

# Taylor model methods vs. interval methods for ODEs: Similarities and distinctions

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# Outline

- 1 Verified Integration of ODEs
- 2 Interval Methods for ODEs
- 3 Taylor Model Methods for ODEs
- 4 Verified Integration of Linear ODEs

# Verified Integration of ODEs

Interval IVP:

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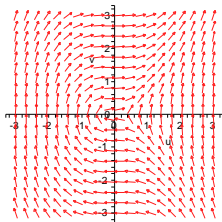
$f : \mathbb{R} \times \mathbb{R}^m \rightarrow \mathbb{R}^m$  sufficiently smooth,  $\mathbf{u}_0 \in \mathbb{IR}^m$ ,  $t_{\text{end}} > t_0$ .

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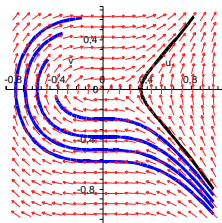


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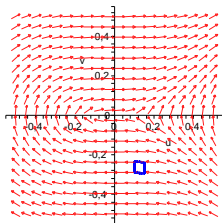


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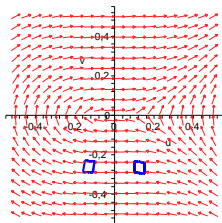


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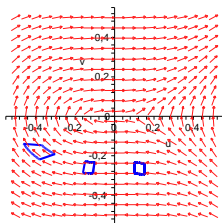


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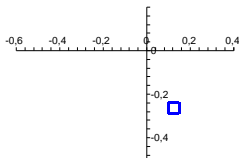


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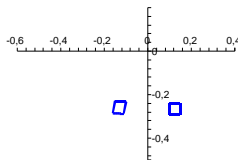


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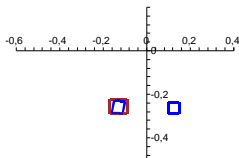


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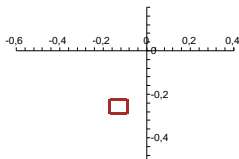


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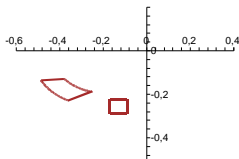


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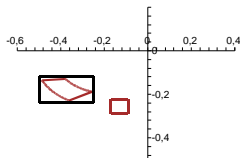


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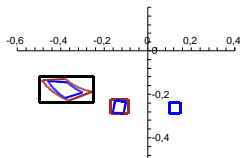


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# Interval Methods for ODEs

## Introduction

Set of autonomous IVPs:

$$u' = f(u), \quad u(t_0) = u_0 \in \mathbf{u}_0, \quad t \in \mathbf{t} = [t_0, t_{\text{end}}],$$

where  $D \subset \mathbb{R}^m$ ,  $f \in C^n(D)$ ,  $f : D \rightarrow \mathbb{R}^m$ ,  $\mathbf{u}_0 \in \mathbb{I}\mathbb{R}^m$ .

Moore's enclosure method:

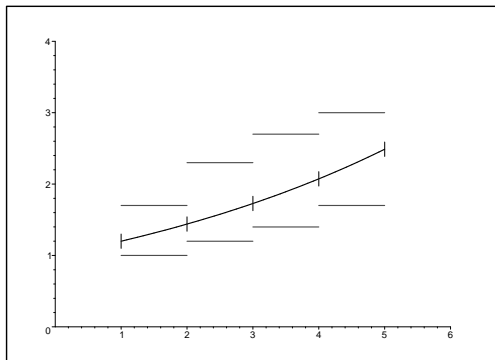
- Automatic computation of Taylor coefficients.
- Interval iteration: For  $j := 1, 2, \dots$ :

A priori enclosure:  $\mathbf{v}_j \supseteq u(t)$  for all  $t \in [t_{j-1}, t_j]$  ("Alg. I").

