

Power method for finding largest eigenvalue of $A \in \mathbb{R}^{n \times n}$

Start $x_0 \in \mathbb{R}^n$, Compute

$$x_{k+1} = Ax_k \quad k=0,1,2,\dots$$

Theorem: λ_1 is a simple eigenvalue of the symmetric matrix $A \in \mathbb{R}^{n \times n}$

$$|\lambda_1| > |\lambda_2| \geq |\lambda_3| \dots \geq |\lambda_n|$$

Let $x_0 \in \mathbb{R}^n$ be a vector that is not orthogonal to the eigenspace of λ_1 . Then the sequence $y_k := \frac{x_k}{\|x_k\|}$ with x_k defined in (*) converges towards a normalized eigenvector of A for λ_1 .

Proof: η_1, \dots, η_n orthonormal basis of \mathbb{R}^n of eigenvectors of A , i.e.

$A\eta_i = \lambda_i \eta_i \quad i=1, \dots, n$, Since η_1, \dots, η_n are a basis,

$$x_0 = \sum_{i=1}^n \alpha_i \eta_i, \quad \alpha_i \in \mathbb{R} \quad \text{and} \quad \alpha_1 = \eta_1^T x_0 \neq 0$$

$$x_k = Ax_{k-1} = A^k x_0 = \sum_{i=1}^n \alpha_i A^k \eta_i = \sum_{i=1}^n \alpha_i \lambda_i^k \eta_i$$

$$= \alpha_1 \lambda_1^k \left(\eta_1 + \sum_{i=2}^n \frac{\alpha_i}{\alpha_1} \left(\frac{\lambda_i}{\lambda_1} \right)^k \eta_i \right) = \alpha_1 \lambda_1^k z_k$$

$$\rightarrow \lim_{k \rightarrow \infty} z_k = \eta_1 \quad \text{since} \quad |\lambda_i| < |\lambda_1|$$

$$y_k = \frac{x_k}{\|x_k\|} = \frac{\alpha_1 \lambda_1^k z_k}{\|\alpha_1 \lambda_1^k z_k\|} = \frac{z_k}{\|z_k\|} \rightarrow \pm \eta_1$$

for $k \rightarrow \infty$. \square

Remarks: •) This method only allows to find the largest eigenvalue & eigenvector.

If I have the eigenvector η_1 , know that

$$A\eta_1 = \lambda_1 \eta_1$$

$$\Rightarrow \eta_1^T A \eta_1 = \lambda_1 \eta_1^T \eta_1$$

$$\Rightarrow \lambda_1 = \frac{\eta_1^T A \eta_1}{\eta_1^T \eta_1} \quad \text{Rayleigh coefficient}$$

•) The speed of convergence depends on the gap between $|\lambda_1|$ and $|\lambda_2|$. If that gap is small, e.g. $\frac{|\lambda_2|}{|\lambda_1|} = 0.9999 \rightarrow$ slow convergence

If gap is larger, e.g. $\frac{|\lambda_2|}{|\lambda_1|} = 0.7 \rightarrow$ fast convergence.

Q4, HW3: $A \in \mathbb{R}^{n \times n}$, λ is an eigenvalue of $A^T A$, $x \in \mathbb{R}^n$, $x \neq 0$
Cor. eigenvector

(a) Show that $\|Ax\|_2^2 = \lambda \|x\|_2^2$ and thus $\lambda \geq 0$.

$$\underbrace{x^T A^T A x}_{\|Ax\|_2^2} = \underbrace{x^T \lambda x}_{\lambda \|x\|_2^2}$$

(b) $\|\cdot\|$ norm in \mathbb{R}^n

$$\|A\|_2 \leq \|A^T A\|_2^{\frac{1}{2}}$$

$$A^T A x = \lambda x \rightarrow \|A^T A x\| = |\lambda| \|x\| = \lambda \|x\|$$

$\Rightarrow \|A^T A\| \|x\|$

$$\rightarrow \lambda \leq \|A^T A\|$$

$$\|A\|_2 = \max_{|\text{eigen}|} \sqrt{\lambda_i(A^T A)} \leq \|A^T A\|_2^{1/2}$$

(c)

$$\|A\|_2 \leq \|A^T A\|_1^{1/2} \leq \|A\|_1^{1/2} \|A^T\|_1^{1/2} \quad \kappa_2(A) \leq (\kappa_1(A) \kappa_\infty(A))^{1/2}$$

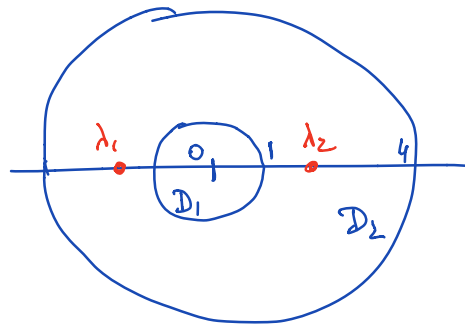
$$= \|A\|_1^{1/2} \|A\|_\infty^{1/2} / \|A\|_2 \|A^{-1}\|_2 \leq (\|A\|_1 \|A^{-1}\|_\infty \|A\|_\infty \|A^{-1}\|_1)^{1/2}$$

$$\|A^{-1}\|_2 \leq \|A^{-1}\|_1^{1/2} \|A^{-1}\|_\infty^{1/2}$$

Gerschgorin:

$$A = \begin{bmatrix} 0 & 1 \\ 4 & 0 \end{bmatrix}$$

eigenvalues are roots of
 $\lambda^2 - 4 \Rightarrow \lambda_{1,2} = \pm 2$



Can the number 6.125 be represented exactly in a computer?

$4 + 2 + \frac{1}{8}$ in binary: 110.001

$\frac{1}{10}$ cannot be represented: 0.0001001001...

Example when Newton does not converge quadratically
 for a C^2 function: At solution $\underline{f'(x^*) \neq 0}$

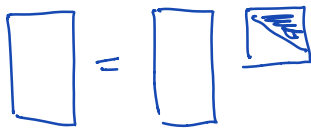
Example: Solve $f(x) = x^3 = 0$, $f'(x) = 3x^2$ at

Solution $f'(x^*) = f'(0) = 0$.

Assumption: $\left| \frac{f''(x)}{f'(y)} \right| \leq A \quad \forall x, y \text{ close to solution}$

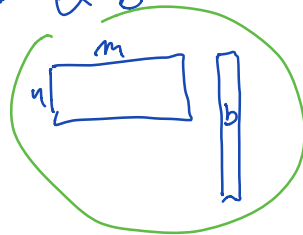
Least squares problem $Ax = b \quad A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m$

$A = \hat{Q} \hat{R}$



How many flops does it take to solve the least squares problem?

$\hat{R}x = \hat{Q}^T b$



m mult
 $m-1$ addition } $\times n$ -times

$n(2m-1)$

$n^2 + 2mn + \dots$