

Spectral radius of a matrix

(1)

$$\rho(A) := \max_{1 \leq i \leq n} |\lambda_i|; \text{ where } \lambda_1, \dots, \lambda_n \text{ are evs of } A \in M_{n \times n}$$

* spectral radius is closely related to the norm of a matrix!

Th^m :- let $A \in M_{n \times n}$;

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

$$(ii) \rho(A) \leq \|A\| \text{ where } \|\cdot\| \text{ is an induced matrix norm.}$$

eg. $A = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ -1 & 1 & 2 \end{pmatrix}$. Find $\|A\|_2$.

$$B = A^T A = \begin{pmatrix} 1 & 1 & -1 \\ 1 & 2 & 1 \\ 0 & 2 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ -1 & 1 & 2 \end{pmatrix} = \begin{pmatrix} 3 & 2 & -1 \\ 2 & 6 & 4 \\ -1 & 4 & 5 \end{pmatrix}$$

Now $|B - \lambda I| = -\lambda^3 + 4\lambda^2 - 42\lambda = 0$

$\Rightarrow \lambda = 0; 7 \pm \sqrt{7}$

$$\|A\|_2 = \sqrt{\rho(A^T A)} = \sqrt{7 + \sqrt{7}} = 3.106$$

A simple matrix decomposition

(2)

Consider any matrix, say

$$A = \begin{pmatrix} 2 & -1 & 5 \\ 0 & 1 & -2 \\ 1 & 5 & -1 \end{pmatrix}$$

$$= \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 5 & 0 \end{pmatrix} + \begin{pmatrix} 0 & -1 & 5 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{pmatrix}$$

Can always
do this
splitting!

$$= \underbrace{\hspace{10em}}_D + \underbrace{\hspace{10em}}_L + \underbrace{\hspace{10em}}_U$$

(strictly lower triangular) (strictly upper triangular)

$$= D - (-L) - (-U)$$

Iterative Schemes to solve systems of linear eqns.

Jacobi iterative method

the idea is to solve $A\vec{x} = \vec{b}$

by rewriting it in the form $\vec{x} = T\vec{x} + \vec{c}$

& then using an iterative scheme of the form $\vec{x}^{(k)} = T\vec{x}^{(k-1)} + \vec{c}$

where $k=1,2,3,\dots$

eg. Let us consider the system of linear eqns

$$\begin{aligned}
 E_1: & 10x_1 - x_2 + 2x_3 = 6 \\
 E_2: & -x_1 + 11x_2 - x_3 + 3x_4 = 25 \\
 E_3: & 2x_1 - x_2 + 10x_3 - x_4 = -11 \\
 E_4: & 3x_2 - x_3 + 8x_4 = 15
 \end{aligned}$$

which has a unique soln $\vec{x} = \begin{pmatrix} 1 \\ 2 \\ -1 \\ 1 \end{pmatrix}$

Consider this in the form $A\vec{x} = \vec{b}$ where $A = \begin{pmatrix} 10 & -1 & 2 & 0 \\ -1 & 11 & -1 & 3 \\ 2 & -1 & 10 & -1 \\ 0 & 3 & -1 & 8 \end{pmatrix}$ and $\vec{b} = \begin{pmatrix} 6 \\ 25 \\ -11 \\ 15 \end{pmatrix}$

Let us re-write $A\vec{x} = b$ as $\vec{x} = T\vec{x} + \vec{c}$ thereby (4)

$$\begin{aligned} x_1 &= \frac{1}{10}x_2 - \frac{1}{5}x_3 + \frac{3}{5} \\ x_2 &= \frac{1}{11}x_1 + \frac{1}{11}x_3 - \frac{3}{11}x_4 + \frac{25}{11} \\ x_3 &= -\frac{1}{5}x_1 + \frac{1}{10}x_2 + \frac{1}{10}x_4 - \frac{11}{10} \\ x_4 &= -\frac{3}{8}x_2 + \frac{1}{8}x_3 + \frac{15}{8} \end{aligned}$$

These unknowns were already computed in the eqns above.

$$\Rightarrow \vec{x} = \begin{pmatrix} 0 & 1/10 & -1/5 & 0 \\ 1/11 & 0 & 1/11 & -3/11 \\ 0 & 1/10 & 0 & 1/10 \\ 0 & -3/8 & 1/8 & 15/8 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} + \begin{pmatrix} 3/5 \\ 25/11 \\ -11/10 \\ 15/8 \end{pmatrix}$$

$$\vec{x} = T\vec{x} + \vec{c}$$

initial guess (say) $\vec{x}_{(0)} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$ then $\vec{x}_{(1)} = \begin{pmatrix} 0.6 \\ 2.2727 \\ -1.1000 \\ 1.8750 \end{pmatrix}$

Carry forth the computations iteratively

$$\vec{x}_{(10)} = \begin{pmatrix} 1.0001 \\ 1.9998 \\ -0.9998 \\ 0.9998 \end{pmatrix}$$

Whence $\frac{\|\vec{x}_{(10)} - \vec{x}_{(9)}\|}{\|\vec{x}_{(10)}\|} < 10^{-3}$

STOP!

Gauss iteration (in matrix form)

(6)

$$A \vec{x} = b$$

$$\Rightarrow (D+L+U) \vec{x} = b$$

$$\Rightarrow D \vec{x} = -(L+U) \vec{x} + \vec{b}$$

$$\Rightarrow \vec{x} = -D^{-1}(L+U) \vec{x} + \vec{b}$$

Hence the iteration

$$\vec{x}_{(k)} = -D^{-1}(L+U) \vec{x}_{(k-1)} + \vec{b}$$

in the form of iterates we have

$$x_{i(k)} = \frac{\sum_{j=1, j \neq i}^n (-a_{ij} x_{j(k-1)}) + b_i}{a_{ii}} ; i = 1, 2, \dots, n$$

$a_{ii} \neq 0 \Rightarrow D$ is invertible!

Gauss-Seidel iterative scheme

(7)

$$x_{i(k)} = \frac{-\sum_{j=1}^{i-1} a_{ij} x_{j(k)} - \sum_{j=i+1}^n a_{ij} x_{j(k-1)} + b_i}{a_{ii}} \quad ; \quad i = 1, 2, \dots, n$$

Why is this a good idea?

Compare w.r.t. the example at the beginning of this lecture!

in matrix form

$$A \vec{x} = \vec{b}$$

$$(D+L) \vec{x} = -U \vec{x} + \vec{b}$$

as iterates $(D+L) \vec{x}_{(k)} = -U \vec{x}_{(k-1)} + \vec{b}$

$$\text{or } \vec{x}_{(k)} = -(D+L)^{-1} U \vec{x}_{(k-1)} + (D+L)^{-1} \vec{b} ; k=1, 2, \dots$$

Gauss-Seidel iteration

*9. Next lecture, we will talk about-
Convergence of iterative schemes & also
SOR scheme!