



INVESTMENTS IN EDUCATION DEVELOPMENT

# Streamlining the Mathematics Studies at the Faculty of Science of Palacky University in Olomouc

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# Numerical Treatment of **Singular** BVPs in ODEs

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Basic concepts – work by H. Keller ✓

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Forward Euler method

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*Consistency and convergence*

*Numerical examples*

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*Existence and uniqueness of the collocation solution*

*Convergence*

*Numerical example*

# Motivation

Complex Ginzburg-Landau equation

$$i \frac{\partial u}{\partial t} + (1 - i\varepsilon) \Delta u + (1 + i\delta) |u|^2 u = 0, \quad t \in (0, \infty), \quad x \in \mathbb{R}^3$$

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We are interested in *self-similar blow-up* solutions

Ansatz

$$u(x, t) = L(\tau) y(\tau)$$

$$\tau = \tau(x, t) = \sqrt{\frac{x_1^2 + x_2^2 + x_3^2}{2a(T - t)}}, \quad \lim_{t \rightarrow T} L(\tau(x, t)) = \infty$$

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Nonlinear optics, models of turbulence, superconductivity

Budd, Koch, W. (2006)

# Similarity reduction

Find  $y : [0, \infty) \rightarrow \mathbb{C}$

$$(1 - i\varepsilon) \left( y''(\tau) + \frac{2}{\tau} y'(\tau) \right) - y(\tau) + ia(\tau y(\tau))' + (1 + i\delta) |y(\tau)|^2 y(\tau) = 0$$

$$y'(0) = 0, \quad \Im y(0) = 0, \quad \lim_{\tau \rightarrow \infty} \tau y'(\tau) = 0$$

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Implicit, parameter dependent ODE with singular points and unknown parameter  $a$  posed on a semi-infinite interval

Need: Code equipped with error estimation for the global error and grid adaptation strategy

# Introduction: Analytical properties

$$z'(t) = \frac{1}{t^\alpha} f(t, z(t)), \quad t \in (0, 1], \quad \alpha \geq 1$$

$$z'(\tau) = \tau^\beta f(\tau, z(\tau)), \quad \tau \in [1, \infty), \quad \beta \geq -1$$

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Problems posed on a *semi-infinite* interval are transformed to a *finite* interval by  $t := 1/\tau$

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## Second order systems

$$z''(t) = \frac{1}{t^\alpha} f(t, z'(t)) + \frac{1}{t^{\alpha+1}} g(t, z(t)), \quad t \in (0, 1], \quad \alpha \geq 1$$

# Introduction: Smooth solutions

Model problem  $z'(t) = \underbrace{\frac{M(t)}{t} z(t)}_{\frac{M(0)+tC(t)}{t} z(t)} + f(t, z(t)), \quad t \in (0, 1]$

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Smoothness of  $z$  depends on smoothness of  $f$  and on the  
**size of the positive real parts of the eigenvalues of  $M(0)$**

de Hoog, Weiss (1976), de Hoog, Weiss (1980), W. (1980)

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$$\|\text{global error} - \text{error estimate}\| = O(h^{m+\gamma}), \quad \gamma > 0$$

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- ▶ Adaptive mesh selection:

Meshes unaffected by the unsmooth (!) direction field

# MATLAB code `bvpsuite` – Scope

Kitzhofer, Koch, Pulverer, Simon, W. (2009)

<http://www.math.tuwien.ac.at/~ewa/>

Implicit nonlinear mixed order system of ODEs

$$F(t, p_1, \dots, p_s, z_1(t), z_1'(t), \dots, z_1^{(l_1)}(t), \dots$$

$$z_n(t), z_n'(t), \dots, z_n^{(l_n)}(t)) = 0$$

$$B(p_1, \dots, p_s, z_1(c_1), \dots, z_1^{(l_1-1)}(c_1), \dots, z_n(c_1), \dots$$

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Unknowns:  $z(t) = (z_1(t), z_2(t), \dots, z_n(t))^T$ ,  $p_i$ ,  $i = 1, \dots, s$

