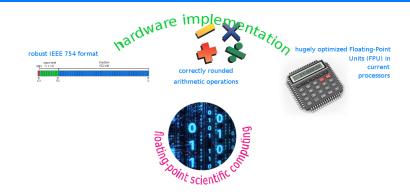






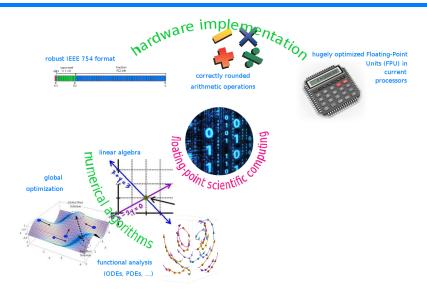
# Power of Floating-Point Scientific Computing...





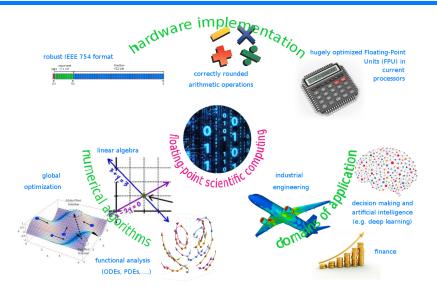
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## rounding errors











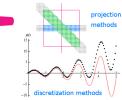
### rounding errors



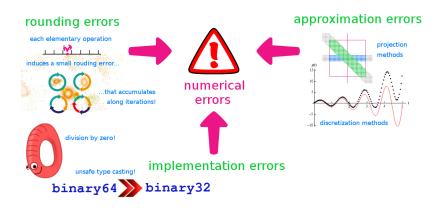




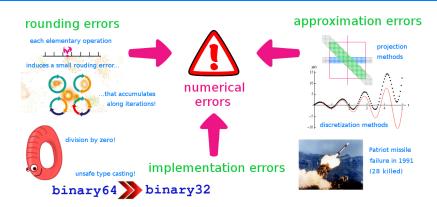




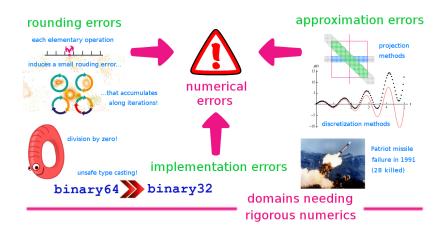




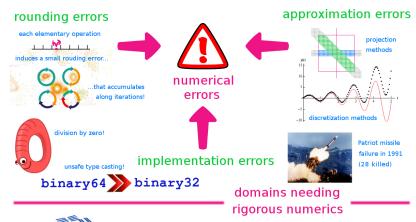








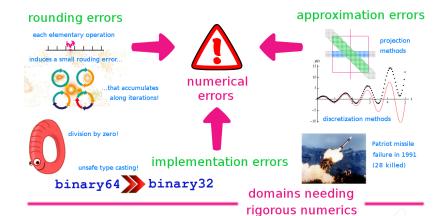






safety-critical engineering







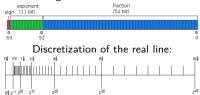
safety-critical engineering

computer-assisted mathematics





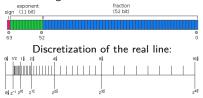
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$$x = (-1)^s \cdot 1. \underbrace{1010011100...1010}_{m} \cdot 2^e$$



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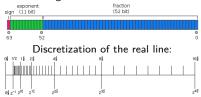


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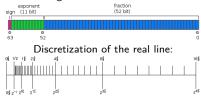
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Overapprox reals by intervals

$$\pi \in [3.14, 3.15]$$
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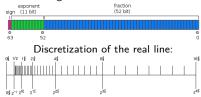
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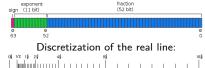
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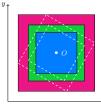
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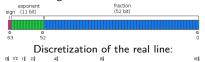
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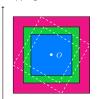
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Loss of correlation

- [-x,x]-[-x,x] $= [-2x, 2x] \neq [0, 0].$
- $\cos([0,2\pi]+\varepsilon)-\cos([0,2\pi])$ = [-1,1] - [-1,1] = [-2,2]but  $|\cos(x+\varepsilon)-\cos(x)| \le \varepsilon$ .





