

Discretization of PDEs and Tools for the Parallel Solution of the Resulting Systems

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Topics

Projection in
Scientific Computing

(lecture 1)

Sparse matrices,
parallel implementations

(lecture 3)

**PDEs, Numerical
solution, Tools, etc.**

(lecture 2)

Iterative Methods

(lectures 4 and 5)

- **Part I**
Partial Differential Equations
- **Part II**
Mesh Generation and Load Balancing
- **Part III**
Tools for Numerical Solution of PDEs

Part I

Partial Differential Equations

Mathematical Model:

- a representation of the **essential aspects** of an existing system which presents knowledge of that system **in usable form** (Eykhoff, 1974)

Mathematical Modeling:

Real world

Model

Navier-Stokes equations:



$$\begin{aligned} \longleftrightarrow \quad \nabla \cdot u &= 0 \\ \frac{\partial u}{\partial t} &= -(u \cdot \nabla)u - \frac{1}{\rho} \nabla p + \nu \nabla^2 u + f \\ B.C. &, \text{ etc.} \end{aligned}$$

We are interested in models that are

- **Dynamic**

i.e. account for changes in time

- **Heterogeneous**

i.e. account for heterogeneous systems

Typically represented with

- **Partial Differential Equations**

How can we model for e.g. **Heat Transfer**?

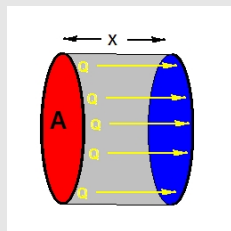
- Heat
 - * a form of energy (thermal)
- Heat Conduction
 - * transfer of thermal energy from a region of higher temperature to a region of lower temperature
- Some notations

Q : amount of heat

k : material conductivity

T : temperature

A : area of cross-section

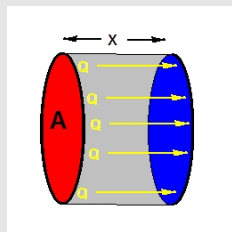


The Law of Heat Conduction

$$\frac{\Delta Q}{\Delta t} = k A \frac{\Delta T}{\Delta x}$$

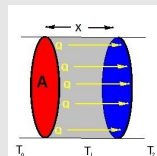
Change of heat is proportional to the gradient of the temperature and the area A of the cross-section.

- Q : amount of heat
- k : material conductivity
- T : temperature
- A : area of cross-section



Consider 1-D heat transfer in a thin wire

- so thin that T is piecewise constant along the slides, i.e. $T_0(t)$, $T_1(t)$, $T_2(t)$, etc.
- ideally insulated



Let us write a balance for the temperature at T_1 for time $t + \Delta t$

$$T_1(t + \Delta t) = ?$$

$$\begin{aligned} T_1(t + \Delta t) &\approx T_1(t) \\ &+ k\Delta t \frac{(T_2(t) - T_1(t))}{(\Delta x)^2} \\ &+ k\Delta t \frac{(T_0(t) - T_1(t))}{(\Delta x)^2} \\ &= T_1(t) + k\Delta t \frac{T_2(t) - 2T_1(t) + T_0(t)}{(\Delta x)^2} \end{aligned}$$

Take $\lim_{\Delta x, \Delta t \rightarrow 0}$

$$\Rightarrow \frac{\partial T}{\partial t} = k \frac{\partial^2 T}{\partial x^2} \quad (\text{Exercise})$$

Extend to 2-D and put a source term f to easily get

$$\frac{\partial T}{\partial t} = k \left(\frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} \right) + f \equiv k \Delta T + f$$

Known as the **Heat equation**

- Poisson equation (**elliptic**)

$$\Delta u = f$$

- Heat equation (**parabolic**)

$$\frac{\partial T}{\partial t} = k \Delta T + f$$

- Wave equation (**hyperbolic**)

$$\frac{1}{\nu^2} \frac{\partial^2 u}{\partial t^2} = \Delta u + f$$

Classification of PDEs

For a general second-order PDE in 2 variables:

$$Au_{xx} + Bu_{xy} + Cu_{yy} + \dots = 0$$

Elliptic:

- if $B^2 - 4AC < 0$
- process in equilibrium (no time dependence)
- easy to discretize but challenging to solve

Parabolic:

- if $B^2 - 4AC = 0$
- processes evolving toward steady state

Hyperbolic:

- if $B^2 - 4AC > 0$
- not evolving toward steady state
- difficult to discretize (support discontinuities) but easy to solve in characteristic form

How do we solve them?