In general,  $t \in [a, b]$  or  $t \in [a, \infty)$ ,  $a \geq 0$

# Main features

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- ▶ Specification of different  $TOL_a$  and  $TOL_r$  parameters for different solution components is possible
- ▶ *Mesh adaptation is the standard choice* but fixed grid is also possible
- ▶ Driver routine for the interval reduction  $[a, \infty)$ ,  $a \geq 0$  to a *finite domain* in case of *Dirichlet BC at infinity* is available

# Introduction: Additional modules

- ▶ Pathfollowing for parameter-dependent problems—turning points covered.

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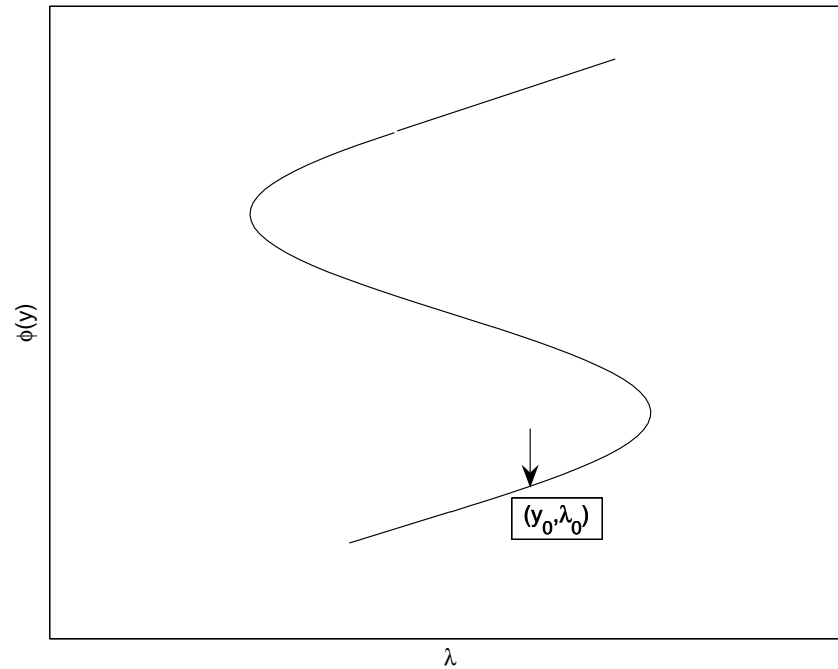
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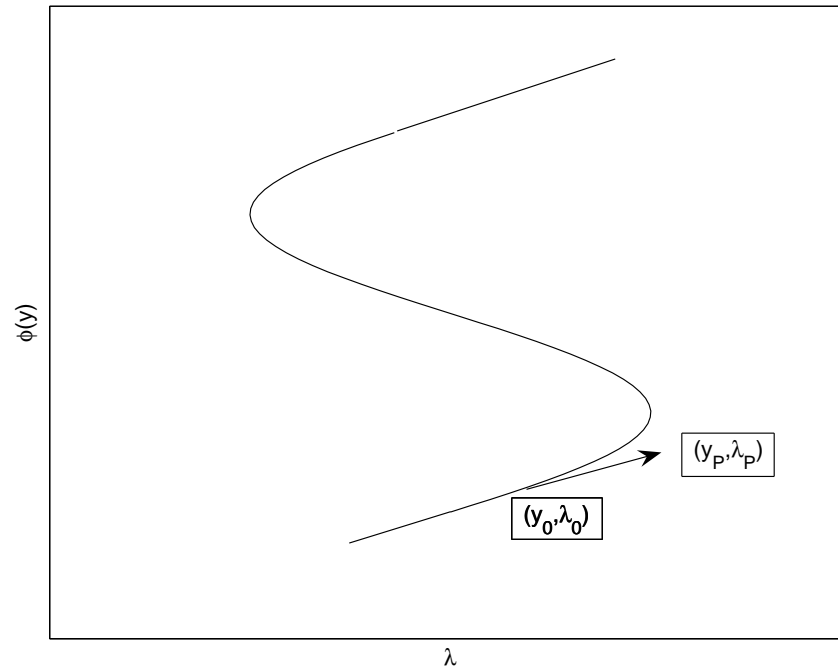
- ▶ Pathfollowing for parameter-dependent problems—turning points covered.
- ▶ Eigenvalue problems—Sturm-Liouville problems with singularities covered.
- ▶ Problems posed on semi-infinite intervals,  
 $t \in [a, \infty], a \leq 0$ .



