A new family of methods for global error control in ODE solvers

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Motivation I: ordinary differential equations

A mechanical system with multiple time-scales: *The Solar System*

- Moon: T = 1/12
- **Earth**: T = 1
- **Pluto:** T = 250
- Multiple time-scales
- Individual time steps





Animation contributed by Johan Jansson

Motivation II: partial differential equations

• Geometry (local refinement)



• Equation (local structures)



Outline

- Basic ideas
- Galerkin formulation
- Error estimates and adaptivity
- Implementation
- Examples and benchmarks
- Current status and future plans

Basic ideas

Objective

Solve the ODE initial value problem

$$\begin{cases} \dot{u}(t) = f(u(t), t), & t \in (0, T], \\ u(0) = u_0, \end{cases}$$

for $u : [0,T] \to \mathbb{R}^N$ with adaptive and individual time steps for the different components $u_i(t)$.

The individual time steps are chosen adaptively based on an a posteriori error estimate of the global error at time t = T.

Key features

- Adaptive individual time steps
- Efficient and reliable control of the global error
- Solution of dual problems, computation of stability factors
- Efficient adaptive iterative methods
- General implementation of arbitrary order mcG(q) and mdG(q) within **DOLFIN**



Individual time steps



Multi-adaptive solution of the bistable equation

Multi-adaptive time steps for the bistable equation

Individual piecewise polynomials

Galerkin formulation

Standard Galerkin

Standard Galerkin, cG(q):

$$\int_0^T (\dot{U}, v) \, \mathrm{dt} = \int_0^T (f(U, \cdot), v) \, \mathrm{dt} \quad \forall v \in \hat{V},$$

with $U \in V$, $U(0) = u_0$, and trial and test spaces given by

$$V = \{ v \in [\mathcal{C}([0,T])]^N : v_i|_{I_j} \in \mathcal{P}^q(I_j) \}, \\ \hat{V} = \{ v : v_i|_{I_j} \in \mathcal{P}^{q-1}(I_j) \}.$$

• Same time steps for all components U_i of U

Multi-adaptive Galerkin

Multi-adaptive Galerkin, mcG(q):

$$\int_0^T (\dot{U}, v) \, \mathrm{dt} = \int_0^T (f(U, \cdot), v) \, \mathrm{dt} \quad \forall v \in \hat{V},$$

with $U \in V$, $U(0) = u_0$, and trial and test spaces given by

$$V = \{ v \in [\mathcal{C}([0,T])]^N : v_i|_{I_{ij}} \in \mathcal{P}^{q_{ij}}(I_{ij}) \}, \\ \hat{V} = \{ v : v_i|_{I_{ij}} \in \mathcal{P}^{q_{ij}-1}(I_{ij}) \}.$$

- Individual time steps for all components U_i of U
- Includes the standard cG(q) method
- Similar extension of the dG(q) method to mdG(q)

The discrete equations for mcG(q)

With the following Ansatz for U_i on I_{ij} ,

$$U_{i}(t) = \sum_{n=0}^{q_{ij}} \xi_{ijn} \lambda_{n}^{[q_{ij}]}(\tau_{ij}(t)),$$

we obtain

$$\xi_{ijn} = \xi_0 + \int_{I_{ij}} w_n^{[q_{ij}]}(\tau_{ij}(t)) f_i(U,t) \,\mathrm{d}t,$$

for certain weight functions $\{w_n^{[q]}\} \subset \mathcal{P}^{q-1}(0,1)$.

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The discrete equations for mdG(q)

Similarly for the multi-adaptive discontinuous Galerkin method, mdG(q), we obtain

$$\xi_{ijn} = \xi_{ij0}^{-} + \int_{I_{ij}} w_n^{[q_{ij}]}(\tau_{ij}(t)) f_i(U,t) \,\mathrm{d}t,$$

for certain weight functions $\{w_n^{[q]}\} \subset \mathcal{P}^q(0,1)$.

