



Supervisor: Prof. Manuel Torrilhon

Comparison of Time Stepping Techniques for Compressible Gas Dynamics

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Outline

- 1 Introduction
- 2 Numerics
- 3 Results
- 4 Conclusion

Introduction

Aim of the Project

Implementation and comparison
of efficient implicit time stepping schemes
for non-linear PDE systems

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- implicit time discretization using adaptive time stepping
- non-linear solver and computation of Jacobian
- preconditioned linear solver
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- implicit time discretization using adaptive time stepping
 - non-linear solver and computation of Jacobian
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 - comparison of methods
- ⇒ runtime speed-up

Example: Heat Equation

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Fourier's law

heat flux q proportional to negative temperature gradient over surface

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Models for heat conductivity κ

- $\kappa = 1 = \text{const}$ (linear)
- $\kappa(T) = c_1 + c_2 \cdot T^2$ (non-linear), $c_1, c_2 \in \mathbb{R}^+$

Explicit vs. Implicit schemes

Explicit schemes

- + implementation
- + no linear or non-linear solver needed
- CFL condition

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Implicit schemes

- implementation
- solution of large non-linear system of equations
- needs Jacobian
- + no time step constraints

Numerics

Spatial Discretization

$$\frac{\partial Q}{\partial t} + L(Q) = 0$$

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Spatial discretization scheme by KAPPER

- least-squares reconstruction of fluxes
- second order in space
- six-point stencil for flux reconstruction
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Apply different time stepping methods to

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Time stepping methods

- implicit Euler method
- implicit midpoint method
- implicit trapezoidal method
- BDF2 method
- Richardson extrapolation

Adaptive Time Stepping

Aim

calculate with largest possible time step Δt subject to given error bound ϵ and stability of the method

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Ingredients

- sensor for the error (error estimate)
- controller for Δt or h (time step adjustment strategy)
- use control theory model

Error Estimation

Comparison with higher order method

$$\hat{r}_{n+1} = \|y_1 - y_2\|$$

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Comparison by step size variation

$$y_{n+1}^h = y(t_{n+1}) + \phi(t_n)h^k + \mathcal{O}(h^{k+1})$$

$$y_{n+1}^{\frac{h}{m}} = y(t_{n+1}) + \phi(t_n)\frac{h^k}{m} + \mathcal{O}\left(\frac{h^{k+1}}{m}\right)$$

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$$\hat{r}_{n+1} = \left\| \frac{y_{n+1}^h - y_{n+1}^{\frac{h}{m}}}{1 - m^{-k}} \right\|$$

Elementary Error Control

Controller, see SÖDERLIND

$$h_{n+1} = \left(\frac{\epsilon}{\widehat{r}_{n+1}} \right)^{\frac{1}{k}} h_n$$

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Properties

- error estimate larger $\epsilon \Rightarrow$ decrease time step size
- error estimate smaller $\epsilon \Rightarrow$ increase time step size

Integral Controller

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- $\log \epsilon - \log \hat{r}_{n+1}$: *control error*
- ϵ *setpoint*
- k_I *integral gain*

Integral Controller

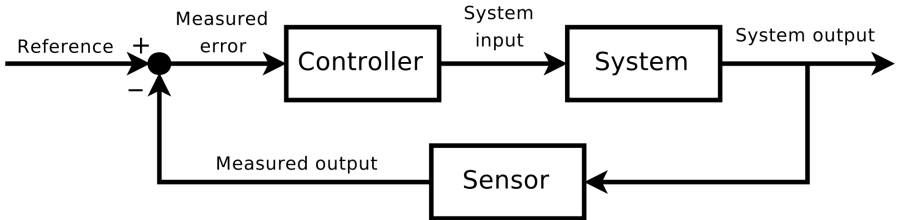
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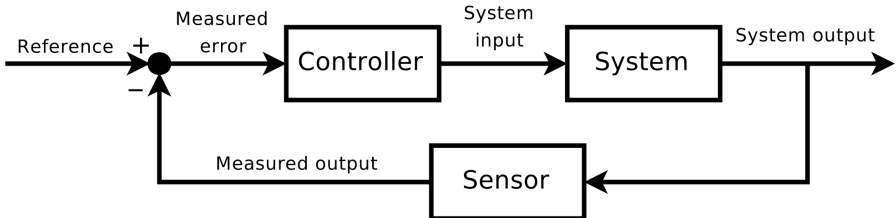
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- k_I *integral gain* ($k_I k \in [0, 2]$ for stability)