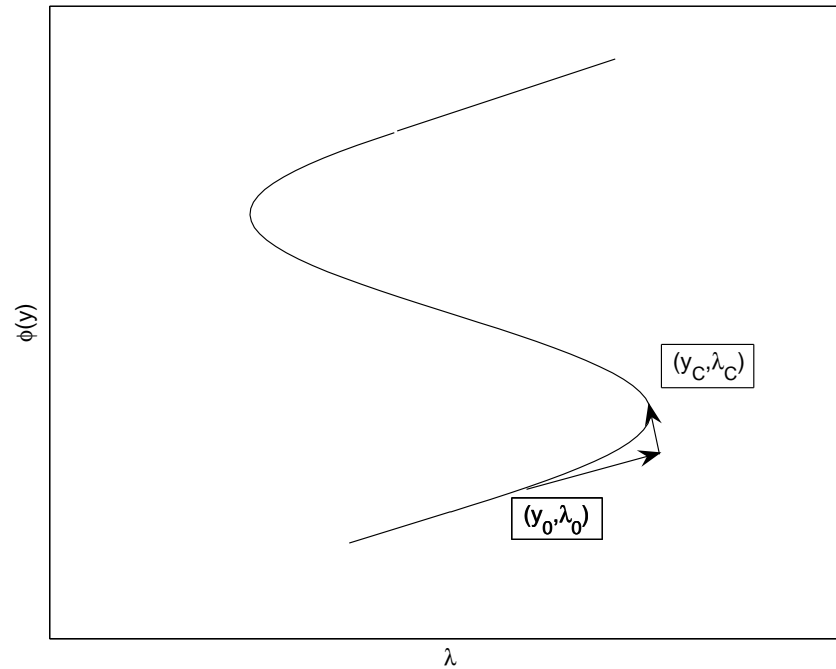
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# Introduction: case study

Case study: Consider linear system

$$z'(t) - \frac{D}{t}z(t) = f(t), \quad t \in (0, 1], \quad D = \text{diag}(-1, 2)$$
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Box scheme on equidistant mesh,  $h = 1/N$  (T2.1-T2.2):

$$\frac{\xi_i - \xi_{i-1}}{h} - \frac{D}{t_{i-1/2}} \frac{\xi_i + \xi_{i-1}}{2} = f_{i-1/2}, \quad i = 1, \dots, N$$
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Question: convergence  $\lim_{h \rightarrow 0} \|\xi_h - z_{\Delta_h}\| = 0?$

$$\xi_h := (\xi_0, \dots, \xi_i, \dots, \xi_N) \quad z_{\Delta_h} := (z(0), \dots, z(t_i), \dots, z(1))$$

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- ▶ Show consistency for  $z$  sufficiently smooth (T2.4)

$$(\Phi_h z_{\Delta_h})_i = \frac{z(t_i) - z(t_{i-1})}{h} - D \frac{z(t_i) + z(t_{i-1})}{2} = f_{i-1/2} + O(h^2).$$

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## ► Difficulty in singular case

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## ► Remedy: $(\Phi_h(\xi_h - z_{\Delta_h}))_i = O\left(\frac{h^2}{t_{i-1/2}}\right) \Rightarrow$

$$(\xi_h - z_{\Delta_h})_i = \Phi_h^{-1} O\left(\frac{h^2}{t_{i-1/2}}\right) = O(h^2)$$



