

Polynomials of matrices

Another **different** way to make polynomials and matrices interact: take a scalar polynomial, and apply a (**square**) matrix to it, e.g.,

$$p(x) = 1 + 3\underline{x} - 5\underline{x}^2 \implies \underbrace{p(A)} = \underbrace{I + 3A - 5A^2}.$$

Lemma

If $A = \underline{S} \text{blkdiag}(J_1, J_2, \dots, J_s) \underline{S}^{-1}$ is a Jordan form, then $p(A) = \underline{S} \text{blkdiag}(\underline{p(J_1)}, p(J_2), \dots, p(J_s)) \underline{S}^{-1}$, and

$$\rightarrow p(J_i) = \begin{bmatrix} p(\lambda_i) & p'(\lambda_i) & \dots & \frac{1}{k!} p^{(k)}(\lambda_i) \\ & p(\lambda_i) & \ddots & \vdots \\ & & \ddots & p'(\lambda_i) \\ & & & p(\lambda_i) \end{bmatrix}.$$

(Proof: Taylor expansion of p around λ .)

Valutare polinomio scalare in un blocco di Jordan:

$$J = \lambda I + N \quad N = \begin{bmatrix} 0 & & & \\ & \ddots & & \\ & & \ddots & \\ & & & 0 \end{bmatrix}$$

$$p(x) = p(\lambda) + p'(\lambda)(x-\lambda) + \dots + \frac{1}{d!} p^{(d)}(\lambda)(x-\lambda)^d$$

$$p(J) = p(\lambda)I + p'(\lambda)(J - \lambda I) + \dots + \frac{1}{d!} p^{(d)}(\lambda)(J - \lambda I)^d =$$

$$= p(\lambda)I + p'(\lambda)N + \frac{1}{2} p''(\lambda)N^2 + \dots + \frac{1}{d!} p^{(d)}(\lambda)N^d =$$

$$= \begin{bmatrix} p(\lambda) & p'(\lambda) & & \\ & \ddots & & \\ & & \ddots & \\ & & & p(\lambda) \end{bmatrix}$$

Functions of matrices [Higham book, '08] *Chap. 1*

We can extend the same definition to arbitrary scalar functions:

Given a function $f : U \subseteq \mathbb{C} \rightarrow \mathbb{C}$, we say that f is **defined on A** if f is defined and differentiable at least $m_g(\lambda_i) - 1$ times on each eigenvalue λ_i of A .

Definition

If $A = S \text{blkdiag}(J_1, J_2, \dots, J_s) S^{-1}$ is a Jordan form, then $f(A) = S \text{blkdiag}(f(J_1), f(J_2), \dots, f(J_s)) S^{-1}$, where

$$f(J_i) = \begin{bmatrix} f(\lambda_i) & f'(\lambda_i) & \dots & \frac{1}{k!} f^{(k)}(\lambda_i) \\ & f(\lambda_i) & \ddots & \vdots \\ & & \ddots & f'(\lambda_i) \\ & & & f(\lambda_i) \end{bmatrix}.$$

(Reasonable doubt: is it independent of the choice of S ?)

Def: Hermite interpolating polynomial:

Dati nodi $\lambda_1, \lambda_2, \dots, \lambda_k$, molteplicità

$1 \leq m_i$ per $i=1, 2, \dots, k$, e una funzione derivabile m_i-1 volte in λ_i , esiste un (unico) polinomio di grado $(\sum m_i) - 1$ tale che $P^{(j)}(\lambda_i) = f^{(j)}(\lambda_i)$

per ogni $i=1, 2, \dots, k$, $j=1, 2, \dots, m_i$

ES: trovare un polinomio tale che

$$P(0) = 1 \quad P(1) = e$$

$$\rightarrow P'(0) = 1 \quad P'(1) = e$$

$$\rightarrow P''(0) = 1 \quad P''(1) = e$$

(coincide con $\exp(x)$
in 0 e 1, con
due derivate uguali)

$$(deg = 5)$$

$$\begin{array}{l}
 x_1 \\
 x_2 \\
 x_3 \\
 x_4
 \end{array}
 \left[\begin{array}{ccc}
 f(x_1) & f[x_1, x_2] & f[x_1, x_2, x_3] \\
 f(x_2) & f[x_2, x_3] & f[x_1, x_2, x_3, x_4] \\
 f(x_3) & f[x_2, x_3, x_4] & \\
 f(x_4) & f[x_3, x_4] &
 \end{array} \right]$$

(forma di Newton)

$$f[x_i, x_{i+1}] = \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i} \quad \leftarrow$$

$$f[x_i, x_{i+1}, x_{i+2}] = \frac{f[x_{i+1}, x_{i+2}] - f[x_i, x_{i+1}]}{x_{i+2} - x_i}$$

⋮

$$P(x) = f[x_1] + f[x_1, x_2](x-x_1) + f[x_1, x_2, x_3](x-x_1)(x-x_2) \\ + \dots + f[x_1, x_2, \dots, x_k](x-x_1)(x-x_2) \dots (x-x_{k-1})$$

(Newton form of interpolating polynomial)

Alternate definition: via Hermite interpolation

Definition

$f(A) = p(A)$, where p is a polynomial such that $f(\lambda_i) = p(\lambda_i)$, $f'(\lambda_i) = p'(\lambda_i), \dots, f^{(m_g(\lambda_i)-1)}(\lambda_i) = p^{(m_g(\lambda_i)-1)}(\lambda_i)$ for each i .