Truncation error:  $\mathbf{z}_j := \frac{h_j^{(n+1)}}{(n+1)!} f^{(n)}(\mathbf{v}_j)$ .

$u(t_j) \in \mathbf{u}_j := \mathbf{u}_{j-1} + \sum_{k=1}^n \frac{h_j^k}{k!} f^{(k-1)}(\mathbf{u}_{j-1}) + \mathbf{z}_j$  ("Algorithm II").

## Piecewise constant a priori enclosure



# A priori Enclosures

- Picard iteration: find  $h_j, \mathbf{v}_j$  such that

$$\mathbf{u}_{j-1} + [0, h_j]f(\mathbf{v}_j) \subseteq \mathbf{v}_j$$

- Step size restrictions: Explicit Euler steps
- More advanced schemes: Lohner 1988, Corliss & Rihm 1996, Nedialkov & Jackson 2001, Makino 1998.

## Modifications of Algorithm II

- Reduction of wrapping effect: Moore, Eijgenraam, Lohner, Rihm, Kuehn, Nedialkov & Jackson, ...
- Nedialkov & Jackson: Hermite-Obreshkov-Method
- Rihm: Implicit methods
- Petras & Hartmann, Bouissou: Runge-Kutta-Methods
- Berz & Makino: Taylor models  
Taylor expansion with respect to initial values and parameters

## Direct method

Direct method (Moore 1965): Apply mean value form to  $f^{(k)}$ :

$$\hat{u}_j \in \mathbf{u}_j, \quad f^{[0]}(u) = u, \quad f^{[k]}(u) = \frac{1}{k} \left( \frac{\partial f^{[k-1]}}{\partial u} f \right) (u) \quad \text{for } k \geq 1.$$

Let

$$\mathbf{S}_j = I + \sum_{k=1}^n h_0^k J(f^{[k]}(\mathbf{u}_{j-1})), \quad \mathbf{z}_j = h_0^{n+1} f^{[n]}(\mathbf{v}_j),$$

( $I$ : identity matrix,  $J(f^{[k]})$ : Jacobian of  $f^{[k]}$ ), then

$$u(t_j; u_0) \in \mathbf{u}_j = \hat{u}_{j-1} + \sum_{k=1}^{n-1} h_j^k f^{[k]}(\hat{u}_{j-1}) + \mathbf{z}_j + \mathbf{S}_j(\mathbf{u}_{j-1} - \hat{u}_{j-1}).$$

## Interval Methods for ODEs

Wrapping effect:  $\mathbf{S}_j(\mathbf{u}_{j-1} - \hat{\mathbf{u}}_{j-1})$  may overestimate

$$\mathcal{S} = \{S_j(u_{j-1} - \hat{u}_{j-1}) \mid S_j \in \mathbf{S}_{j-1}, u_{j-1} \in \mathbf{u}_{j-1}\}$$

Reduction: propagate  $\mathcal{S}$  as a parallelepiped.

Let  $\hat{\mathbf{u}}_0 := \mathbf{m}(\mathbf{u}_0)$ ,  $\mathbf{r}_0 = \mathbf{u}_0 - \hat{\mathbf{u}}_0$ ,  $B_0 = I$ .

For some nonsingular matrices  $B_j$ , do:

$$\left. \begin{aligned} \hat{\mathbf{u}}_j &= \hat{\mathbf{u}}_{j-1} + \sum_{k=1}^{n-1} h_{j-1}^k f^{[k]}(\hat{\mathbf{u}}_{j-1}) + \mathbf{m}(\mathbf{z}_j), \\ \mathbf{u}_j &= \hat{\mathbf{u}}_{j-1} + \sum_{k=1}^{n-1} h_{j-1}^k f^{[k]}(\hat{\mathbf{u}}_{j-1}) + \mathbf{z}_j + (\mathbf{S}_{j-1} B_{j-1}) \mathbf{r}_{j-1}, \end{aligned} \right\}$$

