# Partitioned Adaptive Parallel Integrators for Coupled Stiff Systems

#### Philipp Birken, Peter Meisrimel, Azahar Monge

Lund University, Sweden

Integrating the Integrators for Nonlinear Evolution Equations: from Analysis to Numerical Methods, High-Performance-Computing and Applications Banff, December 3rd, 2018





### 1 3M: Multiphysics, Multicore, Multirate

2 Thermal FSI: Neumann-Neumann waveform iterations

3 Asynchronous waveform iterations



Philipp Birken (Lund University)

## 1 3M: Multiphysics, Multicore, Multirate

2 Thermal FSI: Neumann-Neumann waveform iterations

3 Asynchronous waveform iterations



Philipp Birken (Lund University)

- Complex feedbacks between precipitation, temperature, vegetation
- Climate models incorporate PDE models for air- and waterflow
- Add differential equations for clouds, CO2 sources,...
- Vegetation growth based on PDE as well
- Coupling of different models implemented in different codes

Figure: Output of RCA-GUESS: Leaf area index change



0.2

-0.5



# Example: Steel forging

#### Inductive heating



# Thermo-mechanical forming



Local air-cooling



- Entire process chain is characterized by the exchange of energy
- Crystalline structure of steel influenced by heat

Thermal interaction between air and steel needs to be modelled



# Applications of thermal FSI

- Engines (rocket, car,...)
- Heating of reentry vehicles in space flight
- Turbine blade cooling
- Thermal anti-icing of airplanes
- Generally cooling systems



Figure: Vulcain engine for Ariane 5; CC-by-sa 3.0, Pline, Wikimedia Commons



#### Goal: Solve this fast without too much pain

# What do we want?

- Time adaptivity
- High order in time
- Reuse of existing codes (partitioned approach)
- Parallel execution of coupled codes
- Loadbalancing
- Different time steps in different models
- Fast solvers for equation systems
- Should be usable and robust for large class of models

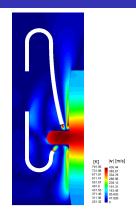


Figure: Navier-Stokes (FV)

+ nonl. heat (FE)



25

For steel forging problem: Dirichlet-Neumann works well, but is not parallel, Monge, B., Comp. Mech. 18

Philipp Birken (Lund University)

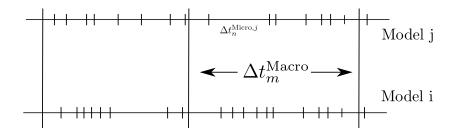
Partitioned Adaptive Parallel

#### Good coders write great code, great coders use good coders code

- Physics and mathematical models is not all
- Software influences design of numerical methods
- Option 1: Full access to code and willingness to edit it
- Option 2: Only access to code of certain functions, allowed to call some specific ones
- Option 3: No access, allowed to call specific interface functions
- Option 4: No access, Allowed only to call main function



## From ODE world: Waveform Iteration



- Synchronize models at macro steps
- In between, subsolvers run with their own time step
- Need to use information from other model at microsteps
- Is black box, but we require possibility to repeat a macrostep



#### 3M: Multiphysics, Multicore, Multirate

## 2 Thermal FSI: Neumann-Neumann waveform iterations

3 Asynchronous waveform iterations



Philipp Birken (Lund University)

## Model Problem: Coupled heat equations

• Coupled PDEs: Nonoverlapping Domain Decomposition

$$\alpha_{m} \frac{\partial u_{m}(\mathbf{x}, t)}{\partial t} - \nabla \cdot (\lambda_{m} \nabla u_{m}(\mathbf{x}, t)) = 0,$$
  

$$t \in [t_{0}, t_{f}], \quad \mathbf{x} \in \Omega_{m} \subset \mathbb{R}^{d}, \quad m = 1, 2$$
  

$$u_{m}(\mathbf{x}, t) = 0, \quad \mathbf{x} \in \partial \Omega_{m} \setminus \Gamma$$
  

