

Week 4- Lecture 1 and 2

Computational Methods II (Elliptic)

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Overview

- We already know the nature of hyperbolic and parabolic PDE's.
- Now we will focus on elliptic PDE.
- The elliptic PDE is a type of PDE for solving incompressible flow.

Overview (cont'd)

- Elliptic PDE's are always smooth
- Easier to obtain accurate solutions compared to hyperbolic problems, the challenge is in getting solutions efficiently.
- Unlike hyperbolic and parabolic problems, information is transmitted everywhere

Overview (cont'd)

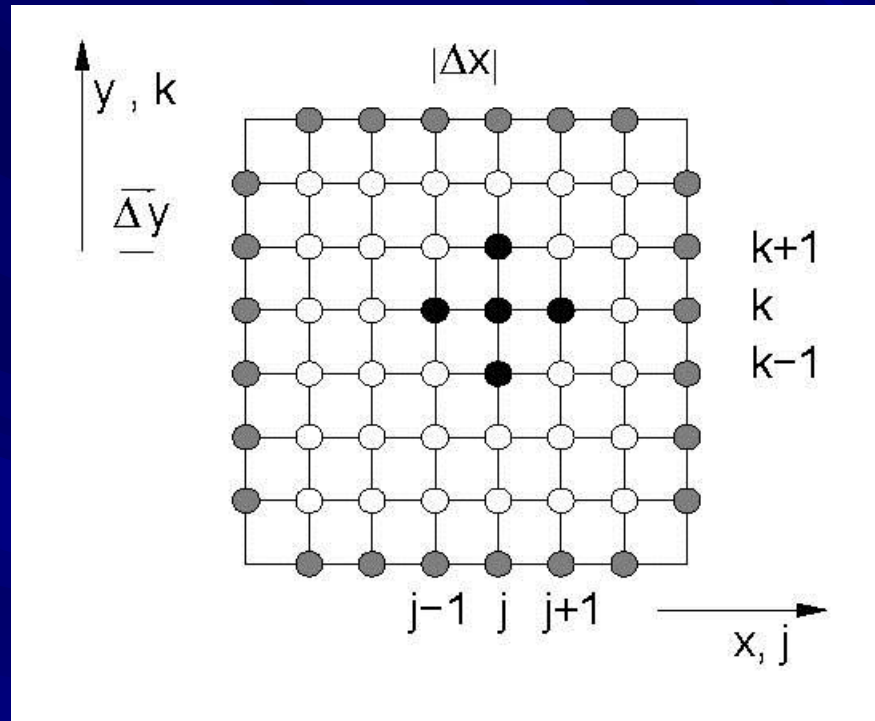
- The model problem that will be discussed is

$$u_{xx} + u_{yy} = f(x, y) \quad \text{Poisson Eqn}$$

$$u_{xx} + u_{yy} = 0 \quad \text{Laplace Equation}$$

- Note that the problem is now 2D but with no time dependence

Overview (cont'd)



$$u_{xx} \approx \frac{u_{j+1,k}^n - 2u_{j,k}^n + u_{j-1,k}^n}{\Delta x^2}$$
$$u_{yy} \approx \frac{u_{j,k+1}^n - 2u_{j,k}^n + u_{j,k-1}^n}{\Delta y^2}$$

Overview (cont'd)

- For special case where the grid sizes in x and y is identical

$$\frac{1}{4}(u_{j-1,k} + u_{j+1,k} + u_{j,k-1} + u_{j,k+1}) - u_{j,k} = r_{j,k} = 0$$

$r_{j,k}$ is the residual associated with cell (j,k)

- Solve for $u(j,k)$ such that $r(j,k)=0$, yielding a system of matrix $Mu = 0$
- Assume M intervals for each direction, Gaussian elimination method needs $O(M^6)$ operations in 2D

Iteration Method

- Solving it directly is too expensive, usually an iteration method is employed.
- The simplest iteration method is the point Jacobi method – based on solving the steady state of parabolic pde

$$u_t = u_{xx} + u_{yy}$$

$$u_{j,k}^{n+1} = u_{j,k}^n + \omega \left[\frac{1}{4} (u_{j-1,k}^n + u_{j+1,k}^n + u_{j,k-1}^n + u_{j,k+1}^n) - u_{j,k}^n \right]$$
$$u_{j,k}^{n+1} = u_{j,k}^n + \omega r_{j,k}^n$$

