

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

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)

Basic definitions and properties

A complex scalar λ 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 λ . The set of all eigenvalues of A is the '**spectrum**' of A . Notation: $\Lambda(A)$.

- λ 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)
- λ is an eigenvalue iff $\boxed{\det(A - \lambda I) = 0}$

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 λ_i is the number of linearly independent eigenvectors associated with λ_i .

- 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.

 1 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?

 2 Same questions if a_{33} is replaced by one.

 3 Same questions if, in addition, a_{12} is replaced by zero.

- 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

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.

➤ 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

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 $PA P^H = \left(\begin{array}{c|c} \lambda & ** \\ \hline 0 & A_2 \end{array} \right)$.
-  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

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.

➤ More refined result: Gershgorin

THEOREM [Gershgorin]

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

➤ In words: eigenvalue λ 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 λ 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

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).

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

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

- 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 λ of A such that:

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

Conditioning of Eigenvalues

- Assume that λ 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 λ 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 :

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

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

➤ 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

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

Definition. The condition number of a simple eigenvalue λ 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 λ .

Example: Consider the matrix

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

- $\Lambda(A) = \{1, 2, 3\}$. 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 $\text{cond}(\lambda) = 1$.

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 $k = 1$.