Control Theory Model



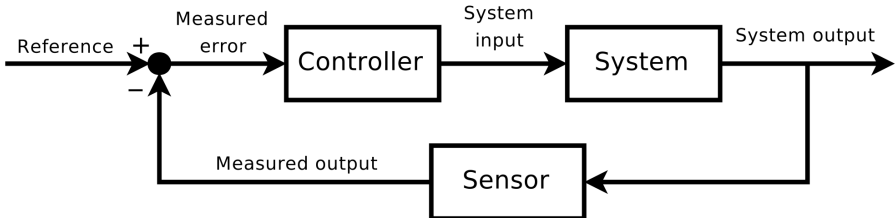
Control Theory Model



Questions

- when is the whole system stable?

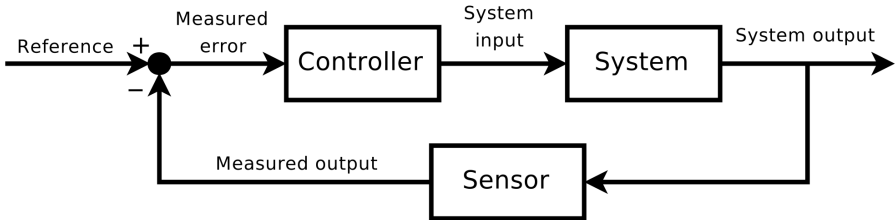
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Predictive controller

$$\frac{h_{n+1}}{h_n} = \left(\frac{\epsilon}{\widehat{r}_{n+1}} \right)^{k_E} \left(\frac{\widehat{r}_n}{\widehat{r}_{n+1}} \right)^{k_R} \frac{h_n}{h_{n-1}}$$

Non-Linear System

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Dual time stepping

$$\left(\frac{1}{\Delta \tau} I + \frac{\partial \tilde{R}}{\partial Q} \right) \Delta Q_{k+1}^{n+1} = -\tilde{R}(Q_k^{n+1})$$

Computation of Jacobian

Non-linear solver needs Jacobian of discretized residual function \tilde{R}

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- + arbitrary right hand side function
- only approximation of derivative
- number of right hand side evaluations increases with unknowns

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Solution

efficient finite differences

Solution of linear system

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Iterative solver

- GMRES
- BiCG
- BiCGSTAB

Solution of linear system

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Preconditioner

- SSOR
- ILU

Results

Testcase

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Linear test

- linear heat equation
- square domain
- quasi 1D setting, Dirichlet BC $Q = 0$, IC $Q = 0$
- right hand side sine function

Testcase

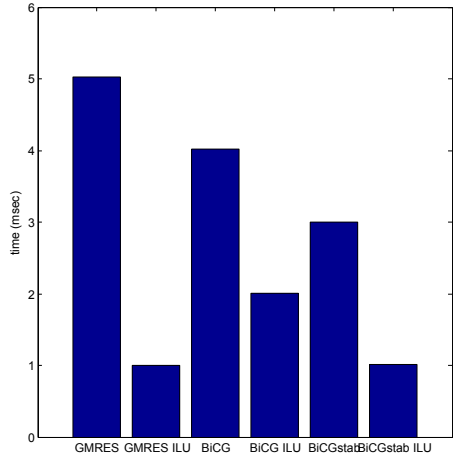
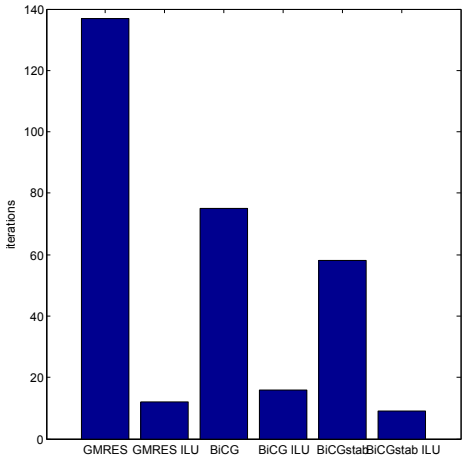
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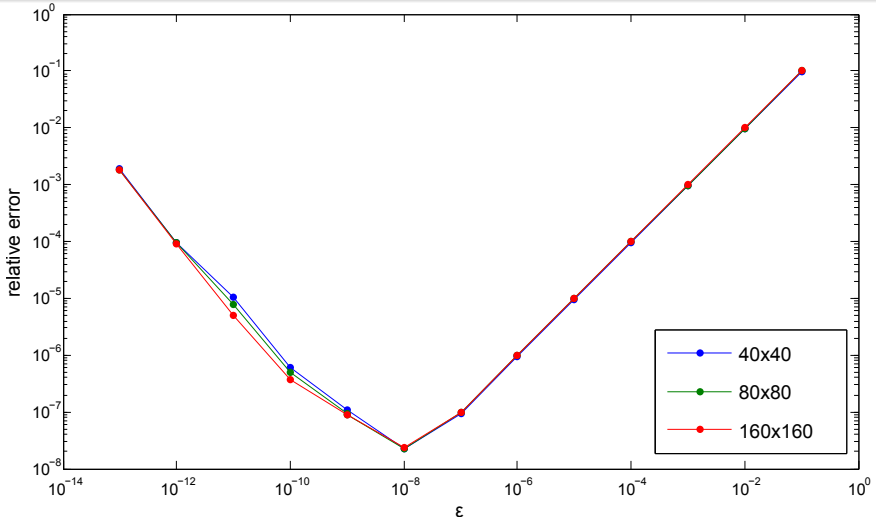
Non-linear test

