

Interval Matrix Exponentiation

Alexandre Goldsztejn¹ Arnold Neumaier²

¹CNRS, Laboratoire d'Informatique de Nantes Atlantique
University of Nantes, France
alexandre.goldsztejn@univ-nantes.fr
www.goldsztejn.com

²University of Vienna, Faculty of Mathematics
Wien, Austria
Arnold.Neumaier@univie.ac.at
www.mat.univie.ac.at/~neum/



CENTRE NATIONAL
DE LA RECHERCHE
SCIENTIFIQUE



universität
wien



Simulation of Linear ODE

From Linear ODE to Matrix Exponentiation

- $x'(t) = Ax(t)$
- ⇒ $x''(t) = (Ax(t))' = A^2x(t)$, and by induction $x^{(n)}(t) = A^n x(t)$
- Taylor expansion:

$$x(t) = x_0 + (tA)x_0 + \frac{(tA)^2x_0}{2} + \frac{(tA)^3x_0}{3!} + \dots = (\exp tA)x_0$$

Matrix Exponentiation

- Truncated Taylor or Padé expansion computed in double precision
 - $\|A\|$ too large ⇒ high expansion order
 - Idea: $\exp(A) = \exp(A/2^s)^{2^s}$
 - Scaling and Squaring: $\exp(A) = (\dots ((r_m(A/2^s))^2)^2 \dots)^2$
- Issue: Find s and m
 - Lower computational cost
 - Improve precision

Simulation of Uncertain Linear ODE

Uncertain Linear ODE

- $x'(t) = Ax(t)$ with $A \in \mathbf{A}$
- $\Rightarrow x(t) \in \{(\exp A) x_0 : A \in \mathbf{A}\}$

Interval Matrix Exponentiation

- $\exp \mathbf{A} := \square\{\exp A : A \in \mathbf{A}\} \in \mathbb{IR}^{n \times n}$
- $\Rightarrow x(t) \in (\exp \mathbf{A}) x_0$
- Issue: sharp enclosure of $\exp \mathbf{A}$

Pessimism in Taylor Series Interval Evaluation

$$\mathbf{A} = \begin{pmatrix} 0 & 1 \\ 0 & [-3, -2] \end{pmatrix} \quad \exp \mathbf{A} \approx \begin{pmatrix} 1 & [0.31, 0.44] \\ 0 & [0.04, 0.14] \end{pmatrix} \quad \mathbf{T}_{10} \approx \begin{pmatrix} 1 & [-1.21, 1.96] \\ 0 & [-6.26, 6.45] \end{pmatrix}$$

Interval Matrix Exponentiation

Issue

- Sharp enclosure of $\exp \mathbf{A}$
- Taylor expansion of $\exp \mathbf{A} \rightarrow$ heavy multi-occurrences of variables a_{ij}
- ≠ Issue wrt matrix exponentiation in numerical analysis

Outline

- 1 NP-hardness
- 2 Polynomial Algorithms
- 3 Experiments
- 4 Experiments

Outline

- 1 NP-hardness
- 2 Polynomial Algorithms
- 3 Experiments
- 4 Experiments

Definition

- Let $\mathbb{A} \subseteq \mathbb{R}^{n \times m}$ and $\mathbf{A} = \square \mathbb{A} \in \mathbb{IR}^{n \times m}$
- $\mathbf{B} \in \mathbb{IR}^{n \times m}$ is an ϵ -enclosure of \mathbb{A} iff $\mathbf{B} \supseteq \mathbf{A}$ and

$$\max \left\{ \max_{ij} |\underline{a}_{ij} - \underline{b}_{ij}|, \max_{ij} |\bar{a}_{ij} - \bar{b}_{ij}| \right\} \leq \epsilon$$

- Idem for sets of reals, sets of vectors

Central Result

- Computing ϵ -accurate enclosure of the range over a box of a bilinear function is NP-hard
(nonconvex quadratic programming is NP-hard)

