

# The matrix exponential

We start our discussion of specific matrix functions from  $\expm(A)$ .

Easy to come up with ways that turn out to be unstable. [Moler, Van Loan, "Nineteen dubious ways to compute the exponential of a matrix", '78 & '03].

Example truncated Taylor series,  $I + A + \frac{1}{2}A^2 + \frac{1}{6}A^3 \cdots + \frac{1}{k!}A^k$ .

Frequent example that this is unstable also for scalars (cancellation if  $x < 0$ ). For scalars, cheap fix via  $\exp(-x) = \exp(x)^{-1}$ . For matrices, often we have both positive and negative eigenvalues.

Se  $x$  (scalare) negativo, ad es.  $-30$ , i termini

$1 + x + \frac{1}{2}x^2 + \frac{1}{3!}x^3 + \dots$  hanno segni alterni e si cancellano

Una matrice  $A$  ha autoval. sia positivi che negativi, non posso "scegliere" uno tra  $\exp(A)$   $\exp(-A)^{-1}$

## Growth in matrix powers

Additional problem in computing matrix power series: intermediate growth of coefficients.

**Example** Even on a nilpotent matrix, entries may grow.

$$\underline{A} = \begin{bmatrix} 0 & 10 & & \\ & 0 & 10 & \\ & & 0 & 10 \\ & & & 0 \end{bmatrix}, \quad \underline{A^2} = \begin{bmatrix} 0 & 0 & \underline{100} & \\ & 0 & 0 & \underline{100} \\ & & 0 & 0 \\ & & & 0 \end{bmatrix}, \quad A^3 = \begin{bmatrix} 0 & 0 & 0 & \boxed{1000} \\ & 0 & 0 & 0 \\ & & 0 & 0 \\ & & & 0 \end{bmatrix}.$$

$\|A^3\| = 1000$

Intrinsic problem on non-normal matrices. Growth + cancellation = trouble.

(On normal matrices,  $\|A^k\| = \|A\|^k = \lambda_{\max}^k$ .)

$AA^T = A^T A$ , oppure le si diagonalizzano con una  $Q$  ortogonale

$$\|(Q \Lambda Q^T)^k\|_F / \|Q \Lambda^k Q^T\| = \|\Lambda^k\| = |\lambda_{\max}|^k$$

# "Humps"

Similarly,  $\exp(\underline{tA})$  may grow for small values of  $t$  before 'settling down'.

## Example

```
>> A = [-0.97 25; 0 -0.3];  
>> t = linspace(0,20,100);  
>> for i = 1:length(t); y(i) = norm(expm(t(i)*A)); end  
>> plot(t, y)
```

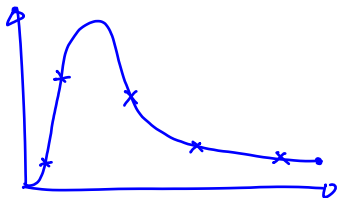
Shows it is also a bad idea to use an ODE solver on

$x(t) = \exp(tA)$  solve  $X'(t) = AX(t), X(0) = I;$   $e^z = \lim_{n \rightarrow \infty} \left(1 + \frac{z}{n}\right)^n$

Remark: explicit Euler produces  $\exp(\underline{At}) \approx (I + \frac{t}{n}A)^n$ .

$$X(t+h) \approx X(t) + h \cdot (AX(t)) = \frac{X(t+h) - X(t)}{h} \approx X'(t) = AX(t)$$
$$= (I + hA)X(t) \quad X(0) = I \quad X(h) \approx I + hA \quad X(2h) \approx (I + hA)^2 \dots$$

(Quindi anche  $\exp(A) \approx \left(I + \frac{1}{n}A\right)^n$  per  $n$  grande  
e problemi di crescita intermedia)



$$\begin{aligned} I + \frac{1}{n}A &\approx \exp\left(\frac{A}{n}\right) \\ \left(I + \frac{1}{n}A\right)^2 &\approx \exp\left(\frac{2A}{n}\right) \\ &\vdots \end{aligned}$$

$$\exp(A) \approx \left(I + \frac{1}{1024}A\right)^{1024}$$

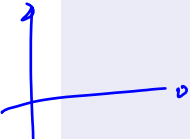
## Padé approximants

Padé approximants to the exponential (in  $x = 0$ ) are known explicitly.

