

Interval Methods and Taylor Model Methods for ODEs

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- Also called: Verified Methods, Rigorous Methods, Guaranteed Methods, Enclosure Methods, . . .
- Aim: Compute guaranteed bounds for the solution of a problem, including
 - Discretization errors (ODEs, PDEs, optimization)
 - Truncation errors (Newton's method, summation)
 - Roundoff errors
- Used for
 - Modelling of uncertain data
 - Bounding of roundoff errors

Outline

- 1 Interval arithmetic
- 2 Interval methods for ODEs
- 3 Taylor model methods for ODEs

Interval Arithmetic

Compact real intervals:

$$\mathbb{IR} = \{\mathbf{x} = [\underline{x}, \bar{x}] \mid \underline{x} \leq \bar{x}\} \quad (\underline{x}, \bar{x} \in \mathbb{R}).$$

Basic arithmetic operations:

$$\mathbf{x} \circ \mathbf{y} := \{x \circ y \mid x \in \mathbf{x}, y \in \mathbf{y}\}, \quad \circ \in \{+, -, *, /\} \quad (0 \notin \mathbf{y} \text{ for } /)$$

$$\mathbf{x} + \mathbf{y} = [\underline{x} + \underline{y}, \bar{x} + \bar{y}]$$

$$\mathbf{x} - \mathbf{y} = [\underline{x} - \bar{y}, \bar{x} - \underline{y}]$$

$$\mathbf{x} * \mathbf{y} = [\min\{\underline{x}\underline{y}, \underline{x}\bar{y}, \bar{x}\underline{y}, \bar{x}\bar{y}\}, \max\{\underline{x}\underline{y}, \underline{x}\bar{y}, \bar{x}\underline{y}, \bar{x}\bar{y}\}],$$

$$\mathbf{x} / \mathbf{y} = \mathbf{x} * [1 / \bar{y}, 1 / \underline{y}]$$

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$$\mathbf{x} + \mathbf{y} = [\underline{x} + \underline{y}, \bar{x} + \bar{y}] \quad \text{FPIA: } [\underline{x} \nabla \underline{y}, \bar{x} \Delta \bar{y}]$$

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Ranges and Inclusion Functions

- Range of $f : D \rightarrow E$: $\text{Rg}(f, D) := \{f(x) \mid x \in D\}$
- Inclusion function $F : \mathbb{IR} \rightarrow \mathbb{IR}$ of $f : D \subseteq \mathbb{R} \rightarrow \mathbb{R}$:

$$F(\mathbf{x}) \supseteq \text{Rg}(f, \mathbf{x}) \quad \text{for all } \mathbf{x} \subseteq D$$

- Examples:

- $\frac{x}{1+x}$, $1 - \frac{1}{1+x}$, are inclusion functions for

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

- $e^{\mathbf{x}} := [e^x, e^{\bar{x}}]$ is an inclusion function for e^x

IA: Dependency

- $f(\mathbf{x}) = \frac{x}{1+x} = 1 - \frac{1}{1+x}, \mathbf{x} = [1, 2]:$

- $\frac{\mathbf{x}}{1+\mathbf{x}} = \frac{[1, 2]}{[2, 3]} = [\frac{1}{3}, 1]$

- $1 - \frac{1}{1+\mathbf{x}} = 1 - \frac{1}{[2, 3]} = 1 - [\frac{1}{3}, \frac{1}{2}] = [\frac{1}{2}, \frac{2}{3}] = \text{Rg}(f, \mathbf{x})$

- Reduced overestimation: centered forms, etc.

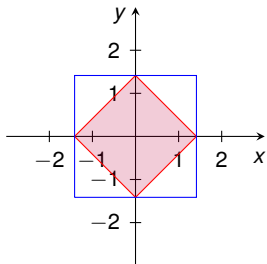
Mean value form: $\text{Rg}(f, \mathbf{x}) \subseteq f(\mathbf{c}) + F'(\mathbf{x})(\mathbf{x} - \mathbf{c}), \mathbf{c} = m(\mathbf{x}).$

IA: Wrapping Effect

Overestimation: Enclose non-interval shaped sets by intervals

Example: $f : (x, y) \rightarrow \frac{\sqrt{2}}{2}(x + y, y - x)$ (Rotation)

Interval evaluation of f on $\mathbf{x} = ([-1, 1], [-1, 1])$:

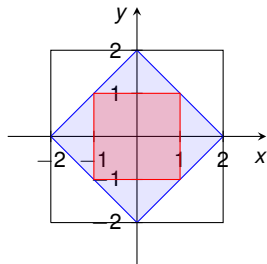
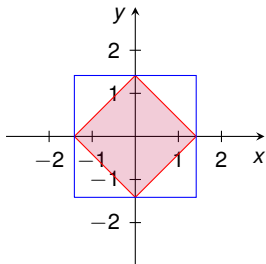


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Interval Methods for ODEs

Interval IVP:

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

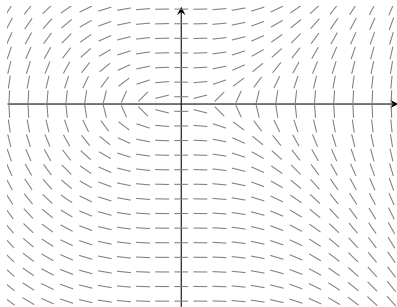
$f : \mathbb{R} \times \mathbb{R}^m \rightarrow \mathbb{R}^m$ sufficiently smooth, $\mathbf{u}_0 \in \mathbb{I}\mathbb{R}^m$, $t_{\text{end}} > t_0$.

Validated Integration of ODEs

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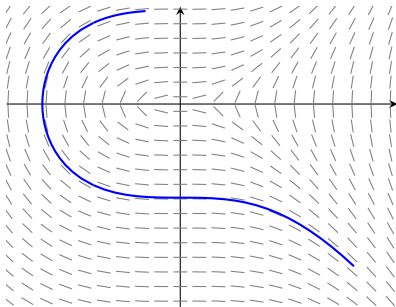


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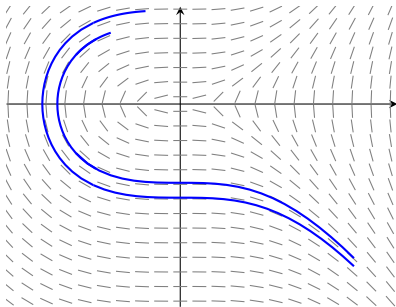


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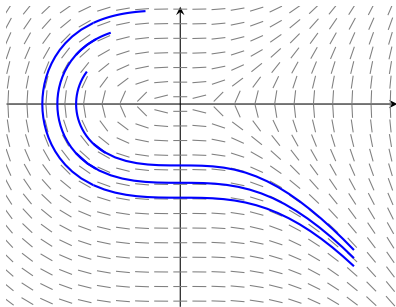


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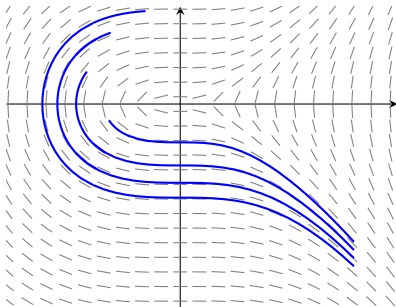


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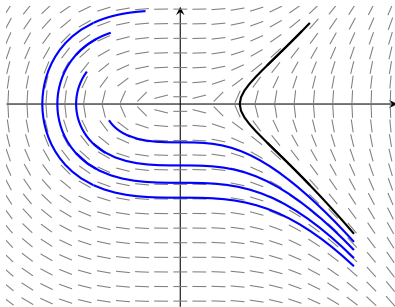


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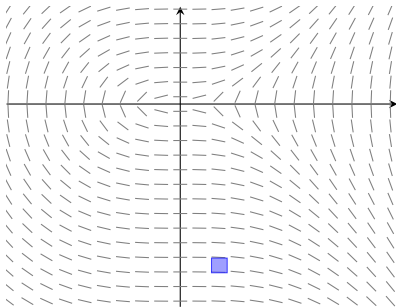


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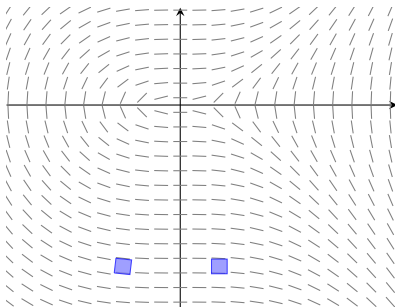


