

ODE/DAE Integrators and Nonlinear Solvers

Presented to
ATPESC 2018 Participants

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Importance of Numerical ODE/DAE Solvers

Needed for almost all time-dependent simulation

- Analytic solutions are rarely of practical use
- Different problems require fundamentally different solution techniques
- Large differences in efficiency depending on the method used

Two main HPC ODE/DAE Solver Packages

- SUNDIALS - Lawrence Livermore National Laboratory
- PETSc - Argonne National Laboratory
- Trilinos - Sandia National Laboratory - has some limited integrators

$$M(t, u)u_t + F(t, u) = G(t, u)$$

$$u(0) = g$$

- $M(t, u)$ - mass matrix
- $F(t, u)$ - stiff portion of equation
- $G(t, u)$ - nonstiff portion

Linear example

$$u_t - Au = 0$$

$$\frac{u^{n+1} - u^n}{\Delta t} = G(t, u^n)$$

$$u^{n+1} = u^n + \Delta t G(t, u^n)$$

$$u_t = u_{xx}$$

$$u(0) = u(2\pi) = 0$$

$$u(0, x) = \sum_m \alpha_m \sin(mx)$$

Analytic solution

$$u(t, x) = \sum_m \alpha_m e^{-m^2 t} \sin(mx)$$

Semi-discrete form

$$(u_i)_t = \frac{u_{i+1} - 2u_i + u_{i-1}}{(\Delta x)^2}$$

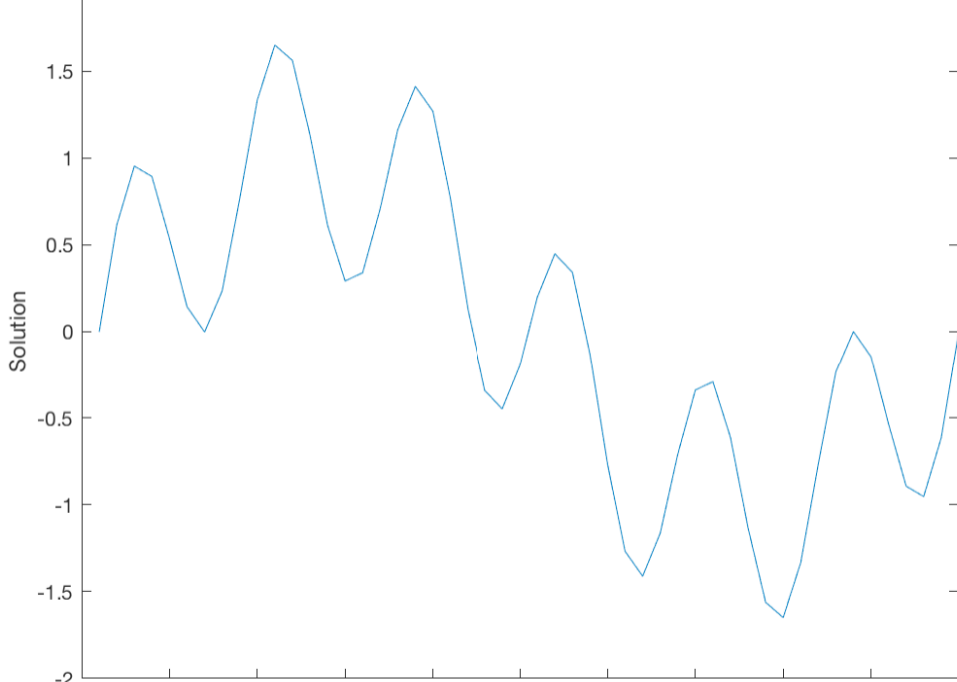
$$u_0 = u_N = 0$$

Fully discrete form with Euler's method

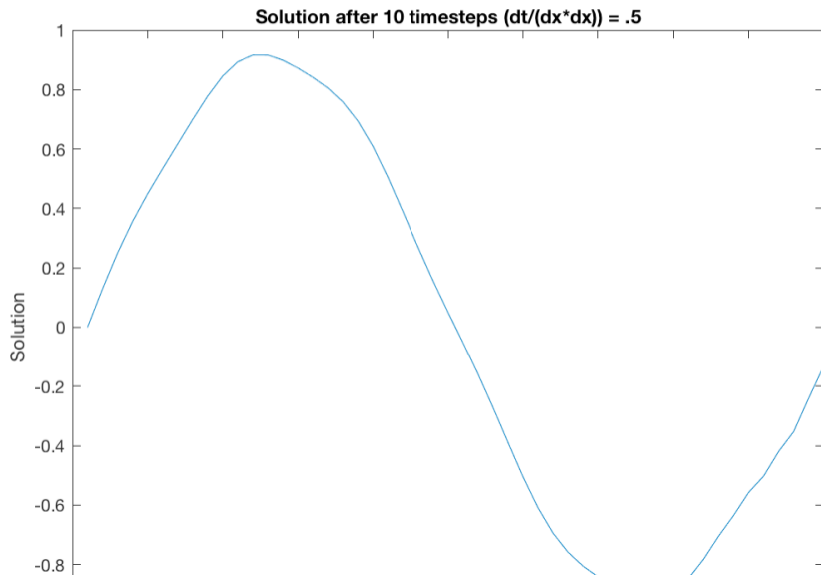
$$u_i^{n+1} = u_i^n + \frac{\Delta t}{(\Delta x)^2} (u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