## Problem setting: IVP

Consider the following linear system and assume that  $\lambda(M) \in \mathbb{R}$  and  $\lambda < 0$ ,

$$z'(t) = \frac{M}{t} z(t) + f(t), \quad t \in (0, 1],$$
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With  $v(t) := E^{-1}z(t)$ ,  $g(t) := E^{-1}f(t)$ , the problem becomes

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$$\underbrace{E^{-1}z'(t)}_{=:v'(t)} = \frac{1}{t} J \underbrace{E^{-1}z(t)}_{=:v(t)} + \underbrace{E^{-1}f(t)}_{=:g(t)}, \quad v(0) = E^{-1}z(0) = 0.$$

# Decoupled problem

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Consider the operator  $F$  defined as

$$F(v) := \begin{pmatrix} v'(t) - \frac{1}{t} Jv(t) - g(t), & t \in (0, 1] \\ v(0) \end{pmatrix},$$

then for the exact solution  $F(v) = 0$  holds.

# Numerical scheme: Forward Euler

Consider a uniform mesh,

$$\Delta_h := (t_{j_0}, t_{j_0+1}, \dots, t_N), \quad t_i = ih, \quad i = j_0, \dots, N, \quad h = \frac{1}{N}, \quad j_0 > 0$$

and  $\lim_{h \rightarrow 0} t_{j_0} = 0$ .

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Applying the forward Euler method to the problem

$$v'(t) = \frac{1}{t} Jv(t) + g(t), \quad t \in (0, 1], \quad v(0) = 0,$$

we obtain

$$v_{j_0} := v(0),$$
$$v_{j+1} := v_j + h \left( \frac{1}{t_j} Jv_j + g(t_j) \right), \quad j = j_0, \dots, N - 1.$$



## Numerical scheme

Let  $v_h = (v_{j_0}, \dots, v_N)$  and consider the discrete operator

$$F_h(v_h) := \begin{pmatrix} \frac{v_{j+1} - v_j}{h} - \frac{1}{t_j} Jv_j - g(t_j) & j = j_0, \dots, N-1 \\ v_{j_0} - v(0) \end{pmatrix},$$

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then the numerical solution satisfies  $F_h(v_h) = 0$ .

Note: We set  $v_{j_0} = v(0)$  not  $v_{j_0} = v(t_{j_0})$ .

## Numerical scheme

Let  $v_h = (v_{j_0}, \dots, v_N)$  and consider the discrete operator

$$F_h(v_h) := \begin{pmatrix} \frac{v_{j+1} - v_j}{h} - \frac{1}{t_j} Jv_j - g(t_j) & j = j_0, \dots, N-1 \\ v_{j_0} - v(0) \end{pmatrix},$$

then the numerical solution satisfies  $F_h(v_h) = 0$ .

Note: We set  $v_{j_0} = v(0)$  not  $v_{j_0} = v(t_{j_0})$ .

Recall

$$F(v) := \begin{pmatrix} v'(t) - \frac{1}{t} Jv(t) - g(t), & t \in (0, 1] \\ v(0) \end{pmatrix}, \quad F(v) = 0.$$

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Aim: Show the convergence of the **approximate solution**  $v_h$  towards the **exact solution**  $v$  as  $h \rightarrow 0$ .

## Towards convergence

Prove that the family  $F_h$  is stable and consistent by decoupling the system into single equations.



# Towards convergence

Consider the system  $v'(t) = \frac{J}{t}v(t) + g(t)$ , where

$$J = \begin{pmatrix} \lambda & 1 & & \\ & \ddots & \ddots & \\ & & \ddots & 1 \\ & & & \lambda \end{pmatrix}, \quad J \in \mathbb{R}^{k \times k}, \quad \lambda < 0.$$

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$$v'_i(t) = \frac{\lambda}{t}v_i(t) + \frac{1}{t}v_{i+1}(t) + g_i(t), \quad i = 1, \dots, k-1.$$