Numerical solution approaches:

- Finite difference method
- Finite element method
- Finite volume method
- Boundary element method

- use finite differences to approximate differential operators
- one of the simplest and extensively used method in solving PDEs
- the error, called truncation error, is due to finite approximation of the Taylor series of the differential operator

A Finite Difference Method Example

Consider the 2-D Poisson equation:

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f$$

The idea, **first in 1-D**:

- Use Taylor series to approximate $\frac{d^2 u}{dx^2}(x)$ with $u(x)$, $u(x+h)$, $u(x-h)$

$$u(x+h) = u(x) + h \frac{du}{dx}(x) + \frac{h^2}{2} \frac{d^2 u}{dx^2}(x) + \frac{h^3}{3!} \frac{d^3 u}{dx^3}(x) + \mathcal{O}(h^4)$$

$$u(x-h) = u(x) - h \frac{du}{dx}(x) + \frac{h^2}{2} \frac{d^2 u}{dx^2}(x) - \frac{h^3}{3!} \frac{d^3 u}{dx^3}(x) + \mathcal{O}(h^4)$$

$$\Rightarrow \frac{d^2 u}{dx^2}(x) = \frac{1}{h^2} (u(x+h) + u(x-h) - 2u(x)) + \mathcal{O}(h^2)$$

Similarly in 2-D

- Use Taylor series to approximate $\Delta u(x, y)$ with $u(x, y), u(x + h, y), u(x - h, y), u(x, y + h), u(x, y - h)$.

$$u(x + h, y) = u(x, y) + h \frac{\partial u}{\partial x}(x, y) + \frac{h^2}{2} \frac{\partial^2 u}{\partial x^2}(x, y) + \frac{h^3}{3!} \frac{\partial^3 u}{\partial x^3}(x, y) + \mathcal{O}(h^4)$$

$$u(x - h, y) = u(x, y) - h \frac{\partial u}{\partial x}(x, y) + \frac{h^2}{2} \frac{\partial^2 u}{\partial x^2}(x, y) - \frac{h^3}{3!} \frac{\partial^3 u}{\partial x^3}(x, y) + \mathcal{O}(h^4)$$

$$u(x, y + h) = u(x, y) + h \frac{\partial u}{\partial y}(x, y) + \frac{h^2}{2} \frac{\partial^2 u}{\partial y^2}(x, y) + \frac{h^3}{3!} \frac{\partial^3 u}{\partial y^3}(x, y) + \mathcal{O}(h^4)$$

$$u(x, y - h) = u(x, y) - h \frac{\partial u}{\partial y}(x, y) + \frac{h^2}{2} \frac{\partial^2 u}{\partial y^2}(x, y) - \frac{h^3}{3!} \frac{\partial^3 u}{\partial y^3}(x, y) + \mathcal{O}(h^4)$$

$$\Rightarrow \Delta u(x, y) = \frac{1}{h^2} (u(x + h, y) + u(x - h, y) + u(x, y + h) + u(x, y - h) - 4u(x, y)) + \mathcal{O}(h^2)$$

- Remember the slides from the previous lecture

▶ <http://www.cs.utk.edu/~dongarra/WEB-PAGES/SPRING-2013/Lect10-2013.pdf>

Main pluses/minuses of FEM vs FDM

- FEM can handle complex geometries
- FDM is easy to implement

A Finite Element Method Example

Consider the 1-D Dirichlet problem:

$$(1) \quad u''(x) = f(x), \quad \text{for } x \in (0, 1)$$

and the Dirichlet boundary condition

$$u(0) = u(1) = 0$$

Weak or Variational formulation:

- Multiply (1) by smooth v and integrate over $(0,1)$

$$\int_0^1 f(x)v(x)dx = \int_0^1 u''(x)v(x)dx$$

- Integrate by parts the above RHS

$$\begin{aligned} \int_0^1 u''(x)v(x)dx &= u'(x)v(x)|_0^1 - \int_0^1 u'(x)v'(x)dx \\ &= - \int_0^1 u'(x)v'(x)dx \equiv -a(u, v) \end{aligned}$$

