

Matrices, Norms, Projections, Gershgorin Circles

This lecture covers

- Interval matrices & eigensystems
- Matrix norms
- Gershgorin circle theorems
- Projections from function spaces to finite-dimensional spaces

Interval matrices

$$A = (a_{ij})_{\substack{1 \leq i \leq n \\ 1 \leq j \leq m}} \quad a_{ij} \in \mathbb{R} \quad A \in \mathbb{R}^{m \times n}$$

Interval matrix arithmetic

$A+B$ has entries $(a_{ij}+b_{ij})$

If $A \in \mathbb{R}^{m \times p}$ $B \in \mathbb{R}^{p \times n}$

$$AB = \left\{ c_{ij} = \sum_{k=1}^p \alpha_{ik} \beta_{kj} : \alpha_{ik} \in a_{ik}, \beta_{kj} \in b_{kj} \right\}$$

- This definition results in overestimation & non-associativity!

Example Overestimation (of AB)

$$A = ([1, 2], [3, 4])$$

$$[\inf_{\text{sup}}(1, 2), \inf_{\text{sup}}(3, 4)]$$

$$B = \begin{pmatrix} [5, 6] & [7, 8] \\ [9, 10] & [11, 12] \end{pmatrix}$$

$$AB = ([31, 52], [39, 64]) \quad \text{according to INTLAB}$$

However, there is no $\alpha \in A, \beta \in B$ such that $\alpha\beta = (32, 64)$

$$\text{mid}(A) * \text{mid}(B) = \underline{A} \underline{B} = (1, 3) \begin{pmatrix} 5 & 7 \\ 9 & 11 \end{pmatrix} = (32, 40)$$

$$\text{mag}(A) * \text{mag}(B) = \bar{A} \bar{B} = (2, 4) \begin{pmatrix} 6 & 8 \\ 10 & 12 \end{pmatrix} = (52, 64)$$

How to get 31 & 39? Further overestimation

because of the INTLAB method (chosen for efficiency)

midpoint-radius expansion

Calculate $\text{mid}(A), \text{mid}(B)$ and add on radii.

Example Non-associativity (of AB)

$$A = \begin{bmatrix} 1 & [0, 1] \\ 1 & -1 \end{bmatrix} \Rightarrow A \times A = \begin{bmatrix} [1, 2] & [-1, 1] \\ 0 & [1, 2] \end{bmatrix}$$

$$(A \times A) \times A = \begin{bmatrix} [0, 3] & [-1, 3] \\ [1, 2] & [-2, -1] \end{bmatrix} \neq A \times (A \times A) = \begin{bmatrix} [1, 2] & [-1, 3] \\ [1, 2] & [-3, 0] \end{bmatrix}$$

Key fact

Theorem $+, -, \times$ are inclusion monotonic for interval matrices. Thus for $\alpha \in A, \beta \in B$,
 $\alpha \circ \beta \subseteq A \circ B$

- Can use $\max, \min, \text{abs}, \text{rad}, \text{diam}$

Solving linear systems

Solve $Ax = b$ ie $\{x : \alpha \in A \nexists \beta \in B, \alpha x = \beta \Rightarrow x \in x\}$

$A \setminus b$

verifies (A, b)

Krawczyk & Hansen methods

- Cannot work if $0 \in \det A$.
- May not work in other cases due to overestimation issues.

Operator norms

p-norm Let $1 \leq p \leq \infty$, $A \in \mathbb{R}^{m \times n}$ $A = (a_{ij})_{\substack{1 \leq i \leq m \\ 1 \leq j \leq n}}$

$$\|A\|_p = \sup_{\|x\|_p = 1} \|Ax\|_p \quad \leftarrow \text{sup} = \text{max in } \mathbb{R}^D$$

p=1 $\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}|$

max absolute column sum

p= ∞ $\|A\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$

max absolute row sum

p=2 $\|A\|_2 = \sqrt{\rho(A^T A)}$

spectral norm

Spectral radius

$$\rho(B) = \max \{ |\lambda| : \lambda \text{ eigenvalue of } B \}$$

Non-operator norm

$$\|A\|_F = \left(\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2 \right)^{\frac{1}{2}}$$

Frobenius norm

Theorem

$$\|A\|_2 \leq \|A\|_F \quad (\text{In fact more such statements hold})$$

Interval matrix norms $A \in \mathbb{R}^{m \times n}$

$$\|A\| = \text{an interval containing } \{\|x\| : x \in A\}$$

Eigensystems

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

$$Av = \lambda v$$

$$\Leftrightarrow \{\lambda, v\}$$

eigenvalue eigenvector pair

Theorem (Gershgorin)