# Stability

We first consider the equation

$$v'(t) - \frac{\lambda}{t}v(t) - g(t) = 0, \quad v(0) = 0,$$

and analyze its discretization

$$F_h(v_h) = \left( \frac{v_{j+1} - v_j}{h} - \frac{\lambda}{t_j}v_j - g(t_j), j = j_0, \dots, N - 1, v_{j_0} - v(0) \right) = 0,$$

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$$F_h(u_h) = \delta_h^1, \quad F_h(w_h) = \delta_h^2 \Rightarrow F_h(u_h - w_h) = \delta_h^1 - \delta_h^2,$$

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Introduce  $\varepsilon_h := u_h - w_h$ ,  $l_h := \delta_h^1 - \delta_h^2$ .

# Stability

Consequently  $\varepsilon_h$  satisfies

$$\frac{\varepsilon_{j+1} - \varepsilon_j}{h} - \frac{\lambda}{t_j} \varepsilon_j = l_{j+1} \quad j = j_0, \dots, N - 1, \quad \varepsilon_{j_0} = l_{j_0}.$$

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The aim is now to show that

$\|\varepsilon_h\|$  is bounded in terms of  $\|l_h\| = \max_{j_0 \leq i \leq N} |l_h|$ ,

$$\|\varepsilon_h\| \leq \text{const} \|l_h\|.$$

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Lemma: The solution of the scheme (T2.7-T2.9)

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is given by

$$\varepsilon_j = \prod_{l=j_0}^{j-1} \left( 1 + \frac{h\lambda}{t_l} \right) l_{j_0} + \sum_{l=j_0}^{j-2} \prod_{k=l+1}^{j-1} \left( 1 + \frac{h\lambda}{t_k} \right) h l_{l+1} + h l_j.$$



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Moreover, the following estimates hold:

$$|\varepsilon_j| \leq \tilde{K} (|l_{j_0}| + t_j \max_{j_0+1 \leq l \leq N} |l_l|) \leq K \|l_h\|, \quad j = j_0, \dots, N.$$

**Proof.**

# Stability

Auxiliary Lemma 1: Let  $\lambda < 0$  and let for  $j \geq k \geq 1$

$$z_{kj}(\lambda) := \prod_{l=k}^{j-1} \left( 1 + \frac{h\lambda}{t_l} \right), \quad 1 \leq k < j, \quad j = 2, 3, \dots, \quad z_{jj} := 1.$$

Then, there exist constants  $\eta > 0$  and  $C \geq 1$  such that

$$|z_{kj}(\lambda)| \leq C \left( \frac{t_k}{t_j} \right)^\eta = C \left( \frac{k}{j} \right)^\eta, \quad 1 \leq k \leq j, \quad j = 1, 2, 3, \dots$$

Proof. (T2.10-T2.11)

# Stability

Auxiliary Lemma 2: Let  $h > 0$ ,  $t_j := jh$ ,  $k > j \geq i_0$  and  $\gamma \in \mathbb{R}$ . Then,

$$\sum_{l=j}^{k-1} ht_l^{\gamma-1} \leq \begin{cases} \text{const} |t_k^\gamma - t_j^\gamma|, & \gamma \neq 0, \\ \text{const} \ln \left( \frac{t_k}{t_j} \right), & \gamma = 0. \end{cases}$$

Proof. (T2.12)

# Stability estimate

Lemma: The solution of the scheme

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Proof. (T2.13-T2.14)

# Consistency

**Theorem:** Let  $v \in C^2[0, 1]$ , be the solution of  $F(v) = 0$ , then  $v(0) = 0$ . Define

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For  $v \in C^2[0, 1]$ , we have

$$\|l_h\| = O(h).$$

**Proof.** (T2.15)



# Convergence

## Corollary

The sequence of solutions  $v_h$  of the problems  $F_h(v_h) = 0$  converges towards  $v_{\Delta_h}$ , where  $v$  is the exact solution of  $F(v) = 0$ .

For  $v \in C^2[0, 1]$  the convergence order is one.

