

Lecture 4

Multigrid

New Directions in Mathematics

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Defect Correction

- ▶ Let $Ax = b$, $A \in \mathbb{R}^{n \times n}$ nonsingular, and $x^{(0)} \simeq x$ with residual

$$r^{(0)} = b - Ax^{(0)} \neq 0$$

- ▶ **Goal:** try to improve the accuracy by writing

$$x^{(1)} = x^{(0)} + \delta^{(0)}$$

- ▶ If the **defect correction** $\delta^{(0)}$ is computed by

$$A\delta^{(0)} = r^{(0)},$$

then $x^{(1)} = x$

- ▶ **Defect correction principle:** let $A_{approx}^{-1} \simeq A^{-1}$ and consider

$$x^{(1)} = x^{(0)} + \delta^{(0)}, \quad \text{with } \delta^{(0)} = A_{approx}^{-1} r^{(0)}$$



Defect Correction Iteration

- ▶ Defect correction iteration:

$$x^{(k+1)} = x^{(k)} + \delta^{(k)} = x^{(k)} + A_{approx}^{-1} r^{(k)}, \quad k = 0, 1, \dots$$

- ▶ The method converges if and only if

$$\rho(I - A_{approx}^{-1}A) < 1$$

- ▶ If $A_{approx}^{-1}A \simeq I$ then $\rho(I - A_{approx}^{-1}A) \simeq 0$ (fast convergence)

- ▶ Defect correction iteration \Leftrightarrow Richardson iteration

- ▶ Richardson iteration: $R = I - P^{-1}A$ and

$$x^{(k+1)} = Rx^{(k)} + P^{-1}b \Leftrightarrow x^{(k+1)} = x^{(k)} + P^{-1}r^{(k)}, \quad k \geq 0$$

- ▶ Defect correction iteration: $P^{-1} = A_{approx}^{-1}$, $R = I - A_{approx}^{-1}A$



Two-Grid Method: Coarse and Fine Grids

- ▶ Idea: use the defect correction principle with

$$A_{approx}^{-1} = A_{coarse}^{-1} \simeq A_{fine}^{-1} = A^{-1}$$

- ▶ What we need: transfer vectors corresponding to the fine grid to vectors corresponding to the coarse grid and vice-versa

- ▶ Applications

- ▶ Improve the accuracy of an approximate solution obtained for example by Gaussian elimination
- ▶ Develop fast iterative solutions of linear system arising in the discretization of differential and integral equations



Example: Elliptic Equation (1D)

- ▶ Boundary value problem:

$$-u_{xx} = f, \quad 0 < x < 1, \quad u(0) = u(1) = 0$$

- ▶ Discretizing the problem (FDM)

$$A_h u_h = f_h,$$

with

$$A_h = \frac{1}{h^2} \begin{bmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & \cdot & \cdot & \cdot & \\ & & & -1 & 2 & -1 \\ & & & & -1 & 2 \end{bmatrix}, \quad h = \frac{1}{n+1},$$

and $f_{h,j} = f(x_j)$, $u_{h,j} \simeq u(x_j)$, $j = 1, \dots, n$, on the grid

$$0 = x_0 < x_1 < \dots < x_n < x_{n+1} = 1, \quad x_j = jh$$



Example: Elliptic Equation (1D)

- ▶ Solve the linear system by the JOR method (see: Lecture 1)

$$u_h^{(k+1)} = u_h^{(k)} + \omega D_h^{-1} \underbrace{(f_h - A_h u_h^{(k)})}_{\text{residual}}, \quad k = 0, 1, \dots$$

- ▶ **Exercise 4.1:** Prove that the eigenvalues of $I - \omega D_h^{-1} A_h$ are

$$\lambda_{h,j} = 1 - 2\omega \sin^2 \frac{\pi j h}{2}, \quad j = 1, \dots, n,$$

and the corresponding eigenvectors are

$$v_{h,j} = [\sin(\pi j h), \dots, \sin(\pi n j h)]^T, \quad j = 1, \dots, n,$$

that form an orthogonal basis of \mathbb{R}^n .



