

**A Collocation Solver  
for  
Singular Boundary Value  
Problems**

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## Problem Class

$$z'(t) = \underbrace{\frac{M(t)}{t}z(t) + f(t, z(t))}_{=:F(t,z(t))}, \quad t \in (0, 1],$$
$$B_a z(0) + B_b z(1) = \beta.$$

*Remark:* 2<sup>nd</sup> order problems of the form

$$y''(t) = \frac{A_1(t)}{t}y'(t) + \frac{A_0(t)}{t^2}y(t) + f(t, y(t))$$

can be transformed to the first order form by setting

$$z(t) := (y(t), ty'(t)).$$

### **Applications:**

- chemical reactor theory
- physics — theory of lasers, computational material science
- mechanics — buckling of (spherical) shells
- ecology — run-up of avalanches

# The Solution Method

- Collocation at **equidistant** or **Gaussian** points.
- New global error estimate  
**in mesh points + collocation points:**  
**Defect Correction Principle!**
- Equidistribution of the global error.

## Motivation

- Order reductions for singular problems occur for various standard methods:
  - explicit Runge-Kutta methods
  - Multistep methods
  - Acceleration techniques  
(IDeC, extrapolation)

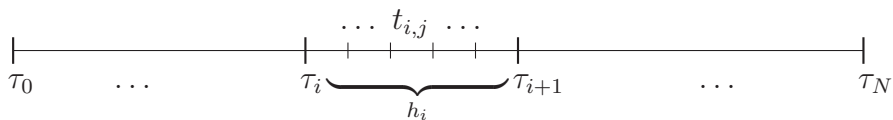
### **for general class of BVPs**

- Shooting + IDeC based on the backward Euler method works, but:

### **Only for a restricted class of BVPs!**

- Estimates of the local error do not work for singular problems in general, e.g. when using:
  - standard three-point discretization for 2<sup>nd</sup> order problems
  - explicit Runge-Kutta pairs

# Collocation



$$p'(t_{i,j}) = F(t_{i,j}, p(t_{i,j})),$$

$$j = 1, \dots, m, \quad i = 0, \dots, N - 1,$$

$$B_a p(0) + B_b p(1) = \beta.$$

The **stage order is retained**, but there is **no superconvergence** in general.

However, the order reduction is actually not observed for most singular problems  $\implies$

- even number of equidistant (interior) collocation points ..... **default**
- Gaussian points ..... **optional**

## **Interior collocation points:**

- No evaluation at the singular point  $t = 0$
- Error estimate requires  $t_{i,m} \neq \tau_{i+1}$

## Global Error Estimation

- We construct a **neighboring problem** using the **Defect Correction Principle**.
- Then we solve for  $j = 1, \dots, m + 1$ ,  
 $i = 0, \dots, N - 1$ , the backward Euler schemes

$$\frac{\pi_{i,j} - \pi_{i,j-1}}{t_{i,j} - t_{i,j-1}} = F(t_{i,j}, \pi_{i,j}) + \bar{d}_{i,j},$$

$$B_a \pi_{0,0} + B_b \pi_{N-1,m+1} = \beta,$$

$$\frac{\xi_{i,j} - \xi_{i,j-1}}{t_{i,j} - t_{i,j-1}} = F(t_{i,j}, \xi_{i,j}),$$

$$B_a \xi_{0,0} + B_b \xi_{N-1,m+1} = \beta,$$

where

$$t_{i,m+1} := \tau_{i+1}$$

denotes the endpoint of each interval.

