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Scientific Computing II

Multigrid Methods

Programming assignment 1: Multigrid for Poisson Equation

We want to solve the two-dimensional finite difference approximation

$$\begin{aligned} a\frac{u_{i+1j} - 2u_{ij} + u_{i-1j}}{h_x^2} + b\frac{u_{ij+1} - 2u_{ij} + u_{ij-1}}{h_y^2} &= 0, \quad i \in \{2, ..., N_x\}, j \in \{2, ..., N_y\} \\ u_{ij} &= 0, \quad i = 1 \vee j = 1 \vee i = N_x + 1 \vee j = N_y + 1 \end{aligned} \tag{1}$$

which is defined on a unit square using a Cartesian mesh $((i-1)h_x, (j-1)h_y)$ with $h_x = 1/N_x$, $h_y = 1/N_y$[1]. The number of grid points in each direction should initially be chosen as multiple of two plus one, that is $N_x = N_y = 2^l$, $l \in \mathbb{N}$, and the size of the grid is $N_x + 1 \times N_y + 1$. The coefficients $a, b \in \mathbb{R}$ can be chosen arbitrarily. For the time being, we set them to unity, $a = b = 1$.
The following two matlab functions are provided on the webpage of this course:

- `gaussSeidel(a,b,u,rhs)` carries out one iteration of the Gauss-Seidel method. The input arguments are given by the scalar coefficients a, b from above, the two-dimensional solution $u \in \mathbb{R}^{N_x+1 \times N_y+1}$ and a right hand side $rhs \in \mathbb{R}^{N_x+1 \times N_y+1}$. The mesh size of the current solution u is determined from the number of grid points $N_x + 1$, $N_y + 1$. The function returns the updated solution of u.
- `residual(a,b,u,rhs)` evaluates the residual for the approximate solution u of Eq. (1). The function returns the residual $r \in \mathbb{R}^{N_x+1 \times N_y+1}$.

In the following, we will prepare the different subroutines required for a multigrid solver and finally put them together in the *V-cycle* algorithm.

(a) Implement a matlab function `restrict_fullweighting(r)` to perform the full-weighting restriction. The argument of the function is the residual $r \in \mathbb{R}^{N_x+1 \times N_y+1}$. The function

[1] The enumeration has been slightly adapted to stay more compatible with the MATLAB enumeration of arrays and matrix structures

should return a restricted residual $r^c \in \mathbb{R}^{\frac{N_x}{2}+1 \times \frac{N_y}{2}+1}$ which arises from

$$\begin{aligned} r^c_{ij} := & \frac{1}{4} r_{2(i-1)+1,2(j-1)+1} \\ & + \frac{1}{8}(r_{2(i-1)+1,2(j-1)} + r_{2(i-1),2(j-1)+1} + r_{2(i-1)+2,2(j-1)+1} + r_{2(i-1)+1,2(j-1)+2}) \\ & + \frac{1}{16}(r_{2(i-1),2(j-1)} + r_{2(i-1)+2,2(j-1)} + r_{2(i-1),2(j-1)+2} + r_{2(i-1)+2,2(j-1)+2}) \end{aligned} \quad (2)$$

for all inner points. The value r^c_{ij} can be considered to be zero for all points on the boundary strip.

(b) Implement a matlab function `restrict_injection(r)`. The function should return a restricted residual $r^c \in \mathbb{R}^{\frac{N_x}{2}+1 \times \frac{N_y}{2}+1}$. For the injection method, see the lecture slides. The value r^c_{ij} can be considered to be zero for all points on the boundary strip.

(c) Implement a matlab function `interpolate(e)` which prolongates the error $e \in \mathbb{R}^{\frac{N_x}{2}+1 \times \frac{N_y}{2}+1}$ to the fine grid. The interpolated error $e^f \in \mathbb{R}^{N_x+1 \times N_y+1}$ is zero on the boundary strip and defined on the inner points via bilinear interpolation as

$$e^f_{ij} := \begin{cases} \frac{1}{4}(e_{i/2,j/2} + e_{i/2+1,j/2} + e_{i/2,j/2+1} + e_{i/2+1,j/2+1}) & \text{if (i,j) is centered between 4 coarse grid points} \\ \frac{1}{2}(e_{(i+1)/2,j/2} + e_{(i+1)/2,j/2+1}) & \text{if (i,j) is centered between 2 y-aligned coarse grid points} \\ \frac{1}{2}(e_{i/2,(j+1)/2} + e_{i/2+1,(j+1)/2}) & \text{if (i,j) is centered between 2 x-aligned coarse grid points} \\ e_{(i+1)/2,(j+1)/2} & \text{if (i,j) coincides with a coarse grid point.} \end{cases} \quad (3)$$