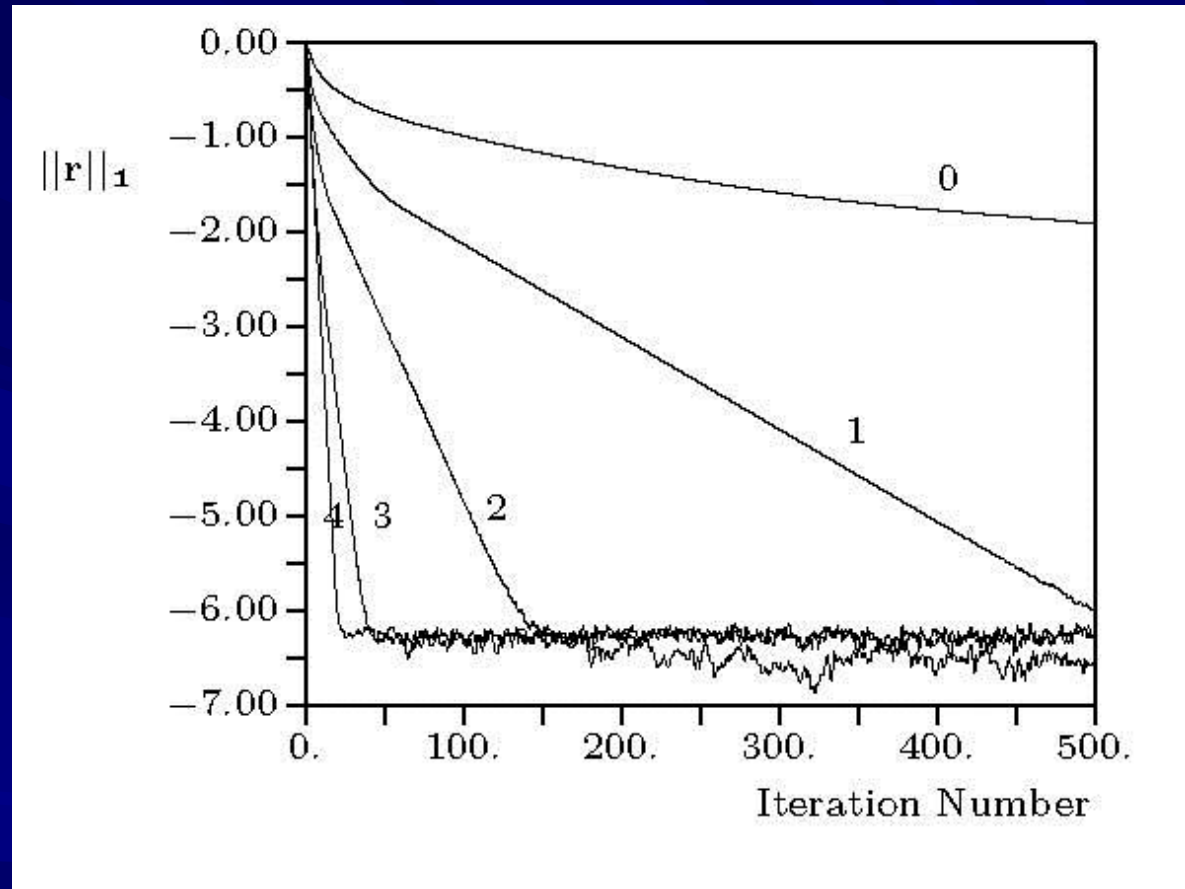
- Solve iteratively for $u_{j,k}^{n+1}$ until

$$u_{j,k}^{n+1} - u_{j,k}^n \approx 0$$

Iteration Method (cont'd)

- The superscript n denotes pseudo-time, not physical time
- ω is the relaxation factor, analogous to μ
- The first thought is to achieve steady-state as quickly as possible, hence choosing a largest ω consistent with stability
- Not a good idea since we need to know what is the best choice for ω

Typical Residual Plot



Residual Pattern

- Plot of convergence history
- Has three distinct phases
- First phase, the residual decays very rapidly
- Second phase decays linearly on the log-plot (or exponentially in real plot)
- Third phase is 'noise', since randomness has set in
 - after a while residuals become so small that they are of the order of round-off errors

Residual Pattern (cont'd)

- Sometimes the residual plot would 'hang' up
- Usually due to a 'bug' in the code
- To save time, apply a 'stopping' criterion to residual
- But not easy to do so, need to understand errors

Analyzing the Errors

- We want know how the iteration errors decay.

