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Overview

Model Problems

- Relaxation Methods
- Error convergence
- Multiple grids
- Performance
- Theoretical Considerations

Differential Equation in 1D

-u''(x) + au(x) = f(x)

for 0 < x < 1, a > 0Boundary: u(0) = u(1) = 0

 Partition continuous problem into n subintervals by "sampling" it at the grid points x_i = jh, with h = 1/n

Grid Ω^h:



Second order finite difference approximation

$$\frac{-v_{j-1} + 2v_j - v_{j+1}}{h^2} + av_j = f(x_j) \text{ with } 1 \le j \le n-1$$

 $v_0 = v_n = 0$; with **v** being the approximate solution to **u** Written in Matrix-Vector form

$$\frac{1}{h^{2}}\begin{bmatrix} 2+ah^{2} & -1 \\ -1 & 2+ah^{2} & -1 \\ & & \ddots \\ & & -1 & 2+ah^{2} \end{bmatrix} \cdot \begin{bmatrix} v_{1} \\ \vdots \\ \vdots \\ v_{n-1} \end{bmatrix} = \begin{bmatrix} f(x_{1}) \\ \vdots \\ \vdots \\ f(x_{n-1}) \end{bmatrix} = \begin{bmatrix} f_{1} \\ \vdots \\ \vdots \\ f_{n-1} \end{bmatrix}$$

Written compactly: Av = f

(Elliptic) Partial Differential Equation

 Sampled with a two-dimensional grid (n-1,m-1 interior grid points)



Sampling results in difference approximation

$$\frac{-v_{i-1,j} + 2v_{ij} - v_{i+1,j}}{h_x^2} + \frac{-v_{i,j-1} + 2v_{ij} - v_{i,j+1}}{h_y^2} + av_{ij} = f_{ij} \quad \text{with } 1 \le j \le n-1$$

$$v_{i0} = v_{in} = v_{0j} = v_{mj} = 0, \qquad 1 \le i, j \le m-1$$

Written in Matrix-Vector form



B looks almost like matrix from 1D - Problem - dimension is $(n-1) \times (n-1)$ I is a $(n-1) \times (n-1)$ identity matrix \Rightarrow Dimension of matrix is $(m-1) \cdot (n-1) \times (m-1) \cdot (n-1)$

Model Problem 2D

Example: System for a=0, n=4, h=1



Again, written compactly: Av = f

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- To solve the PDE, $\mathbf{u} = A^{-1}\mathbf{f}$ is too complicated
- Based on an estimated solution $\mathbf{v}^{(0)} \rightarrow$ find better solution $\mathbf{v}^{(1)}$ in next step
- Reduces norm of the error $\mathbf{e} = \mathbf{u} \mathbf{v}$
- Use residual r = f − Av as a measure Relationship error / residual: Ae = r
 For exact solution v = u ⇒ r = 0
- For the following, split matrix A = D L U
 D: diagonal of A; L/U: lower/upper triangular part of A

- General approximation: $\mathbf{v}^{(1)} = \mathbf{v}^{(0)} + B\mathbf{r}^0$
 - Try to find a B "close" to A^{-1} , as $\mathbf{u} \mathbf{v} = A^{-1}\mathbf{r}$
- Jacobi scheme / Simultaneous displacement
 - jth component of v is calculated using the two neighbours from previous step

 $v_{j}^{(1)} = \frac{1}{2} (v_{j-1}^{(0)} + v_{j+1}^{(0)} + h^{2} f_{j}), \qquad 1 \le j \le n-1$ Jacobi iteration matrix : $\mathbf{R}_{J} = D^{-1} (L+U)$ $v^{(1)} = R_{J} v^{(0)} + D^{-1} f$

 Solves the PDE locally (compare original problem: -u_{j-1} + 2u_j - u_{j+1} = h²f_j)

- Weighted or damped Jacobi method
 - Weighting factor $0 < \omega < 1$

 $v_{j}^{(1)} = (1 - \omega)v_{j}^{(0)} + \omega v_{j}^{*}, \qquad 1 \le j \le n - 1$ with $v_{j}^{*} = \frac{1}{2}(v_{j-1}^{(0)} + v_{j+1}^{(0)} + h^{2}f_{j})$ (like above) Weighted Jacobi iteration matrix : $\mathbf{R}_{\omega} = (1 - \omega)I + \omega R_{J}$ $v^{(1)} = R_{\omega}v^{(0)} + \omega D^{-1}f$