Padé approximants to  $\exp(x)$

$p, q \in \mathbb{N}$  gradi:  $p = \deg N_{pq}(x)$   
 $q = \deg D_{pq}(x)$

$|\exp(x) - N_{pq}(x)/D_{pq}(x)| = \underline{O(x^{p+q+1})}$ , where


$$N_{pq}(x) = \sum_{j=0}^p \frac{(p+q-j)! p!}{(p+q)! j! (p-j)!} x^j, \quad \approx \exp\left(\frac{1}{2}x\right)$$

$$D_{pq}(x) = \sum_{j=0}^q \frac{(p+q-j)! q!}{(p+q)! j! (q-j)!} (-x)^j. \quad \approx \exp\left(-\frac{1}{2}x\right)$$

$$\exp(A) \approx \underline{(D_{pq}(A))^{-1}} N_{pq}(A).$$

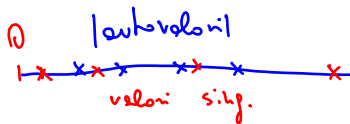
The main danger comes from  $D_{pq}(A)^{-1}$ .

For large  $p, q$ ,  $\underline{D_{pq}(A)} \approx \exp(-\frac{1}{2}A)$ .  $\underline{\kappa(D_{pq}(A))} \approx \frac{e^{-\frac{1}{2}\lambda_{\min}}}{e^{-\frac{1}{2}\lambda_{\max}}}$ .

Per una matrice  $M$ ,  $\sigma_i = \|M\| \geq |\lambda_i|$

Per  $M^{-1}$ ,  $\sigma_{\min}^{-1}(M) = \|M^{-1}\| \geq |\lambda_{\min}(M)|^{-1}$

$$\kappa(M) = \|M\| \cdot \|M^{-1}\| \geq |\lambda(M) \cdot \lambda_{\min}^{-1}(M)|$$



## Backward error of Padé approximants

Are Padé approximants reliable when  $\|A\|$  is small, at least?

Recall: perfect scalar approximation does not imply good matrix approximation.

Let  $H = f(A)$ , where  $f(x) = \log\left(\frac{e^{-x} N_{pq}(x)}{D_{pq}(x)}\right)$ .  $H$  is a matrix function, so it commutes with  $A$ . *= 1 + O(x^{p+q+1}), quindi il log esiste in un intorno di 0*

(Note that  $e^{-x} \frac{N_{pq}(x)}{D_{pq}(x)} = 1 + O(x^{p+q+1})$ , so the log exists for  $x$  sufficiently small).

One has  $\exp(H) = \exp(-A) (D_{pq}(A))^{-1} N_{pq}(A)$ , so

$$(D_{pq}(A))^{-1} N_{pq}(A) = \exp(A) \exp(H) = \exp(A + H)$$

(since  $A$  and  $H$  commute).

We can regard  $H$  as a sort of 'backward error': the Padé approximant  $(D_{pq}(A))^{-1} N_{pq}(A)$  is the exact exponential of a certain perturbed matrix  $A + H$ .

Can one bound  $\frac{\|H\|}{\|A\|}$ ?

$$\underline{H} = f(A)$$

$$f(x) = \log \left( e^{-x} \frac{N_{pq}(x)}{D_{pq}(x)} \right)$$

$$\exp(H) = g(A) =$$

$$g(x) = e^{-x} \frac{N_{pq}(x)}{D_{pq}(x)}$$

$$= \exp(-A) \cdot D_{pq}^{-1}(A) N_{pq}(A)$$

(perché se lo  
 $(f_1 \cdot f_2)(A) = f_1(A) \cdot f_2(A)$ )

$$\exp(A) \exp(H) = \boxed{D_{pq}^{-1}(A) N_{pq}(A)}$$

$= \exp(A+H)$  perché  $A$  e  $H$  commutano

La quantità calcolata  $D_{pq}^{-1}(A) N_{pq}(A)$  è l'esponentiale



esatto di una versione perturbata di  $A$ ,  $A+H$

Se riuscissi a fare in modo che  $\|H\| \leq \underline{u} \cdot \|A\|$

$\underline{u} \approx 2 \cdot 10^{-16}$ , prec. di macchina,

allora la mia approssimazione sarebbe tanto accurata quanto possibile su un computer con precisione  $\underline{u}$ .