We may use this as a definition of $f(A)$ (and it does not depend on S).

Obvious from the definitions that it coincides with the previous one.

Remark: if $A \in \mathbb{C}^{m \times m}$ has multiple Jordan blocks with the same eigenvalue, these may be fewer than m conditions.

Remark: be careful when you say "all matrix functions are polynomials", because p depends on A .

$$f(A) = p(A) \quad A p(A) = p(A) A$$

$$f(B) = q(B) \quad B q(B) = q(B) B$$

Se f è un polinomio, allora $f(A)$ definita in questo modo coincide con il valore $f(x)$ in A :

$$A = SJS^{-1}$$

valore $f(x)$ in A produce $S \cdot \text{blkdiag}(f(J_1), \dots, f(J_k)) \cdot S^{-1}$

dove $f(J_i) = \begin{bmatrix} f(\lambda_i) & f'(\lambda_i) & \dots & \frac{1}{(m_i-1)!} f^{(m_i-1)}(\lambda_i) \\ & \ddots & \ddots & \vdots \\ & & f(\lambda_i) & \\ & & & f(\lambda_i) \end{bmatrix}$

che coincide con $p(J_i)$ se $p(x)$ è il polinomio di interpolazione.

Quindi $f(A)$ coincide con $p(A)$

Se voglio valutare $f(A)$ per

$$A = S \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix} S^{-1}, \text{ mi basta sapere } f(1)$$

e $f(2)$, 2 condizioni
anziché 4 che avrei in generale.

$$\sin(A) \quad \cos(A)$$

$$\sin(A) \cos(A) = \cos(A) \sin(A)$$

$$f(x) = \sin(x) \cos(x)$$

Some properties

- ▶ If the eigenvalues of A are $\lambda_1, \dots, \lambda_s$, the eigenvalues of $f(A)$ are $f(\lambda_1), \dots, f(\lambda_s)$. (Remark: geometric multiplicities may drop)
- ▶ $\underline{f(A)g(A)} = \underline{g(A)f(A)} = \underline{(fg)(A)}$ (since they are all polynomials in A).
- ▶ If $\underline{f_n \rightarrow f}$ together with 'enough derivatives' (for instance because they are analytic and the convergence is uniform), then $\underline{f_n(A)} \rightarrow \underline{f(A)}$.
- ▶ If a sequence of matrix $\underline{A_n \rightarrow A}$, then $\underline{f(A_n)} \rightarrow \underline{f(A)}$.
Proof: let p_n be the (Hermite) interpolating polynomial on the eigenvalues of A_n . Interpolating polynomials are continuous in the nodes, so $p_n \rightarrow p$ (coefficient by coefficient). Then
 $\underline{\|p_n(A_n) - p(A)\|} \leq \|p_n(A_n) - p_n(A)\| + \|p_n(A) - p(A)\| \leq \dots$

$$f(A_n) = p_n(A_n)$$

↑ ↑

Examples

- ▶ $\exp(A) = I + A + \frac{1}{2}A^2 + \frac{1}{3!}A^3 + \dots$. The identity holds also for non-diagonalizable matrices (by continuity).
- ▶ $\operatorname{sgn}(A)$ from

$$\operatorname{sgn}(x) = \begin{cases} -1 & \operatorname{Re} x < 0, \\ 1 & \operatorname{Re} x > 0, \\ \text{undefined} & \operatorname{Re} x = 0. \end{cases}$$

- ▶ $A^{1/2}$ from $f(x) = \sqrt{x}$. Note that we can choose signs (branch) independently on each eigenvalue. All the various ways satisfy $(A^{1/2})^2 = A$.

Nonprimary matrix functions

If a matrix A has multiple eigenvalues, one could also define a 'square root' by choosing different signs on Jordan blocks with the same eigenvalue: for instance, $\begin{bmatrix} 1 & \\ & -1 \end{bmatrix}$ as a square root of I_2 (or also $V \begin{bmatrix} 1 & \\ & -1 \end{bmatrix} V^{-1}$ for any invertible $V \dots$).

These are called **nonprimary** matrix functions (and they are **not** matrix functions with our definition).

(They all satisfy $f(A)^2 = A$.)

(They are **not** polynomials in A .)

Cauchy integrals

If f is analytic on and inside a contour Γ that encloses the eigenvalues of A ,

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

Generalizes the analogous scalar formula.

Proof If $A = V\Lambda V^{-1} \in \mathbb{C}^{m \times m}$ is diagonalizable, the integral equals

$$V \begin{bmatrix} \frac{1}{2\pi i} \int_{\Gamma} \frac{f(z)}{z-\lambda_1} dz & & \\ & \ddots & \\ & & \frac{1}{2\pi i} \int_{\Gamma} \frac{f(z)}{z-\lambda_m} dz \end{bmatrix} V^{-1} = V \begin{bmatrix} f(\lambda_1) & & \\ & \ddots & \\ & & f(\lambda_m) \end{bmatrix} V^{-1}.$$

By continuity, the equality holds also for non-diagonalizable A .

Methods

Matrix functions arise in several areas: solving ODEs (e.g. $\exp A$), matrix analysis (square roots), physics, ...

Main methods to compute them:

- ▶ Factorizations (eigendecompositions, Schur...),
- ▶ Matrix versions of scalar iterations (e.g., Newton on $x^2 = a$),
- ▶ Interpolation / approximation,
- ▶ Complex integrals.