

MAT 228A Notes

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November 1, 2016

1 Iterative Methods for the Poisson Equation

Linear fixed-point iteration problem: $u_{k+1} = Tu_k + c$. When should we stop the iteration? There are two standard ways to figure this out:

1. Stop based on the size of the residual, $r_k = f - Au_k$
2. Stop based on the size of $u_{k+1} - u_k$

Use the residual

- absolute tolerance $\|r_k\| < \text{tol} =: \varepsilon$
- relative tolerance $\|r_k\| \leq \text{tol}\|f\|$

Relative tolerance is better.. if $u_0 = 0$, $r_0 = f$.

Reminder: residual equation: $Ae_k = r_k$ where e_k is the algebraic error on the k th iterate. We want to control e_k ,

$$\|e_k\| = \|A^{-1}r_k\| \leq \|A^{-1}\|\|r_k\| \leq \|A^{-1}\|\varepsilon$$

$\varepsilon = Ch^2 \dots$ use $u_{k+1} - u_k$

- Absolute: $\|u_{k+1} - u_k\| < \text{tol}$
- Relative: $\|u_{k+1} - u_k\| < \text{tol}\|u_k\|$

2 Jac, GS, SOR, Multigrid

$$\begin{aligned} u_{k+1} &= u_k + Br_k, & B &\approx A^{-1} \\ u_{k+1} - u_k &= Br_k \end{aligned}$$

Want to control the error - $\|e_k\| = \|A^{-1}r_k\| = \|A^{-1}B^{-1}(u_{k+1} - u_k)\| \leq \|A^{-1}B^{-1}\|\|u_{k+1} - u_k\|$.

2.1 Jacobi

$B = -\frac{h^2}{4}I$, which implies $\|B^{-1}\| = \frac{4}{h^2}$.

2.2 Ordering of Unknowns

Jacobi does depend on the ordering of unknowns. GS does. We did GS-Lex, which is a sweep through of the unknowns - gives some sort of diagonal structure (with streaks at n th, n^2 th, etc. super- and sub-diagonals). Another way to order them is label them **red** or **blue** (even/odd) so $i + j$ is either even or odd. Then lexicographic ordering for **red** points and THEN for **blue** points. This is called GS-RB. Update **red** points first, and then **blue** points.

2.2.1 GS-Lex Pseudocode

- loop k
 - loops i, j
 - * $u_{ij} = \frac{1}{4}(u_{i-1,j} + u_{i,j-1} + u_{i+1,j} + u_{i,j+1} - h^2 F_{ij})$

2.2.2 GS-RB Pseudocode

- loop k
 - loop **red**
 - * $u_{ij} = \frac{1}{4}(u_{i-1,j} + u_{i,j-1} + u_{i+1,j} + u_{i,j+1} - h^2 F_{ij})$
 - loop **blue**
 - * $u_{ij} = \frac{1}{4}(u_{i-1,j} + u_{i,j-1} + u_{i+1,j} + u_{i,j+1} - h^2 F_{ij})$

2.2.3 Other variations

Block or line relaxation methods - update groups of points at once. For example, in 2D, each row is a group - do a 1D solve on each row of points. This is useful for problems such as

$$u_{xx} + \varepsilon u_{yy} = f$$

3 Successive Over Relaxation (SOR) Method

This is a generalization of GS by including a relaxation parameter.

3.1 GS

- $u_{ij} = \frac{1}{4}(u_{i-1,j} + u_{i,j-1} + u_{i+1,j} + u_{i,j+1} - h^2 F_{ij})$

3.2 SOR

- $u_{ij} = \frac{\omega}{4}(u_{i-1,j} + u_{i,j-1} + u_{i+1,j} + u_{i,j+1} - h^2 F_{ij}) + (1 - \omega)u_{ij}$

Choosing $\omega < 1$ is “under-relaxation” and $\omega > 1$ is “over relaxation”. SOR requires $\omega \in (0, 2)$ for convergence.

$$A = M - N$$

where $Mu_{k+1} = Nu_k + F$ and so $u_{k+1}M^{-1}Nu_k + M^{-1}F$. Also,

$$M = \frac{1}{\omega}D - L \quad \text{and} \quad N = \frac{1 - \omega}{\omega}D + U$$

$$\begin{aligned} T_{\text{SOR}} &= M^{-1}N = \left(\frac{1}{\omega}D - L\right)^{-1} \left(\left(\frac{1 - \omega}{\omega}\right)D + U\right) = (D - \omega L)^{-1}((1 - \omega)D + \omega U) \\ \implies \det(T_{\text{SOR}}) &= \det((D - \omega L)^{-1}) \det((1 - \omega)D + \omega U) = \frac{\det(1 - \omega)D}{\det(D)} = (1 - \omega)^N \end{aligned}$$

where N is the number of grid points. Therefore we require $|1 - \omega| < 1$ (i.e. $\omega \in (0, 2)$).

3.2.1 Convergence analysis

Use the same trick to compute the eigenvalues of the update matrix in terms of the eigenvalues of the Jacobi update. So let μ be an eigenvalue of the Jacobi update. Then

$$\mu = \frac{\lambda + \omega - 1}{\omega \lambda^{\frac{1}{2}}}$$

Rearrange this equation to get

$$\begin{aligned} \lambda - \omega \mu \lambda^{\frac{1}{2}} + (\omega - 1) &= 0 \\ \implies 2\lambda^{\frac{1}{2}} &= \omega \mu \pm (\omega^2 \mu^2 - 4(\omega - 1))^{\frac{1}{2}} \end{aligned}$$

As $\omega \rightarrow 0$, we see $\lambda \rightarrow 1$. As $\omega \rightarrow 2$, we have $(4\mu^2 - 4)^{\frac{1}{2}}$ where $\mu < 1$ and thus we have complex eigenvalues. But if $\lambda \in \mathbb{C} \setminus \mathbb{R}$, then we have an explicit formula for the modulus of the eigenvalues:

$$\left|\lambda^{\frac{1}{2}}\right| = |\omega - 1|$$

and so we decrease ω for better convergence. If $\lambda \in \mathbb{R}$, then

$$\frac{\partial \lambda^{\frac{1}{2}}}{\partial \omega} < 0$$

which takes some work, but increasing ω gives better convergence. So the optimal ω satisfies

$$(\omega^*)^2 \mu^2 - 4(\omega^* - 1) = 0 \quad (1)$$

$$\implies \omega^* = \frac{2}{1 + (1 - \rho_J^2)^{\frac{1}{2}}} \quad (2)$$

where ρ_J is the spectral radius of the Jacobi update matrix, so $\rho_{\text{SOR}} = 1 - \omega^*$.

$$\rho_J = \cos(\pi h) \quad (3)$$

$$w^* = \frac{2}{1 + (1 - \cos^2(\pi h))^{\frac{1}{2}}} = \frac{2}{1 + \sin(\pi h)} = 2(1 - \pi h) + \mathcal{O}(h^2) \quad (4)$$

so

$$\rho_{\text{SOR}} = 1 - 2\pi h \quad (5)$$

3.3 Number of iterations to reduce error by 10^{-1}

$N \times N$	GS	SOR
32×32	254	12
64×64	985	24
128×128	3882	47
256×256	15404	94

(6)