- Gauss-Seidel
 - Like Jacobi, but components updated immediately
 - Reduces storage requirements

$$v_{j} \leftarrow \frac{1}{2}(v_{j-1} + v_{j+1} + h^{2}f_{j}), \leftarrow \text{meaning "overwrite"}$$

formally $: v_{j}^{(1)} = \frac{1}{2}(v_{j-1}^{(1)} + v_{j+1}^{(0)} + h^{2}f_{j})$

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- Simplified problem: Au = 0
 ⇒ v should converge to 0, and e = v
- In what way does weighted Jacobi decrease the error?
 Analyse eigenvectors of iteration matrix
- Eigenvectors w_k of matrices A and R_{ω} $w_{k,j} = \sin\left(\frac{jk\pi}{n}\right)$, with $1 \le k \le n-1$, $0 \le j \le n$
 - Vector w_k is also the kth Fourier mode
- Eigen values λ_k of matrix R_ω (generally: $R_\omega w_k = \lambda_k w_k$) $\lambda_k(R_\omega) = 1 - 2\omega \sin^2 \left(\frac{k\pi}{2n}\right)$, with $1 \le k \le n - 1$
 - For $0 \le \omega < 1 \Rightarrow |\lambda_k| < 1$, iteration converges

- **Eigen values** $\lambda_k(R_\omega) = 1 2\omega \sin^2\left(\frac{k\pi}{2n}\right)$, with $1 \le k \le n-1$
- Smooth, low-frequency Fourier modes of $e: 1 \le k \le \frac{1}{2}n$
 - $|\lambda_k|$ is close to 1 \Rightarrow no satisfactory damping
- Oscillatory, high-frequency modes: $\frac{1}{2}n \le k \le n-1$
 - For the right ω , $|\lambda_k|$ is close to $0 \Rightarrow$ good damping
 - Optimal damping for $\omega = \frac{2}{3}$



Damping diagram for the weighted Jacobi method



- Oscillatory modes of the error are removed quite well
- Smooth modes are hardly damped.

Error convergence

- Example code in MATLAB
 - Grid n = 64
 - Initial error modes 2 and 16
 - Solves –u''(x) = 0

```
n = 64;
% components of A
D = 2 * diag(ones(n-1,1),0); U = d
% iteration matrices
w=2/3; RJ = inv(D) * (L+U); RW =
% init f=0, v with modes 2 and 16
f = zeros(n-1,1);
v = transpose(sin((1:n-1) * 2 * pi
plot(v); hold on
% do 10 iterations
for i = 1:10
v = RW*v + 0; end
plot(v);
```



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- Fundamental idea of multigrid
 - Make smooth modes look oscillatory!
 - Smooth mode on Ω^h looks oscillatory on grid Ω^{nh}
 - A "hierarchy of discretizations" is used to solve the problem of small damping for smooth modes



Multiple Grids

- Intergrid Transfer coarse \rightarrow fine: Interpolation
 - $\Omega^{2h} \rightarrow \Omega^{h}$, "Upsampling"
 - Linear interpolation is effective

$$\begin{aligned} v_{2j}^{h} &= v_{j}^{2h} \\ v_{2j+1}^{h} &= \frac{1}{2} (v_{j}^{2h} + v_{j+1}^{2h}), \quad 0 \leq j \leq \frac{n}{2} - 1 \end{aligned}$$

$$I_{2h}^{h} \mathbf{v}^{2h} = \frac{1}{2} \begin{bmatrix} 1 & & \\ 2 & & \\ 1 & 1 & \\ & 2 & \\ & 1 & 1 \\ & & 2 \\ & & & 1 \end{bmatrix} \cdot \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}_{2h} = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \\ v_6 \\ v_7 \end{bmatrix}_{h} = \mathbf{v}^{h}$$

Multiple Grids

- Intergrid Transfer fine \rightarrow coarse: **Restriction**
 - $\Omega^h \rightarrow \Omega^{2h}$, "Downsampling"
 - Simplest method: Injection
 - Better: Full weighting

$$v_j^{2h} = \frac{1}{4} (v_{2j-1}^h + 2v_{2j}^h + v_{2j+1}^h), \quad 1 \le j \le \frac{n}{2} - 1$$

