

LECTURE 9A

A with minimal polynomial $\mu(x)$ of degree ν
 ($\mu(A) = 0$, it divides the characteristic polynomial)
 Let $p_{\nu-1}(x)$ the interpolating polynomial of $f(x)$
 in the Hermite sense at the roots of $\mu(x)$.

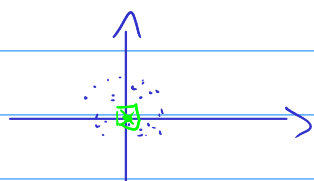
Then $f(A) = p_{\nu-1}(A)$.

It can be used as a definition.

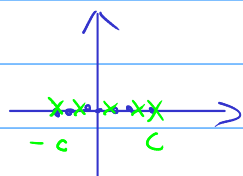
It can be unfeasible to compute the eigenvalues
 of A .

Truncated Taylor series is an interpolation of e^x
 at $m+1$ zeros.

$$A \rightarrow A - \mu I, \quad \mu = \frac{\text{trace}(A)}{N} \approx \text{average eigenvalue}$$



$L_{m,c}$ interpolation in $[-c, c]$ (Gershgorin, ...)



$$A \approx \mathcal{D}_{xx}$$

Krylov method $e^A v$ $A \in \mathbb{R}^{N \times N}$

$k_m(A, v)$ Krylov space $= \langle \{v, Av, A^2v, \dots, A^{m-1}v\} \rangle$

$$m \ll N$$

$$K_m(A, v) = \langle \{v_1, v_2, \dots, v_m\} \rangle$$

$$v_1 = \frac{v}{\|v\|_2}$$

$$V_m = [v_1, v_2, \dots, v_m] \in \mathbb{R}^{N \times m}$$

$$V_m^T V_m = I_m$$

$$V_m V_m^T = ?$$

Arnoldi factorization

$$AV_m = V_m H_m + h_{m+1, m} v_{m+1} e_m^T \quad H_m \in \mathbb{R}^{m \times m}$$

↖ Hessenberg

$$\begin{array}{c} N \\ \boxed{\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}} \\ N \end{array} = \begin{array}{c} \boxed{\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}} \\ m \end{array} = \begin{array}{c} \boxed{\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}} \\ m \end{array} \begin{array}{c} \boxed{\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}} \\ m \end{array} + \begin{array}{c} \boxed{\begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array}} \\ m \end{array}$$

$$A \text{ is sparse } (nnz(A) = \mathcal{O}(N))$$

$$= V_{m+1} \bar{H}_m$$

$$\bar{H}_m \in \mathbb{R}^{(m+1) \times m}$$

$$V_m^T A V_m = H_m$$

"A is similar to H_m "

$$\text{Cost: } \mathcal{O}(m^2) + \mathcal{O}(mN)$$

Av costs $\mathcal{O}(N)$
because A is sparse

↑
matrix-vector products

(Arnoldi factorization is used in linear system solution with iterative methods GMR, FOM, ...)

$$AV_m \approx V_m H_m$$

$$(\lambda I_N - A)V_m \approx V_m (\lambda I_m - H_m)$$

$$V_m (\lambda I_m - H_m)^{-1} e_1 \approx (\lambda I_N - A)^{-1} V_m e_1 = (\lambda I_N - A)^{-1} v_1$$

field of values

$$\lambda \in \mathcal{F}(A) = \{ x^* A x : x \in \mathbb{C}^N, \|x\|_2 = 1 \} \supset \mathcal{F}(H_m)$$

$$e^A v = \frac{1}{2\pi i} \int_{\Gamma} e^{\lambda} (\lambda I_N - A)^{-1} v d\lambda \quad \Gamma \text{ exterior to } \mathcal{F}(A)$$

Cauchy definition of $e^A v$ (contour integral formulas rational)

$$\approx \frac{1}{2\pi i} \int_{\Gamma} e^{\lambda} V_m (\lambda I_m - H_m)^{-1} e_1 \|v\| d\lambda$$

$$= \|v\| V_m e^{H_m} e_1$$

$$\boxed{e^A v \approx \|v\| V_m \left[e^{H_m} \right] e_1} \quad \text{Taylor approx.}$$

Taylor, Padé, diagonalization, ...

$h_{j+1,j} \neq 0 \Rightarrow$ the geometric multiplicity of each eigenvalue is 1.

minimal polynomial \equiv characteristic polynomial

$$e^{H_m} = p_{m-1}(H_m)$$

$p_{m-1}(x)$ interpolates e^x at the eigenvalues of H_m

$$\text{Lemma: } p_j(A) v_1 = V_m p_j(H_m) e_1 \quad j \leq m-1$$

BY INDUCTION

PROOF: $j=0$ $v_1 = V_m e_1$ ✓

Suppose true for $j \leq m-2$.

$$V_m V_m^T w = w \quad \text{for} \quad w \in K_m(A, v)$$

$$(w = \alpha_1 v_1 + \dots + \alpha_m v_m)$$

$$V_m^T w = \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_m \end{pmatrix} \quad V_m \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_m \end{pmatrix} = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_m v_m = w$$

$$\underbrace{A^{j+1} v_1}_{\in K_m} = V_m V_m^T A^{j+1} v_1 = V_m V_m^T A \underbrace{A^j v_1}_{\in K_m} =$$

$$\in K_m = \underbrace{V_m V_m^T A V_m V_m^T}_{H_m} A^j v_1 =$$

H_m

$$= V_m \underbrace{H_m V_m^T}_{H_m^j} A^j v_1 = V_m H_m H_m^j e_1$$

$$= V_m H_m^{j+1} e_1 = A^{j+1} v_1.$$

Theorem.

$$A \underbrace{e v}_{\approx \|v\|} V_m e^{H_m} e_1 = p_{m-1}(A) v$$

PROOF:

$$\|v\| V_m e^{H_m} e_1 = \|v\| V_m p_{m-1}(H_m) e_1 =$$

$$\|v\| p_{m-1}(A) v_1 = p_{m-1}(A) v \quad \square$$

Krylov with incomplete orthogonalization: $O(m) + O(mN)$

rational Krylov $K_m(A, v) = \text{span} \{ v, (A - \lambda_1 I)^{-1} v, (A - \lambda_1 I)^{-2} v, \dots \}$

If I have to solve linear systems to apply exp. int.
to solve $y'(t) = f(t, y(t))$
then I can use implicit methods to solve it.

$$(I - \tau A)^{-1} y_{n+1} = \dots$$