Choice of the Damping Factor

- ▶ From Theorem 1.7

$$\omega_{opt} = \frac{2}{2 - \lambda_{h,max} - \lambda_{h,min}} \Rightarrow \omega_{opt} \simeq 1$$

- ▶ Use **Damped/Underrelaxed Jacobi Iteration (DJI)** (i.e, JOR with $0 < \omega < 1$)
- ▶ DJI with $\omega = 0.5$ is a smoothing iteration
 - ▶ Since v_j is an orthogonal basis of \mathbb{R}^n , the error satisfies

$$u_h - u_h^{(k)} = \sum_{j=1}^n \alpha_j^{(k)} v_{h,j},$$

and it may proven that $\alpha_j^{(k+1)} = \lambda_{h,j} \alpha_j^{(k)}$

- ▶ **Exercise 4.2:** Prove that, for $\omega = 0.5$, we have

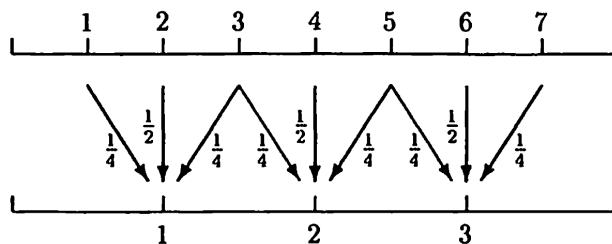
$$\left| \alpha_j^{(k+1)} \right| = \cos^2 \frac{\pi j h}{2} \left| \alpha_j^{(k)} \right| \leq \frac{1}{2} \left| \alpha_j^{(k)} \right|, \quad j = \underbrace{(n+1)/2, \dots, n}_{\text{high frequencies}}$$

- ▶ Fast convergence for high frequencies

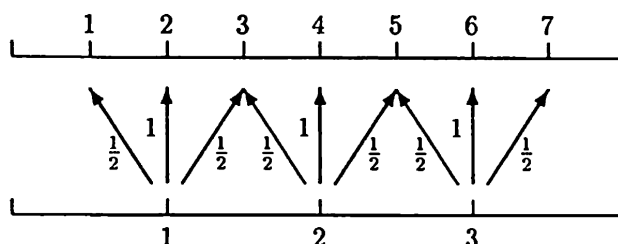


Coarse Grid Correction

- ▶ We consider a **fine grid** with mesh size h and a **coarse grid** with mesh size $2h$ (n is odd)
- ▶ **Restriction**



- ▶ **Prolongation**



Defect Correction Iteration

- ▶ Let

$$A_{coarse}^{-1} = P_{2h} A_{2h}^{-1} R_h$$

and compute

$$u_h^{(k+1)} = u_h^{(k)} + A_{coarse}^{-1} (f_h - A_h u_h^{(k)}), \quad k = 0, 1, \dots$$

- ▶ How to compute $\delta_h^{(k)} = A_{coarse}^{-1} r_h^{(k)} = P_{2h} A_{2h}^{-1} R_h r_h^{(k)}$

1. Restrict $r_h^{(k)} \in \mathbb{R}^n$ to $r_{2h}^{(k)} = R_h r_h^{(k)} \in \mathbb{R}^{\frac{n-1}{2}}$
2. Solve the $\frac{n-1}{2} \times \frac{n-1}{2}$ system $A_{2h} \delta_{2h}^{(k)} = r_{2h}^{(k)}$
3. Prolong the solution $\delta_{2h}^{(k)} \in \mathbb{R}^{\frac{n-1}{2}}$ to $\delta_h^{(k)} = P_{2h} \delta_{2h}^{(k)} \in \mathbb{R}^n$



Two-Grid Iteration Scheme

- ▶ **Two-Grid Iteration Scheme:** Combining this coarse grid correction with N steps of the DJI we obtain

$$u_h^{(k+1)} = u_h^{(k,N)} + P_{2h} A_{2h}^{-1} R_h (f_h - A_h u_h^{(k,N)}),$$

for $k = 0, 1, \dots$, where $u_h^{(k,N)} = \mathcal{J}^N(A_h, u^{(k)}, f_h)$ denotes the result of N steps of DJI starting with $u^{(k)}$.

- ▶ **Exercise 4.3:** Prove that the iteration matrix corresponding to this two-grid method is given by

$$T_N = \underbrace{(I - P_{2h} A_{2h}^{-1} R_h A_h)}_{\text{coarse-grid correction}} \left(I - \frac{1}{2} D_h^{-1} A_h \right)^N.$$

- ▶ $N = 1$ ($T_1 = T$): alternate one step of the DJI (fine grid) with a coarse-grid correction by elimination (coarse grid)



HW Exercises

- ▶ **Exercise 4.4:** Prove that, for $j = 1, \dots, (n-1)/2$,

$$R_h v_{h,j} = c_j^2 v_{2h,j}, \quad R_h v_{h,n+1-j} = -s_j^2 v_{2h,j},$$

and

$$P_{2h} v_{2h,j} = c_j^2 v_{h,j} - s_j^2 v_{h,n+1-j},$$

with

$$c_j = \cos \frac{j\pi h}{2}, \quad s_j = \sin \frac{j\pi h}{2}.$$

- ▶ **Exercise 4.5:** Prove that, for $j = 1, \dots, (n-1)/2$,

$$\begin{bmatrix} T v_{h,j} \\ T v_{h,n+1-j} \end{bmatrix} = s_j^2 c_j^2 \underbrace{\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}}_Q \begin{bmatrix} v_{h,j} \\ v_{h,n+1-j} \end{bmatrix},$$

and, since $R_h v_{h,(n+1)/2} = 0$,

$$T v_{h,(n+1)/2} = \frac{1}{2} v_{h,(n+1)/2}.$$



Convergence of Two-Grid Method

Theorem 4.1

For the spectral radius of T we have that $\rho(T) = 0.5$.
Consequently, the two-grid iterations converge.