2 Rigorous Polynomial Approximations

3 A Posteriori Validation with Fixed-Points

4 Validated Solutions of Linear Differential Equations

5 Conclusion and Future Work

6 Some Extras



## Outline

- 1 Introduction
- 2 Rigorous Polynomial Approximations
- 3 A Posteriori Validation with Fixed-Points
- 4 Validated Solutions of Linear Differential Equations
- 5 Conclusion and Future Work
- 6 Some Extra



■ a class  $\mathcal{F}$  of functions, a reference norm  $\|\cdot\|$ , and a *computable* family  $\mathcal{P} = (P_n)$  to approximate them.

### Theorem (Stone Weierstrass)

The family of polynomials is dense in the set of continuous functions over a compact interval.



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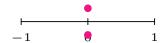
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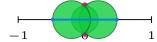
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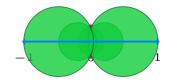
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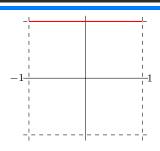




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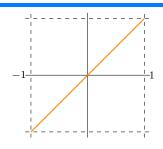
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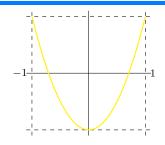
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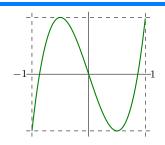
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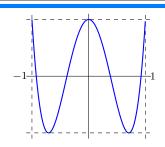
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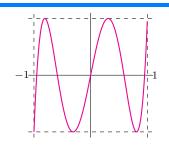
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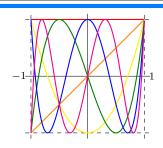
#### Chebyshev Family of Polynomials

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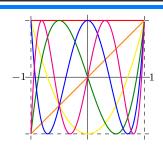
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### **Convergence Theorems**

- If  $f \in C^k$ ,  $\widehat{f}^{[N]} \to f$  in  $O(N^{-k})$ .
- If f analytic,  $\widehat{f}^{[N]} \to f$  exponentially fast.



#### Definition

A pair  $(P,\varepsilon)\in\mathbb{R}[X]\times\mathbb{R}_+$  is a rigorous polynomial approximation (RPA) of f for a given norm  $\|\cdot\|$  if  $\|f-P\|\leq \varepsilon$ .

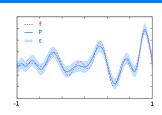


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$$f \in (P, \varepsilon) \Leftrightarrow |f(t) - P(t)| \le \varepsilon \quad \forall t \in [-1, 1]$$



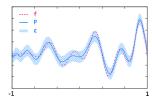


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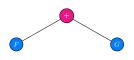


## Example:

$$r(t) = f(t) + g(t)$$

## Some elementary operations:

$$(P,\varepsilon) + (Q,\eta) \coloneqq (P+Q,\varepsilon+\eta),$$





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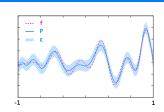
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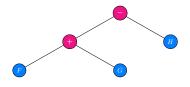
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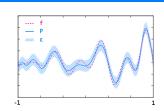
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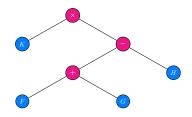
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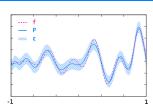
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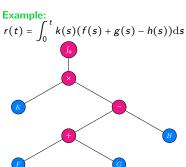
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  \text{if } \| \cdot \| = \| \cdot \|_{\infty, \lceil -1, 1 \rceil}.$







#### Definition

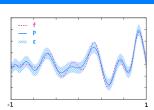
A pair  $(P,\varepsilon)\in\mathbb{R}[X]\times\mathbb{R}_+$  is a rigorous polynomial approximation (RPA) of f for a given norm  $\|\cdot\|$  if  $\|f-P\|\leq\varepsilon$ .

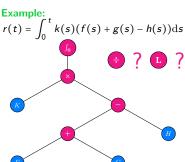
Example: sup-norm over [-1,1]:

$$f \in (P, \varepsilon) \Leftrightarrow |f(t) - P(t)| \le \varepsilon \quad \forall t \in [-1, 1]$$

#### Some elementary operations:

- $(P,\varepsilon) + (Q,\eta) := (P + Q,\varepsilon + \eta),$
- $(P,\varepsilon)-(Q,\eta)\coloneqq (P-Q,\varepsilon+\eta),$
- $(P,\varepsilon)\cdot (Q,\eta)\coloneqq (PQ,\|Q\|\eta+\|P\|\varepsilon+\eta\varepsilon)$  provided that  $\|fg\|\leq \|f\|\|g\|$ ,
- $\int_0 (P, \varepsilon) := \left( \int_0^t P(s) ds, \varepsilon \right) \\
  \text{if } \| \cdot \| = \| \cdot \|_{\infty, \lceil -1, 1 \rceil}.$