(errore della stessa grandezza di quello commesso approssimando  $A \in \mathbb{R}^{n \times n}$  con  $fl(A)$ .)





Bounding  $\|H\|$   $f(x) = \log(1 + O(x^{p+q+1})) = O(x^{p+q+1})$

$H = f(A)$ , where  $f(x) = \log(e^{-x} \frac{N_{pq}(x)}{D_{pq}(x)})$ .

$f$  is analytic, so  $f(x) = \underline{c_1}x^{p+q+1} + \underline{c_2}x^{p+q+2} + \underline{c_3}x^{p+q+3} + \dots$

$$H = f(A) = \underline{c_1}A^{p+q+1} + \underline{c_2}A^{p+q+2} + \underline{c_3}A^{p+q+3} + \dots$$

$r = \|A\|$

$$\|H\| \leq |c_1| \|A\|^{p+q+1} + |c_2| \|A\|^{p+q+2} + |c_3| \|A\|^{p+q+3} + \dots$$

All these quantities can be computed, explicitly or with Mathematica (but it's a lot of work).

Luckily, someone did it for us. For instance:

[Higham book '08, p. 244]

If  $\underline{p = q = 13}$  and  $\|A\| \leq \underline{5.4}$ , then  $\frac{\|H\|}{\|A\|} \leq \mathbf{u}$  (machine precision).

(tabella con valori per ogni  $p=q$  su [Higham])  
 $\Rightarrow$  se scegliamo  $p=q=13$ , per ogni  $A$  tale che  $\|A\| \leq 5.4$ , l'errore all'indietro commesso in  $\exp(A) \approx D^{-1}(A)N(A)$  è  $\leq \underline{u}$

$$\|H\| \leq |c_1| \cdot r^{p+q+1} + |c_2| \cdot r^{p+q+2} + |c_3| \cdot r^{p+q+3} + \dots$$

dove  $r = \|A\|$

$$\frac{\|H\|}{\|A\|} = \frac{\|H\|}{r} = |c_1| \cdot r^{p+q} + |c_2| \cdot r^{p+q+1} + \dots \quad \longleftarrow$$

Riesco a trovare numericamente il valore  $r$

$$\text{per cui } |c_1| r^{p+q} + |c_2| r^{p+q+1} + \dots \leq \underline{u}$$

## Scaling and squaring

$$e^x = (e^{\frac{1}{3}x})^3 \cdot (e^{\frac{1}{2^k}x})^{2^k}$$

What if  $\|A\| > 5.4$ ? Trick:  $\exp(A) = (\exp(\frac{1}{s}A))^s$ .

### Algorithm (scaling and squaring)

1. Find  $s = 2^k$  such that  $\|\frac{1}{s}A\| \leq 5.4$ .

2. Compute  $F = D_{13,13}(B)^{-1}N_{13,13}(B)$ , where  $D_{13,13}$  and  $N_{13,13}$  are given polynomials and  $B = \frac{1}{s}A$ .

3. Compute  $F^{2^k}$  by repeated squaring.

$$B = \frac{1}{s}A \quad \|B\| \leq 5.4$$

$$F \approx \exp(B) = \exp(\frac{1}{s}A)$$

This is what is used in practice on Matlab.

Why 13? Chosen to minimize number of operations.

Note that we can evaluate  $D_{13,13}$  and  $N_{13,13}$  with 6 matmuls, using Paterson-Stockmeyer.

This is Matlab's `expm`, currently. (Warning: `exp(A)` is componentwise).

$A, A^2, A^3$

$$\begin{aligned} & (c_0 + c_1 A + c_2 A^2 + c_3 A^3) + (c_4 A + c_5 A^2 + c_6 A^3) A^3 \\ & + \dots \left( \quad \right) A^3 \times A^3 + \left( \dots \quad \right) \Rightarrow A^3 \times A^3 \times A^3 \end{aligned}$$

## Is scaling and squaring stable?

Note that 'humps' may still give problems:  $\exp(B)$  may be much larger than  $\exp(A) = \exp(B)^{2^k}$ , leading to cancellation in the squares.

Is scaling and squaring stable for all matrices? Yes numerically, but no definitive answer.

