

$$f: U \subseteq \mathbb{C} \rightarrow \mathbb{C} \quad A = SJS$$

$$f(A) = S \begin{bmatrix} f(J_1), & \\ & f(J_2), \\ & & \ddots \\ & & & f(J_r) \end{bmatrix} S^{-1}$$

$$f(J_i) = f \left( \begin{bmatrix} \lambda_i & & \\ & \ddots & \\ & & \lambda_i \end{bmatrix} \right) = \begin{bmatrix} f(\lambda_i) & f'(\lambda_i) & \dots & \frac{f^{(n_i-1)}(\lambda_i)}{(n_i-1)!} \\ & f(\lambda_i) & & \\ & & \ddots & \\ & & & f(\lambda_i) \end{bmatrix} \quad (*)$$

$f$  defined in  $A$  if  $f$  regular enough to write (\*) for all blocks.

Properties of matrix functions

1) if  $A$  has eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$ , then

$f(A)$  has eigenvalues  $f(\lambda_1), f(\lambda_2), \dots, f(\lambda_n)$ .

Algebraic multiplicities are preserved

Geometric multiplicities may increase

$$f(S) = \begin{bmatrix} f(\lambda) & f(\lambda) & f''(\lambda) & \dots \\ & f(\lambda) & f'(\lambda) & \dots \\ & & f(\lambda) & \dots \\ & & & f'(\lambda) \\ & & & & f(\lambda) \end{bmatrix}$$

$$= \begin{bmatrix} 0 & f'(\lambda) & f''(\lambda) & \dots \\ & 0 & f'(\lambda) & \dots \\ & & 0 & \dots \\ & & & 0 \end{bmatrix}$$

or  $f(\lambda)$   
geometric multiplicity in  $f(S)$

$$\dim \ker f(S) - f(\lambda) I =$$

If  $f'(\lambda) \neq 0$ , then  $\dim \ker = 1$   
and  $m_g(f(\lambda), f(S)) = m_g(\lambda, S) = 1$

If  $f'(A) = 0$ , then  $\mathcal{M}_g(f(A), f(S)) > 1$

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Given  $f(x), g(x)$  functions,

$$h(x) := f(x) + g(x)$$

$$h(A) = f(A) + g(A)$$

$$P_h(A) = P_f(A) + P_g(A)$$

And the interpolating polynomial of  $h$  is the sum of the two interpolating polynomials of  $f$  &  $g$ .

Similarly, if  $h(x) := f(x)g(x) = g(x)f(x)$

$$h(A) = f(A)g(A) = g(A)f(A)$$

In particular, functions of the same matrix commute.

$$\exp(A) A^2 = A^2 \exp(A).$$

Similarly,  $h(x) = f(g(x))$ ,  $h(A) = f(g(A))$ .

Continuity: 1) If  $f_1, f_2, \dots \rightarrow f$ , is it true that  $f_n(A) \rightarrow f(A)$ ?

2) If  $A_n \rightarrow A$ ,  $f(A_n) \rightarrow f(A)$ ?

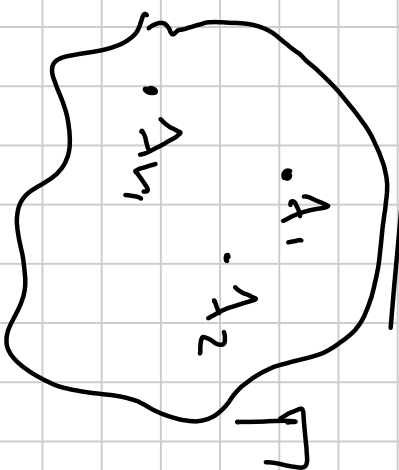
1): If  $f_n \rightarrow f$ ,  $f'_n \rightarrow f'$ ,  $f''_n \rightarrow f''$  with enough derivatives

for  $f(A)$  to be defined, how  $f_n(A) \rightarrow f(A)$

$$f_n(x) = \begin{bmatrix} f_n(x) & f'_n(x) & \dots & \frac{f_n^{(k-1)}(x)}{(k-1)!} \\ \vdots & \vdots & \vdots & \vdots \\ 0 & \vdots & \vdots & \vdots \end{bmatrix} \rightarrow \begin{bmatrix} f(x) & f'(x) & \dots & \frac{f^{(k-1)}(x)}{(k-1)!} \\ \vdots & \vdots & \vdots & \vdots \\ 0 & \vdots & \vdots & \vdots \end{bmatrix} = f(x)$$