$$u_{1}(\mathbf{x}, t) = u_{2}(\mathbf{x}, t), \quad \mathbf{x} \in \Gamma$$
  

$$\lambda_{2} \frac{\partial u_{2}(\mathbf{x}, t)}{\partial \mathbf{n}_{2}} = -\lambda_{1} \frac{\partial u_{1}(\mathbf{x}, t)}{\partial \mathbf{n}_{1}}, \quad \mathbf{x} \in \Gamma$$
  

$$u_{m}(\mathbf{x}, 0) = g_{m}(\mathbf{x}) \quad \mathbf{x} \in \Omega_{m}$$

• In the weak sense equivalent to

$$\alpha(\mathbf{x})\frac{\partial u_m(\mathbf{x},t)}{\partial t} - \nabla \cdot (\lambda(\mathbf{x})\nabla u_m(\mathbf{x},t)) = 0 \qquad \mathbf{x} \in \Omega$$
$$u(\mathbf{x},t) = 0, \quad \mathbf{x} \in \partial\Omega, \ u(\mathbf{x},0) = g_m(\mathbf{x}), \qquad \mathbf{x} \in \partial\Omega$$



#### Neumann-Neumann method

Given initial guess for values at boundary, iterate

- Solve Dirichlet problems in both domains with that data
- ② Solve Neumann problems in both domains with derivatives from above
- Get new Dirichlet data, do relaxation step
  - Can be done in parallel!
- Gander, Kwok, Mandal, ETNA '16: Fully continuous version with all material parameters one
- Prove optimal relaxation parameter of 1/4
- With this, exact solution at interface after one step!!



## Neumann-Neumann waveform relaxation (NNWR)

1. Dirichlet :

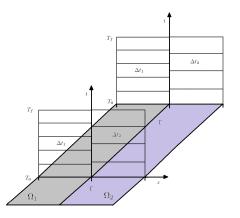
$$\begin{cases} \alpha_m \frac{\partial u_m^{k+1}(\mathbf{x},t)}{\partial t} - \nabla \cdot (\lambda_m \nabla u_m^{k+1}(\mathbf{x},t)) = 0, \quad \mathbf{x} \in \Omega_m, \\ u_m^{k+1}(\mathbf{x},t) = 0, \quad \mathbf{x} \in \partial \Omega_m \setminus \Gamma, \\ u_m^{k+1}(\mathbf{x},t) = g^k(\mathbf{x},t), \quad \mathbf{x} \in \Gamma, \\ u_m^{k+1}(\mathbf{x},0) = u_1^0(\mathbf{x}), \quad \mathbf{x} \in \Omega_m. \end{cases}$$

2.Neumann: 
$$\begin{cases} \alpha_m \frac{\partial \psi_m^{k+1}(\mathbf{x},t)}{\partial t} - \nabla \cdot (\lambda_m \nabla \psi_m^{k+1}(\mathbf{x},t)) = 0, \quad \mathbf{x} \in \Omega_m, \\ \psi_m^{k+1}(\mathbf{x},t) = 0, \quad \mathbf{x} \in \partial \Omega_m \setminus \Gamma, \\ \lambda_m \frac{\partial \psi_m^{k+1}(\mathbf{x},t)}{\partial \mathbf{n}_1} = \lambda_1 \frac{\partial u_1^{k+1}(\mathbf{x},t)}{\partial \mathbf{n}_1} + \lambda_2 \frac{\partial u_2^{k+1}(\mathbf{x},t)}{\partial \mathbf{n}_2}, \quad \mathbf{x} \in \Gamma, \\ \psi_m^{k+1}(\mathbf{x},0) = 0, \quad \mathbf{x} \in \Omega_m. \end{cases}$$

3. *Relaxation*:  $g^{k+1}(\mathbf{x},t) = g^k(\mathbf{x},t) - \Theta(\psi_1^{k+1}(\mathbf{x},t) + \psi_2^{k+1}(\mathbf{x},t)),$ 