(d) Put the algorithmic components together in a function

`vCycle(a,b,u,rhs,preSmoothing,postSmoothing,level,mgID).`

Besides the coefficients a, b, the approximate solution u and the right hand side *rhs*, the parameters *preSmoothing* and *postSmoothing* determine the number of pre- and post-smoothing Gauss-Seidel iterations. The parameter *level* corresponds to $N_x = N_y = 2^{level}$. If $level = 1$, we only have one inner grid point (and 3×3 points in total). In this case, the function `vCycle(...)` should carry out one Gauss-Seidel iteration on the current data u and return the new solution. For $level > 1$, the function should carry out the *recursive* v-cycle algorithm, see slide 8 of the multigrid lecture slides. The additional parameter $mgID$ can be used to define which restriction-interpolation combination should be used. You may introduce an additional alternative which abstains from the v-cycle and only carries out one Gauss-Seidel iteration (e.g. for testing purposes).

(e) Use your implementations to solve the 2D Poisson problem. Initialise the solution u according to the function $u(x, y) = \sin(\pi x)\sin(\pi y)$ on the unit square. Use two pre- and two post-smoothing steps and iterate until the maximum norm of the residual has

reached an accuracy of 10^{-5}. Run the simulation for different grid levels $l \in \{2, 3, 4, 5, 6\}$ and both full-weighting restriction or injection. How many iteration steps do you need for each grid resolution/ solver scheme? How many iteration steps do you need if you use pure Gauss-Seidel iterations to solve the Poisson problem? Create error plots (iteration steps vs. maximum norm of residual) for both multigrid schemes and the Gauss-Seidel scheme for grid level $l = 5$, assuming a tolerance $TOL = 1e - 12$.

Programming assignment 2: Multigrid for Anisotropic Poisson Equation

So far, we considered the coefficients $a = b = 1$. In the following, we want to consider the case where $a \gg b$ and study the respective behaviour of the multigrid scheme. For this purpose, we set $a = 1$, $b = 1e - 4$ and obtain an anisotropic Poisson-like equation.

(a) Use your matlab code (using the settings $TOL = 1e - 5$, pre-/post-smoothing=2 steps etc.) to solve the anisotropic Poisson problem. How many iteration steps do you need now for the different v-cycle schemes and the pure Gauss-Seidel method?

(b) Implement new functions `restrict_semicoarsening(r)` and `interpolate_semi(e)` which allow for a semi-coarsening and semi-prolongation of the residual along x-direction:

- `restrict_semicoarsening(r)` restricts a residual $r \in \mathbb{R}^{N_x+1 \times N_y+1}$ only along x-direction and returns the restricted residual $r^c \in \mathbb{R}^{\frac{N_x}{2}+1 \times N_y+1}$. The restriction rule reads

$$r^c_{ij} := \frac{1}{4}(r_{2(i-1)+1,j} + r_{2(i-1),j}) + \frac{1}{2} r_{2(i-1)+1,j}. \tag{4}$$

- `interpolate_semi(e)` prolongates an error $e \in \mathbb{R}^{\frac{N_x}{2}+1 \times N_y+1}$ to a finer grid, resulting in a prolongated error $e^f \in \mathbb{R}^{N_x+1 \times N_y+1}$. The interpolation rules read

$$e^f_{ij} := \begin{cases} \frac{1}{2}(e_{i/2,j} + e_{i/2+1,j}) & \text{if (i,j) is centered between 2 x-aligned coarse grid points} \\ e_{(i+1)/2,j} & \text{if (i,j) coincides with a coarse grid point.} \end{cases} \tag{5}$$

Incorporate the new coarsening and interpolation schemes into the function `vCycle(...)`. The coarsest grid contains more than one inner grid point in case of semi-coarsening; what do you have to change in your current v-cycle implementation so that you obtain convergence? Measure the number of iteration steps for different grid levels. What do you observe?