Scientific Computing Lecture 2 Delft University of Technology

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Recap Last Week

- Model problem
- Uniqueness
- Real-world Applications
- Setting up the Numerical Model (Ch. 3)

Sources of Errors

STEP 1

Setting up a Model (mimick reality using ODEs/PDEs)

STEP 2

Numerical treatment (discretization, translating continous model to discrete version and design algorithms) STEP 3

Implementation (computer language, data structures, distributed/parallel computing)

- Modelling error
- Error in data

- Discretization error
- Truncation error
- Round-off error

 Programming errors

Sources of Errors - Discretization

Discretization error



Only n values are obtained: gives an approximate solution curve.

§ 3.4: Model Problem (MP) - 2D Poisson

- Continuous Poisson operator: $-\triangle u := -\frac{\partial^2 u}{\partial x^2} \frac{\partial^2 u}{\partial u^2} = f$ with $u(x, y), f(x, y) \in \mathbb{R}$
- Continuous Domain



- Boundary conditions:
 - Dirichlet: u(x, y) = b(x, y)Neumann: $\frac{\partial u}{\partial n} = c(x, y)$

§ 3.4: Continuous 1D Poisson MP

Given the domain $\Omega = (0, 1)$ with boundary $\Gamma = \partial \Omega$ and outward normal **n**, solve for u(x)

$$-rac{d^2u(x)}{dx^2}=f(x)$$
 on Ω

with either Dirichlet BCs

$$u(x) = b(x)$$
 on Γ

or Neumann BCs

$$rac{\partial u(x)}{\partial n} =
abla u(x) \cdot \mathbf{n} = c(x) ext{ on } \Gamma$$

Continuous eigenfunction and eigenvalues with homog. Dirichlet

$$u^{[k]}(x)=\sin(k\pi x),\lambda_k=k^2\pi^2 ext{ for }k\in\mathbb{N},k
eq 0$$

§ 3.5: Grid Construction

• 1D Discrete Grid
$$G_h$$

 $x_i = (i-1)h, i = 1, ..., N+1$ where $h = \frac{1}{N}$ st $x_1 = 0, x_{N+1} = 1$
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• 2D Discrete Grid G_h $x_i = (i-1)h, y_j = (j-1)h, i, j = 1, 2, ..., N + 1$ $(x_1, y_1) = (0, 0), (x_1, y_{N+1}) = (0, 1),$ etc.





How can we represent continuous differential operators on a discrete grid? Finite Differences!

- Numerical approximation using FD only defined in the grid points
- Grid functions for all $(x_i, y_j) \in G_h$: $u(x, y) \approx u(x_i, y_j) \approx u_{i,j}^h, f(x, y) \approx f(x_i, y_j) \approx f_{i,j}^h$ $b(x_i, y_j) \approx b_{i,j}^h, c(x_i, y_j) \approx c_{i,j}^h$



We now (recall) several finite difference formulas Forward difference:

$$\left.\frac{\partial u}{\partial x}\right|_{i,j} = \frac{u_{i+1,j} - u_{i,j}}{\Delta x} + O(h)$$

Backward difference:

$$\left.\frac{\partial u}{\partial x}\right|_{i,j} = \frac{u_{i,j} - u_{i-1,j}}{h} + O(h)$$

Central differences:

$$\frac{\partial u}{\partial x}\Big|_{i,j} = \frac{u_{i+1,j} - u_{i-1,j}}{2h} + O\left[(h^2)\right]$$
$$\frac{\partial^2 u}{\partial x^2}\Big|_{i,j} = \frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{h^2} + O\left[(h^2)\right]$$

Similar formulas for
$$\frac{\partial u}{\partial y}$$
 and $\frac{\partial^2 u}{\partial y^2}$
Forward difference:
 $\frac{\partial u}{\partial y}\Big|_{i,j} \simeq \frac{u_{i,j+1} - u_{i,j}}{h}$

Backward difference:

$$\left.\frac{\partial u}{\partial y}\right|_{i,j} \simeq \frac{u_{i,j} - u_{i,j-1}}{h}$$

Central differences:

$$\frac{\partial u}{\partial y}\Big|_{i,j} \simeq \frac{u_{i,j+1} - u_{i,j-1}}{2h}$$
$$\frac{\partial^2 u}{\partial y^2}\Big|_{i,j} \simeq \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{(h)^2}$$