- Variational formulation: Find $u \in H_0^1(0, 1)$ such that

$$\int_0^1 f(x)v(x)dx = -a(u, v) \text{ for } \forall v \in H_0^1(0, 1)$$

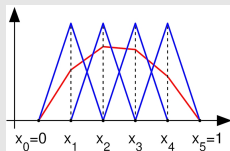
A Finite Element Method Example

Discretization (Galerkin FE problem):

- Replace $H_0^1(0,1)$ with finite dimensional subspace V

Shown is a 4 dimensional space V (basis in blue) and a linear combination (in red)

$$v_k(x) = \begin{cases} \frac{x-x_{k+1}}{x_k-x_{k+1}} & \text{if } x \in [x_{k-1}, x_k], \\ \frac{x_{k+1}-x}{x_{k+1}-x_k} & \text{if } x \in [x_k, x_{k+1}], \\ 0 & \text{otherwise,} \end{cases}$$



What is the matrix form of the problem
(Exercise)

Part II

Mesh Generation and Load Balancing

slides at: [▶ http://www.cs.utk.edu/~dongarra/WEB-PAGES/SPRING-2013/Lect12-p2.pdf](http://www.cs.utk.edu/~dongarra/WEB-PAGES/SPRING-2013/Lect12-p2.pdf)

Part III

Tools for Numerical Solution of PDEs

Challenges:

- Software Complexity
- Data Distribution and Access
- Portability, Algorithms, and Data Redistribution

Read more in Chapter 21

There is software; to mention a few packages:

■ Overture

OO framework for PDEs in complex moving geometry

■ PARASOL

Parallel, sparse matrix solvers; in Fortran 90

■ SAMRAI

OO framework for parallel AMR applications

■ Hype

Large sparse linear solvers and preconditioners

■ PETSc

Tools for numerical solution of PDEs

■ FFTW

parallel FFT routines

■ Diffpack

OO framework for solving PDEs

■ Doug

FEM for elliptic PDEs

■ POOMA

OO framework for HP applications

■ UG

PDEs on unstructured grids using multigrid

See also:

▶ <http://www.mgnet.org/>

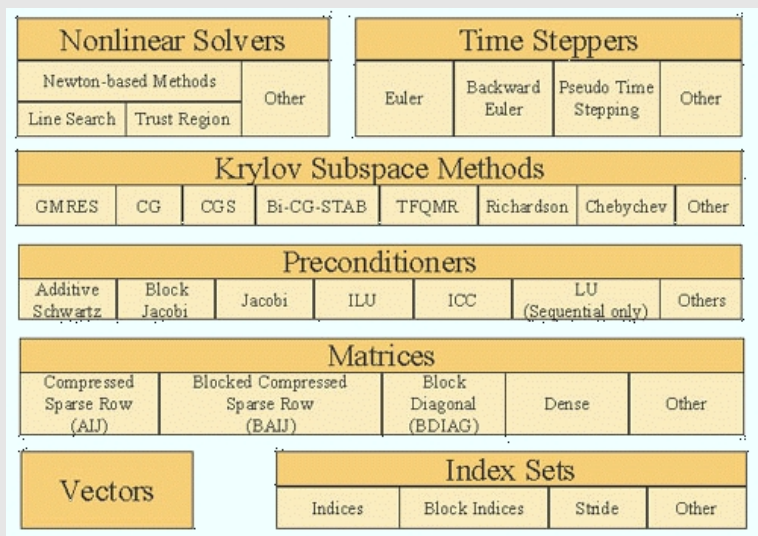
▶ <http://www.nhse.org/>

▶ <http://www.netlib.org/>

PETSc: Portable, Extensible Toolkit for Scientific computation

- for large-scale sparse systems
- facilitate extensibility
- provides interface to external packages, e.g.
BlockSolve95, ESSL, Matlab, ParMeTis,
PVODE, and SPAI.
- programed in C, usable from Fortran and C++
- uses MPI for all parallel communication
 - in a distributed-memory model
 - user do communication on level higher than MPI
- Computation and communication kernels:
MPI, MPI-IO, BLAS, LAPACK

PETSc's Main Numerical Components



more info at: <http://acts.nersc.gov/petsc/>

A brief overview of Numerical PDEs and related issues

- Mathematical modeling
- PDEs for describing changes in physical processes
- More specific discretization examples
 - Finite Differences (natural)
 - FEM
 - reinforce the idea and application of Petrov-Galerkin conditions
- Issues related to mesh generation and load balancing and importance in HPC
 - Adaptive methods
- Software