# Outline

- 1 Introduction
- 2 Rigorous Polynomial Approximations
- 3 A Posteriori Validation with Fixed-Points
- 4 Validated Solutions of Linear Differential Equations
- 5 Conclusion and Future Work
- 6 Some Extras

Banach Fixed-Point Theorem



### Main Idea: A Posteriori Validation

Reformulate the problem as a fixed-point equation  $\mathbf{T} \cdot x = x$  over metric space (X, d) and obtain x candidate approximation of exact solution  $x^*$ .

▶ Find **rigorous** error bound  $||x - x^*||$ .



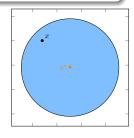
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#### Banach Fixed-Point Theorem

If (X, d) is complete and T contracting of ratio  $\mu < 1$ ,

▶ Then T admits a unique fixed-point  $x^*$ ,





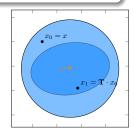
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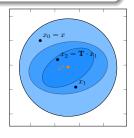
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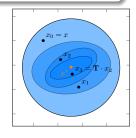
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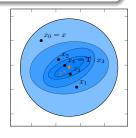
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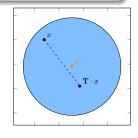
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#### Banach Fixed-Point Theorem

If (X, d) is complete and T contracting of ratio  $\mu < 1$ ,

- ▶ Then T admits a unique fixed-point  $x^*$ , and
- ▶ For all  $x \in X$ ,

$$\frac{d(x,\mathbf{T}\cdot x)}{1+\mu} \leq d(x,x^*) \leq \frac{d(x,\mathbf{T}\cdot x)}{1-\mu}.$$





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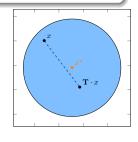
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### Quasi-Newton Method for $\mathbf{F} \cdot \mathbf{x} = \mathbf{0}$

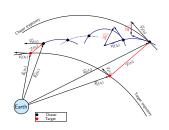
Obtain  $\mathbf{A} \approx (D\mathbf{F})^{-1}$  in order to define:

$$\mathbf{T} \cdot \mathbf{x} = \mathbf{x} - \mathbf{A} \cdot \mathbf{F} \cdot \mathbf{x}$$

Relative Motion in Keplerian Dynamics



$$z'' + \left(4 - \frac{3}{1 + e\cos\nu}\right)z = c$$

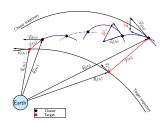


Relative Motion in Keplerian Dynamics



## Reduced Equation

$$z'' + \left(4 - \frac{3}{1 + e\cos\nu}\right)z = c$$



#### To Do List

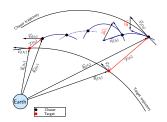
Approximate coefficient

Relative Motion in Keplerian Dynamics



## Reduced Equation

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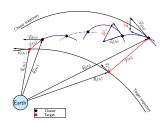
- Approximate coefficient
- 2 Approximate solution with a Chebyshev series

Relative Motion in Keplerian Dynamics



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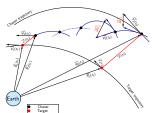
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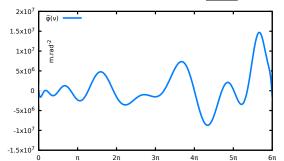
- Approximate coefficient
- Approximate solution with a Chebyshev series
- Validate the obtained solution

Relative Motion in Keplerian Dynamics



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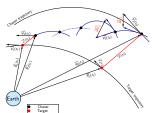


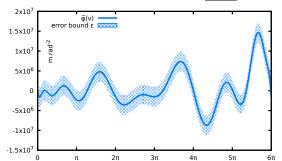


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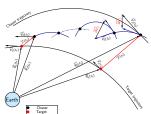