$$u_i^0 = g_i$$

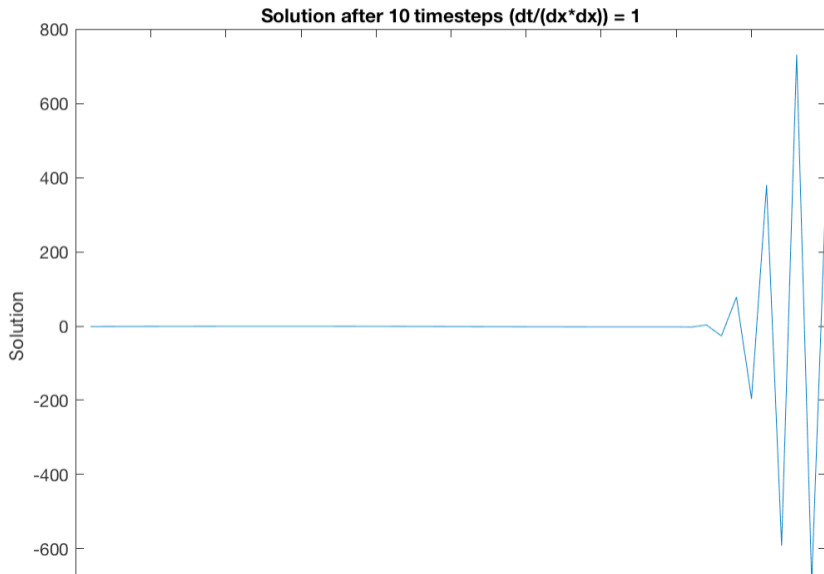
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Stability: Stable Solution



Stability: Unstable Solution



$$u_i^{n+1} = u_i^n + \frac{\Delta t}{(\Delta x)^2} (u_{i+1}^n - 2u_i^n + u_{i-1}^n)$$

Consider a solution of the form $u_i^n = \sin(mi\Delta x)$. Does it grow or shrink as we integrate in time?

$$\sin(mi\Delta x) + \frac{\Delta t}{(\Delta x)^2} (\sin(m(i+1)\Delta x) - 2\sin(mi\Delta x) + \sin(m(i-1)\Delta x))$$

$$\begin{aligned} \sin(mi\Delta x) + \frac{2\Delta t}{(\Delta x)^2} (\sin(mi\Delta x) \cos(m\Delta x) + \cos(mi\Delta x) \sin(m\Delta x) - \\ 2\sin(mi\Delta x) + \sin(mi\Delta x) \cos(m\Delta x) - \cos(mi\Delta x) \sin(m\Delta x)) \end{aligned}$$

$$\sin(mi\Delta x) + \frac{2\Delta t}{(\Delta x)^2} \sin(mi\Delta x) (\cos(m\Delta x) - 1)$$

$$-1 \leq \sin(mi\Delta x) \left(1 + \frac{2\Delta t}{(\Delta x)^2} (\cos(m\Delta x) - 1) \right) \leq 1$$

$$-1 \leq 1 + \frac{2\Delta t}{(\Delta x)^2} (-2)$$

$$\frac{\Delta t}{(\Delta x)^2} \leq 1/2.$$

Courant-Friedrichs-Lewy (CFL) condition

- “The stepsize needed to maintain stability of the forward Euler method is much smaller than that required to represent the solution accurately” – U. Ascher and L. Petzold

Backward Euler: Implicit Schemes

$$M(t, u^n) \frac{u^{n+1} - u^n}{\Delta t} + F(t, u^{n+1}) = 0$$

Unconditionally stable, proof, do the same analysis as for Euler and observe that the solution never grows independent of Δt and Δx .