$\hat{\mathbf{u}}_j$ : approximate point solution for the central IVP

$\mathbf{z}_j$ : local error;  $\mathbf{r}_j$ : global error

## Global error

Global error propagation:

$$\mathbf{r}_j = \left( B_j^{-1}(\mathbf{S}_{j-1}B_{j-1}) \right) \mathbf{r}_{j-1} + B_j^{-1}(\mathbf{z}_j - m(\mathbf{z}_j))$$

- Moore's direct method:  $B_j = I$
- Pep method (Eijgenraam, Lohner):  $B_j = m(\mathbf{S}_{j-1}B_{j-1})$
- QR method (Lohner):  $B_j$  is chosen as the orthogonal matrix  $Q$  in the QR factorization of  $m(\mathbf{S}_{j-1}B_{j-1})$
- Blunting method (Berz, Makino): modify  $B_j$  in the pep method such that condition numbers remain small

# Taylor Model Methods for ODEs

# Quadratic Model Problem

$$\begin{aligned}u' &= v, & u(0) &\in [0.95, 1.05], \\v' &= u^2, & v(0) &\in [-1.05, -0.95].\end{aligned}$$

Taylor model method: initial set described by parameters  $a$  and  $b$ :

$$\begin{aligned}u_0(a, b) &:= 1 + a, & a \in \mathbf{a} &:= [-0.05, 0.05], \\v_0(a, b) &:= -1 + b, & b \in \mathbf{b} &:= [-0.05, 0.05].\end{aligned}$$

## Naive Taylor Model Method

Picard iteration ( $n = 3$ ,  $h = 0.1$ ):

$$u^{(0)}(\tau, a, b) = 1 + a, \quad v^{(0)}(\tau, a, b) = -1 + b,$$

$$u^{(1)}(\tau, a, b) = u_0(a, b) + \int_0^\tau v^{(0)}(s, a, b) ds$$

$$v^{(1)}(\tau, a, b) = v_0(a, b) + \int_0^\tau (u^{(0)}(s, a, b))^2 ds$$

$$u^{(3)}(\tau, a, b) = 1 + a - \tau + b\tau + \frac{1}{2}\tau^2 + a\tau^2 - \frac{1}{3}\tau^3,$$

$$v^{(3)}(\tau, a, b) = -1 + b + \tau + 2a\tau - \tau^2 + a^2\tau - a\tau^2 + b\tau^2 + \frac{2}{3}\tau^3.$$

## Naive Taylor Model Method: Remainder Bounds

Remainder bounds by fixed point iteration (Makino, 1998):

Find  $\mathbf{i}_0$  and  $\mathbf{j}_0$  s.t.

$$u_0 + \int_0^\tau \left( v^{(3)}(s, a, b) + \mathbf{j}_0 \right) ds \subseteq u^{(3)}(\tau, a, b) + \mathbf{i}_0,$$

$$v_0 + \int_0^\tau \left( u^{(3)}(s, a, b) + \mathbf{i}_0 \right)^2 ds \subseteq v^{(3)}(\tau, a, b) + \mathbf{j}_0$$

for all  $a \in \mathbf{a}$ ,  $b \in \mathbf{b}$ ,  $\tau \in [0, 0.1]$ .

## Naive Taylor Model Method: Enclosure of the Flow

Flow for  $\tau \in [0, 0.1]$ :

$$\tilde{\mathcal{U}}_1(\tau, a, b) := u^{(3)}(\tau, a, b) + \mathbf{i}_0,$$

$$\tilde{\mathcal{V}}_1(\tau, a, b) := v^{(3)}(\tau, a, b) + \mathbf{j}_0.$$

Flow at  $t_1 = 0.1$ :

$$\mathcal{U}_1(a, b) := \tilde{\mathcal{U}}_1(0.1, a, b) = 0.905 + 1.01a + 0.1b + \mathbf{i}_0,$$

$$\mathcal{V}_1(a, b) := \tilde{\mathcal{V}}_1(0.1, a, b) = -0.909 + 0.19a + 1.01b + 0.1a^2 + \mathbf{j}_0$$

(nonlinear boundary).

