

MA398 Matrix Analysis and Algorithms: Exercise Sheet 8

1. (Jacobi) Consider the Jacobi method for solving

$$Ax = b, \quad \text{with } A = \begin{pmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{pmatrix}, \quad b = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

and with start value $x^{(0)} = (0, 0, 0)^T$.

- (a) State the iteration matrix $R = -D^{-1}(L + U)$, compute its spectral radius $\rho(R)$ and deduce that the Jacobi method converges.
 (b) Recall the estimate

$$k \geq k^\# = \frac{\log(\|A\|_2 \|e^{(0)}\|_2 / \|b\|_2) - \log(\varepsilon_r)}{\log(\|R\|_2^{-1})}$$

for the number of steps in order to achieve that $\|r^{(k)}\|_2 \leq \varepsilon_r \|b\|_2$.

For the above specific data, give an upper bound for the number of steps required to get the relative error of the residual below 10^{-6} .

- (c) Derive the estimate

$$\frac{\|e^{(k)}\|_2}{\|x\|_2} \leq \kappa_2(A) \frac{\|r^{(k)}\|_2}{\|b\|_2}$$

and give an upper bound for the number of steps required to get the relative error of the solution below 10^{-6} .

- (d) State the definition of the graph $G(B)$ of a matrix $B \in \mathbb{C}^{n \times n}$.
 Prove that $B \in \mathbb{C}^{n \times n}$ is irreducible if and only if its graph $G(B)$ is connected.

2. (SSOR) The symmetric successive over relaxation consists in performing the following iteration:

$$\begin{aligned} i = 1, \dots, n : \quad a_{ii}x_i^{(k+\frac{1}{2})} &= \omega \left(- \sum_{j=1}^{i-1} a_{ij}x_j^{(k+\frac{1}{2})} - \sum_{j=i+1}^n a_{ij}x_j^{(k)} + b_i \right) - (\omega - 1)a_{ii}x_i^{(k)}, \\ i = n, \dots, 1 : \quad a_{ii}x_i^{(k+1)} &= \omega \left(- \sum_{j=1}^{i-1} a_{ij}x_j^{(k+\frac{1}{2})} - \sum_{j=i+1}^n a_{ij}x_j^{(k+1)} + b_i \right) - (\omega - 1)a_{ii}x_i^{(k+\frac{1}{2})}. \end{aligned}$$

Here, $x^{(k)}$ stands for the k^{th} iterate, and $x^{(k+\frac{1}{2})}$ is an intermediate value.

Show that SSOR is a linear iterative method with

$$M_{\text{SSOR}}^{-1} = \omega(2 - \omega)(D + \omega U)^{-1}D(D + \omega L)^{-1}.$$

Remark: Recalling that SOR uses $M_{\text{SOR}} = \frac{1}{\omega}D + L$ we see that SSOR essentially consists in performing an SOR step followed by a reverse SOR step with $\frac{1}{\omega}D + U$, which explains its name. A couple of SSOR steps sometimes are applied as a preconditioner in CG.

3. Implement the Gauss-Seidel method, a variant of the Jacobi method, where $M := L + D$ and $N := U$. Write a Python function with the signature `def gauss_seidel_method(A, b, x0, max_iter, tol, omega=1.0)`, where `omega` is the relaxation parameter.

- (a) Test your function on a system of linear equations with a diagonally dominant matrix A and different choices of `omega`. Plot the norm of the residual vector as a function of the number of iterations for different choices of `omega`.

- (b) Use your implementation to investigate the impact of the relaxation parameter on the convergence of the method. For which values of ω does the method converge fastest?
- (c) If possible, compare the performance (in terms of both accuracy and computational cost) of your Gauss-Seidel implementation with a basic Jacobi method implementation.