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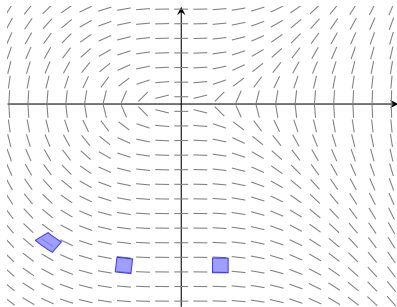


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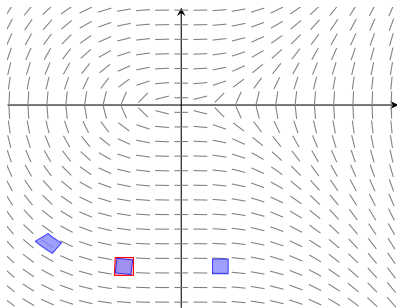


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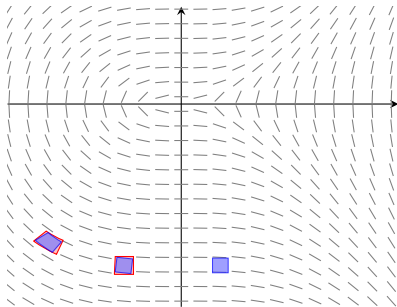


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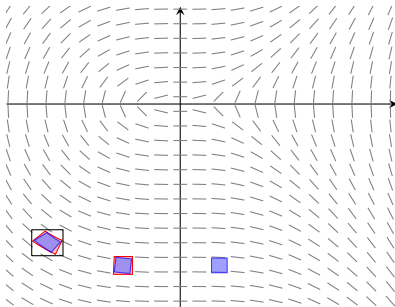


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$$u' = f(u), \quad u(t_0) = u_0 \in \mathbf{u}_0, \quad t \in \mathbf{t} = [t_0, t_{\text{end}}],$$

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:

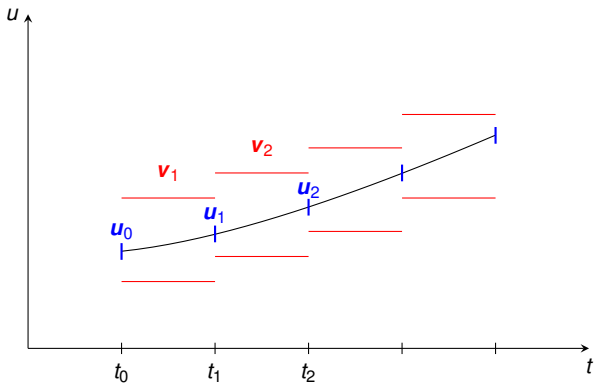
- Automatic computation of Taylor coefficients
- Interval iteration: For $j = 1, 2, \dots$:

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



- 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
- Improvements: Lohner 1988, Corliss & Rihm 1996, Makino 1998, Nedialkov & Jackson 2001
- Alternatives: Neumaier 1994, N. 1999, N. 2007, Delanoue & Jaulin 2010, Kin, Kim & Nakao 2011

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 of solution w.r.t. time and initial values

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

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

Let

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

(I : identity matrix, $J(f^{[k]})$: Jacobian of $f^{[k]}$), then for some $\hat{u}_{j-1} \in \mathbf{u}_{j-1}$

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

Global Error Propagation

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

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

→ propagate \mathcal{S} as a parallelepiped.

$\hat{\mathbf{u}}_0 := \mathbf{m}(\mathbf{u}_0)$, $\mathbf{r}_0 = \mathbf{u}_0 - \hat{\mathbf{u}}_0$, $\mathbf{B}_0 = \mathbf{I}$; for some nonsingular \mathbf{B}_{j-1} :

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

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

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

Global error 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): $m(\mathbf{S}_{j-1} B_{j-1}) = QR$, $B_j := Q$
- Blunting method (Berz, Makino): modify B_j in the pep method such that condition numbers remain small

Autonomous Linear ODE

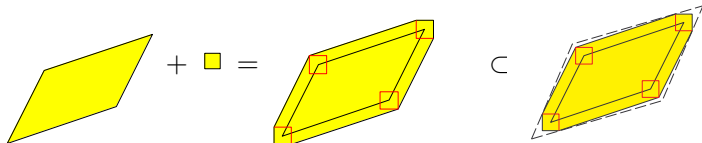
Autonomous linear system ($A \in \mathbb{R}^{m \times m}$):