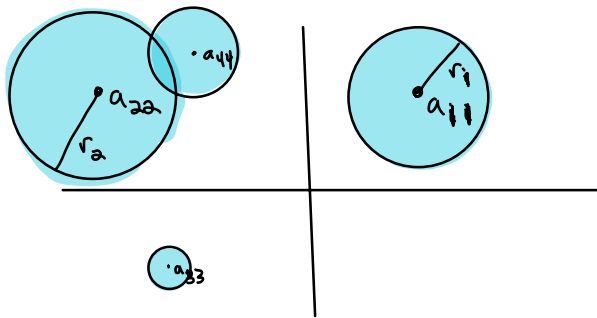
$$\text{Let } A \in \mathbb{C}^{n \times n}$$

Gershgorin disks

$$B_i(A) = B(a_{ii}, r_i)$$

$$B_G = \bigcup B_i$$

$$r_i = \sum_{j=1, j \neq i}^n |a_{ij}|$$



$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & a_{2n} \\ a_{31} & & a_{33} & \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & & & a_{nn} \end{pmatrix}$$

Then

- all eigenvalues lie in B_G .
- If A is real & symmetric, each B_i contains at least one eigenvalue.

Finding eigenpairs in INTLAB

Finding approximate eigenpairs

$$[V, D] = \text{eig}(m(A)) \quad A \in \mathbb{R}^{n \times n}$$

Verifying eigenpairs

$$\text{verifyeig}(A, D(j,j), V(:,j))$$

The method for `verifyeig` is iterative, more sophisticated than Gershgorin, which would yield overly conservative estimates.

Projections

Let $l_p = \{x = (x_k)_{k=1}^{\infty} : \sum_{k=1}^{\infty} |x_k|^p < \infty\}$

with norm $\|x\|_p = \left(\sum_{k=1}^{\infty} |x_k|^p\right)^{\frac{1}{p}}$
 l_2 is a Hilbert space.

Projections

Define projection $P_N : l_p \rightarrow l_p^N$

by $P_N(x) = (x_1, \dots, x_N)$

l_p^N is isometrically isomorphic to $\mathbb{R}^N$.

$Q_N = I - P_N$ is also a projection from l_p into an infinite-dimensional subspace of l_p , which we call l_p^{∞}

Lemma Given $x \in l_2$ and $N \in \mathbb{N}$,

there is a unique $x_N \in l_p^N$, $x_{\infty} \in l_p^{\infty}$ such that

$$x = x_N + x_{\infty}$$

Furthermore $x_N = P_N x$, $x_{\infty} = (I - P_N)x$.

Keeping in mind the above lemma,

$$\|x\|_2 \leq \|x_N\|_2 + \|x_\infty\|_2$$

Now consider a bounded linear operator

$$T: l_2 \rightarrow l_2$$

$$T \in \mathcal{L}(l_2, l_2) = \mathcal{L}(l_2)$$

$$y = Tx$$

For emphasis

$$Q_N = P_\infty$$

$$y = T(x_N + x_\infty)$$

$$= Tx_N + Tx_\infty$$

$$= P_N(Tx_N) + P_\infty(Tx_N) + P_N(Tx_\infty) + P_\infty(Tx_\infty)$$

$$= \underbrace{T_N x_N}_{\in \mathbb{R}^N} + \underbrace{T_\infty x_N}_{l_2^\infty} + \underbrace{T_N x_\infty}_{\in \mathbb{R}^N} + \underbrace{T_\infty x_\infty}_{l_2^\infty}$$

$$\|y\| \leq \underbrace{\|T_N x_N\|}_{\substack{N\text{-dim'l norm} \\ \text{computer}}} + \underbrace{\|T_\infty x_N\|}_{\substack{\infty\text{-dim'l} \\ \text{analysis}}} + \underbrace{\|T_N x_\infty\|}_{\substack{N\text{-dim'l norm} \\ \text{analysis}}} + \underbrace{\|T_\infty x_\infty\|}_{\substack{\infty\text{-dim'l} \\ \text{analysis}}}$$

Matrix form

$$T \sim A = \begin{pmatrix} \begin{matrix} a_{11} & a_{12} & \dots & a_{1N} \\ \vdots & \vdots & & \vdots \\ a_{N1} & a_{N2} & \dots & a_{NN} \end{matrix} & \begin{matrix} a_{N+1} & \dots \\ a_{NN+1} & \dots \\ \vdots & \vdots \end{matrix} \\ \begin{matrix} a_{N+1} & \dots & a_{N+N} \\ \vdots & & \vdots \end{matrix} & \begin{matrix} a_{N+1N+1} & \dots \\ \vdots & \vdots \end{matrix} \end{pmatrix}$$