## Cauchy integral formula

$$f(A) = \frac{1}{2\pi i} \int_{\Gamma} f(z) (zI - A)^{-1} dz$$



where  $\Gamma$  is a contour that encloses the eigenvalues of  $A$

Proof: Jordan Form:  $A = SJS^{-1}$

$$\frac{1}{2\pi i} \int_{\Gamma} f(z) (zI - A)^{-1} dz = \frac{1}{2\pi i} \int_{\Gamma} f(z) S(zI - J)^{-1} S^{-1} dz =$$

$$= S \left[ \frac{1}{2\pi i} \int_{\Gamma} f(z) (zI - J)^{-1} dz \right] S^{-1} = S \begin{bmatrix} f(J_1) & & \\ & \ddots & \\ & & f(J_n) \end{bmatrix} S^{-1} = f(A).$$

block diagonal with blocks  $\downarrow$

$$\frac{1}{2\pi i} \int_{\Gamma} f(z) \begin{bmatrix} z^{-1} - 1 & & & \\ & z^{-1} - 1 & & \\ & & \ddots & \\ & & & z^{-1} - 1 \end{bmatrix}^{-1} dz = \frac{1}{2\pi i} \int_{\Gamma} f(z)$$

$$\begin{bmatrix} (z-\lambda)^{-1} & (z-\lambda)^{-2} & (z-\lambda)^{-3} & \dots & (z-\lambda)^{-n} \\ & \circ & (z-\lambda)^{-1} & & \\ & & & \ddots & \\ & & & & (z-\lambda)^{-1} \end{bmatrix} dz$$

$$\frac{1}{2\pi i} \int_{\Gamma} \frac{f(z)}{z-\lambda} dz$$

$$\frac{1}{2\pi i} \int_{\Gamma} \frac{f(z)}{(z-\lambda)^2} dz$$

$$\frac{1}{2\pi i} \int_{\Gamma} \frac{f(z)}{(z-\lambda)^n} dz$$





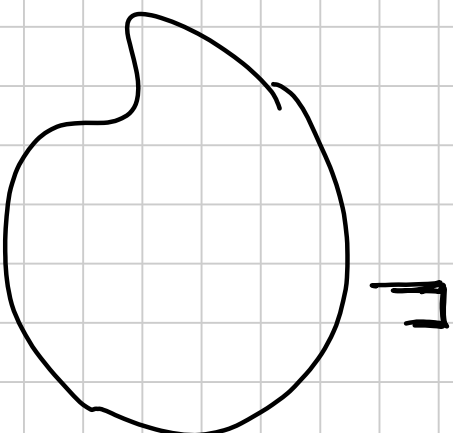
$$\leq \frac{1}{2\pi i} \int_{\Gamma} \|R(z)\| \left\| (zI - A_n)^{-1} - (zI - A)^{-1} \right\| \cdot dz \leq$$

$$\leq \frac{1}{2\pi i} \int_{\Gamma} \|R(z)\| \cdot \|(zI - A)^{-1}\| \cdot \|A - A_n\| \cdot \|(zI - A_n)^{-1}\| \cdot dz$$

bounded by  
 $\max_{\Gamma} \|R(z)\|$

$\forall \epsilon$

bounded if all eigenvalues of the  $A_n$  are separated from  $\Gamma$



Sampling  $R(A)$

(For general  $R$ )



1) If  $A$  diagonalizable,  $A = V \cdot \Lambda \cdot V^{-1}$   $\Lambda = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$

$$f(A) = V \begin{bmatrix} f(\lambda_1) & & \\ & \ddots & \\ & & f(\lambda_n) \end{bmatrix} V^{-1}$$

All ok if  $V$  is orthogonal (unitary)