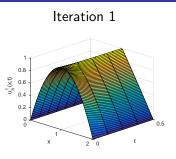
## Going discrete: Multirate

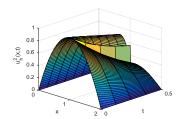
- Nonmatching time grids at the interface, linear interpolation
- Time discretization: Implicit Euler and SDIRK2 Order is acchieved





## Multirate 1D solution using NNWR algorithm

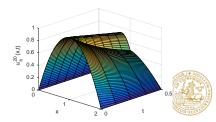


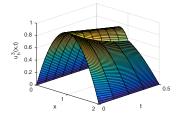


Iteration 2

Iteration 3

Iteration 20





Philipp Birken (Lund University)

Integrating Banff 15 / 25

**Q**: How to choose the relaxation parameter  $\Theta$ ?

• Write iteration in terms of interface unknowns at final time,  $\mathbf{u}_{\Gamma}(T_f)$ :

$$\mathbf{u}_{\Gamma}^{k+1,T_{f}} = \Sigma(\Theta)\mathbf{u}_{\Gamma}^{k,T_{f}} + \Psi$$

- Do only one time step only
- Find  $\Theta$  that minimizes the spectral radius of  $\Sigma(\Theta)$
- $\Theta_{opt} = \rho(2 + \mathbf{S}^{(1)^{-1}}\mathbf{S}^{(2)} + \mathbf{S}^{(2)^{-1}}\mathbf{S}^{(1)})$
- Give exact formula for model discretization
- Space discretization: equidistant FE/FE in 1D
- Matching space grid at the interface, unknowns on interface
- Time integration: nonmultirate Implicit Euler or SDIRK2



Use eigendecomposition of tridiagonal Toeplitz matrices  $\mathbf{M}/\Delta t + \mathbf{A}$  to get

$$\Theta_{opt} = \left(2 + \frac{(6\Delta x(\alpha_2 \Delta x^2 + 3\lambda_2 \Delta t) - (\alpha_2 \Delta x^2 - 6\lambda_2 \Delta t)^2 s_2)}{(6\Delta x(\alpha_1 \Delta x^2 + 3\lambda_1 \Delta t) - (\alpha_1 \Delta x^2 - 6\lambda_1 \Delta t)^2 s_1)} + \frac{(6\Delta x(\alpha_1 \Delta x^2 + 3\lambda_1 \Delta t) - (\alpha_1 \Delta x^2 - 6\lambda_1 \Delta t)^2 s_1)}{(6\Delta x(\alpha_2 \Delta x^2 + 3\lambda_2 \Delta t) - (\alpha_2 \Delta x^2 - 6\lambda_2 \Delta t)^2 s_2)}\right)^{-1}$$

with

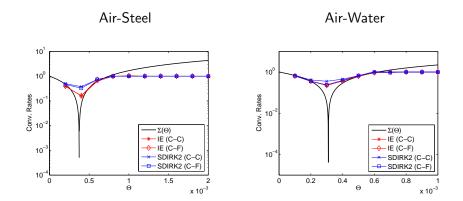
$$s_m = \sum_{i=1}^{N} \frac{3\Delta x^2 \sin^2(i\pi\Delta x)}{2\alpha_m \Delta x^2 + 6\lambda_m \Delta t + (\alpha_m \Delta x^2 - 6\lambda_m \Delta t) \cos(i\pi\Delta x)}$$

Asymptotics: 1D FEM/FEM

- Temporal limit of  $\Theta_{opt}$ :  $\Theta_{\{c \to 0\}} = \frac{\alpha_1 \alpha_2}{(\alpha_1 + \alpha_2)^2}$ .
- Spatial limit of  $\Theta_{opt}$ :  $\Theta_{\{c \to \infty\}} = \frac{\lambda_1 \lambda_2}{(\lambda_1 + \lambda_2)^2}$ .