The estimate for the global error is given by

$$\mathcal{E}_{i,j} := \pi_{i,j} - \xi_{i,j}.$$

## Global Error Estimation (2)

The defect is chosen as

$$\bar{d}_{i,j} := \frac{p(t_{i,j}) - p(t_{i,j-1})}{t_{i,j} - t_{i,j-1}} - \sum_{k=1}^{m+1} \alpha_{j,k} F(t_{i,k}, p(t_{i,k})).$$

Here, the  $\alpha_{j,k}$  define a quadrature rule,

$$\frac{1}{t_{i,j} - t_{i,j-1}} \int_{t_{i,j-1}}^{t_{i,j}} \varphi(\tau) d\tau = \sum_{k=1}^{m+1} \alpha_{j,k} \varphi(t_{i,k}) + O(h_i^{m+1}).$$

**Theorem:** For regular problems and collocation at an even number of interior points, this estimate is asymptotically correct,

$$|(p(t_{i,j}) - z(t_{i,j})) - (\pi_{i,j} - \xi_{i,j})| = O(h_i^{m+1}), \\ j = 0, \dots, m+1, \quad i = 0, \dots, N-1.$$

**Proof:** Auzinger, Koch, Weinmüller (2001).

# Adaptive Mesh Selection

**Monitor function**  $\Theta_{i,j}$  based on our error estimate,

$$\begin{aligned}\mathcal{E}_{i,j} &:= \pi_{i,j} - \xi_{i,j}, \\ \Theta_{i,j} &:= \sqrt[m]{|\mathcal{E}_{i,j}|}.\end{aligned}$$

*Remark:* Actually, a smoothed version of  $\Theta_{i,j}$  is used.

$(\bar{\tau}_0, \dots, \bar{\tau}_{\bar{N}})$  ... grid aiming at the equidistribution of the integral of the monitor function on the interval  $[0, 1]$ :

$$\begin{aligned}I &:= \int_0^1 \Theta(s) ds := \sum_{i,j} \frac{\Theta_{i,j} + \Theta_{i,j-1}}{2} (t_{i,j} - t_{i,j-1}), \\ \int_{\bar{\tau}_i}^{\bar{\tau}_{i+1}} \Theta(s) ds &= \frac{I}{\bar{N}}, \quad i = 0, \dots, \bar{N} - 1.\end{aligned}$$



# Damped Newton Iteration

Let

$$x := x_{i,j}, \quad j = 1, \dots, m + 1, \quad i = 0, \dots, N - 1$$

be the vector of the coefficients of the collocation solution w.r.t. a basis of piecewise polynomial functions,

$$p(t) = \sum_{j=1}^{m+1} x_{i,j} \varphi_j \left( \frac{t - \tau_i}{h_i} \right), \quad t \in [\tau_i, \tau_{i+1}],$$

e. g. Lagrange basis, Runge-Kutta basis.

$n$ -th approximation  $x^{(n)}$ :

$$x^{(n)} = x^{(n-1)} + \lambda^{(n)} \Delta x^{(n)},$$

$\Delta x^{(n)}$  ..... classical Newton increment

$\lambda^{(n)}$  ..... damping factor

See Deuffhard, Hohmann (1995).

# Example

$$z'(t) = \frac{1}{t} \begin{pmatrix} 0 & 1 \\ 0 & -1 \end{pmatrix} z(t) - \begin{pmatrix} 0 \\ tz_1^5(t) \end{pmatrix},$$
$$\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} z(0) + \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} z(1) = \begin{pmatrix} \frac{\sqrt{3}}{2} \\ 0 \end{pmatrix}.$$