# Naive Taylor Model Method: Second Integration Step

Find  $\mathbf{i}_1$  and  $\mathbf{j}_1$  s.t.

$$\mathcal{U}_1(a, b) + \int_0^\tau (v^{(3)}(s, a, b) + \mathbf{j}_1) ds \subseteq u^{(3)}(\tau, a, b) + \mathbf{i}_1,$$

$$\mathcal{V}_1(a, b) + \int_0^\tau (u^{(3)}(s, a, b) + \mathbf{i}_1)^2 ds \subseteq v^{(3)}(\tau, a, b) + \mathbf{j}_1$$

for all  $a, b \in [-0.05, 0.05]$  and for all  $\tau \in [0, 0.1]$ .

# Naive Taylor Model Method: Second Integration Step

Find  $\mathbf{i}_1$  and  $\mathbf{j}_1$  s.t.

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for all  $a, b \in [-0.05, 0.05]$  and for all  $\tau \in [0, 0.1]$ .

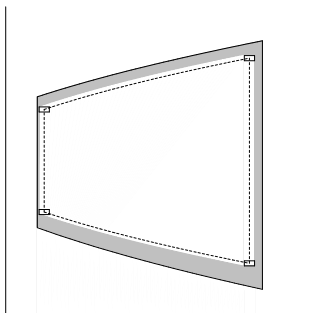
Since  $\mathbf{i}_0$  and  $\mathbf{j}_0$  are contained in  $\mathcal{U}_1$  and  $\mathcal{V}_1$ , diameters of interval terms are nondecreasing!

# Shrink Wrapping

Absorb interval term into polynomial part via shrink wrap factor  $q$   
 (Makino and Berz 2002):

$$\left. \begin{aligned}
 \mathcal{U}(a, b) &:= 2 + 4a + \frac{1}{2}a^2 + [-0.2, 0.2], \\
 \mathcal{V}(a, b) &:= 1 + 3b + 1ab + [-0.1, 0.1], \\
 \mathcal{U}_{\text{sw}}(a, b) &:= 2 + \frac{89}{20}a + \frac{89}{160}a^2, \\
 \mathcal{V}_{\text{sw}}(a, b) &:= 1 + \frac{287}{80}b + \frac{89}{80}ab.
 \end{aligned} \right\} \begin{aligned}
 &a, b \in [-1, 1], \\
 &(q = \frac{89}{80}).
 \end{aligned}$$

# Shrink Wrapping



$$\begin{pmatrix} \mathcal{U} \\ \mathcal{V} \end{pmatrix} \text{ (white) vs. } \begin{pmatrix} \mathcal{U}_{\text{sw}} \\ \mathcal{V}_{\text{sw}} \end{pmatrix}.$$

# Integration with Preconditioned Taylor Models

Preconditioned integration: flow at  $t_j$ :

$$\mathcal{U}_j = \mathcal{U}_{l,j} \circ \mathcal{U}_{r,j} = (p_{l,j} + \mathbf{i}_{l,j}) \circ (p_{r,j} + \mathbf{i}_{r,j}).$$

Purpose: stabilize integration similar to QR interval method.

## Theorem (Makino and Berz 2004)

*If the initial set of an IVP is given by a preconditioned Taylor model, then integrating the flow of the ODE only acts on the left Taylor model.*

