatrix}$$

Eigenvalues of  $A$ ? their algebraic multiplicities? their geometric multiplicities? Is one a semi-simple eigenvalue?

**Ex2** Same questions if  $a_{33}$  is replaced by one.

**Ex3** Same questions if, in addition,  $a_{12}$  is replaced by zero.

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### Transformations that preserve eigenvectors

**Shift**  $B = A - \sigma I$ :  $Av = \lambda v \iff Bv = (\lambda - \sigma)v$   
eigenvalues move, eigenvectors remain the same.

**Polynomial**  $B = p(A) = \alpha_0 I + \dots + \alpha_n A^n$ :  $Av = \lambda v \iff Bv = p(\lambda)v$   
eigenvalues transformed, eigenvectors remain the same.

**Invert**  $B = A^{-1}$ :  $Av = \lambda v \iff Bv = \lambda^{-1}v$   
eigenvalues inverted, eigenvectors remain the same.

**Shift & Invert**  $B = (A - \sigma I)^{-1}$ :  $Av = \lambda v \iff Bv = (\lambda - \sigma)^{-1}v$   
eigenvalues transformed, eigenvectors remain the same. spacing between eigenvalues can be radically changed.

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► THEOREM (Schur form): Any matrix is unitarily similar to a triangular matrix, i.e., for any  $A$  there exists a unitary matrix  $Q$  and an upper triangular matrix  $R$  such that

$$A = QRQ^H$$

► Any Hermitian matrix is unitarily similar to a **real diagonal** matrix, (i.e. its Schur form is real diagonal).

► It is easy to read off the eigenvalues (including all the multiplicities) from the triangular matrix  $R$

► Eigenvectors can be obtained by back-solving

### Localization theorems and perturbation analysis

► Localization: where are the eigenvalues located in  $\mathbb{C}$ ?

► Perturbation analysis: If  $A$  is perturbed how does an eigenvalue change? How about an eigenvector?

► Also: sensitivity of an eigenvalue to perturbations

► Next result is a “localization” theorem

► We have seen one such result before. Let  $\|\cdot\|$  be a matrix norm.

Then:

$$\forall \lambda \in \Lambda(A) : |\lambda| \leq \|A\|$$

► All eigenvalues are located in a disk of radius  $\|A\|$  centered at 0.

### Schur Form – Proof

◻4 Show that there is at least one eigenvalue and eigenvector of  $A$ :  $Ax = \lambda x$ , with  $\|x\|_2 = 1$

◻5 There is a unitary transformation  $P$  such that  $Px = e_1$ . How do you define  $P$ ?

◻6 Show that  $PAP^H = \begin{pmatrix} \lambda & ** \\ 0 & A_2 \end{pmatrix}$ .

◻7 Apply process recursively to  $A_2$ .

◻8 What happens if  $A$  is Hermitian?

◻9 Another proof altogether: use Jordan form of  $A$  and QR factorization

► More refined result: Gershgorin

THEOREM [Gershgorin]

$$\forall \lambda \in \Lambda(A), \quad \exists i \text{ such that } |\lambda - a_{ii}| \leq \sum_{\substack{j=1 \\ j \neq i}}^n |a_{ij}|.$$

► In words: eigenvalue  $\lambda$  is located in one of the closed discs of the complex plane centered at  $a_{ii}$  and with radius  $\rho_i = \sum_{j \neq i} |a_{ij}|$ .

**Proof:** By contradiction. If contrary is true then there is one eigenvalue  $\lambda$  that does not belong to any of the disks, i.e., such that  $|\lambda - a_{ii}| > \rho_i$  for all  $i$ . Write matrix  $A - \lambda I$  as:

$$A - \lambda I = D - \lambda I - [D - A] \equiv (D - \lambda I) - F$$

where  $D$  is the diagonal of  $A$  and  $-F = -(D - A)$  is the matrix of off-diagonal entries. Now write

$$A - \lambda I = (D - \lambda I)(I - (D - \lambda I)^{-1}F).$$

From assumptions we have  $\|(D - \lambda I)^{-1}F\|_\infty < 1$ . (Show this). The Lemma in P. 5-3 of notes would then show that  $A - \lambda I$  is nonsingular – a contradiction  $\square$

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### Bauer-Fike theorem

**THEOREM [Bauer-Fike]** Let  $\tilde{\lambda}, \tilde{u}$  be an approximate eigenpair with  $\|\tilde{u}\|_2 = 1$ , and let  $r = A\tilde{u} - \tilde{\lambda}\tilde{u}$  ('residual vector'). Assume  $A$  is diagonalizable:  $A = XDX^{-1}$ , with  $D$  diagonal. Then

$$\exists \lambda \in \Lambda(A) \text{ such that } |\lambda - \tilde{\lambda}| \leq \text{cond}_2(X)\|r\|_2.$$

➤ Very restrictive result - also not too sharp in general.