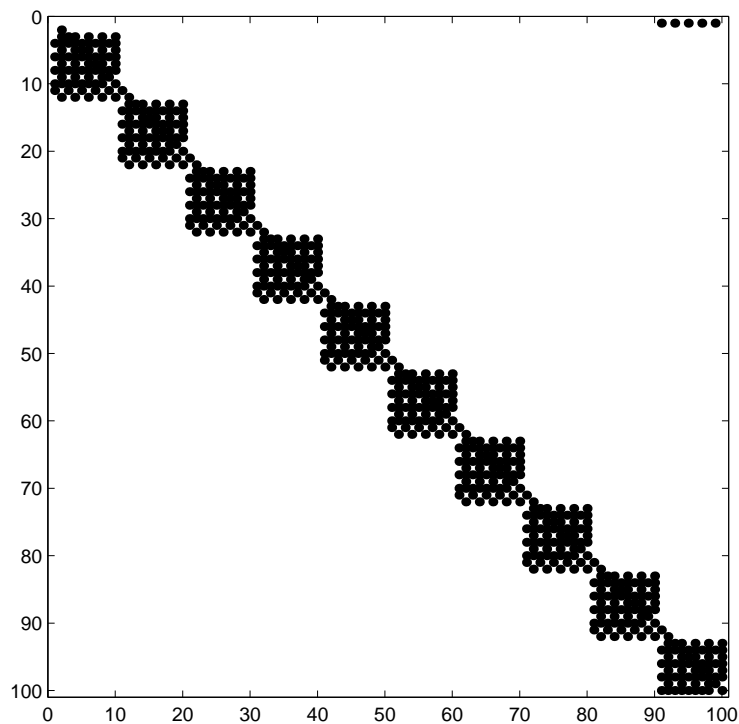
Exact solution:

$$z(t) = (z_1(t), z_2(t))^T = \left( \frac{1}{\sqrt{1+t^2/3}}, -\frac{t^2}{3\sqrt{(1+t^2/3)^3}} \right)^T.$$

$n$	$\text{res}^{(n)}$	$ \Delta x^{(n)} $	$p^{(n)}$	$\lambda^{(n)}$
1	6.34e+01	1.88e+00		0.500
2	3.17e+01	9.35e-01	-0.19	0.125
3	2.77e+01	1.07e+00	-3.02	0.250
4	2.08e+01	7.17e-01	1.70	0.500
5	1.04e+01	3.64e-01	2.51	1.000
6	3.94e-01	6.65e-02	1.01	1.000
7	4.01e-02	1.20e-02	2.30	1.000
8	9.40e-04	2.38e-04	1.92	1.000
9	4.23e-07	1.25e-07	2.01	1.000
10	1.07e-13	3.25e-14		1.000

# Conditioning of the Jacobian

The nonlinear system of equations for  $x = x_{i,j}$  leads to a Newton iteration with a Jacobian with the following sparse structure:





## Choice of the Basis

The choice of the polynomial basis  $\varphi_j$  also has an important influence on the maximal attainable accuracy of the collocation solution.

Let

$$0 < \rho_1 < \cdots < \rho_m < 1.$$

Lagrange basis:

$$\varphi_j(0) = \delta_{j,1}, \quad \varphi_j(\rho_i) = \delta_{j,i+1}.$$

Runge-Kutta basis:

$$\varphi_j(0) = \delta_{j,1}, \quad \varphi_j'(\rho_i) = \delta_{j,i+1}.$$

$N$	$m = 4$ , RK	$m = 4$ , La	$m = 6$ , RK	$m = 6$ , La
16	1.18e+08	1.18e+08	1.89e+04	1.88e+04
32	7.41e+06	7.41e+06	2.98e+02	1.37e+02
64	4.63e+05	4.63e+05	2.36e+00	4.36e+02
128	2.89e+04	2.87e+04	4.00e+00	8.79e+02
256	1.81e+03	1.31e+03	1.52e+00	1.75e+03
512	1.12e+02	8.95e+02	4.06e+00	3.50e+03
1024	8.00e+00	2.00e+03	4.51e+00	7.01e+03

Error in multiples of  $\epsilon$

## Implementation Details

- Efficient storage and referencing of variables taking into account MATLAB's memory representation of arrays.
- Vectorization:

Avoid loops where possible!

Our implementation provides the possibility to specify the data functions for the right-hand side and the boundary conditions of the BVP in a vectorized format.

- Indexing — use of precomputed indices.