- ▶ **Proof:** From Exercise 4.1 we have, for $j = 1, \dots, (n-1)/2$

$$A_{2h} v_{2h,j} = \frac{1}{h^2} \sin^2(\pi j h) v_{2h,j} = \frac{4}{h^2} c_j^2 s_j^2 v_{2h,j},$$

and then

$$A_{2h}^{-1} v_{2h,j} = \frac{h^2}{4c_j^2 s_j^2} v_{2h,j}.$$

From Exercise 4.5, since the matrix Q has eigenvalues 0 and 2, it can be proved that the matrix T has the eigenvalues

$$2c_j^2 s_j^2 = \frac{1}{2} \sin^2 \pi j h, \quad j = 1, \dots, (n+1)/2,$$

and the eigenvalue zero of multiplicity $(n-1)/2$. \square



Two-Grid Method: General Formulation

- ▶ At the highest level (finest grid) a mesh-size of h is used

$$A_h u_h = f_h$$

- ▶ Requirement: a system similar to the one above must be solved at the coarser levels
- ▶ A coarse grid with mesh size H is used (coarser mesh Ω_H)
- ▶ Galerkin projection:

$$A_H = R_h A_h P_H, \quad f_H = R_h f_h$$



Smoothing Step

- ▶ Define a smoother \mathcal{S} (e.g., DJI, SOR, ...) and consider

$$u_h^{(N)} = \mathcal{S}^N(A_h, u_h^{(0)}, f_h)$$

- ▶ Smoothing iterations

$$u_h^{(j+1)} = S_h u_h^{(j)} + g_h, \quad j = 0, \dots, N-1,$$

where S_h is the smoothing iteration matrix

- ▶ **Exercise 4.6:** Prove that

$$u_h^{(j+1)} = u_h^{(j)} + B_h(f_h - A_h u_h^{(j)}), \quad j = 0, \dots, N-1,$$

with $B_h = (I - S_h)A_h^{-1}$, $S_h = I - B_h A_h$, $g_h = B_h f_h$.

- ▶ The error $e_h^{(N)}$ and residual $r_h^{(N)}$ after N smoothing steps are

$$e_h^{(N)} = S_h^N e_h^{(0)} = (I - B_h A_h)^N e_h^{(0)}, \quad r_h^{(N)} = (I - A_h B_h)^N r_h^{(0)}$$



Two-Grid Cycle

- ▶ Algorithm: Two-Grid Cycle

$$u_h = \mathcal{S}^N(A_h, u_h^{(0)}, f_h) \quad \% \text{ Pre-smooth}$$

$$r_h = f_h - A_h u_h \quad \% \text{ Get residual}$$

$$r_H = R_h r_h \quad \% \text{ Coarsen}$$

$$\text{Solve } A_H \delta_H = r_H$$

$$u_h = u_h + P_H \delta_H \quad \% \text{ Correct}$$

$$u_h = \mathcal{S}^M(A_h, u_h, f_h) \quad \% \text{ Post-smooth}$$



Two-Grid Cycle

- ▶ One iteration of the 2-grid algorithms corresponds to

$$u_h^{(1)} = T_h u_h^{(0)} + g_{M_h}$$

- ▶ If $f_h \equiv 0$

$$u_h^{(1)} = S_h^M (S_h^N u_h^{(0)} + P_H A_H^{-1} R_h (-A_h S_h^N u_h^{(0)}))$$

- ▶ The 2-grid iteration operator is

$$T_h = S_h^M \underbrace{(I - P_h A_H^{-1} R_h A_h)}_{\text{coarse-grid correction}} S_h^N$$



W-cycle Multigrid

- ▶ More corrections, say γ times on a coarse grid before returning to a finer level
- ▶ Algorithm: W-cycle Multigrid