- non-linear heat equation
- square domain
- Dirichlet BC $Q = 0$, IC $Q = 1$
- with constant right hand side function

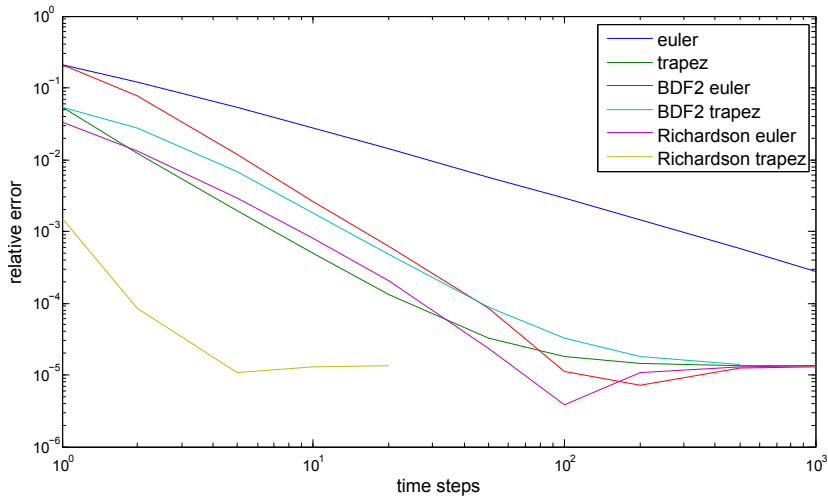
Linear solver iterations and runtime



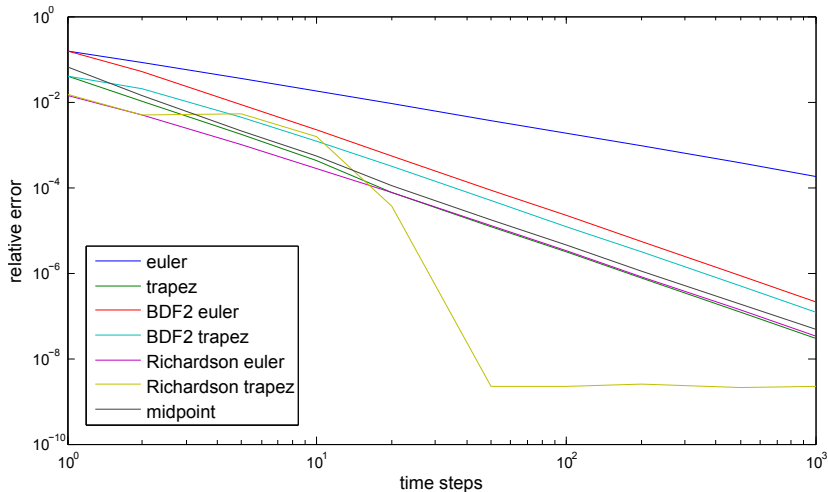
Error of Jacobian calculation



Linear test case



Non-linear test case



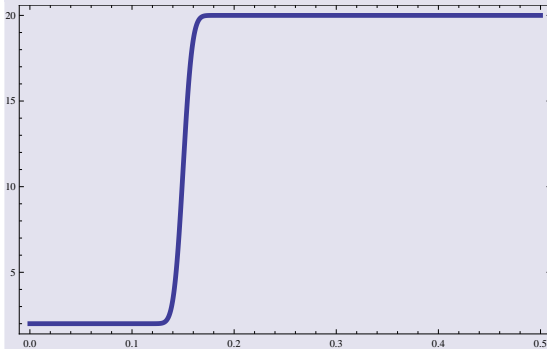
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Linear heat equation with right hand side function

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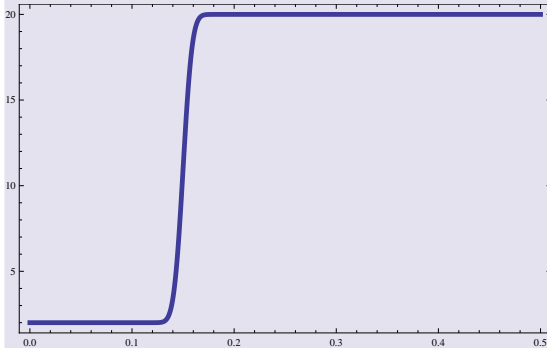