Suppose we compute  $\tilde{f}(\lambda_i)$  with error

$$|f(\lambda_i) - \tilde{f}(\lambda_i)| \leq \varepsilon. \text{ Then}$$

$$f(A) - \tilde{f}(A) = V \left( \begin{bmatrix} f(\lambda_1) & & \\ & \ddots & \\ & & f(\lambda_n) \end{bmatrix} - \begin{bmatrix} \tilde{f}(\lambda_1) & & \\ & \ddots & \\ & & \tilde{f}(\lambda_n) \end{bmatrix} \right) V^{-1}$$

$$\|f(A) - \tilde{f}(A)\| \leq \|V\| \cdot \left\| \begin{bmatrix} f(\lambda_1) - \tilde{f}(\lambda_1) & & \\ & \ddots & \\ & & f(\lambda_n) - \tilde{f}(\lambda_n) \end{bmatrix} \right\| \cdot \|V^{-1}\|$$

$$= \|V\| \cdot \varepsilon \cdot \|V^{-1}\| = \kappa(V) \cdot \varepsilon$$

Ex  $f \left( \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix} \right)$   $f = \text{sqrt}$

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Polynomial s.t.  $f(2) = \sqrt{2}$

$$f'(x) = \frac{1}{2\sqrt{x}}$$

$$f'(2) = \frac{1}{2\sqrt{2}}$$

$$f(x) = \sqrt{2} + (x-2) \frac{1}{2\sqrt{2}}$$

$$= \frac{1}{2\sqrt{2}} x + \sqrt{2} - \frac{1}{\sqrt{2}}$$

$$\begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} P_0 \\ P_1 \end{bmatrix} = \begin{bmatrix} \sqrt{2} \\ \frac{1}{2\sqrt{2}} \end{bmatrix}$$

$$P_1 = \frac{1}{2\sqrt{2}}$$

$$P_0 = \sqrt{2} - 2 \cdot \frac{1}{2\sqrt{2}} =$$

$$= P_0 = \sqrt{2} - \frac{1}{\sqrt{2}}$$

$$\left(\sqrt{2} - \frac{1}{\sqrt{2}}\right) + \frac{1}{2\sqrt{2}} X$$

2) Finding irreducible polynomials and evaluating them.

Question: How does one evaluate a polynomial in a matrix?

1)  $\varphi(A) = \varphi_0 I + \varphi_1 A + \varphi_2 A^2 + \dots + \varphi_d A^d$  after evaluating  $A, A^2, \dots, A^d$   
 via successive products  $O(dn^2)$

2)  $\varphi_0 I + \varphi_1 A + \dots + \varphi_{d-2} A^{d-2} + \varphi_{d-1} A + \varphi_d A^d$  (Horner rule)

$O(dn^2)$  for  $\varphi(A)$

For scalar arguments, 1) needs  $d-1$  products for  $A_1, A_2, \dots, A_d$   
 $d-1$  products &  $d-1$  sums for the formula

2) needs  $d-1$  sums &  $d-1$  products

For matrices, 1) needs  $d-1$  matrix products for  $A_1, A_2, \dots, A_d$   $O(n^3)$   
 $d-1$  scalar-matrix products  $O(n^2)$   
 $d-1$  sums of matrices  $O(n^2)$

2) needs  $d-1$  matrix products  $O(n^3)$   
1 scalar-matrix product  $O(n^2)$   
 $d-1$  matrix sums  $O(n^2)$

Same cost (of least up to lower order terms)

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Another method: Paterson-Stockmayer method:  
divide terms in chunks of size  $\sqrt{d}$ ; e.g.

$$(p_8 A^2 + p_7 A + p_6 I) A^6 + (p_5 A^2 + p_4 A + p_3 I) A^3 + (p_2 A^2 + p_1 A + p_0)$$

mat muls: 1 to form  $A^2$  | 3 muls for degree 8  
 1 to form  $A^2 \cdot A = A^3$   
 1 to form  $A^3 \cdot A^3 = A^6$

$\sqrt{d}$  muls for  $A^1, A^2, \dots, A^{\sqrt{d}}$   $\sqrt{d}$  muls for  $A^{2\sqrt{d}}, A^{3\sqrt{d}}, \dots, A^d$

Storage of  $\sqrt{d}$  matrices (larger than the other methods  $O(1)$ )

The value  $\tilde{y}$  computed with this method satisfies

$$|\tilde{y} - \phi(A)| \leq (|p_0|I + |p_1| \cdot |A| + |p_2| \cdot |A|^2 + \dots + |p_d| \cdot |A|^d) O(\epsilon \cdot u)$$

$|X|$  means the square matrix  
with elements  $|X_{ij}|$

$\delta$   
machine precision

Other part of the question: if I use polynomials to approximate a matrix function, e.g.

$$\exp(A) \approx I + A + \frac{1}{2}A^2 + \dots + \frac{1}{k!}A^k,$$

does the result converge to the right matrix?