• Restriction operator:

$$I_{h}^{2h} = \frac{1}{4} \begin{bmatrix} 1 & 2 & 1 & & \\ & 1 & 2 & 1 & \\ & & 1 & 2 & 1 \end{bmatrix}$$
$$\mathbf{v}^{2h} = I_{h}^{2h} \mathbf{v}^{h}$$

• Transfer Operations $\Omega^h \leftrightarrow \Omega^{2h}$ sufficient

- Aliasing: Oscillatory modes on Ω^h will be represented as smooth modes on Ω^{2h}
- A basic two-grid correction scheme
 - On grid Ω^h , relax υ_1 times on $A^h \mathbf{v}^h = \mathbf{0}$ with initial guess $\mathbf{v}^{(0)h}$
 - Restrict fine-grid residual **r**^h to the coarse grid
 - On grid Ω^{2h} , relax υ_2 times on $A^{2h}e^{2h} = r^{2h}$ with initial guess $e^{(0)h} = 0$
 - Interpolate the coarse-grid error
 - Correct the fine-grid approximation: $\mathbf{v}^{h} \leftarrow \mathbf{v}^{h} + \mathbf{e}^{h}$
 - On grid Ω^h , relax υ_1 times on $A^h \mathbf{v}^h = \mathbf{0}$ with initial guess \mathbf{v}^h

- Multigrid strategies
 - Nested iteration: Use coarse grids to generate improved initial guesses
 - Coarse grid correction: Approximate the error by relaxing on the residual equation on a course grid



- The V-Cycle Scheme (Coarse Grid Correction)
 - V-Cycle(v^h, f^h)
 - Relax v_1 times on $A^h v^h = 0$ with initial guess v^h
 - If (current grid = coarsest grid) goto last point
 - Else: $\mathbf{f}^{2h} = \text{Restrict}(\mathbf{f}^h A^h \mathbf{v}^h)$
 - $v^{2h} = 0$
 - Call v^{2h} = V-Cycle(v^{2h}, f^{2h})
 - Correct v^h += Interpolate(v^{2h})
 - Relax v_2 times on $A^h v^h = 0$ with initial guess v^h
 - Recursive algorithm



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Performance

- Storage requirements
 - Vectors v and f for n = 16 with boundary values

V	Ω ^h	Ω^{2h}	Ω^{4h}	Ω^{8h}
	17	9	5	3
f	Ω ^h	Ω^{2h}	Ω^{4h}	Ω^{8h}
	17	9	5	3

StorageSpace $< \frac{2n^d}{1-2^{-d}}$, with d: Dimension of problem

 For d = 1, memory requirement is less then twice that of the fine-grid problem alone

- Computational costs
 - 1 work unit (WU): one relaxation sweep on Ω^h
 - O(WU) = O(N), with N: Total number of grid points
 - Intergrid transfer is neglected
 - One relaxation sweep per level ($v_i = 1$)

$$Cost_{\text{V-Cycle}} < \frac{2}{1 - 2^{-d}} WU$$
$$Cost_{\text{FMG Cycle}} < \frac{2}{(1 - 2^{-d})^2} WU$$

 1D problem: Single V-Cycle costs ~4WU, Complete FMG cycle ~8WU

- Diagnostic Tools
 - Help to debug your implementation
 - Methodical Plan for testing modules
 - Starting Simply with small, simple problems
 - *Exposing Trouble* difficulties might be hidden
 - Fixed Point Property relaxation may not change exact solution
 - Homogenous Problem: norm of error and residual should decrease to zero

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- The Impact of Intergrid Transfer and the iterative method may be expressed and proven in a formal way
- Two-grid correction TG consists of matrices for Interpolation, Restriction and Relaxation
- Spectral picture of multigrid
 - Relaxation damps oscillator modes
 - Interpolation & Restriction damp smooth modes
- *Algebraic* picture of multigrid
 - Decompose Space of the error: $\Omega^h = R \oplus N$
 - $R = Range(I_{2h}^{h}) = Ker(TG)$ $N = N(I_{h}^{2h}A^{h}) = Ker(I_{h}^{2h}A^{h})$
 - L similar to R, H similar to N

Theory

- Operations of multigrid, visualized
 - Plane represents Ω^h
 - Error e^h is successively projected on one of the axes
 - Relaxations on the fine grid (1)
 - Two-grid correction (2)
 - Again, relaxation on the fine grid (3)



Thanks for your attentionAny questions?