Treat the stiff portion of the equation implicitly and the rest explicitly

$$M(t, u^n) \frac{u^{n+1} - u^n}{\Delta t} + F(t, u^{n+1}) = G(t, u^n)$$

$$M(t, u, w)u_t + F(t, u, w) = G(t, u, w)$$

$$H(t, u, w) = 0$$

$$u(0) = g$$

$$w(0) = h$$

Multistep and Multistage Schemes

Approximate the time derivative with higher order finite differences

- multistep - Use previous solutions (steps) to approximate the time derivatives
- multistage - Use new intermediate solutions (stages) to approximate the time derivatives

Adaptive Time Stepping:

- Estimate the local truncation error induced by the finite differencing in time at each time step by integrating again with a higher order scheme
- Adjust the time-step to keep the local truncation error below a prescribed value
- May decrease or increase the time step
- Can lead to much more efficient (and accurate) computation of the solution

Exercise I: Properties of Time Steppings:

https://xsdk-project.github.io/ATPESC2018HandsOnLessons/lessons/time_integrators

Newton's Method:

For implicit methods one needs to solve nonlinear systems

$$Q(w) = 0.$$

From Taylor series

$$Q(w + \delta w) = Q(w) + J_Q(w)\delta w + \dots = 0$$

$$\delta w = -J_Q(w)^{-1}Q(w)$$

$$w^{m+1} = w^m - J_Q(w^m)^{-1}Q(w^m)$$

Exercise II: Quadratic and Mesh Independence convergence of Newton's method

PETSc/TAO:

Portable, Extensible Toolkit for Scientific
Computation / Toolkit for Advanced Optimization

Scalable algebraic solvers for PDEs. Encapsulate parallelism in high-level objects. Active & supported user community. Full API from Fortran, C/C++, Python.

Optimization

Time Integrators

Nonlinear Algebraic Solvers

Krylov Subspace Solvers

Preconditioners

Domain-
Specific
Interfaces

Networks

Quadtree / Octree

Unstructured Mesh

Structured Mesh

Vectors

Index Sets

Matrices

Computation &
Communication Kernels

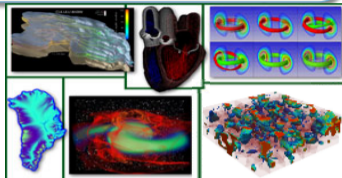
- **Easy customization and composability of solvers at runtime**

- Enables optimality via flexible combinations of physics, algorithmics, architectures
- Try new algorithms by composing new/existing algorithms (multilevel, domain decomposition, splitting, etc.)

- **Portability & performance**

- Largest DOE machines, also clusters, laptops
- Thousands of users worldwide

Argonne
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PETSc provides the backbone of diverse scientific applications. clockwise from upper left: hydrology, cardiology, fusion, multiphase steel, relativistic matter, ice sheet modeling



<https://www.mcs.anl.gov/petsc>

PETSc: Platform for experimentation

- *No optimality without interplay among physics, algorithmics, and architectures*
- **Need algebraic solvers to be:**
 - **Composable:** Separately developed solvers may be easily combined, by non-experts, to form a more powerful solver.
 - **Hierarchical:** Outer solvers may iterate over all variables for a global problem, while inner solvers handle smaller subsets of physics, smaller physical subdomains, or coarser meshes.
 - **Nested:** Outer solvers call nested inner solvers.
 - **Extensible:** Users can easily customize/extend.
- Many solver configurations can be set at runtime to avoid needing to recompile.

PETSc/TAO capabilities

Functionality

Optimization	
Time Integrators	
Nonlinear Algebraic Solvers	
Krylov Subspace Solvers	
Preconditioners	
Domain-Specific Interfaces	Networks
	Quadtree / Octree
	Unstructured Mesh
	Structured Mesh
Vectors	Index Sets Matrices
Computation & Communication Kernels	

More Details (Algorithms, Data Structures, etc.)

PDE Constrained	Adjoint Based	Derivative Free	Others
Pseudo-transient General Linear	Runge-Kutta IMEX	Strong Stability Preserving Rosenbrock-W	Others
Line Search Newton Trust Region Newton	Quasi-Newton (BFGS) Nonlinear Multigrid (FAS)	Nonlinear Gauss Seidel Successive Substitutions	Nonlinear CG Active Set VI
Pipeline methods Hierarchical Krylov	GMRES LSQR	Chebyshev SYMMLQ	BiCG-Stabilized TFQMR Others
Blocks (by field) Algebraic Multigrid	Additive Schwarz Geometric Multigrid	ILU/ICC App-specific	Schur Complement Others
Infrastructure networks, e.g., electrical, gas, water			
Structured mesh refinement			
Complex domains with finite element and finite volume discretizations			
Simple domains and discretizations, e.g., finite difference methods			
Compressed Sparse Row (AIJ) Symmetric Block AIJ		Block AIJ Dense	Matrix Blocks (MatNest) GPU and Pthread Matrices
MPI, OpenMP, MPI-IO, CUDA, Pthreads, BLAS, LAPACK, etc.			

Take Away

- PETSc and SUNDIALS provide a wide variety of high quality, scalable ODE/DAE integrators
- PDEs can be converted to ODEs/DAE via discretization in space and then solved using ODE/DAE libraries
- Stiffness is an important property of ODEs and effects the appropriate schemes to use
- Adaptive time-stepping provides an inexpensive way to to efficiently integrate ODEs/DAEs