## Convergence Rates, observed in 2D

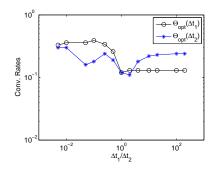


Simplifications in analysis don't affect location of  $\Theta_{opt}$ .



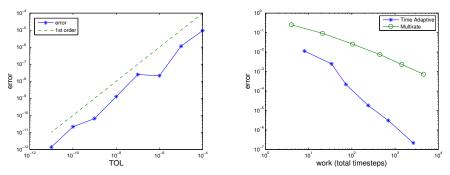
## And now once more, with time adaptivity!

- Add two step size controllers on the Dirichlet problems
- Problem:  $\Theta_{opt}$  depends on  $\alpha_m$ ,  $\lambda_m$ ,  $m = 1, 2, \Delta x$  and  $\Delta t$
- Initial iteration:  $\Theta = (\Theta_{\{c \to 0\}} + \Theta_{\{c \to \infty\}})/2$
- Then take average of time steps on each domain



• If  $\overline{\Delta t_1} < \overline{\Delta t_2}$ ,  $\Theta_{k+1} = \Theta_{opt}(\overline{\Delta t_2})$ ; else,  $\Theta_{k+1} = \Theta_{opt}(\overline{\Delta t_1})$ 





• Multirate uses respective minimal timestep from adaptive method

 See Monge, B.: A time adaptive Neumann-Neumann waveform relaxation method for thermal fluid-structure interaction, DD25, submitted



- Time adaptivity
- High order in time
- Parallel execution of coupled codes
- Loadbalancing to be done
- Different time steps in different models
- Fast solvers for equation systems
- Is being implemented in open source coupling software framework PreCICE (with B. Rueth, B. Ueckermann, M. Mehl)
- Robustness problematic: Relaxation parameter sensitive. Now look at Dirichlet-Neumann again with waveform relaxation with pipeline implementation
- Specific to thermal FSI

More at: Monge, B., A multirate Neumann-Neumann waveform relaxation method for heterogeneous coupled heat equations, SISC, submitted, arXiv:1805.04336

### 3M: Multiphysics, Multicore, Multirate

2 Thermal FSI: Neumann-Neumann waveform iterations

3 Asynchronous waveform iterations



Philipp Birken (Lund University)

## Discretely asynchronous

- Consider one macro step and solution operators  $\phi_x$ ,  $\phi_y$
- Use adaptive numerical method for both problems
- Gauß-Seidel waveform iteration not parallel
- Jacobi waveform iteration parallel, but in fact double the work
- Use continuous interpolations  $\mathbf{x}^{[t_n,t_{n+1}],k}(t)$ ,  $\mathbf{y}^{[t_n,t_{n+1}],k}(t)$  in time.
- Update interpolations after every microstep

$$\mathbf{x}^{m+1} = \mathbf{x}^m + \Phi_x(\mathbf{x}^m, \mathbf{y}^{[t_m, t_{m+1}], k}(t))$$

$$\mathbf{y}^{m+1} = \mathbf{y}^m + \Phi_y(\mathbf{y}^m, \mathbf{x}^{[t_m, t_{m+1}], k}(t))$$

- Is multiphysics, multirate, parallel and adaptive
- See also Frommer, Szyld 2000

## Preliminary results using MPI Window, Put, Get

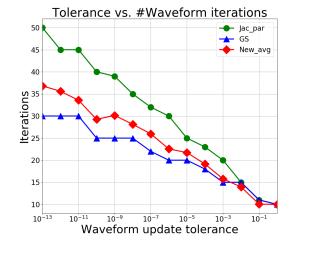


Figure: Averaged iterations for 1000 runs for 2-component linear system

- Time adaptivity
- High order in time
- Parallel execution of coupled codes
- Loadbalancing to be done
- Different time steps in different models
- Fast solvers for equation systems: Not clear
- Robustness unclear

