**Exercise 1: Poisson's Equation** Consider the linear system  $A_n\phi = \rho$ , where  $A_n$  is an  $n \times n$  matrix with 2's on the main diagonal, -1's directly above and below the main diagonal and 0's everywhere else. For instance,  $A_5$  is

$$A_5 = \begin{bmatrix} 2 & -1 & 0 & 0 & 0 \\ -1 & 2 & -1 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 \\ 0 & 0 & -1 & 2 & -1 \\ 0 & 0 & 0 & -1 & 2 \end{bmatrix}.$$

This is a discretized version of Poisson's equation:

$$\frac{\partial^2 \phi}{\partial x^2} = \rho.$$

This equation appears very often in physics.

Construct the matrix  $A_{50}$  in Matlab. Try reading the documentation for diag(). Make the vector  $\rho$  according to the formula

$$\rho_j = 2 (1 - \cos(23\pi/51)) \sin(23\pi j/51)$$
.

- (a) Write down the matrix form of the Jacobi iteration  $\phi_{k+1} = M\phi_k + \mathbf{c}$ . Concatenate the matrix M and the vector  $\mathbf{c}$  and save the resulting  $50 \times 51$  matrix as  $\mathbf{A1.dat}$ .
- (b) Use Jacobi iteration to solve for  $\phi$  given an initial guess of a column of ones. Continue to iterate the Jacobi method until every term in the vector  $\phi$  is within  $10^{-4}$  of the previous iteration. I.e.,

$$norm(phi(:,k+1) - phi(:,k), Inf) \le 1e-4.$$

Save the final iteration as a column vector in **A2.dat** and save the total number of iterations as **A3.dat**.

- (c) Now write down the matrix form of the Gauss-Seidel iteration  $\phi_{k+1} = M\phi_k + \mathbf{c}$ . Note that these are not the same M and  $\mathbf{c}$  as in part (a). Concatenate the matrix M and the vector  $\mathbf{c}$  and save the resulting  $50 \times 51$  matrix as  $\mathbf{A4.dat}$ .
- (d) Use Gauss-Seidel iteration to solve for  $\phi$  given an initial guess of a column of ones. Continue to iterate the Gauss-Seidel method until every term in the vector  $\phi$  is within  $10^{-4}$  of the previous iteration (as in part (b)). Save the final iteration as a column vector in **A5.dat** and save the total number of iterations as **A6.dat**.

**Exercise 2: Successive Over-Relaxation** The successive over-relaxation method (SOR method) is another iterative method related to Jacobi and Gauss-Seidel iteration. Suppose we wish to solve the system  $A\mathbf{x} = \mathbf{b}$ , and the matrix A is decomposed into diagonal, upper, and lower parts:

$$D = \begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn} \end{bmatrix}, U = \begin{bmatrix} 0 & a_{12} & \dots & a_{1n} \\ 0 & 0 & \ddots & \vdots \\ \vdots & \ddots & \ddots & a_{n-1,n} \\ 0 & 0 & \dots & 0 \end{bmatrix},$$

$$L = \begin{bmatrix} 0 & 0 & \dots & 0 \\ a_{21} & 0 & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ a_{n1} & \dots & a_{n,n-1} & 0 \end{bmatrix}, \quad A = D + U + L.$$

The SOR method for a relaxation factor  $\omega$  is defined as:

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

Note that for  $\omega = 1$ , this is equivalent to the Gauss-Seidel method. Consider the system  $A_n \phi = \rho$  from Exercise 1.

- (a) For  $\omega = 1.5$ , write down the matrix form of  $\phi_{k+1} = M\phi_k + \mathbf{c}$ . Concatenate the matrix M and the vector  $\mathbf{c}$  and save the resulting  $50 \times 51$  matrix as  $\mathbf{A7.dat}$ .
- (b) If we define the error of iteration k as  $e_k = \phi \phi_k$ , we can show that the error evolves like  $e_{k+1} = Be_k$ . The error will decay if the eigenvalues of B are all less than one in absolute value. For  $\omega = 1$  to  $\omega = 1.99$  (in increments of 0.01), compute the eigenvalues of the matrix B. Save the absolute values of these eigenvalues as a  $100 \times 50$  matrix in **A8.dat**. Find the choice of  $\omega$  that yields the smallest maximal eigenvalue (in absolute value). Save this  $\omega$  and the absolute value of its maximal eigenvalue as a column vector (with  $\omega$  as the first entry) in **A9.dat**.
- (c) For the same set of  $\omega$  choices, compute 200 SOR iterations from an initial guess of a column of ones. Compute the residual  $A_{50}\phi_{200} \rho$  of the final iteration  $\phi_{200}$  for each case. Save the 2-norm of each residual as a  $100 \times 1$  column vector in **A10.dat**. Does the smallest residual correspond to the same optimal  $\omega$  as you found in part (b)?
- (d) Use SOR iteration with the optimal  $\omega$  from part (b) to solve for  $\phi$  given an initial guess of a column of ones. Continue to iterate the SOR method until every term in the vector  $\phi$  is within  $10^{-4}$  of the previous iteration (as in Exercise 1, part (b)). Save the final iteration as a column vector in **A11.dat** and save the total number of iterations as **A12.dat**. Which

iterative method required the fewest iterations (think about this, but you do not need to submit an answer).