

The Power method in  $x_p$  form.  
'different scaling'

$x_p$  notation: Given  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$

Define  $x_p$  as the 1-st component  
such that  $|x_p| = \|\vec{x}\|_\infty$ ,  $p$  is the  
index. e.g.

$$x = \begin{bmatrix} 2 \\ 1 \\ -3 \\ 3 \\ 2 \end{bmatrix},$$

$$x_p = -3, \quad p = 3$$

$$\|\vec{x}\|_\infty = 3$$

Power method I

Given  $x_0$ ,

for  $k=1$  until converges

$$y_{k+1} = A x_k \rightarrow \text{the same}$$

$$x_{k+1} = \frac{y_{k+1}}{\|y_{k+1}\|_p} \quad \|x_{k+1}\|_\infty = 1$$

$$M_{k+1} = (y_{k+1})_p$$

$$x_{k+1} = \frac{A^{k+1} x_0}{\alpha}$$

end

Q: How about  $\|\cdot\|_1$  norm?

Convergence speed, linear  $\left| \frac{\lambda_2}{\lambda_1} \right|$

If  $A=A^T$ , then the original quadratic  $\left( \frac{\lambda_2}{\lambda_1} \right)^2$ .

Power method is quadratically convergent, otherwise it is linearly convergent.

linear convergent

$$x_{k+1} = \alpha_k A^{k+1} x_0 \rightarrow 0$$

$$= \alpha_k \left( \beta_1 v_1 + \beta_2 v_2 \left( \frac{\lambda_2}{\lambda_1} \right)^{k+1} \right)$$

$$A v_i = \beta_i v_i, \quad i=1, 2, \dots, n. \quad + \dots + \beta_n v_n \left( \frac{\lambda_n}{\lambda_1} \right)^{k+1} \rightarrow 0$$

$$A, \lambda_1, \lambda_2, \dots, \lambda_n$$

$$A - \sigma I \quad \lambda_1 - \sigma, \lambda_2 - \sigma, \dots, \lambda_n - \sigma$$

$$\det(\lambda I - (A - \sigma I)) = 0$$

If we want to approximate  $\lambda_k$ , we can find a  $\sigma$  such that

$$|\sigma - \lambda_k| = \min_{1 \leq i \leq n} |\sigma - \lambda_i(A)|$$

$$\lambda_i(A) \quad 9 \quad 8 \quad 7 \quad 5 \quad 2 \quad 0 \quad -1$$

$$\sigma = 4.6 \quad 4.4 \quad 3.4 \quad 2.4 \quad 0.4 \quad 1.4$$

$$\sigma - \lambda_i(A) \quad \quad \quad \underline{\underline{0.4}}$$

We use the shifted power method

For  $k=1$  until converges

$$(A - \sigma I) y_{k+1} = x_k$$

$$x_{k+1} = y_{k+1} / \|y_{k+1}\|_2$$

$$\mu_{k+1} = x_{k+1}^T A x_{k+1}$$

$$x_k \rightarrow v_k$$

$$\mu_k \rightarrow \lambda_k$$

end

How do we roughly locate eigenvalues.

$$D = \begin{pmatrix} 5 & & \\ & -4 & 0 \\ & & \ddots \\ 0 & & & 1 \end{pmatrix} \quad \lambda_i(A) = a_{ii}.$$

$$\underline{D+E} = \begin{pmatrix} 5 & 0.01 & & \epsilon \\ -0.03 & -4 & & \\ & & \ddots & \\ & & & \end{pmatrix}$$

$$|\lambda_i(A) - \lambda_i(A+E)| \leq C \|E\|, \text{ perturbation of eigenvalues.}$$