# Examples

Example 1: Consider the **linear** problem

$$y''(t) = -\frac{2}{t}y'(t) - \underbrace{n^2 \cos(nt) - \frac{2}{t}n \sin(nt)}_{\text{has a limit for } t \rightarrow 0}, \quad t \in (0, 1],$$

$$y(0) = 2, \quad y'(0) = 0,$$

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Example 2: Consider the **nonlinear** problem

$$y''(t) = -\frac{2}{t}y'(t) - y^5(t), \quad t \in (0, 1],$$

$$y(0) = 1, \quad y'(0) = 0,$$

where  $y(t) = \frac{1}{\sqrt{1+t^2/3}}$ .

# Example 1

$h$	$\ \varepsilon_h\ $	$p$	$c$
$1/5$	8.5	0.970	$-4.0 \cdot 10^{+01}$
$1/5 \cdot 2^{-1}$	4.3	0.982	$-4.1 \cdot 10^{+01}$
$1/5 \cdot 2^{-2}$	$2.1 \cdot 10^{-01}$	0.990	$-4.2 \cdot 10^{+01}$
$1/5 \cdot 2^{-3}$	$1.1 \cdot 10^{-01}$	0.995	$-4.3 \cdot 10^{+01}$
$1/5 \cdot 2^{-4}$	$5.5 \cdot 10^{-01}$	0.997	$-4.3 \cdot 10^{+01}$
$1/5 \cdot 2^{-5}$	$2.7 \cdot 10^{-02}$	0.998	$-4.4 \cdot 10^{+01}$
$1/5 \cdot 2^{-6}$	$1.3 \cdot 10^{-02}$	0.999	$-4.4 \cdot 10^{+01}$
$1/5 \cdot 2^{-7}$	$6.9 \cdot 10^{-02}$	0.999	$-4.4 \cdot 10^{+01}$
$1/5 \cdot 2^{-8}$	$3.4 \cdot 10^{-02}$	0.999	$-4.4 \cdot 10^{+01}$
$1/5 \cdot 2^{-9}$	$1.7 \cdot 10^{-03}$	0.999	$-4.4 \cdot 10^{+01}$

## Example 2

$h$	$\ \varepsilon_h\ $	$p$	$c$
$1/5$	$2.4 \cdot 10^{-02}$	0.834	$-9.2 \cdot 10^{-01}$
$1/5 \cdot 2^{-1}$	$1.3 \cdot 10^{-02}$	0.921	$-1.1 \cdot 10^{-01}$
$1/5 \cdot 2^{-2}$	$7.1 \cdot 10^{-02}$	0.960	$-1.2 \cdot 10^{-01}$
$1/5 \cdot 2^{-3}$	$3.6 \cdot 10^{-02}$	0.980	$-1.3 \cdot 10^{-01}$
$1/5 \cdot 2^{-4}$	$1.8 \cdot 10^{-03}$	0.990	$-1.4 \cdot 10^{-01}$
$1/5 \cdot 2^{-5}$	$9.3 \cdot 10^{-03}$	0.995	$-1.4 \cdot 10^{-01}$
$1/5 \cdot 2^{-6}$	$4.7 \cdot 10^{-03}$	0.997	$-1.4 \cdot 10^{-01}$
$1/5 \cdot 2^{-7}$	$2.3 \cdot 10^{-04}$	0.998	$-1.4 \cdot 10^{-01}$
$1/5 \cdot 2^{-8}$	$1.1 \cdot 10^{-04}$	0.999	$-1.5 \cdot 10^{-01}$
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# Summary

We discussed convergence of the forward Euler method applied to

$$z'(t) = \frac{M}{t}z(t) + f(t), \quad t \in (0, 1], \quad z(0) = 0, \quad \lambda(M) < 0.$$

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- ▶ For stability we invert the discrete operator.
- ▶ Consistency of order  $p = 1$  holds. Therefore that classical approach is applicable and the global error converges of order one on the interval  $[t_0, 1]$ , where  $t_0 \rightarrow 0$  for  $h \rightarrow 0$ .