Note that if you have a sequence of polynomials  $p_n$  s.t.

$$p_n(x) \rightarrow f(x), \text{ this may not be sufficient to ensure} \\ \text{(for all } x) \\ \varphi(A) \rightarrow f(A):$$



If  $|p_n(\lambda) - f(\lambda)| \leq \varepsilon$  for each eigenvalue  $\lambda$ , then


$$\|f(A) - p_n(A)\| \leq \|V\| \cdot \left\| \begin{bmatrix} p_n(\lambda_1) - f(\lambda_1) \\ \vdots \\ p_n(\lambda_s) - f(\lambda_s) \end{bmatrix} \right\| \|V^{-1}\|$$

$$\leq \|V\| \cdot \|V^{-1}\| \cdot \varepsilon$$

Suppose you have polynomials s.t.

$$p_n(x) \rightarrow f(x) \quad \forall x \in [a, b]$$

$$p_n'(x) \neq f'(x)$$



$$p_n \left( \begin{bmatrix} A & 1 \\ 0 & A \end{bmatrix} \right) = \begin{bmatrix} p_n(A) & p_n'(A) \\ 0 & p_n(A) \end{bmatrix} \neq \begin{bmatrix} f(A) & f'(A) \\ 0 & f(A) \end{bmatrix}$$

Thm: [Higham book Th. 4.7]

Suppose  $f(x) = \sum_{k=0}^{\infty} P_k(x-\alpha)^k$        $P_k = \frac{f^{(k)}(\alpha)}{k!}$

Taylor series centered in  $\alpha$ , with convergence radius  $r$   
(converges inside the open ball  $\{z \in \mathbb{C} : |z-\alpha| < r\}$ )

Then  $\lim_{d \rightarrow \infty} \sum_{k=0}^d P_k(A-\alpha I)^k = f(A)$

For each matrix with eigenvalues s.t.  $|A_i - \alpha| < r$

Proof:  $r = \limsup (P_k)^{1/k}$

Let us call  $g_d(x) = \sum_{k=0}^d P_k (x-a)^k$

$$g_d^{(n)} = \sum_{k=0}^d P_k k(k-1) \dots (k-n+1) (x-a)^{k-n} = \sum_{k=0}^d \frac{f^{(k)}(a)}{k!} \cdot k(k-1) \dots (k-n+1) (x-a)^{k-n}$$

$$= \sum_{k=0}^d \frac{f^{(k)}(a)}{(k-n)!} (x-a)^{k-n} = \text{Taylor polynomials of } f^{(n)}(x)$$

w/ the same radius of convergence: indeed, the first deriv. series

$$f'(x) = \sum_{k=1}^{\infty} P_{k-1} k (x-a)^{k-1}$$

has radius of conv. given by

$$\limsup (k P_{k-1})^{1/k} = \limsup (P_k)^{1/k}$$

$$(\lim k^{1/k} = 1)$$

→ Taylor polynomials  $g_d(x) \rightarrow f(x)$

$$g'_d(x) \rightarrow f'(x)$$

$$g_d^{(k)}(x) \rightarrow f^{(k)}(x) \quad \forall k$$

→  $g_d(A) \rightarrow f(A)$  when  $d \rightarrow \infty$

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Why can't we use Taylor series for everything?

Problem 1: not everything has a Taylor series, e.g.  $f(x) = \sqrt{x}$   
in  $x=0$

Problem 2: intermediate growth of coefficients

$$f(x) = \exp(x)$$

$$A = \begin{bmatrix} 0 & 30 \\ -30 & 0 \end{bmatrix}$$

Intermediate growth:  $\frac{A^{30}}{30!} \approx 10^{11}$

Same problem for scalars:

$$\exp(50) = 1 + 50 + \frac{50^2}{2} + \frac{50^3}{3!} + \dots$$

Terms are large, but  $\exp(50)$  is even larger

$$\exp(-50) = 1 - 50 + \frac{50^2}{2} - \frac{50^3}{3!} + \dots$$

Terms are larger,  $\exp(-50)$  is tiny  
 $\Rightarrow$  very large (relative) error

For scalars, one can solve the  
issue by computing

$$\exp(-50) = \frac{1}{\exp(50)} \quad (\text{no cancellation any more})$$

For matrices, you compute the series  
on "all eigenvalues of the same time"  
and these tricks do not work.