Home Assignments in Numerical Methods

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HA08 (12p)

Preamble. A finite-difference approximation of a partial differential equation often results in a system of linear algebraic equations, which can in general be written as

$$\mathbf{A}\mathbf{x} = \mathbf{b} \text{ or } \sum_{i=1}^{n} a_{ij} x_{j} = b_{i}.$$

The matrix here is characterized by having many zero elements (sparse) and high order n especially in two- and three-dimensional problems. An iterative method often suggests an effective and cheap alternative to so-called direct methods that require $O(n^3)$ arithmetic operations and additional computer memory. Perhaps, the simplest family of iterative methods is so-called one-step (first order) stationary algorithms with the generic form

$$\mathbf{x}^{(k+1)} = \mathbf{B}\mathbf{x}^{(k)} + \mathbf{f}$$

Here, $\mathbf{x}^{(0)}$ is an arbitrary starting vector (guess), and \mathbf{B} is referred to as the iteration matrix. The necessary and sufficient condition for convergence is

$$\rho(\mathbf{B}) < 1$$
.

A systematic way to construct such iterative methods begins with splitting the original matrix **A** into a preconditioner **P** and the rest **N**:

$$\mathbf{A} = \mathbf{P} - \mathbf{N}$$
 and $\mathbf{N} = \mathbf{P} - \mathbf{A}$.

Next, the original system of equations rewrites as follows

$$Px = Nx + b$$
.

It is then logical to distribute iterative indices as

$$\mathbf{P}\mathbf{x}^{(k+1)} = \mathbf{N}\mathbf{x}^{(k)} + \mathbf{b}$$

Thus, we arrive at

$$\mathbf{x}^{(k+1)} = \mathbf{P}^{-1}\mathbf{N}\mathbf{x}^{(k)} + \mathbf{P}^{-1}\mathbf{b} = (\mathbf{I} - \mathbf{P}^{-1}\mathbf{A})\mathbf{x}^{(k)} + \mathbf{f}$$

Clearly, the preconditioner should have two properties: (1) **P** must be easily invertible; (2) **P** must be as close (in some sense) to **A** as possible. Selecting the preconditioner can be based on studying the algebraic structure of the original matrix **A**. One general way of doing so is to consider **A** as consisting of the diagonal, **D**, lower triangular, **L**, and upper triangular, **U**, matrices

$$\mathbf{A} = \mathbf{D} - \mathbf{L} - \mathbf{U}.$$

Considered in this course, the finite-difference approximation of the neutron diffusion equation, leads to matrices **D**, **L**, and **U** all having non-negative elements. The original equation can thus be rewritten as

$$Dx = Lx + Ux + b.$$

The Jacobi method is based on the following distribution of iteration indices

$$\mathbf{D}\mathbf{x}^{(k+1)} = \mathbf{L}\mathbf{x}^{(k)} + \mathbf{U}\mathbf{x}^{(k)} + \mathbf{b}$$
.

Equivalently, it is defined by

$$\mathbf{x}^{(k+1)} = \mathbf{D}^{-1} (\mathbf{L} + \mathbf{U}) \mathbf{x}^{(k)} + \mathbf{D}^{-1} \mathbf{b}$$

The iteration matrix becomes

$$\mathbf{B} = \mathbf{B}_{I} \equiv \mathbf{D}^{-1} (\mathbf{L} + \mathbf{U})$$

Alternatively, in the coordinate form, the Jacobi method runs as

$$x_i^{(k+1)} = \frac{1}{a_{ii}} \left(b_i - \sum_{\substack{j=1\\j \neq i}}^n a_{ij} x_j^{(k)} \right)$$

The Gauss-Seidel method is based on the following distribution of iteration indices

$$\mathbf{D}\mathbf{x}^{(k+1)} = \mathbf{L}\mathbf{x}^{(k+1)} + \mathbf{U}\mathbf{x}^{(k)} + \mathbf{b}$$
.