Properties of mcG(q) and mdG(q)

- mcG(q) conserves energy if $k_{U_i} = k_{V_i}$
- mdG(q) is *B*-stable: if *f* is monotone,

$$(f(u, \cdot) - f(v, \cdot), u - v) \le 0 \quad \forall u, v \in \mathbb{R}^N,$$

then

 $||U(\bar{t}^{-}) - V(\bar{t}^{-})|| \le ||U(0^{-}) - V(0^{-})||$

Iterative method

- Arrange elements in time slabs
- Adaptive fixed-point iteration on time slabs
- Control the computational error

Basic strategy

The last component steps first

Error estimates and adaptivity

A priori error estimates

• The order of mcG(q) is 2q (locally $2q_{ij}$):

 $||e(T)|| \le CS(T) ||k^{2q} u^{(2q)}||_{L_{\infty}([0,T],l_1)}.$

• The order of mdG(q) is 2q + 1 (locally $2q_{ij} + 1$):

 $||e(T)|| \le CS(T) ||k^{2q+1} u^{(2q+1)}||_{L_{\infty}([0,T],l_1)}.$

S(T) is a stability factor obtained from the discrete dual problem.

A posteriori error estimates

The global error at final time is controlled using an a posteriori error estimate of the form

$$|L_{\psi,g}(e)| \le E_G + E_C + E_Q,$$

where $L_{\psi,g}(e) \equiv (e(T), \psi) + \int_0^T (e, g) dt$ is a functional of the error e = U - u, and

- E_G : Galerkin error
- E_C : Computational error
- E_Q : Quadrature error

The Galerkin error: E_G

• Residual:

$$R_i(U,t) = \dot{U}_i(t) - f_i(U(t),t)$$

• Stability factor:

$$S_i^{[q]}(T) = \int_0^T |\phi_i^{(q)}| \,\mathrm{dt}$$

• Error estimate (for mcG(q)):

$$E_{G} = \left| \int_{0}^{T} (R, \phi) \, \mathrm{dt} \right| = \left| \sum_{i=1}^{N} \sum_{j=1}^{M_{i}} \int_{I_{ij}} R_{i} (\phi_{i} - \pi_{k} \phi_{i}) \, \mathrm{dt} \right|$$
$$\leq \sum_{i=1}^{N} CS_{i}^{[q]}(T) \max_{[0,T]} \{k_{i}^{q_{i}} | R_{i}(U) | \}$$

The Computational Error: E_C

• Computational residual:

$$R_{i}^{C}(U,t) = \frac{1}{k_{ij}} \left[U(t_{ij}) - U(t_{i,j-1}) - \int_{I_{ij}} f_{i}(U,\cdot) \,\mathrm{dt} \right], \quad t \in I_{ij}$$

• Stability factor:

$$S_i^{[0]}(T) = \int_0^T |\phi_i| \,\mathrm{dt}$$

• Error estimate:

$$E_C \approx \sum_{i=1}^{N} S_i^{[0]} \max_{[0,T]} |R_i^C|$$

The Quadrature Error: E_Q

• Quadrature residual:

$$R_i^Q = \frac{1}{k_{ij}} \left[\tilde{\int}_{I_{ij}} f_i(U, \cdot) \, \mathrm{dt} - \int_{I_{ij}} f_i(U, \cdot) \, \mathrm{dt} \right], \quad t \in I_{ij}$$

• Stability factor:

$$S_i^{[0]}(T) = \int_0^T |\phi_i| \,\mathrm{dt}$$

• Error estimate:

$$E_Q \approx \sum_{i=1}^N S_i^{[0]} \max_{[0,T]} |R_i^Q|$$

Computational cost (complexity of output)

- Computational cost given by the product $S(T) \| u^{(p)} \|$
- Determined both by the stability/sensitivity of the model and the regularity of the solution

Quantitative classification according to stability:

The dual problem

The dual problem is given by

where

$$J(v_1, v_2, \cdot) = \int_0^1 \frac{\partial f}{\partial u} (sv_1 + (1 - s)v_2, \cdot) \, ds.$$

By choosing ψ and g, different functionals $L_{\psi,g}(e)$ can be estimated. Basic examples:

•
$$\psi \approx e(T)/\|e(T)\|$$
 and $g = 0$ gives $L_{\psi,g}(e) \approx \|e(T)\|$

•
$$\psi = (0, ..., 0, 1, 0, ..., 0)$$
 and $g = 0$ gives $L_{\psi,g}(e) = e_i(T)$

•
$$\psi = 0$$
 and $g = (1, \dots, 1)/(NT)$ gives $L_{\psi,g}(e) = \overline{e}$

The adaptive algorithm

1. Solve the primal problem with $S_i(T) = 1$ and

$$k_{ij} = \left(\frac{\text{TOL}}{CNS_i(T) \|R_i\|_{I_{ij}}}\right)^{1/q_i}$$

- 2. Solve the dual problem
- 3. Compute new stability factors $S_i(T)$
- 4. Compute the error estimate E
- 5. If $E \leq \text{TOL}$ then stop, otherwise go back to 1

Implementation

Implementation

- Implemented as a C++ library (part of DOLFIN)
- Mono-adaptive or multi-adaptive
- Newton or fixed-point
- User implements interface specified by ODE base class:

```
class ODE
{
public:
```

```
ODE(uint N);
```

virtual real f(const real u[], real t, uint i); virtual void f(const real u[], real t, real y[]); ...

Implementation

- Data stored in a "minimal" set of C arrays
- Build time slab: O(# elements)
- Interpolate $U_i(t)$: $\mathcal{O}(1)$

real* sa; // s --> start time t of sub slab s
real* sb; // s --> end time t of sub slab s
uint* ei; // e --> component index i of element e
uint* es; // e --> time slab s containing element e
uint* ee; // e --> previous element e of element e
uint* ed; // e --> first dependency d of element e
real* jx; // j --> value of dof j
int* de; // d --> element e of dependency d

Multi-adaptive profile (mcG(1))

Examples and benchmarks

Examples

- A mechanical multi-scale system
- The heat equation
- A system of reaction—diffusion equations
- Wave propagation through a narrow slit

A mechanical multi-scale system

$$\begin{cases} m_i \ddot{x}_i &= k(x_{i+1} - x_i) - kx_i, & i = 1, \\ m_i \ddot{x}_i &= k(x_{i+1} - x_i) - k(x_i - x_{i-1}), & 1 < i < N, \\ m_i \ddot{x}_i &= -k(x_i - x_{i-1}), & i = N \end{cases}$$

A mechanical multi-scale system

[solid: cG(q) dashed: mcG(q)]

The heat equation

 $\dot{u}(x,t) - \Delta u(x,t) = f(x,t)$

A system of reaction-diffusion equations

Two substances, A and B, distributed along [0,1] with concentrations u_1 and u_2 . A reacts to form B with B working as a catalyst.

 $A + 2B \rightarrow B + 2B$ $\begin{cases} \dot{u}_1 - \epsilon u_1'' = -u_1 u_2^2 \\ \dot{u}_2 - \epsilon u_2'' = u_1 u_2^2 \end{cases}$

A system of reaction-diffusion equations

A system of reaction-diffusion equations

Wave propagation through a narrow slit

Wave propagation through a narrow slit

- $k \sim h$
- Multi-adaptive speedup: 3.7 (theoretical 27)

Current status and future plans

Current status and future plans

• A new improved multi-adaptive solver is currently being developed as part **DOLFIN**:

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http://www.fenics.org/dolfin/
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- (Re-)implement dual problems and global error control
- Improve multi-adaptive preconditioners
- Integrate multi-adaptive solver with FFC/DOLFIN
- Testing, benchmarking, optimization

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