ϵ -Accurate Interval Matrix Exponential Enclosure

Theorem

Computing an ϵ -accurate enclosure of the interval matrix exponential is NP-hard

Proof

- Consider $x'By$ and $\mathbf{x}, \mathbf{y} \in \mathbb{IR}^n$ and

$$A := \left(\begin{array}{c|c|c|c} 0 & x' & 0 & 0 \\ \hline 0 & 0 & B & 0 \\ \hline 0 & 0 & 0 & y \\ \hline 0 & 0 & 0 & 0 \end{array} \right) \quad A^2 = \left(\begin{array}{c|c|c|c} 0 & 0 & x'B & 0 \\ \hline 0 & 0 & 0 & By \\ \hline 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 \end{array} \right) \quad A^3 = \left(\begin{array}{c|c|c|c} 0 & 0 & 0 & x'By \\ \hline 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 \end{array} \right) \quad A^4 = 0$$

- A is nilpotent $\Rightarrow (\exp A)_{1,2n+2} = \frac{1}{6}x'By$
- ϵ -accurate enclosure of $\exp \mathbf{A}$ gives rise to ϵ -accurate enclosure of $\{x'By : x \in \mathbf{x}, y \in \mathbf{y}\}$

Outline

- 1 NP-hardness
- 2 Polynomial Algorithms**
- 3 Experiments
- 4 Experiments

1 NP-hardness

2 Polynomial Algorithms

- Taylor Series Interval Evaluation
- Taylor Series Horner Interval Evaluation
- Interval Scaling and Squaring
- State of the Art

3 Experiments

4 Experiments

Taylor Series Interval Evaluation

Definition

$$\begin{aligned}\tilde{\mathcal{T}}_K(\mathbf{A}) &:= I + \mathbf{A} + \frac{1}{2}\mathbf{A}^2 + \dots + \frac{1}{K!}\mathbf{A}^K \\ \mathcal{T}_K(\mathbf{A}) &:= \tilde{\mathcal{T}}_K(\mathbf{A}) + \mathcal{R}_K(\mathbf{A})\end{aligned}$$

where the interval remainder $[\mathcal{R}](\mathbf{A}, K)$ is

$$\mathcal{R}_K(\mathbf{A}) := \rho(\|\mathbf{A}\|, K) [-E, E] \quad \text{with} \quad \rho(\alpha, K) = \frac{\alpha^{K+1}}{(K+1)! \left(1 - \frac{\alpha}{K+2}\right)}$$

$\rho(\alpha, K)$: well known upper bound truncation error **valid for $\|\mathbf{A}\| \leq K+2$**

Example

$$\mathcal{T}_{16}(\mathbf{A}) \approx \begin{pmatrix} 1 \pm 9 \times 10^{-7} & [-1.21, 1.96] \\ \pm 9 \times 10^{-7} & [-6.26, 6.45] \end{pmatrix} \supseteq \begin{pmatrix} 1 & [0.31, 0.44] \\ 0 & [0.04, 0.14] \end{pmatrix}$$

Higher order for the expansion do not improve the entries (1, 2) and (2, 2) anymore

1 NP-hardness

2 Polynomial Algorithms

- Taylor Series Interval Evaluation
- **Taylor Series Horner Interval Evaluation**
- Interval Scaling and Squaring
- State of the Art

3 Experiments

4 Experiments

Definition

$$\begin{aligned}\tilde{\mathcal{H}}_K(\mathbf{A}) &:= I + \mathbf{A} \left(I + \frac{\mathbf{A}}{2} \left(I + \frac{\mathbf{A}}{3} \left(\cdots \left(I + \frac{\mathbf{A}}{K} \right) \cdots \right) \right) \right) \\ \mathcal{H}_K(\mathbf{A}) &:= \tilde{\mathcal{H}}_K(\mathbf{A}) + \mathcal{R}_K(\mathbf{A})\end{aligned}$$