# Integration with Preconditioned Taylor Models

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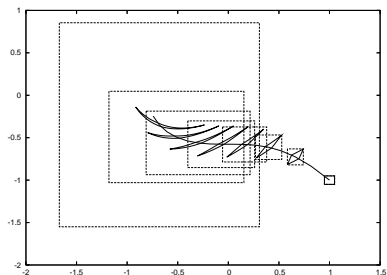
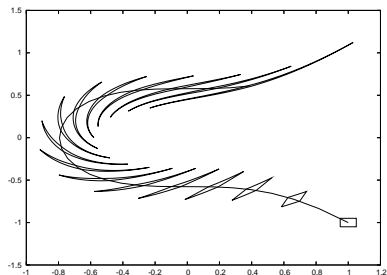
In practice:

$$\mathcal{U}_j = (\tilde{\mathcal{U}}_{l,j} \circ S_j) \circ (S_j^{-1} \circ \tilde{\mathcal{U}}_{r,j}) \quad S_j : \text{scaling matrix.}$$

such that

$$\text{Rg} \left( S_j^{-1} \circ \tilde{\mathcal{U}}_{r,j} \right) \approx [-1, 1]^m.$$

# Integration of Model Problem with COSY Infinity and AWA



# Verified Integration of Linear ODEs

## Interval Methods: Enclosure Representation

$$\begin{aligned}
 u(t_j; \mathbf{u}_0) &= \{u(t_j; u_0) \mid u_0 \in \mathbf{u}_0\} \\
 &\subseteq \{u_j + S_j w + B_j r \mid w \in \mathbf{u}_0 - m(\mathbf{u}_0), r \in \mathbf{r}_j\},
 \end{aligned}$$

where

- $u_j, w, r \in \mathbb{R}^m$ ,  $\mathbf{r}_j \in \mathbb{IR}^m$ ,
- $S_j, B_j \in \mathbb{R}^{m \times m}$ ,  $B_j$  nonsingular,
- $\{u_j + S_j w \mid w \in \mathbf{u}_0 - m(\mathbf{u}_0)\}$ : approximation to  $u(t_j; \mathbf{u}_0)$ ,
- $\{B_j r \mid r \in \mathbf{r}_j\}$ : bound on global error.

$$j = 0: \quad u_0 = m(\mathbf{u}_0), \quad \mathbf{r}_0 = 0, \quad S_0 = B_0 = I.$$

# Taylor Method for Linear Model Problem

Linear model problem:

$$u' = Au, \quad (A \in \mathbb{R}^{m \times m}, m \geq 2)$$

$$u(0) = u_0 \in \mathbf{u}_0.$$

Taylor method (constant order  $n$ , stepsize  $h$ ):

$$u_j := Tu_{j-1} \quad , \quad j = 1, 2, \dots$$

$$\left( T = T_{n-1}(hA) = \sum_{\nu=0}^{n-1} \frac{(hA)^\nu}{\nu!} \right)$$

# Taylor Method for Linear Model Problem

Linear model problem:

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$$u(0) = u_0 \in \mathbf{u}_0.$$

**Interval** Taylor method (constant order  $n$ , stepsize  $h$ ):

$$u_j := Tu_{j-1} + z_j, \quad j = 1, 2, \dots$$

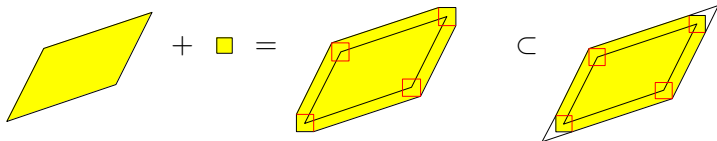
$$\left( T = T_{n-1}(hA) = \sum_{\nu=0}^{n-1} \frac{(hA)^\nu}{\nu!}; \quad z_j : \text{local error} \right)$$

# Propagation of the Global Error

$$\mathbf{r}_j = (B_j^{-1}TB_{j-1})\mathbf{r}_{j-1} + B_j^{-1}(\mathbf{z}_j - m(\mathbf{z}_j)).$$

Required:  $B_j$  for tight enclosure:

$$\{TB_{j-1}r + z \mid r \in \mathbf{r}_{j-1}, z \in \mathbf{z}_j - m(\mathbf{z}_j)\} \subseteq \{B_j r \mid r \in \mathbf{r}_j\}.$$