➤ Alternative formulation. If  $E$  is a perturbation to  $A$  then for any eigenvalue  $\tilde{\lambda}$  of  $A + E$  there is an eigenvalue  $\lambda$  of  $A$  such that:

$$|\lambda - \tilde{\lambda}| \leq \text{cond}_2(X)\|E\|_2.$$

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### Gershgorin's theorem - example

 Find a region of the complex plane where the eigenvalues of the following matrix are located:

$$A = \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 2 & 0 & 1 \\ -1 & -2 & -3 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 & -4 \end{pmatrix}$$

- Refinement: if disks are all disjoint then each of them contains one eigenvalue
- Refinement: can combine row and column version of the theorem (column version: apply theorem to  $A^H$ ).

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### Conditioning of Eigenvalues

➤ Assume that  $\lambda$  is a simple eigenvalue with right and left eigenvectors  $u$  and  $w^H$  respectively. Consider the matrices:

$$A(t) = A + tE$$

Eigenvalue  $\lambda(t)$ ,  
Eigenvector  $u(t)$ .

➤ Conditioning of  $\lambda$  of  $A$  relative to  $E$  is  $\left| \frac{d\lambda(t)}{dt} \right|_{t=0}$ .

➤ Write  $A(t)u(t) = \lambda(t)u(t)$  Then multiply both sides to the left by  $w^H$ :

$$\begin{aligned} w^H(A + tE)u(t) &= \lambda(t)w^H u(t) \rightarrow \\ \lambda(t)w^H u(t) &= w^H A u(t) + t w^H E u(t) \\ &= \lambda w^H u(t) + t w^H E u(t). \end{aligned}$$

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$$\rightarrow \frac{\lambda(t) - \lambda}{t} w^H u(t) = w^H E u(t)$$

► Take the limit at  $t = 0$ ,

$$\lambda'(0) = \frac{w^H E u}{w^H u}$$

► Note: the left and right eigenvectors associated with a simple eigenvalue cannot be orthogonal to each other.

► Actual conditioning of an eigenvalue, given a perturbation “in the direction of  $E$ ” is  $|\lambda'(0)|$ .

► In practice only estimate of  $\|E\|$  is available, so

$$|\lambda'(0)| \leq \frac{\|Eu\|_2 \|w\|_2}{|(u, w)|} \leq \|E\|_2 \frac{\|u\|_2 \|w\|_2}{|(u, w)|}$$

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►  $\Lambda(A) = \{1, 2, 3\}$ . Right and left eigenvectors associated with  $\lambda_1 = 1$ :

$$u = \begin{pmatrix} 0.3162 \\ -0.9487 \\ 0.0 \end{pmatrix} \quad \text{and} \quad w = \begin{pmatrix} 0.6810 \\ 0.2253 \\ 0.6967 \end{pmatrix}$$

So:  $\text{cond}(\lambda_1) \approx 603.64$

► Perturbing  $a_{11}$  to  $-149.01$  yields the spectrum:

$$\{0.2287, 3.2878, 2.4735\}.$$

► as expected..

► For Hermitian (also normal matrices) every simple eigenvalue is well-conditioned, since  $\text{cond}(\lambda) = 1$ .

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**Definition.** The condition number of a simple eigenvalue  $\lambda$  of an arbitrary matrix  $A$  is defined by

$$\text{cond}(\lambda) = \frac{1}{\cos \theta(u, w)}$$

in which  $u$  and  $w^H$  are the right and left eigenvectors, respectively, associated with  $\lambda$ .

**Example:** Consider the matrix

$$A = \begin{pmatrix} -149 & -50 & -154 \\ 537 & 180 & 546 \\ -27 & -9 & -25 \end{pmatrix}$$

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### Perturbations with Multiple Eigenvalues - Example

► Consider  $A = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix}$

► Worst case perturbation is in 3,1 position: set  $A_{31} = \epsilon$ .

► Eigenvalues of perturbed  $A$  are the roots of

$$p(\mu) = (\mu - 1)^3 - 4 \cdot \epsilon.$$

► Roots:  $\mu_k = 1 + (4\epsilon)^{1/3} e^{\frac{2ki\pi}{3}}, \quad k = 1, 2, 3$

► Hence eigenvalues of perturbed  $A$  are  $1 + O(\sqrt[3]{\epsilon})$ .

► If index of eigenvalue (dimension of largest Jordan block) is  $k$ , then an  $O(\epsilon)$  perturbation to  $A$  leads to  $O(\sqrt[k]{\epsilon})$  change in eigenvalue. Simple eigenvalue case corresponds to  $k = 1$ .

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