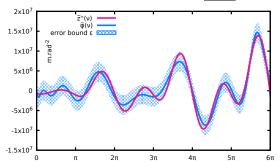


Relative Motion in Keplerian Dynamics



$$z'' + \left(4 - \frac{3}{1 + e\cos\nu}\right)z = c$$







**Approximation of**  $x \mapsto 4 - \frac{3}{1 + e \cos x}$ 

$$\checkmark$$
 RPA for  $x \mapsto \cos x$ :

$$0.77 T_0(x) - 0.23 T_2(x) + 0.005 T_4(x) \pm 4.2 \cdot 10^{-5}$$

# ${\sf Fixed-Point\ Based\ Validation}$

## Application to Division



# **Approximation of** $x \mapsto 4 - \frac{3}{1 + e \cos x}$

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 RPA for  $x \mapsto \cos x$ :

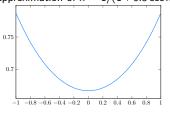
$$0.77T_0(x) - 0.23T_2(x) + 0.005T_4(x) \pm 4.2 \cdot 10^{-5}$$

$$\checkmark$$
 RPA for  $x \mapsto 1 + 0.5 \cos x$ :

$$1.38\,T_0(x) - 0.11\,T_2(x) + 0.002\,T_4(x) \pm 2.1 \cdot 10^{-5}$$



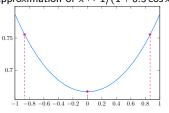
**Approximation of** 
$$x \mapsto 4 - \frac{3}{1 + e \cos x}$$



Application to Division



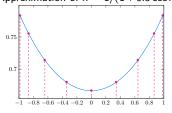
**Approximation of**  $x \mapsto 4 - \frac{3}{1 + e \cos x}$ 



Application to Division

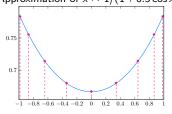


# **Approximation of** $x \mapsto 4 - \frac{3}{1 + e \cos x}$





# **Approximation of** $x \mapsto 4 - \frac{3}{1 + e \cos x}$



$$\varphi = 0.73 T_0(x) + 0.06 T_2(x) \approx 1/(1+0.5\cos x)$$

#### Application to Division



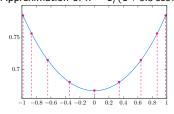
÷

Division: g/f ( $f \neq 0$ )

► Solve 
$$\mathbf{F} \cdot \boldsymbol{\varphi} = \boldsymbol{f} \boldsymbol{\varphi} - \boldsymbol{g} = 0$$

$$(\mathbf{DF})_{\varphi} \cdot h = fh$$
  $(\mathbf{DF})_{\varphi}^{-1} \cdot h = f^{-1}h$ 

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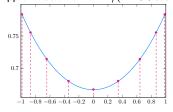
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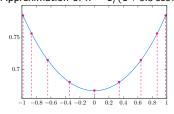
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▶ **T** affine,  $\mu = \|D\mathbf{T}\| = \|1 - f_0 f\| < 1$ ?

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### Fixed-Point Based Validation

#### Application to Division



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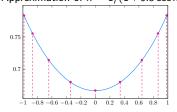
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# **Approximation of** $x \mapsto 4 - \frac{3}{1 + e \cos x}$

✓ Approximation of  $x \mapsto 1/(1 + 0.5 \cos x)$ :



$$\varphi = 0.73 T_0(x) + 0.06 T_2(x) \approx 1/(1+0.5\cos x)$$

$$\mu = 3.7 \cdot 10^{-3} << 1$$

# ${\sf Fixed-Point\ Based\ Validation}$

### Application to Division



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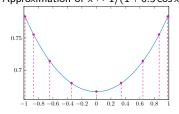
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 $\sqrt{\text{RPA for } x \mapsto 1/(1 + 0.5 \cos x)}$ :

$$0.73T_0(x) + 0.06T_2(x) \pm 1.3 \cdot 10^{-3}$$

# ${\sf Fixed-Point\ Based\ Validation}$

### Application to Division



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### Division: g/f ( $f \neq 0$ )

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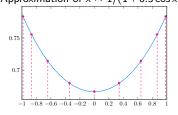
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✓ RPA for 
$$x \mapsto 1/(1 + 0.5 \cos x)$$
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$$0.73T_0(x) + 0.06T_2(x) \pm 1.3 \cdot 10^{-3}$$

$$\checkmark$$
 RPA for  $x \mapsto 4 - 3/(1 + 0.5 \cos x)$ :

$$1.82T_0(x) - 0.18T_2(x) \pm 3.8 \cdot 10^{-3}$$



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### LODE and Initial Value Problem

$$y^{(r)}(t) + \alpha_{r-1}(t)y^{(r-1)}(t) + \dots + \alpha_1(t)y'(t) + \alpha_0(t)y(t) = g(t)$$

$$y(-1) = v_0 \qquad y'(-1) = v_1 \qquad \dots \qquad y^{(r-1)}(-1) = v_{r-1}$$
(D)