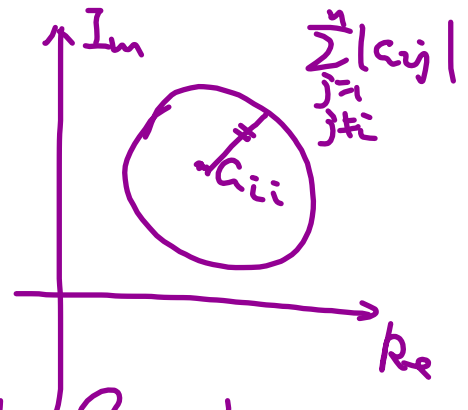
$$Ax = b$$

$$(A+E)\tilde{x} = b + \delta b$$

$$\frac{\|x - \tilde{x}\|}{\|\tilde{x}\|} \leq$$

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

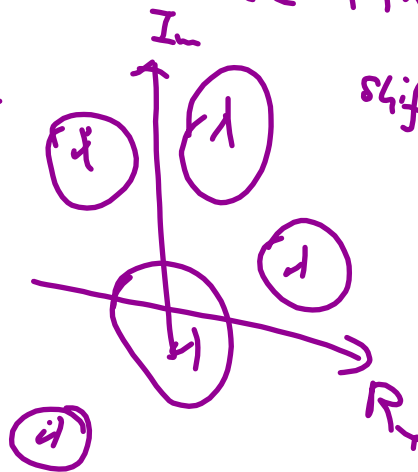
$$|\lambda - a_{ii}| \leq \sum_{\substack{j=1 \\ j \neq i}}^n |a_{ij}|$$



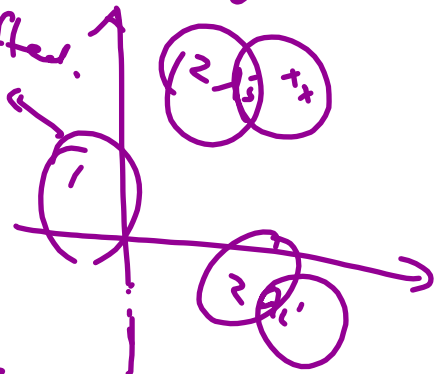
The set is called the  $i$ th Gershgorin Circle.

shifed  
for any

Ex



shifed.



Th: Gershgorin.

1. Any eigenvalue has to be in one of Gershgorin circle
2. The union of  $k$  Gershgorin circles, which do not intersect with other  $n-k$  circles, contains precisely  $k$  eigenvalues of  $A$ .



Proof:  $Ax^* = \lambda^* x^*$ ,  $x_p^* = 1$

$$(Ax^*)_p = \lambda^* x_p^*$$

$$\sum_{j=1}^n a_{ij} x_j^* = \lambda^* x_p^*$$

$$\left| (a_{pp} - \lambda^*) x_p^* \right| = \left| \sum_{\substack{j=1 \\ j \neq p}}^n a_{ij} x_j^* \right|$$

$$\left| a_{pp} - \lambda^* \right| \leq \sum_{\substack{j=1 \\ j \neq p}}^n |a_{ij}|$$

Wellposedness of algebraic

eigenvalue problems.

$$Ax = \lambda x$$

existence<sup>✓</sup>, uniqueness<sup>✓</sup>, sensitivity,  $x \neq 0$

$$\det(\lambda I - A) = 0, \quad \lambda^n \pm (\text{trace}) \lambda^{n-1} + \dots + (-1)^n \det(A) = 0$$

Solvability.  $\sum_{i=1}^n a_{ii}$

If  $\det(A) = 0$ ,  $\lambda = 0$  is an eigenvalue.  
 $A^{-1}$  does not exist.

$$(A + E)x = \lambda x$$

$$\|E\| \leq C \|A\| \epsilon$$

$Ax = \lambda x$ ,  $x$  is called the right eigenvector.

$$\|x\| = 1$$

$y^T A = \lambda y^T$ ,  $y$  is called the

$$\|y\| = 1$$

left eigenvector. If  $A = A^T$ , then

$$x = y.$$

$$|\lambda_i(A) - \lambda_i(A+E)| \leq \frac{C \|E\|}{y^T x}$$

$\left| \frac{1}{y^T x} \right|$  is called the condition number of the eigenvalue problem.

$A = A^T \in \mathbb{R}^{n \times n}$ , then  $|\lambda_i(A) - \lambda_i(A+E)| \leq C \|E\|$

Well-conditioned.

eigen-vectors.

$\lambda_i(A)$  are all real.

$$Q = [x_1, x_2, \dots, x_n]$$

$$Q^T Q = I$$

$$Q^T A Q = D = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{bmatrix} x_i^T x_j = \delta_i^j = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$$

$A = A^T$ , then the Power method is quadratically convergent.

$$\begin{aligned}
 M_k &= X_k^T A X_k = X_k^T \left( \beta_k \alpha_1 v_1 + \bar{\beta}_k \alpha_2 \left( \frac{\lambda_2}{\lambda_1} \right)^k + \dots \right) \\
 &= \left( \beta_k \alpha_1 v_1 + (\dots) \right)^T \left( \beta_k \alpha_1 \lambda_1 v_1 + \bar{\beta}_k \alpha_2 \left( \frac{\lambda_2}{\lambda_1} \right)^k v_2 + \dots \right) \\
 &= \lambda_1 + \bar{\beta}_k \left( \frac{\lambda_2}{\lambda_1} \right)^{2k} v_2 + (\dots) \rightarrow \lambda_1
 \end{aligned}$$

Ex. of Gershgorin Thm.