Non-linear right hand side function



Test case

Linear heat equation with right hand side function

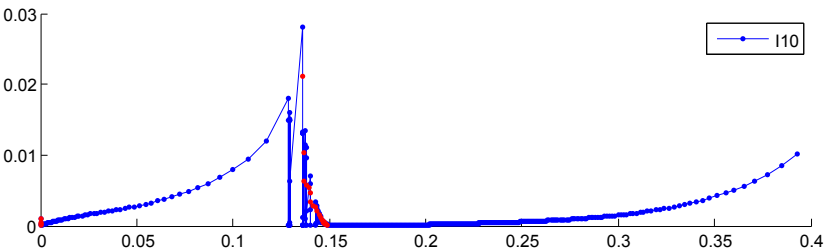
Non-linear right hand side function



Time step development

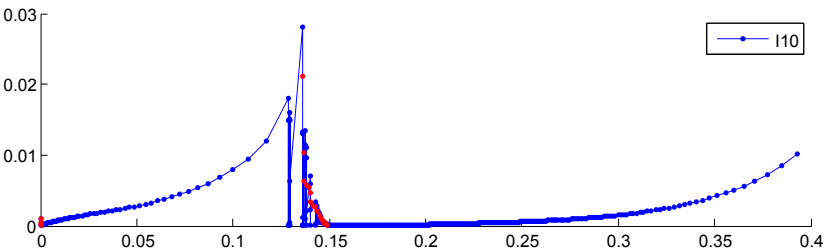
- increasing time step at the beginning
- strong non-linearity near jump at $Q = 0.15$ leads to small timesteps
- large time steps in the end

I vs. PI controller

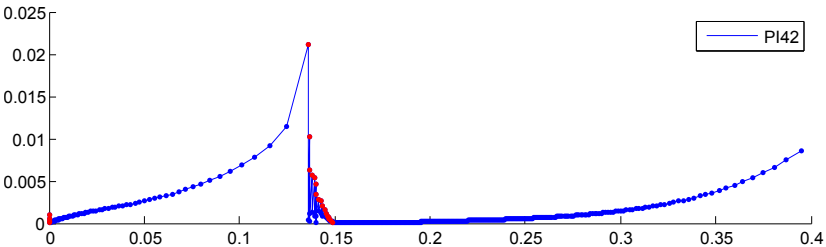


timesteps:
1170
rejects:
34

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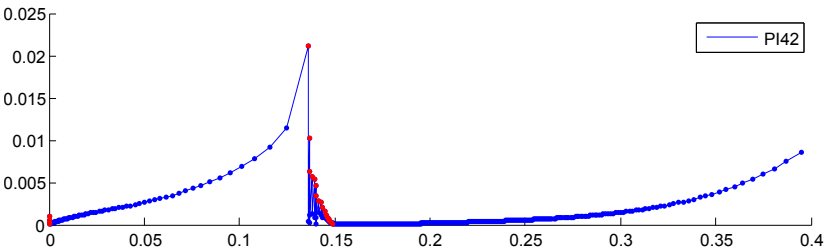