 $t \in [-1,1]$   $\alpha_i, g$  sufficiently regular ( $\mathcal{C}^0$ , RPA, polynomial)



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### Integral Reformulation

$$\varphi + \mathbf{K} \cdot \varphi = \psi, \tag{I}$$



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### LODE and Initial Value Problem

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### LODE and Initial Value Problem

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- $\psi(t) = g(t) + \text{(some function depending on the } v_j\text{'s)}$



### LODE and Initial Value Problem

$$y^{(r)}(t) + \alpha_{r-1}(t)y^{(r-1)}(t) + \dots + \alpha_1(t)y'(t) + \alpha_0(t)y(t) = g(t)$$

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 $t \in [-1,1]$   $\alpha_i, g$  sufficiently regular ( $\mathcal{C}^0$ , RPA, polynomial)

#### Integral Reformulation

Let  $\varphi = y^{(r)}$ , (D) becomes:

$$\varphi + \mathbf{K} \cdot \varphi = \psi, \tag{I}$$

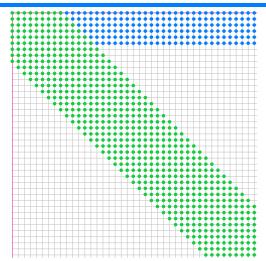
- $\mathbf{K} \cdot \varphi(t) = \sum_{j=0}^{r-1} \beta_j(t) \int_{-1}^t T_j(s) \varphi(s) ds \Rightarrow \text{compact operator}$
- $\psi(t) = g(t) + \text{(some function depending on the } v_j\text{'s)}$

#### Theorem (Picard-Lindelöf)

(I) (and hence (D)) has a unique solution.

Matrix Representation in Chebyshev Basis

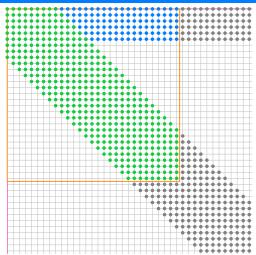




The infinite-dimensional operator K.

Matrix Representation in Chebyshev Basis





The finite-dimensional truncation  $\mathbf{K}^{[\mathit{N}]}.$ 

Example with Tschauner-Hempel Equation



$$\mathbf{K} \cdot \varphi = t \left( 4 - \frac{3}{1 + e \cos t} \right) \int_{t_0}^t \varphi(s) \mathrm{d}s + \left( -4 + \frac{3}{1 + e \cos t} \right) \int_{t_0}^t s \varphi(s) \mathrm{d}s$$

Example with Tschauner-Hempel Equation



$$\mathbf{K} \cdot \boldsymbol{\varphi} \approx t \big(1.82 - 0.18\,T_2(t)\big) \int_{t_0}^t \boldsymbol{\varphi}(s) \mathrm{d}s + \big(-1.82 + 0.18\,T_2(t)\big) \int_{t_0}^t s \boldsymbol{\varphi}(s) \mathrm{d}s$$



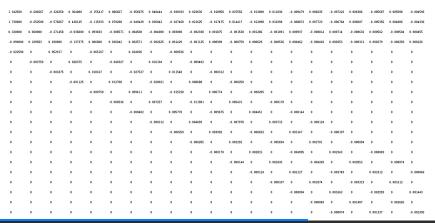


$$\mathbf{K} \cdot \varphi \approx \underbrace{\left(1.73 T_1(t) - 0.09 T_3(t)\right)}_{\beta_0(t)} \int_{t_0}^t \varphi(s) \mathrm{d}s + \underbrace{\left(-1.82 + 0.18 T_2(t)\right)}_{\beta_1(t)} \int_{t_0}^t s \varphi(s) \mathrm{d}s$$