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

Propagation of the Global Error

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

$$T = \sum_{v=0}^{n-1} \frac{(hA)^v}{v!}.$$



Interval Methods for Linear ODEs

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

- Optimal for local error, bad for global error (rotation)

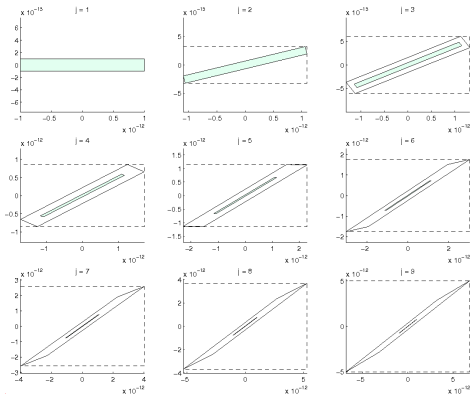
Parallelepiped method: $\mathbf{r}_j = \mathbf{r}_{j-1} + (T^{-j})(\mathbf{z}_j - m(\mathbf{z}_j))$.

- Optimal for global error; suitable for local error, if $\text{cond}(T^j)$ is small
- Bad for local error in presence of shear

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

- Handles rotation, contraction, shear

Wrapping Effect: Direct Interval Method

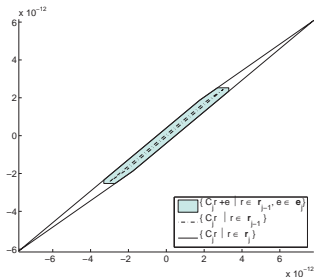
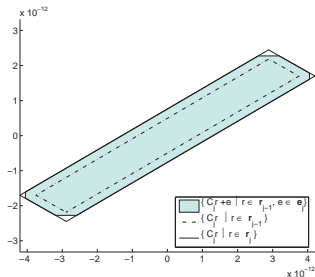


(Plots by Ned Nedialkov)

Huge overestimations in general

Wrapping Effect: Parallelepiped Method

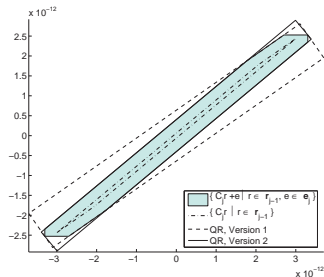
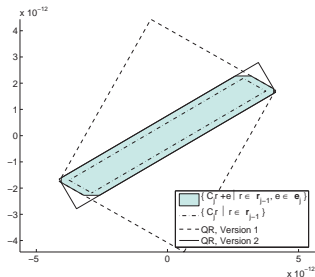
$$B_j = TB_{j-1}$$



B_j often ill-conditioned, large overestimations

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}

Taylor Model Methods for ODEs

Taylor Models (I)

- $\mathbf{x} \subset \mathbb{R}^m$, $f : \mathbf{x} \rightarrow \mathbb{R}$, $f \in \mathcal{C}^{n+1}$, $x_0 \in \mathbf{x}$;

$$f(x) = p_{n,f}(x - x_0) + R_{n,f}(x - x_0), \quad x \in \mathbf{x}$$

($p_{n,f}$ Taylor polynomial, $R_{n,f}$ remainder term)

- **Interval remainder bound** of order n of f on \mathbf{x} :

$$\forall x \in \mathbf{x} : R_{n,f}(x - x_0) \in \mathbf{i}_{n,f}$$

- **Taylor model** $T_{n,f} = (p_{n,f}, \mathbf{i}_{n,f})$ of order n of f :

$$\forall x \in \mathbf{x} : f(x) \in p_{n,f}(x - x_0) + \mathbf{i}_{n,f}$$

Taylor Models (II)

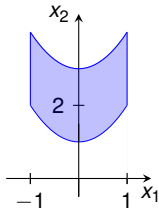
Taylor model: $\mathcal{U} := p_n(x) + \mathbf{i}, \quad x \in \mathbf{x}, \mathbf{x} \in \mathbb{IR}^m, \mathbf{i} \in \mathbb{IR}^m$
(p_n : vector of m -variate polynomials of order n)

Function set: $\mathcal{U} = \{f \in \mathcal{C}^0(\mathbf{x}) : f(x) \in p_n(x) + \mathbf{i} \text{ for all } x \in \mathbf{x}\}$