Equivalently, it is defined by

$$\mathbf{x}^{(k+1)} = \left(\mathbf{D} - \mathbf{L}\right)^{-1} \mathbf{U} \mathbf{x}^{(k)} + \left(\mathbf{D} - \mathbf{L}\right)^{-1} \mathbf{b}$$

The iteration matrix becomes

$$\mathbf{B} = \mathbf{B}_{GS} \equiv \left(\mathbf{D} - \mathbf{L}\right)^{-1} \mathbf{U}$$

Alternatively, in the coordinate form, the Gauss-Seidel method runs as

$$x_i^{(k+1)} = \frac{1}{a_{ii}} \left(b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^{n} a_{ij} x_j^{(k)} \right)$$

The Successive Over-Relaxation method, SOR, involves an accelerating parameter, ω , and is based on the following distribution of iteration indices

$$\mathbf{D}\mathbf{x}^{(k+1)} = \omega \left(\mathbf{L}\mathbf{x}^{(k+1)} + \mathbf{U}\mathbf{x}^{(k)} + \mathbf{b} \right) + (1 - \omega)\mathbf{x}^{(k)}.$$

Skipping some simple algebra, the iteration matrix becomes

$$\mathbf{B} = \mathbf{B}_{\omega} \equiv \left(\mathbf{I} - \omega \mathbf{D}^{-1} \mathbf{L}\right)^{-1} \left[\left(1 - \omega\right) \mathbf{I} + \omega \mathbf{D}^{-1} \mathbf{U} \right]$$

Alternatively, in the coordinate form, the SOR method runs as

$$x_i^{(k+1)} = \frac{\omega}{a_{ii}} \left(b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^n a_{ij} x_j^{(k)} \right) + (1 - \omega) x_i^{(k)}$$

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Exercise 1 (1p). Find by hand the first two iterations of the Jacobi method for the following linear system, using $\mathbf{x}^{(0)} = \mathbf{0}$:

$$\begin{cases} 3x_1 - x_2 + x_3 = 1 \\ 3x_1 + 6x_2 + 2x_3 = 0 \\ 3x_1 + 3x_2 + 7x_3 = 4 \end{cases}$$

Exercise 2 (1p). Repeat Exercise 1 using the Gauss-Seidel method.

Exercise 3 (2p). Find by hand the first two iterations of the SOR method with $\omega = 1.1$ for the following linear system, using $\mathbf{x}^{(0)} = \mathbf{0}$:

$$\begin{cases} 3x_1 - x_2 + x_3 = 1 \\ 3x_1 + 6x_2 + 2x_3 = 0 \\ 3x_1 + 3x_2 + 7x_3 = 4 \end{cases}$$

Exercise 4 (2p). The linear system

$$\begin{cases} 2x_1 - x_2 + x_3 = -1 \\ 2x_1 + 2x_2 + 2x_3 = 4 \\ -x_1 - x_2 + 2x_3 = -5 \end{cases}$$

has the solution $(1, 2, -1)^T$.

- a) Show that $\rho(\mathbf{B}_I) = \frac{\sqrt{5}}{2}$.
- b) Show that $\rho(\mathbf{B}_{GS}) = \frac{1}{2}$.
- c) What can we say about the convergence?

Exercise 5 (2p). The linear system

$$\begin{cases} x_1 + 2x_2 - 2x_3 = 7 \\ x_1 + x_2 + x_3 = 2 \\ 2x_1 + 2x_2 + x_3 = 5 \end{cases}$$

has the solution $(1, 2, -1)^T$.

- a) Show that $\rho(\mathbf{B}_I) = 0$.
- b) Show that $\rho(\mathbf{B}_{GS}) = 2$.
- c) What can we say about the convergence?

Exercise 6 (2p). Using the Jacobi, Gauss-Seidel and SOR ($\omega = 1.1$) iterative methods, write and execute a computer program to solve the following linear system to four decimal places (rounded) of accuracy:

$$\begin{bmatrix} 7 & 1 & -1 & 2 \\ 1 & 8 & 0 & -2 \\ -1 & 0 & 4 & -1 \\ 2 & -2 & -1 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 3 \\ -5 \\ 4 \\ -3 \end{bmatrix}$$

Compare the number of iterations needed in each case. Hint: The exact solution is

$$\mathbf{x} = \begin{bmatrix} 1 & -1 & 1 & -1 \end{bmatrix}^T$$

Exercise 7 (2p). Solve the following linear system to four decimal places of accuracy using the SOR iterative method with values of $\omega_i = 1 + 0.1j$ where j = 1, 2, ..., 9.

$$\begin{bmatrix} 7 & 3 & -1 & 2 \\ 3 & 8 & 1 & -4 \\ -1 & 1 & 4 & -1 \\ 2 & -4 & -1 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -1 \\ 0 \\ -3 \\ 1 \end{bmatrix}$$

Plot the number of iterations for convergence versus the value of ω . Which value of ω results in the fastest convergence? *Hint*: The exact solution is

$$\mathbf{x} = \begin{bmatrix} -1 & 1 & -1 & 1 \end{bmatrix}^T$$