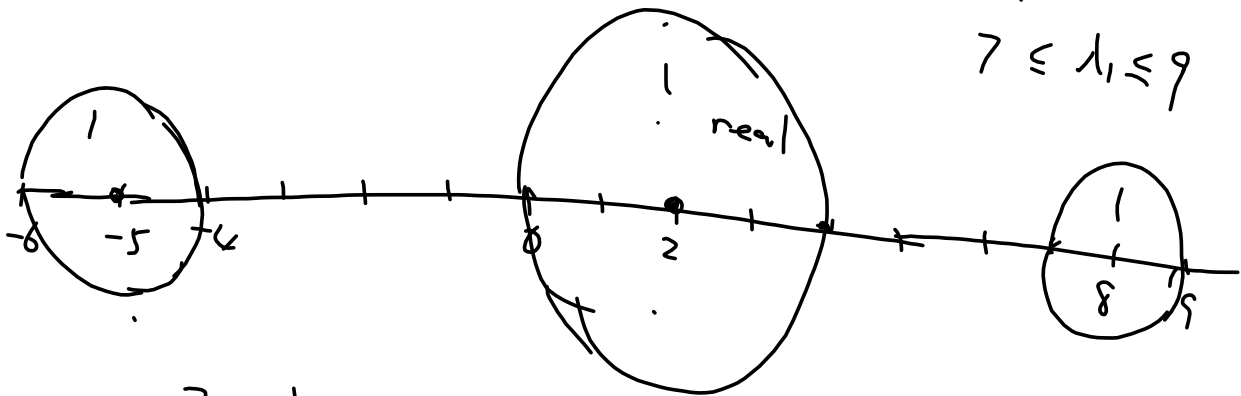
$$A = \begin{bmatrix} -5 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & 1 & 8 \end{bmatrix}$$

$$|z + 5| \leq 1 \quad R_1$$

$$R_2 \quad |z - 2| \leq 2$$

$$|z - 8| \leq 1,$$

$$7 \leq \lambda_1 \leq 9$$



$$7 \leq \lambda_1 \leq 9, \quad 0 \leq \lambda_3 \leq 4, \quad -6 \leq \lambda_2 \leq -4$$

$$(A - \sigma I) y_{k+1} = x_{k+1} \quad \sigma = -5$$

$$(A + 5I) y_{k+1} = x_{k+1} \quad x_{k+1} = y_{k+1} / \|y_{k+1}\|_2$$

$$= x_{k+1}, \quad \mu_{k+1} = x_{k+1}^T A x_{k+1}$$

$$x_{i-1} - 2x_i + x_{i+1} = f(x_i) = \cos \frac{k\pi}{n+1}, \quad k=1, 2, \dots$$

Find all eigenvalues,  $\rightarrow$  QR method  
for eigenvalue problems.

iterative  
method

$$A \rightarrow \dots D \quad A \text{ and } D$$

Should have  
same eigenvalues

$$S^{-1} A S = D \quad \det(S) \neq 0.$$

Similarity transformation, Not practical  
for general  $S$ , except for orthogonal

matrices,  $Q^T = Q^{-1}$

$$\text{Cond}_2(QA) = \text{Cond}_2(A)$$

$$\|QA\|_2 = \|A\|_2$$

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \dots & a_{nn} \end{bmatrix}$$

$$\|x\|_2 = \|y\|_2$$

$$QA = \begin{bmatrix} \bar{a}_{11} & & & \\ 0 & & & \\ \vdots & & & \\ 0 & & & \end{bmatrix}$$

$Q^T$   $\bar{a}_{ij}$

$$Q \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} |x| \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$x$   $y$

$$Q^T Q = I$$

$$\bar{x}_1 = \pm \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

Householder matrix in  $\mathbb{R}^n$

$$P = I - 2ww^T$$

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

$$\|w\|_2 = 1$$

$$\begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} \begin{pmatrix} w_1 & w_2 & \dots & w_n \end{pmatrix}$$

$$\begin{pmatrix} w_1^2 & w_1 w_2 & \dots & w_1 w_n \end{pmatrix}$$