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- Direct method:  $B_j = I.$
- Parallelepiped method:  $B_j = TB_{j-1}.$
- QR method:  $B_j = Q_j, Q_j R_j = TB_{j-1}.$
- Blunting method:  $B_j = TB_{j-1} + \varepsilon Q_j, \varepsilon > 0.$

## Linear ODE: Naive TMM

Linear autonomous system ( $A \in \mathbb{R}^{m \times m}$ ):

$$u' = A u, \quad u(0) \in \mathbf{u}_0 = \mathcal{U}_0.$$

Direct interval method ( $\mathbf{z}_j$ : local error,  $T = \sum_{\nu=0}^n \frac{(hA)^\nu}{\nu!}$ ):

$$\mathbf{r}_j = T \mathbf{r}_{j-1} + \mathbf{z}_j - m(\mathbf{z}_j), \quad j = 1, 2, \dots$$

Naive Taylor model method:

$$\mathcal{U}_j = T^j \mathcal{U}_0 + \sum_{k=1}^j (T \circ)^{j-k} \mathbf{i}_k, \quad j = 1, 2, \dots,$$

where  $(T \circ)^0 \mathbf{x} := \mathbf{x}$ ,  $(T \circ)^k \mathbf{x} := T \cdot ((T \circ)^{k-1} \mathbf{x})$ ,  $k \in \mathbb{N}$ .

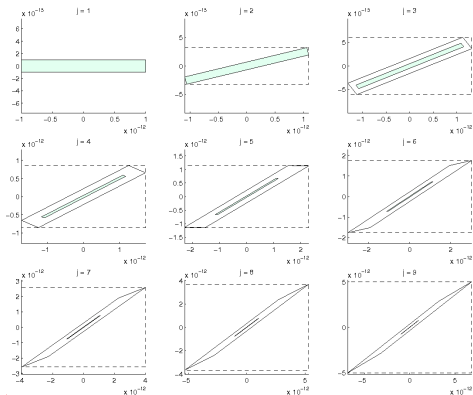
## Wrapping Effect: Direct Method

$$\mathbf{r}_j = (B_j^{-1} T B_{j-1}) \mathbf{r}_{j-1} + B_j^{-1} (\mathbf{z}_j - \mathbf{m}(\mathbf{z}_j)), \quad B_j = I :$$

$$\mathbf{r}_j = T \mathbf{r}_{j-1} + \mathbf{z}_j - \mathbf{m}(\mathbf{z}_j).$$

- Optimal coordinates for local error.
- Bad coordinates for global error (rotation).

# Wrapping Effect: Direct Method



(Plots by Ned Nedialkov)

Huge overestimations in general.

## Linear ODE: Naive TMM with Shrink Wrapping

Linear autonomous system ( $A \in \mathbb{R}^{m \times m}$ ):

$$u' = A u, \quad u(0) \in \mathbf{u}_0 = \mathcal{U}_0.$$

Parallelepiped method ( $\mathbf{z}_j$ : local error,  $\mathbf{r}_0 := \mathbf{u}_0 - m(\mathbf{u}_0)$ ):

$$\mathbf{r}_j := \mathbf{r}_{j-1} + (T^{1-j})(\mathbf{z}_j - (m(\mathbf{z}_j))), \quad j = 1, 2, \dots$$

Naive Taylor model method with shrink wrapping:

$$d_j := \|\mathbb{w}(T^{-j} \mathbf{i}_j)\|_\infty, \quad q_j := 1 + d_j/2, \quad \tilde{p}_{sw,j} := \left( \prod_{k=1}^j q_k \right) \tilde{p}_0(x).$$