Remark: Multi-occurrences decreased by Horner scheme

Example

$$\mathcal{H}_{16}(\mathbf{A}) \approx \begin{pmatrix} 1 \pm 1.1 \times 10^{-6} & [-0.08, 0.74] \\ \pm 1.1 \times 10^{-6} & [-1.21, 1.22] \end{pmatrix} \supseteq \begin{pmatrix} 1 & [0.31, 0.44] \\ 0 & [0.04, 0.14] \end{pmatrix}$$

Horner scheme reduces the pessimism (well known for real polynomials, e.g. Ceberio & Granvilliers 2002)

1 NP-hardness

2 **Polynomial Algorithms**

- Taylor Series Interval Evaluation
- Taylor Series Horner Interval Evaluation
- **Interval Scaling and Squaring**
- State of the Art

3 Experiments

4 Experiments

Interval Scaling and Squaring

Definition

$$S_{L,K}(\mathbf{A}) := \left(\mathcal{H}_K(\mathbf{A}/2^L) \right)^{2^L}$$

Theorem

$S_{L,K}(\mathbf{A})$ contains $\exp A$ for every $A \in \mathbf{A}$

Proof: • $\mathcal{H}_K(\mathbf{A}/2^L)$ contains $\exp A/2^L$
 $\Rightarrow S_{L,K}(\mathbf{A})$ contains $(\exp A/2^L)^{2^L} = \exp A$

Example

$$S_{10,6}(\mathbf{A}) \approx \begin{pmatrix} 1 \pm 1. \times 10^{-11} & [0.31, 0.44] \\ \pm 2.4 \times 10^{-19} & [0.04, 0.14] \end{pmatrix} \supseteq \begin{pmatrix} 1 & [0.31, 0.44] \\ 0 & [0.04, 0.14] \end{pmatrix}$$

Much sharper than Taylor and Horner Taylor

1 NP-hardness

2 **Polynomial Algorithms**

- Taylor Series Interval Evaluation
- Taylor Series Horner Interval Evaluation
- Interval Scaling and Squaring
- **State of the Art**

3 Experiments

4 Experiments

State of the Art

Michel and Oppenheimer (IEEE Trans. Circuits and Systems 1988)

- Specific centered form: $\exp \mathbf{A} \subseteq \mathcal{H}_K(\hat{\mathbf{A}}) + [\underline{\Delta}, \overline{\Delta}]$
- Tedious proofs with gaps, very complex computation of $[\underline{\Delta}, \overline{\Delta}]$
- Horner not sharp even for a thin matrix (cf. Experiments)

Gambill and Skeel (SINUM 1988)

- Specific method for non autonomous non homogeneous linear ODE:

$$((\mathbf{G}_8 \mathbf{G}_7)(\mathbf{G}_6 \mathbf{G}_5))((\mathbf{G}_4 \mathbf{G}_3)(\mathbf{G}_2 \mathbf{G}_1)) \text{ instead of } \mathbf{G}_8(\mathbf{G}_7(\cdots(\mathbf{G}_1)\cdots))$$

- Similar to scaling and squaring when applied to autonomous homogeneous linear ODE
- + Applies to more general systems
- Use $(\mathcal{T}_K(\mathbf{A}/2^L))^{2^L}$ which is less sharp than $(\mathcal{H}_K(\mathbf{A}/2^L))^{2^L}$ (necessary in the general case)
- Repeated computations (necessary in the general case)

Outline

- 1 NP-hardness
- 2 Polynomial Algorithms
- 3 Experiments
 - Exponentiation of a Thin Matrix
 - Exponentiation of an Interval Matrix
 - Tridiagonal Matrices
- 4 Experiments

Outline

- 1 NP-hardness
- 2 Polynomial Algorithms
- 3 Experiments**
 - **Exponentiation of a Thin Matrix**
 - Exponentiation of an Interval Matrix
 - Tridiagonal Matrices
- 4 Experiments