- Expand the solution as

$$u_{j,k}^n = u_{j,k}^\infty + \epsilon_{j,k}^n$$

- The first term on the RHS is the solution after infinite number of iterations
- The second term on the RHS is the error between the solution at iterative level n and after infinite iterations
- Get the best solution if error is removed

Analyzing the Errors (cont'd)

- Substitute the error relation into the iteration method.

$$u_{j,k}^{\infty} + \epsilon_{j,k}^{n+1} = u_{j,k}^{\infty} + \epsilon_{j,k}^n + \omega \left[\frac{1}{4} (u_{j-1,k}^{\infty} + u_{j+1,k}^{\infty} + u_{j,k-1}^{\infty} + u_{j,k+1}^{\infty}) - u_{j,k}^{\infty} \right. \\ \left. + \frac{1}{4} (\epsilon_{j-1,k}^n + \epsilon_{j+1,k}^n + \epsilon_{j,k-1}^n + \epsilon_{j,k+1}^n) - \epsilon_{j,k}^n \right]$$

- Converge solution gives zero residual, hence

$$\epsilon_{j,k}^{n+1} = \epsilon_{j,k}^n + \omega \left[\frac{1}{4} (\epsilon_{j-1,k}^n + \epsilon_{j+1,k}^n + \epsilon_{j,k-1}^n + \epsilon_{j,k+1}^n) - \epsilon_{j,k}^n \right]$$

- Iterations for error

Analyzing the Errors (cont'd)

- Shows that error itself follow an evolutionary pattern
- This is true for all elliptic problems
- Remarkably, we can determine how the error would behave even if we do not know the solution

Analyzing the Errors using VA

- In 2D, von Neumann analysis (VA)

$$\epsilon_{j,k}^n = \text{Real}(g^n \exp(i[j\theta_x + k\theta_y]))$$

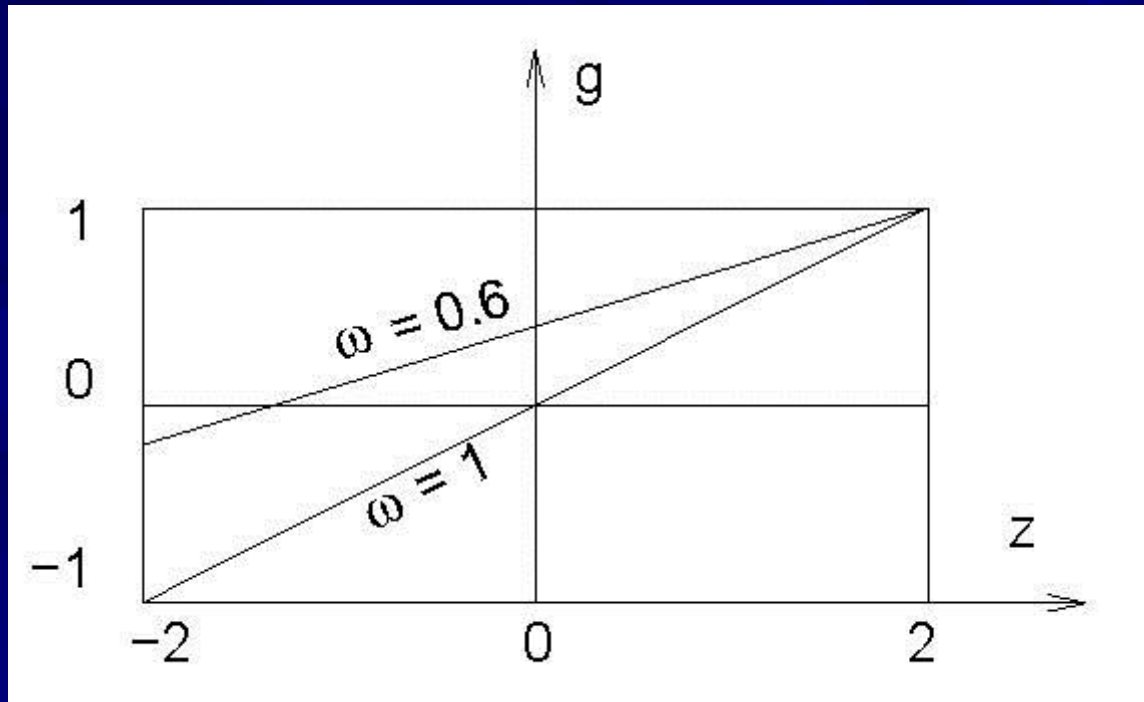
- Insert this into the error equation

$$\epsilon_{j,k}^{n+1} = \epsilon_{j,k}^n + \omega \left[\frac{1}{4} (\epsilon_{j-1,k}^n + \epsilon_{j+1,k}^n + \epsilon_{j,k-1}^n + \epsilon_{j,k+1}^n) - \epsilon_{j,k}^n \right]$$

- Yields the amplification factor for the errors

$$g = 1 - \frac{\omega}{2} [2 - (\cos\theta_x + \cos\theta_y)]$$

Amplification factor for Point Jacobi Method



$$z = (\cos\theta_x + \cos\theta_y)$$
$$g = 1 - \frac{\omega}{2}[2 - z]$$

What does the figure tell you ???