Range of a TM: $\text{Rg}(\mathcal{U}) = \{z = p(x) + \zeta \mid x \in \mathbf{x}, \zeta \in \mathbf{i}\} \subset \mathbb{IR}^m$

Ex.: $\mathcal{U} := \left(\begin{array}{c} x_1 \\ 2 + x_1^2 + x_2 \end{array} \right), \quad x_1, x_2 \in [-1, 1]$

$\text{Rg}(\mathcal{U})$:



Paradigm for TMA:

- $p_{n,f}$ is processed symbolically to order n
- Higher order terms are enclosed into the remainder interval of the result

Taylor Model Methods for ODEs

- Taylor expansion of solution w.r.t. time and initial values
- Computation of Taylor coefficients by Picard iteration:
Parameters describing initial set treated symbolically
- Interval remainder bounds by fixed point iteration (Makino, 1998)

Example: Quadratic Problem

$$u' = v, \quad u(0) \in [0.95, 1.05],$$

$$v' = u^2, \quad v(0) \in [-1.05, -0.95].$$

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

$$u_0(a, b) := 1 + a, \quad a \in \mathbf{a} := [-0.05, 0.05],$$

$$v_0(a, b) := -1 + b, \quad b \in \mathbf{b} := [-0.05, 0.05].$$

3rd order TM Method: Enclosure of the Flow

$h = 0.1$, flow for $\tau \in [0, 0.1]$:

$$\tilde{\mathcal{U}}_1(\tau, a, b) := 1 + a - \tau + b\tau + \frac{1}{2}\tau^2 + a\tau^2 - \frac{1}{3}\tau^3 + \mathbf{i}_0,$$

$$\tilde{\mathcal{V}}_1(\tau, a, b) := -1 + b + \tau + 2a\tau - \tau^2 + a^2\tau - a\tau^2 + b\tau^2 + \frac{2}{3}\tau^3 + \mathbf{j}_0.$$

Flow at $t_1 = 0.1$:

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

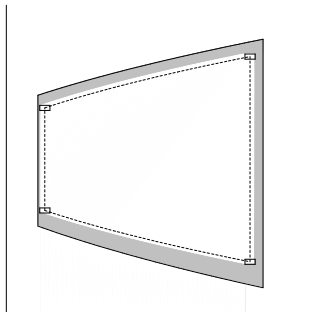
$$\mathcal{V}_1(a, b) := \tilde{\mathcal{V}}_1(0.1, a, b) = -0.909 + 0.19a + 1.01b + 0.1a^2 + \mathbf{j}_0$$

(nonlinear boundary).

Naive TM Method

- Interval remainder terms accumulate
- Linear ODEs:
Naive TM method performs similarly to the direct interval method
- → Shrink wrapping, preconditioned TM methods

Absorb interval term into polynomial part (Makino and Berz 2002):



$$\begin{pmatrix} u \\ v \end{pmatrix} \text{ (white) vs. } \begin{pmatrix} u_{sw} \\ v_{sw} \end{pmatrix}.$$

Linear ODEs: Shrink wrapping performs similarly to the pep method.

Preconditioned integration: flow at t_j :

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

Purpose: stabilize integration as in the 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.

Global error:

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

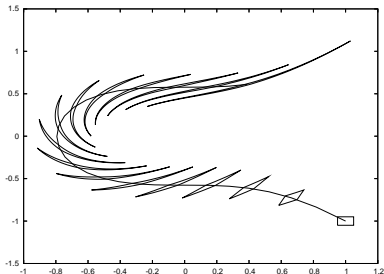
$C_{l,j+1} = TC_{l,j}$: parallelepiped preconditioning

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

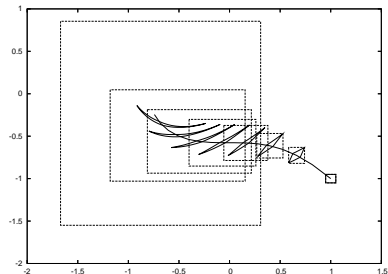
Other choices: curvilinear coordinates, blunting
(Makino and Berz 2004)

Integration of Quadratic Problem

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



COSY Infinity



AWA

Summary

- Interval arithmetic
- Interval methods for ODEs
- Taylor model methods for ODEs

Open problems

- Treatment of high-dimensional problems
- Validated implicit methods