## Wrapping Effect: Parallelepiped Method

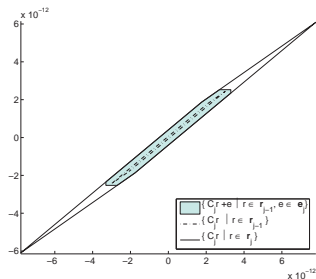
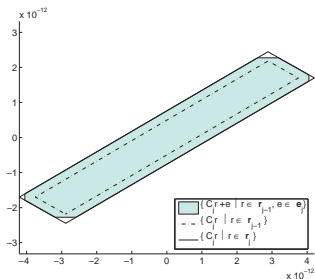
$$\mathbf{r}_j = (B_j^{-1} T B_{j-1}) \mathbf{r}_{j-1} + B_j^{-1} (\mathbf{z}_j - \mathbf{m}(\mathbf{z}_j)), \quad B_j = T B_{j-1} :$$

$$\mathbf{r}_j = \mathbf{r}_{j-1} + T^{-j} (\mathbf{z}_j - \mathbf{m}(\mathbf{z}_j)).$$

- Optimal coordinates for global error.
- Suitable coordinates for local error, if  $\text{cond}(T^j)$  is small.
- Bad coordinates for local error in presence of shear ( $T^j$  becomes singular for  $j \rightarrow \infty$ ).

# Wrapping Effect: Parallelepiped Method

$$B_j = TB_{j-1}$$



$B_j$  often ill-conditioned, large overestimations.

## Wrapping Effect: QR Method

Linear autonomous system ( $A \in \mathbb{R}^{m \times m}$ ):

$$u' = A u, \quad u(0) \in \mathbf{u}_0 = \mathcal{U}_0.$$

QR method ( $\mathbf{z}_j$ : local error,  $\mathbf{r}_0 := \mathbf{u}_0 - m(\mathbf{u}_0)$ ):

$$\mathbf{r}_j = R_j \mathbf{r}_{j-1} + Q_j^T (\mathbf{z}_j - m(\mathbf{z}_j)), \quad j = 1, 2, \dots$$

- Suitable, but not optimal coordinates for global error.
- Kühn: Numerical example for exponential overestimation.
- Good coordinates for local error.
- Handles rotation, contraction, shear.

# Preconditioned Taylor Model Method

Linear autonomous system ( $A \in \mathbb{R}^{m \times m}$ ):

$$u' = A u, \quad u(0) \in \mathbf{u}_0 = \mathcal{U}_0.$$

Initial set:  $p_{l,0}(x) = c_0 + C_0 x, \quad p_{r,0}(x) = x, \quad \mathbf{i}_{l,0} = \mathbf{i}_{r,0} = 0.$

$j$ th initial set:  $\mathcal{U}_j = (c_{l,j} + C_{l,j} x + \mathbf{i}_{l,j}) \circ (c_{r,j} + C_{r,j} x + \mathbf{i}_{r,j}),$

$$c_{l,j}, c_{r,j} \in \mathbb{R}^m, \quad C_{l,j}, C_{r,j} \in \mathbb{R}^{m \times m}.$$

Integrated flow:  $\tilde{\mathcal{U}}_j := (T C_{l,j} + T C_{l,j} x + \mathbf{i}_{l,j+1}) \circ (p_{r,j} + \mathbf{i}_{r,j}).$

## Preconditioned Taylor Model Method

$$\begin{aligned}
 C_{l,j+1} \text{ nonsingular: } \quad \tilde{\mathcal{U}}_j &= (TC_{l,j} + C_{l,j+1}x + [0,0]) \\
 &\circ \left\{ C_{l,j+1}^{-1} TC_{l,j} c_{r,j} + C_{l,j+1}^{-1} TC_{l,j} C_{r,j} x + C_{l,j+1}^{-1} TC_{l,j} \mathbf{i}_{r,j} + C_{l,j+1}^{-1} \mathbf{i}_{l,j+1} \right\} \\
 &=: (c_{l,j+1} + C_{l,j+1}x + [0,0]) \circ (c_{r,j+1} + C_{r,j+1}x + \mathbf{i}_{r,j+1}) =: \mathcal{U}_{j+1}
 \end{aligned}$$