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 1170
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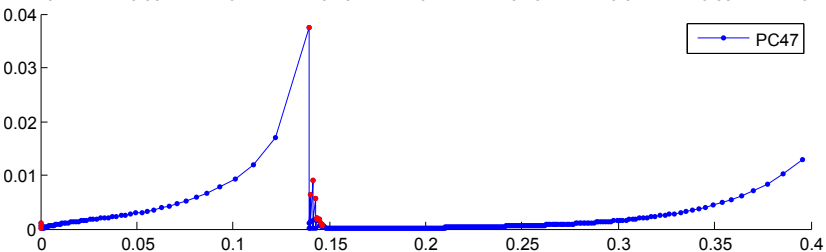
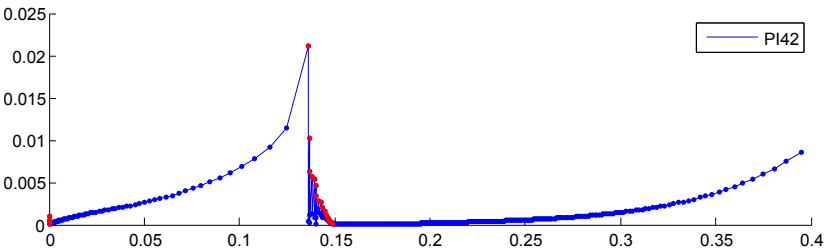
timesteps:
 1169
 rejects:
 32

PI vs. PC controller



timesteps:
1169
rejects:
32

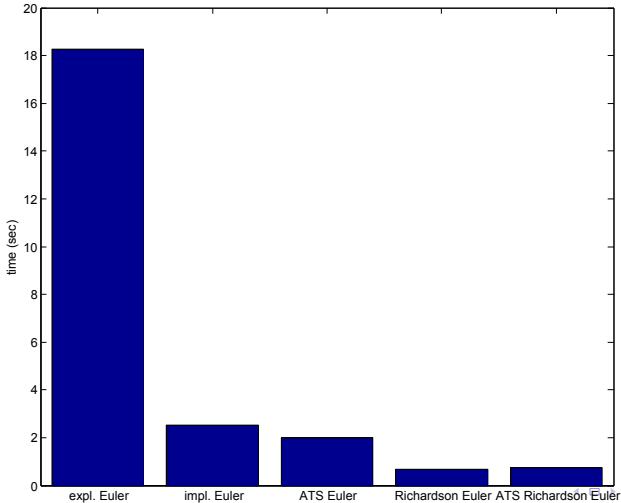
PI vs. PC controller



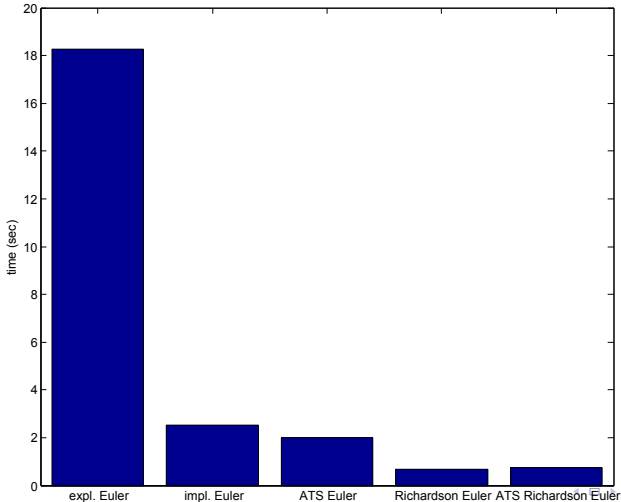
Test setting

- non-linear heat equation with constant right hand side
- calculate Q at time $t_{end} = 1\text{sec}$
- expl. method: time step size limited by stability
- impl. methods: maximum time step size for 1% relative error

Runtime Measurements



Runtime Measurements



expl. Euler

18.2 sec

impl. Euler

2.5 sec

ATS impl. Euler

2.0 sec

Richardson
Extrapolation

0.7 sec

Conclusion

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Successful implementation and comparison of different implicit time stepping techniques

- different implicit time discretizations
- adaptive time stepping
- non-linear and linear solvers as well as preconditioners
- comparison of methods for acceleration of simulation
- speedup of more than 90% with respect to explicit method

Thank you for your attention