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$$\begin{split} \varphi &= -0.6T_0 - 1.19T_1 + 0.62T_2 + 0.17T_3 - 0.05T_4 - 0.01T_5 \\ &+ 2.1 \cdot 10^{-3}T_6 + 3.2 \cdot 10^{-3}T_7 - 5.8 \cdot 10^{-5}T_8 - 7.6 \cdot 10^{-6}T_9 + 1.2 \cdot 10^{-6}T_{10} \\ &+ 1.4 \cdot 10^{-7}T_{11} - 1.9 \cdot 10^{-8}T_{12} - 2.0 \cdot 10^{-9}T_{13} + 2.6 \cdot 10^{-10}T_{14} + 2.5 \cdot 10^{-11}T_{15} \\ &- 3.0 \cdot 10^{-12}T_{16} - 2.6 \cdot 10^{-13}T_{17} + 3.0 \cdot 10^{-14}T_{18} + 2.5 \cdot 10^{-15}T_{19} - 2.6 \cdot 10^{-16}T_{20} \end{split}$$

## Designing the Newton-like Operator ${f T}$



### Construct T: To-Do List

■ Truncation order *N*.



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- Approx inverse:

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### Decomposition of the Operator Norm

$$\|\mathbf{D}\mathbf{T}\| = \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\| \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\| + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|.$$



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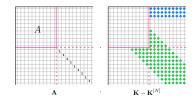
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#### Truncation error:

Determines the minimal value of N we can choose.





$$\mathbf{L} \cdot y = y^{(r)}(t) + \alpha_{r-1}(t)y^{(r-1)}(t) + \dots + \alpha_1(t)y'(t) + \alpha_0(t)y(t) = g(t)$$

$$y(-1) = v_0 \qquad y'(-1) = v_1 \qquad \dots \qquad y^{(r-1)}(-1) = v_{r-1}$$
(D)

#### Rigorous Solving - Overview

- Integral reformulation:  $\varphi + \mathbf{K} \cdot \varphi = \psi$  with  $\varphi = y^{(r)}$ ,
- **2** Numerical solving: approximation  $\varphi$  of  $\varphi^*$ ,
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- Obtaining  $\mu \geq \|D\mathbf{T}\|$ ,
- 5 If  $\mu < 1$ ,  $\|\varphi \varphi^*\| \le \varepsilon := \|\varphi \mathbf{T} \cdot \varphi\|/(1 \mu)$ ,
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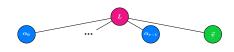


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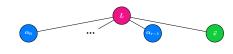


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- ► Extension to *Boundary Value Problems* (BVP)



# Solution to Tschauner and Hempel Equations

Bring our Example to the End



■ Approximation error  $\leq 1.5 \cdot 10^{-3}$ .



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- Hence:

$$\frac{6.48 \cdot 10^{-16}}{1 + \textcolor{red}{\mu}} \hspace{1cm} \leq \hspace{1cm} \| \varphi - \varphi^* \| \hspace{1cm} \leq \hspace{1cm} \frac{6.48 \cdot 10^{-16}}{1 - \textcolor{red}{\mu}}$$



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Take into account approximation error of coefficient!



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## $\label{eq:Application} \mbox{Application to Division and Square Root}$





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#### Application to Division and Square Root





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▶ **T** non-linear,  $\|(\mathbf{DT})_{\varphi}\| = \|1 - f_{0}\varphi\|$ .

$$\exists r > 0 \cdot \mathbf{T} : \overline{B}(\varphi, r) \to \overline{B}(\varphi, r)$$
?

$$\|\mathbf{T} \cdot \boldsymbol{\varphi} - \boldsymbol{\varphi}\| + r \sup_{\psi \in \overline{B}(\boldsymbol{\varphi}, r)} \|(\mathbf{D}\mathbf{T})_{\psi}\| < r$$

dessin inclusion boule image

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$$\frac{\|f_0(f\varphi-g)\|}{1+\mu} \leq \|\varphi-\frac{g}{f}\| \leq \frac{\|f_0(f\varphi-g)\|}{1-\mu}$$

dessin inclusion boule image



# **Square Root:** $\sqrt{f}$ (f > 0)

Solve  $\mathbf{F} \cdot \varphi = \varphi^2 - f = 0$  Two solutions!

$$(\mathbf{DF})_{\varphi} \cdot h = 2\varphi h$$
  $(\mathbf{DF})_{\varphi}^{-1} \cdot h = \frac{\varphi^{-1}}{2} h$ 

▶ Use  $f_0 \approx \varphi^{-1}$  (≈  $1/\sqrt{f}$ ):

$$\mathbf{T} \cdot \varphi = \varphi$$
  $\mathbf{T} \cdot \varphi = \varphi - \frac{f_0}{2} (\varphi^2 - f)$ 