## Comparisons

- **bvp4c** (MATLAB 6.0 standard routine)
  - Collocation at 3 Lobatto points
  - (Fixed) order 4
  - Provides no error estimate
  - Modifications for singular problems:
    - Definition of right-hand side at  $t = 0$
    - Jacobian by finite differences
- **COLNEW** (Fortran 90 package)
  - Collocation at Gaussian points
  - Orders 2–14
  - Estimate:  $h, h/2$  principle  $\implies$  finer grid
- **sbvp** (our package)
  - Coll. at equidistant/Gaussian points
  - Orders 2–16
  - Estimate:  $2\times$  cheap auxiliary method

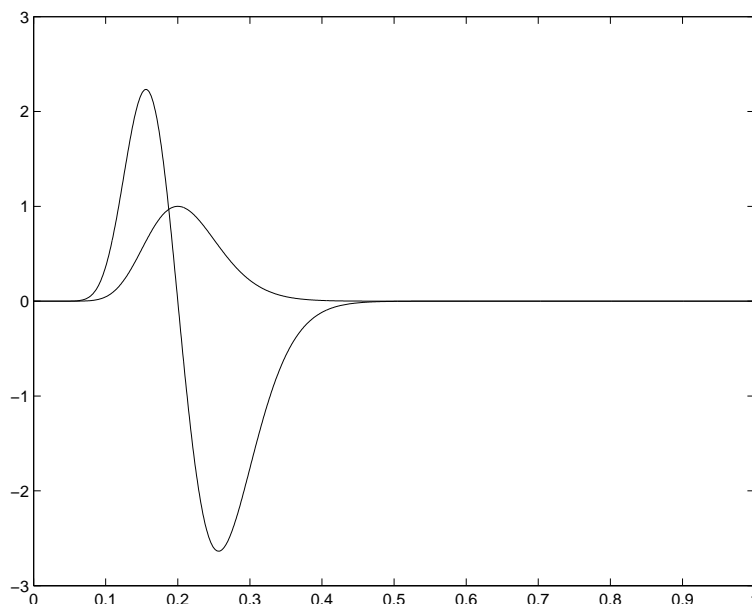
## Comparisons (2)

$$z'(t) = \frac{1}{t} \begin{pmatrix} 0 & 1 \\ 1 + \alpha^2 t^2 & 0 \end{pmatrix} z(t) + \\ + \begin{pmatrix} 0 \\ ct^{k-1} e^{-\alpha t} (k^2 - 1 - \alpha t(1 + 2k)) \end{pmatrix}, \\ \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} z(0) + \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} z(1) = \begin{pmatrix} 0 \\ ce^{-\alpha} \end{pmatrix},$$

where  $\alpha = 80$ ,  $k = 16$  and  $c = \left(\frac{\alpha}{k}\right)^k e^k$ .

Exact solution:

$$z(t) = (ct^k e^{-\alpha t}, ct^k e^{-\alpha t} (k - \alpha t))^T.$$





## **Comparisons (3)**

Hierher kommt die erste landscape-seite!