Transformed MP: 2D Poisson (internal nodes) Grid: $G_h = \{(x_i, y_i) \mid x_i = (i-1)h, y_i = (j-1)h; 1 \le i, j \le N+1\},\$ Continuous MP: $-\frac{\partial^2 u}{\partial x^2} - \frac{\partial^2 u}{\partial x^2} = f$ $\tfrac{\partial^2 u}{\partial x^2}\left(x_i, y_i\right) = \tfrac{u_{i-1,j}^h - 2u_{i,j}^h + u_{i+1,j}^h}{h^2} + \mathcal{O}\left(h^2\right) \text{ for } 2 \leq i,j \leq N, (x_i, y_j) \in \mathcal{G}_h$ $\frac{\partial^{2} u}{\partial v^{2}}\left(x_{i}, y_{j}\right) = \frac{u_{i,j-1}^{h} - 2u_{i,j}^{h} + u_{i,j+1}^{h}}{h^{2}} + \mathcal{O}\left(h^{2}\right) \text{ for } 2 \leq i, j \leq N, (x_{i}, y_{j}) \in G_{h}$ Transformed MP: $\frac{-u_{i,j-1}^{h}-u_{i-1,j}^{h}+4u_{i,j}^{h}-u_{i+1,j}^{h}-u_{i,j+1}^{h}}{\frac{1}{2}}=f_{i,i}^{h} \text{ for } 2\leq i,j\leq N$

§ 3.5: Stencil Notation On a subset of notes, Transformed MP:

 $\frac{-u_{i,j-1}^{h}-u_{i-1,j}^{h}+4u_{i,j}^{h}-u_{i+1,j}^{h}-u_{i,j+1}^{h}}{h^{2}}=f_{i,j}^{h} \text{ for } 2\leq i,j\leq N$

is represented by a 5-point stencil for each internal $(x_i, y_j) \in G_h$:



§ 3.5: Boundary Conditions



• Set of boundary points in x-direction

•
$$y = 0: \left\{ u_{1,1}^h, u_{2,1}^h, \dots u_{N+1,1}^h \right\}$$

• $y = 1: \left\{ u_{1,N+1}^h, u_{2,N+1}^h, \dots u_{N+1,N+1}^h \right\}$
Set of boundary points in v-direction

•
$$x = 0: \left\{ u_{1,1}^h, u_{1,2}^h, \dots, u_{1,N+1}^h \right\}$$

§ 3.5: Boundary Conditions

No elimination: each boundary node becomes an equation for which the RHS $f_{i,i}^h$ is replaced by $b_{i,i}^h$



With elimination: the known boundary values are directly substituted and $b_{i-1,i}^{h}$ is added to the RHS



No elimination

 x−line lexicographic ordening of internal and boundary nodes node (i, j) is assigned global index I = i + (j − 1)(N + 1) for 1 ≤ i, j ≤ N + 1



Figure 3.4: x-lexicographic using I

No elimination

• Group known and unknown grid values $f_{i,j}^h$ and $u_{i,j}^h$ into column vectors \mathbf{u}^h and \mathbf{f}^h of size $(N+1)^2$

Example: 2D Poisson + Dirichlet BCs

Given the domain $\Omega=(0,1)$ with boundary $\Gamma=\partial\Omega,$ discretize

$$-\frac{\partial^2 u}{\partial x^2} - \frac{\partial^2 u}{\partial y^2} = f \text{ on } \Omega$$
$$u(x, y) = b(x, y) \text{ on } \Gamma$$

using second-order FD and write it in the form $A^h u^h = f^h$ using no elimination.

Take N = 2 and symmetrize A^h .

Example: 2D Poisson + Dirichlet BCs

We get system of equations (red: internal node):

 $u_{11} = b_{11}$ $u_{21} = b_{21}$ $u_{31} = b_{31}$ $u_{12} = b_{12}$ $-u_{12} - u_{21} - 4u_{22} - u_{23} - u_{32} = f_{22}$ h^2 $u_{23} = b_{23}$ $u_{13} = b_{13}$ $u_{23} = b_{23}$ $u_{33} = b_{33}$





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STEP 2

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Summary and Next Week

1 Transform continuous model to discrete numerical model

- Transform domain Ω into discrete domain/grid G_h with grid points, i.e. from (x, y) to (x_i, y_j)
- Discretize solution function u(x, y) to $u(x_i, y_j)$ on the grid points
- Discretize derivatives of the PDE at grid points using Finite Differences
- Rewrite system of equations into matrix-vector format
- 2 Analyze matrix properties (this dictates which numerical algorithm to use (next week))
 - Symmetry
 - Eigenvalues
 - Positive Definiteness
 - Conditioning (remember round-off error example last week!)