▶ **T** non-linear,  $\|(\mathbf{DT})_{\varphi}\| = \|1 - f_{0}\varphi\|$ .

$$\exists r > 0 \cdot \mathbf{T} : \overline{B}(\varphi, r) \to \overline{B}(\varphi, r)$$
?

$$\|\mathbf{T} \cdot \boldsymbol{\varphi} - \boldsymbol{\varphi}\| + r \sup_{\psi \in \overline{B}(\boldsymbol{\varphi}, r)} \|(\mathbf{D}\mathbf{T})_{\psi}\| < r$$

- ▶ Check  $\Delta > 0$   $\rightarrow$   $0 < r_{min} < r_{max}$ .
- ► Check  $\mu = \|1 f_0 \varphi\| + \|f_0\|_{r_{\min}} < 1$ .

#### Application to Division and Square Root



# Division: g/f ( $f \neq 0$ )

Solve  $\mathbf{F} \cdot \boldsymbol{\varphi} = f \boldsymbol{\varphi} - \boldsymbol{\varrho} = 0$ 

$$(\mathbf{DF})_{\varphi} \cdot h = fh$$
  $(\mathbf{DF})_{\varphi}^{-1} \cdot h = f^{-1}h$ 

▶ Use  $f_0 \approx f^{-1}$ :

$$\mathbf{T} \cdot \boldsymbol{\varphi} = \boldsymbol{\varphi}$$
  $\mathbf{T} \cdot \boldsymbol{\varphi} = \boldsymbol{\varphi} - f_0 (\boldsymbol{f} \boldsymbol{\varphi} - \boldsymbol{g})$ 

▶ **T** affine,  $\mu = \|D\mathbf{T}\| = \|1 - f_0 f\| < 1$ ?

$$\frac{\|f_0(f\varphi-g)\|}{1+\mu} \leq \|\varphi-\frac{g}{f}\| \leq \frac{\|f_0(f\varphi-g)\|}{1-\mu}$$

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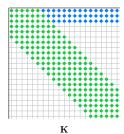
$$\frac{\|f_0(\varphi^2 - f)/2\|}{1 + \mu} \le \|\varphi - \sqrt{f}\| \le \frac{\|f_0(\varphi^2 - f)/2\|}{1 - \mu}$$



$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\|=\sup_{i\geq 0}\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot T_i\|$$

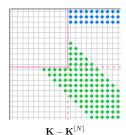


$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\|=\sup_{i\geq 0}\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot\mathcal{T}_i\|$$



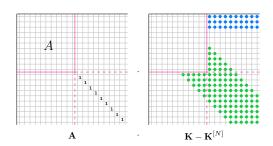


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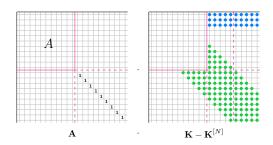


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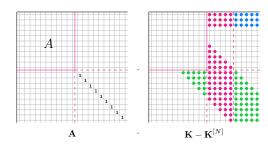
$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\| = \sup_{i\geq 0} \|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot T_i\|$$

Direct computation.



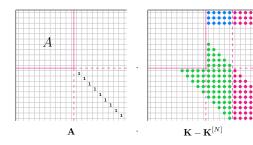
$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\| = \sup_{i\geq 0} \|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot T_i\|$$

- Direct computation.
- Direct computation.





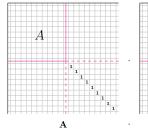
$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\| = \sup_{i\geq 0} \|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot T_i\|$$

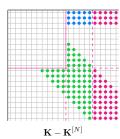


- Direct computation.
- Direct computation.
- Bound the remaining *infinite* number of columns:



$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\| = \sup_{i\geq 0} \|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot T_i\|$$



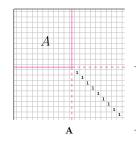


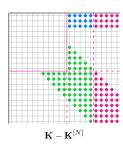
- Direct computation.
- Direct computation.
- Bound the remaining *infinite* number of columns:
  - Using the bounds in 1/i and  $1/i^2$ : possibly large overestimations.

$$diag(i) \le \frac{C}{i} \quad init(i) \le \frac{D}{i^2}$$



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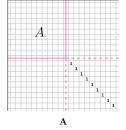
$$diag(i) \le \frac{C}{i} \quad init(i) \le \frac{D}{i^2}$$

 Using a first order difference method: differences in 1/i<sup>2</sup> and 1/i<sup>4</sup>.