# Comparisons (4)

2. landscape seite

# Comparisons (5)

Meshes to reach the tolerances

$$aTOL=rTOL=10^{-5}$$

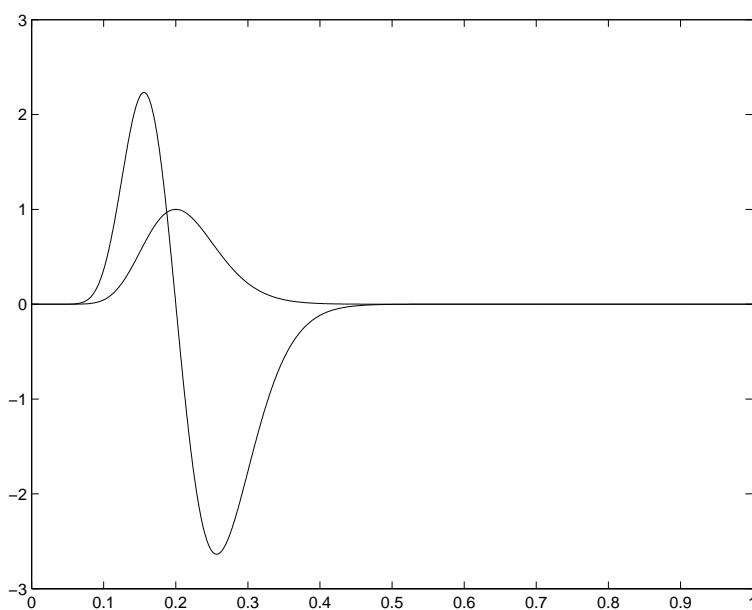


Figure 1: bvp4c,  $N = 116$

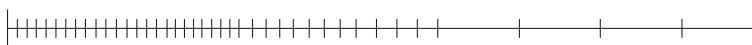


Figure 2: COLNEW-4,  $N = 41$

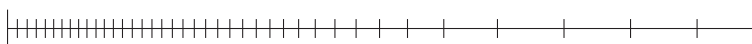


Figure 3: sbvp4g,  $N = 40$

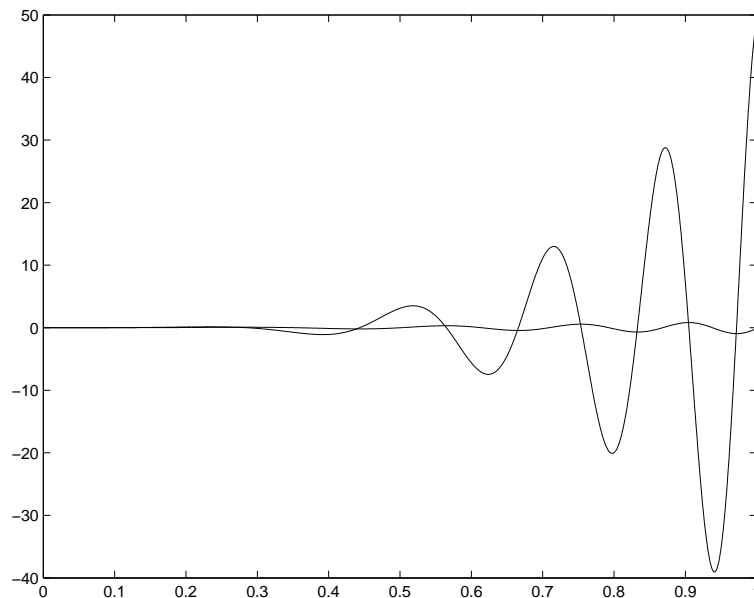
## Comparisons (6)

$$z'(t) = \frac{1}{t} \begin{pmatrix} 0 & 1 \\ 2 & 6 \end{pmatrix} z(t) - \begin{pmatrix} 0 \\ 4k^4t^5 \sin(k^2t^2) + 10t \sin(k^2t^2) \end{pmatrix},$$
$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} z(0) + \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} z(1) = \begin{pmatrix} 0 \\ \sin(k^2) \end{pmatrix},$$

where  $k = 5$ .

Exact solution:

$$z(t) = (t^2 \sin(k^2t^2), 2k^2t^4 \cos(k^2t^2) + 2t^2 \sin(k^2t^2))^T.$$



## Comparisons (7)

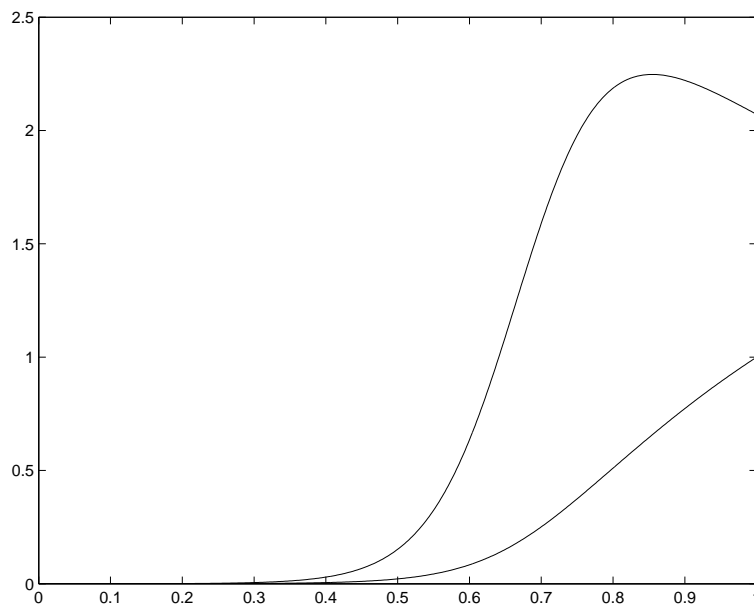
3. landscape seite

## Comparisons (8)

$$z'(t) = \frac{1}{t} \begin{pmatrix} 0 & 1 \\ 0 & -1 \end{pmatrix} z(t) + \begin{pmatrix} 0 \\ t\phi^2 z_1(t) \exp\left(\frac{\gamma\beta(1-z_1(t))}{1+\beta(1-z_1(t))}\right) \end{pmatrix},$$
$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} z(0) + \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} z(1) = \begin{pmatrix} 0 \\ 1 \end{pmatrix},$$

where  $\phi = 0.6$ ,  $\gamma = 40$ ,  $\beta = 0.2$ .

This problem has multiple solutions!



## Comparisons (9)

4. landscape seite!

## Conclusions & Future Work

- **sbvp** is competitive for **singular problems with smooth solutions**
- The results are most favorable for linear problems.
- We encountered problems when solving singular problems with unsmooth solutions.
- Regular problems with steep slopes:  
We are experimenting with a new mesh selection strategy.



## Conclusions & Future Work (2)

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$$z'(t) = \begin{pmatrix} z_2(t) \\ n \sinh(nz_1(t)) \end{pmatrix},$$
$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} z(0) + \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} z(1) = \begin{pmatrix} 0 \\ 1 \end{pmatrix},$$

where  $n = 8$ .

aTOL=rTOL= $10^{-3}$ ,  $m = 6$ .

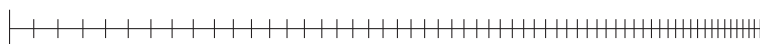
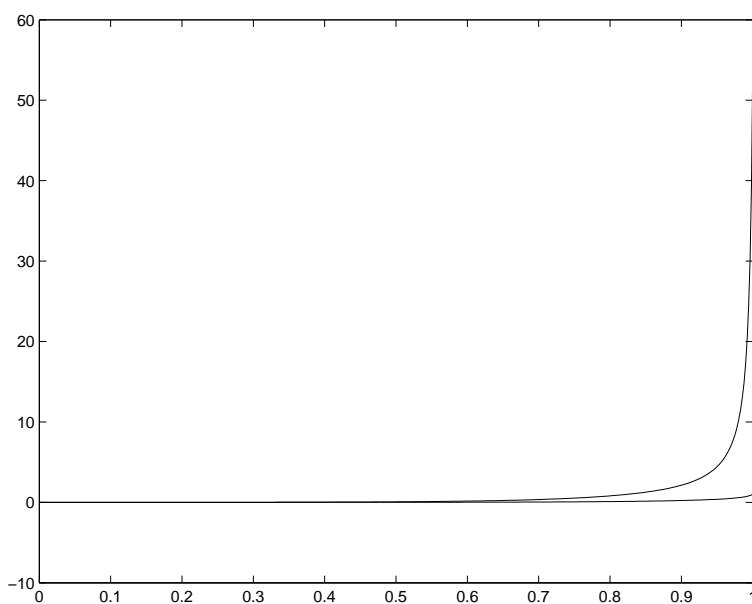


Figure 4: sbvp,  $N = 61$ ,  $h_{\max}/h_{\min} = 4.32$

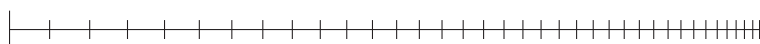


Figure 5: New strategy,  $N = 38$ ,  $h_{\max}/h_{\min} = 6.78$