Exponentiation of a Thin Matrix

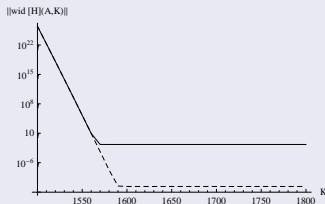
Matrix from Bochev and Markov (Computing 1999)

$$A := \begin{pmatrix} -131 & 19 & 18 \\ -390 & 56 & 54 \\ -387 & 57 & 52 \end{pmatrix}$$

Difficult to exponentiate:

- Significant eigenvalue separation
- Poorly conditioned eigenvector set

Horner Evaluation



- $\|A\| = 500$ implies $K \geq 502$
- 110 and 120 digits interval arithmetic (Mathematica)

Scaling and Squaring

$$\| \text{wid } S_{12,12}(A) \| \approx 10^{-6}$$

- Standard double precision interval arithmetic

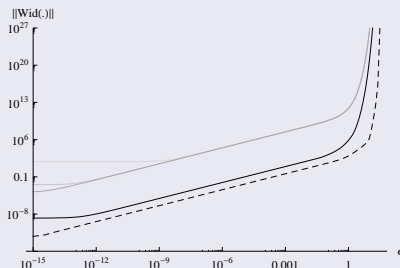
Outline

- 1 NP-hardness
- 2 Polynomial Algorithms
- 3 Experiments**
 - Exponentiation of a Thin Matrix
 - Exponentiation of an Interval Matrix**
 - Tridiagonal Matrices
- 4 Experiments

Exponentiation of an Interval Matrix

Matrix from Bochev and Markov (Computing 1999)

$$\mathbf{A}_\epsilon := 0.1\mathbf{A} + [-\epsilon, \epsilon]$$



- Gray: Horner evaluation
- Black: Scaling and squaring
- Dashed: Inner approximation

Outline

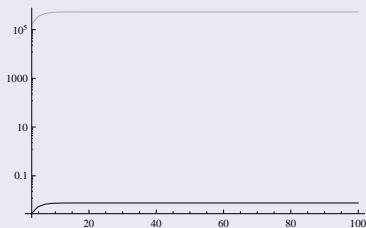
- 1 NP-hardness
- 2 Polynomial Algorithms
- 3 Experiments**
 - Exponentiation of a Thin Matrix
 - Exponentiation of an Interval Matrix
 - **Tridiagonal Matrices**
- 4 Experiments

Tridiagonal Matrices (1/2)

Exponentiation of an Interval Matrix

- \mathbf{T}_n defined by $t_{ii} = [-11, -9]$ and $t_{i+1,i} = t_{i,i+1} = [0, 2]$

$$[\mathbf{T}_3] = \begin{pmatrix} [-11, -9] & [0, 2] & 0 \\ [0, 2] & [-11, -9] & [0, 2] \\ 0 & [0, 2] & [-11, -9] \end{pmatrix}$$

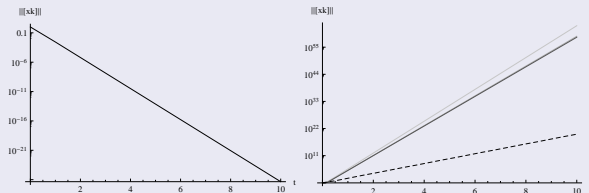


- n varies from 3 to 100
 - Gray: Horner evaluation
 - Black: Scaling and squaring

Tridiagonal Matrices (2/2)

Simulation of Linear ODE

- Simulation of $x'(t) = Ax(t)$ with $A \in \mathbf{T}_3$



- Left: Scaling and squaring
- Right: Basic method (full lines), and VNODE-LP (dashed line)

Originality vs State of the Art

- Most is known to the experts
- Gambill and Skeel:
 - Experiments with thin matrices
 - In this case, parallelotope methods better
 - Our experiments: Better than parallelotope methods for uncertain linear ODE
- Michel and Oppenheimer: Scaling and squaring simpler and sharper

Future Work

- Use parallelotope and interval matrix exponential
- Generalize Gambill and Skeel to nonlinear ODE