Global error:

$$\mathbf{i}_{r,j+1} := C_{l,j+1}^{-1} TC_{l,j} \mathbf{i}_{r,j} + C_{l,j+1}^{-1} \mathbf{i}_{l,j+1}, \quad j = 0, 1, \dots$$

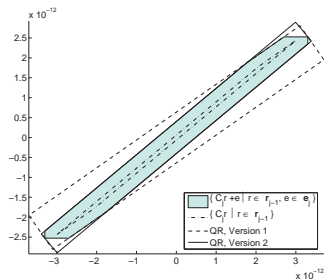
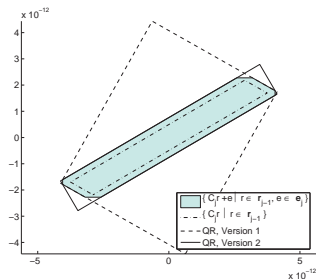
$C_{l,j+1} = TC_{l,j}$ : parallelepiped preconditioning

$C_{l,j+1} = Q_j$ : QR preconditioning

Other choices: curvilinear coordinates, blunting  
 (Makino and Berz 2004)

# Wrapping Effect: QR Method

$$B_j = Q_j, \quad Q_j R_j = T B_{j-1}$$



Overestimation depends on column permutations of  $B_{j-1}$ .

## Example 1: Pure Contraction

$$A = \begin{pmatrix} -0.4375 & 0.0625 & -0.2652 \\ 0.0625 & -0.4375 & -0.2652 \\ -0.2652 & -0.2652 & -0.375 \end{pmatrix} \approx \begin{pmatrix} -\frac{1}{2} & 0 & 0 \\ 0 & -\frac{3}{4} & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Method	Steps	$y_1(100)$	
AWA iv	216	$1.4_{5593}^{7301}E-001$	✓
AWA pe	131	abort at $t = 52.6$	fail
AWA QR	216	$1.4_{5593}^{7301}E-001$	✓
TM na	391	$[-2.4E+26, 2.4E+26]$	fail
TM sw	272	$[-2.3E+112, 2.3E+112]$	fail
TM QR	122	$1.4_{5593}^{7301}E-001$	✓

$$(\mathbf{u}_0 = [0.999, 1.001] \cdot (1 \ 1 \ 1)^T)$$

## Example 2: Pure Rotation

$$A = \begin{pmatrix} 0 & -0.7071 & -0.5 \\ 0.7071 & 0 & 0.5 \\ 0.5 & -0.5 & 0 \end{pmatrix} \approx \begin{pmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Method	Steps	$y_1(100)$	
AWA iv	393	abort at $t = 76.5$	fail
AWA pe	449	$1.49_{222}^{522}E+000$	✓
AWA QR	449	$1.49_{222}^{522}E+000$	✓
TM na	396	$[-1.6E+45, 1.6E+45]$	fail
TM sw	343	$1.49_{222}^{522}E+000$	✓
TM QR	343	$1.49_{222}^{522}E+000$	✓

## Example 3: Contraction and Rotation

$$A = \begin{pmatrix} -0.125 & -0.8321 & -0.3232 \\ 0.5821 & -0.125 & 0.6768 \\ 0.6768 & -0.3232 & -0.25 \end{pmatrix} \approx \begin{pmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & -\frac{1}{2} \end{pmatrix}$$

Method	Steps	$y_1(100)$	
AWA iv	507	abort at $t = 85.5$	fail
AWA pe	404	abort at $t = 75.2$	fail
AWA QR	516	$1.34_{592}^{862}E+000$	✓
TM na	397	$[-1.7E+55, 1.7E+55]$	fail
TM sw	357	$[-3.6E+106, 3.6E+106]$	fail
TM QR	362	$1.34_{592}^{862}E+000$	✓

# Summary

- Verified integration methods
- Interval methods vs. TM methods
- Performance for linear ODEs
- Future work: Analysis of TM methods for nonlinear ODEs

Thank you.

Questions or Remarks?