

Spectral radius of a matrix (1)

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

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

Thm :- Let $A \in M_{n \times n}$;

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

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

e.g. $A = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix}$. Find $\|A\|_2$. Now $|8 - \lambda I| = -\lambda^3 + 14\lambda^2 - 42\lambda = 0$

$$B = A^T A = \begin{pmatrix} 1 & 1 & -1 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix} = \begin{pmatrix} 3 & 2 & -1 \\ 2 & 6 & 4 \\ -1 & 4 & 5 \end{pmatrix} \Rightarrow \lambda = 0; 7 \pm \sqrt{7}$$

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

A simple matrix decomposition

(2)

Consider any matrix, say

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

$$\begin{matrix} \nearrow & & \\ & \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} & + & \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 5 & 0 \end{pmatrix} & + & \begin{pmatrix} 0 & -1 & 5 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{pmatrix} \\ & \underbrace{\qquad\qquad\qquad}_{D} & + & \underbrace{\qquad\qquad\qquad}_{L} & + & \underbrace{\qquad\qquad\qquad}_{U} \\ & & & & & \\ & & & & & \\ & & & & & \text{(strictly lower triangular)} \\ & & & & & \text{(strictly upper triangular)} \end{matrix}$$

Can always do this splitting!

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

B)

Iterative Schemes to solve systems of linear eqns

Jacobi iterative method

Eg. Let us consider the system of linear eqns

$$E_1: 10x_1 - x_2 + 2x_3 = 6$$

$$E_2: -x_1 + 11x_2 - x_3 + 3x_4 = 25$$

$$E_3: 2x_1 - x_2 + 10x_3 - x_4 = -11$$

$$E_4: 3x_2 - x_3 + 8x_4 = 15$$

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

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

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

$$\vec{b} = \begin{pmatrix} 6 \\ 25 \\ -11 \\ 15 \end{pmatrix}$$

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

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

These unknowns were already computed in the eqns above.

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

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

(5)

Carry forth the computations iteratively

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

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

Stop!

Gauss iteration (in matrix form)

(6)

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

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

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

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

D^{-1} is missing

Hence the iteration

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

in the form of iterates we have

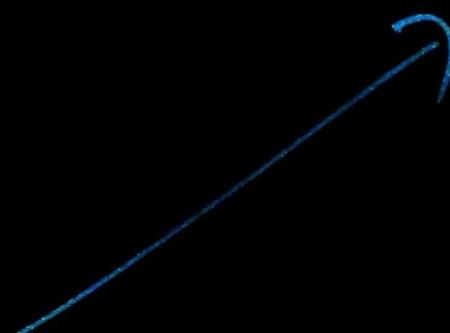
$$x_{i(k)} = \frac{\sum_{\substack{j=1 \\ j \neq i}}^n (-a_{ij} x_{j(k-1)}) + b_i}{a_{ii}} ; \quad i = 1, 2, \dots, n$$

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

Gauss - Seidel iterative Scheme

(7)

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



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

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

Gauss-Seidel iteration

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