$$diag(i) \le diag(i_0) + \frac{C'}{i^2}$$
  
 $init(i) \le init(i_0) + \frac{D'}{i^4}$ 



$$\|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\| = \sup_{i\geq 0} \|\mathbf{A}\cdot(\mathbf{K}-\mathbf{K}^{[N]})\cdot\mathcal{T}_i\|$$







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## Coupled LODEs and Initial Value Problem

$$Y^{(r)}(t) + A_{r-1}(t) \cdot Y^{(r-1)}(t) + \dots + A_{1}(t) \cdot Y'(t) + A_{0}(t) \cdot Y(t) = G(t) \qquad (p-D)$$

$$A_{k}(t) = \begin{pmatrix} a_{k11}(t) & \dots & a_{k1p}(t) \\ \vdots & \ddots & \vdots \\ a_{kp1}(t) & \dots & a_{kpp}(t) \end{pmatrix} \qquad G(t) = \begin{pmatrix} g_{1}(t) \\ \vdots \\ g_{p}(t) \end{pmatrix}$$

$$t \in [-1, 1] \qquad Y_{i}^{(k)}(-1) = v_{ik} \qquad i \in [1, p], k \in [0, r-1]$$



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# Integral Reformulation

Posing  $\Phi = Y^{(r)}$ , System (p-D) is transformed into:

$$\Phi(t) + \int_{t_0}^t \begin{pmatrix} k_{11}(t,s) & \cdots & k_{1p}(t,s) \\ \vdots & \ddots & \vdots \\ k_{p1}(t) & \cdots & k_{pp}(t) \end{pmatrix} \cdot \Phi(s) \mathrm{d}s = \Psi(t) \tag{p-I}$$

# The Almost-Banded Structure of the Operator ${f K}$

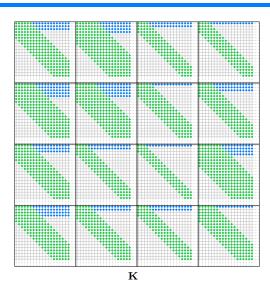




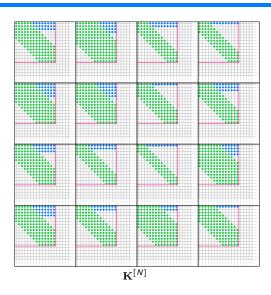
$\mathbf{K}_{1,1}$	$\mathbf{K}_{1,2}$	$\mathbf{K}_{1,3}$	$\mathbf{K}_{1,4}$
$\mathbf{K}_{2,1}$	$\mathbf{K}_{2,2}$	$\mathbf{K}_{2,3}$	$\mathbf{K}_{2,4}$
$\mathbf{K}_{3,1}$	$\mathbf{K}_{3,2}$	$\mathbf{K}_{3,3}$	$\mathbf{K}_{3,4}$
$\mathbf{K}_{4,1}$	$\mathbf{K}_{4,2}$	$\mathbf{K}_{4,3}$	$\mathbf{K}_{4,4}$

 $\mathbf{K}$ 

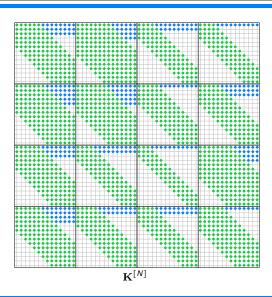




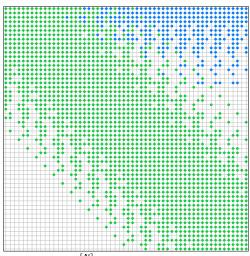












 $\mathbf{K}^{[N]}$  (rearranged basis)



 $(X_1,d_1),\ldots,(X_p,d_p)$  complete metric spaces.

- $d(x,y) = (d_1(x_1,y_1),\ldots,d_p(x_p,y_p)) \in \mathbb{R}_+^p$  vector-valued metric.
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$$\varepsilon = d(x, x^*)$$
 and  $\eta = d(x, \mathbf{T} \cdot x)$ :

$$(1 - \Lambda) \cdot \varepsilon \le \eta \tag{P}$$

$$\big(1+{\color{red}\Lambda}\big)\cdot {\color{blue}\varepsilon} \geq {\color{blue}\eta}$$

$$\varepsilon \geq 0$$



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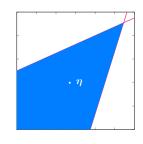
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