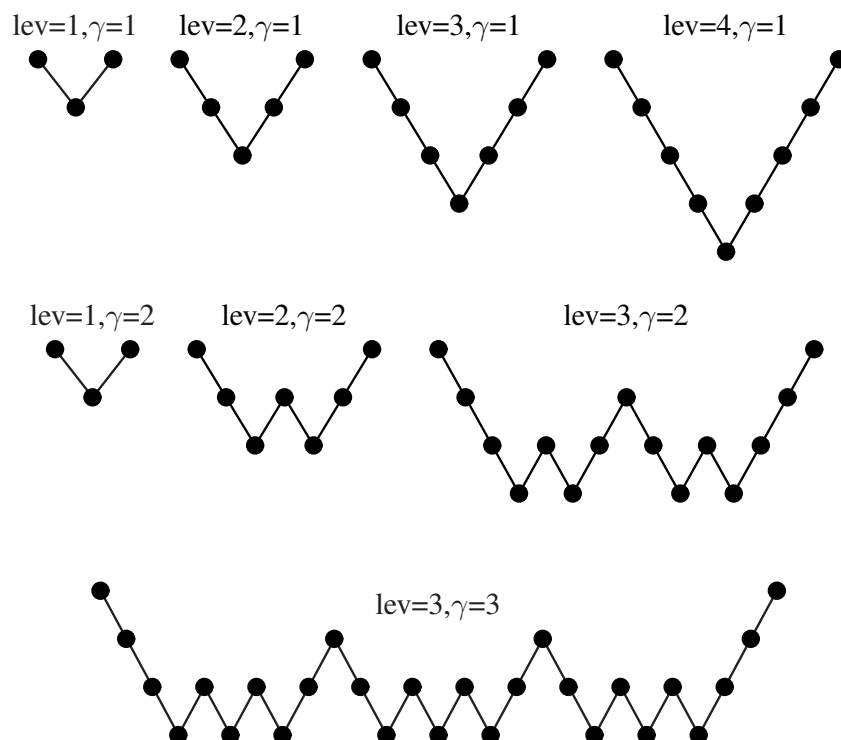
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function  $u_h = MG(A_h, u_h^{(0)}, f_h, N, M, \gamma)$ 
     $u_h = \mathcal{S}^N(A_h, u_h^{(0)}, f_h)$  % Pre-smooth
     $r_h = f_h - A_h u_h$  % Get residual
     $r_H = R_h r_h$  % Coarsen
    If ( $H == h_0$ )
        Solve  $A_H \delta_H = r_H$ 
    else
         $\delta_H = MG^\gamma(A_h, 0, f_h, N, M, \gamma)$  % Recursion
    end if
     $u_h = u_h + P_H \delta_H$  % Correct
     $u_h = \mathcal{S}^M(A_h, u_h, f_h)$  % Post-smooth
end function

```



V-cycles and W-cycle



Full Multigrid

- ▶ **Slightly different approach:** find an approximation to the solution with only one sweep through the levels, going from bottom to top
- ▶ The system is first solved (or smoothed) on a very coarse grid, then one goes to the next finer grid and smoothes the system on this grid and so on, until the finest grid is reached
- ▶ **Algorithm: Full Multigrid**

Set $h = h_0$ and solve $A_h u_h = f_h$

for $k = 1$ to p do

$$u_{h/2} = \hat{P}_h u_h$$

$$h = h/2$$

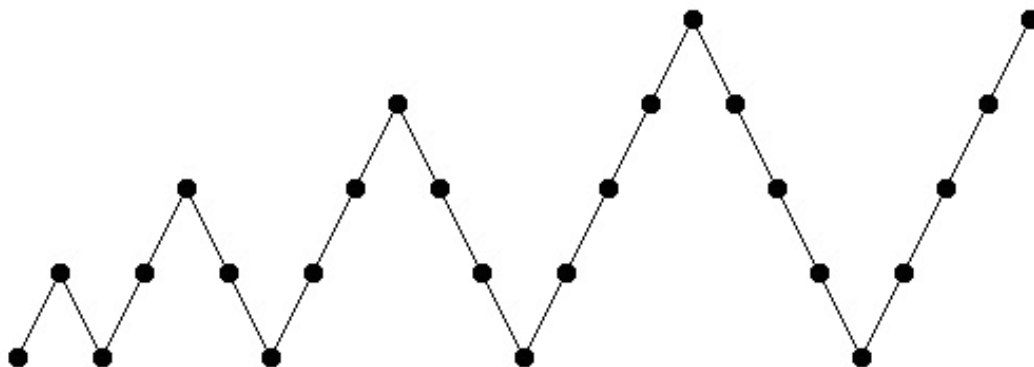
$$u_h = MG^\mu(A_h, u_h, f_h, N, M, \gamma)$$

end for

- ▶ **Note:** the interpolation operator \hat{P}_h is typically of a lower order than P_h



Full Multigrid V-cycle



Basic Convergence Results

- ▶ Cost (storage and computing time)
 - ▶ overall costs are dominated by the costs of the finest grid
- ▶ Speed of convergence
 - ▶ significant acceleration compared with relaxation methods
 - ▶ the convergence rate is independent of the number of unknowns
 - ▶ constant number of multigrid steps to obtain a given number of digits
 - ▶ overall computational work increases only linearly with the number of unknowns