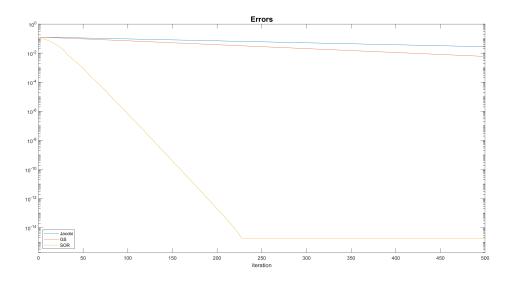
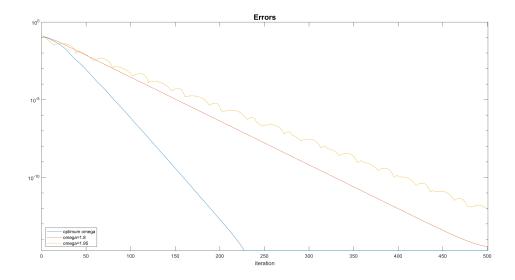
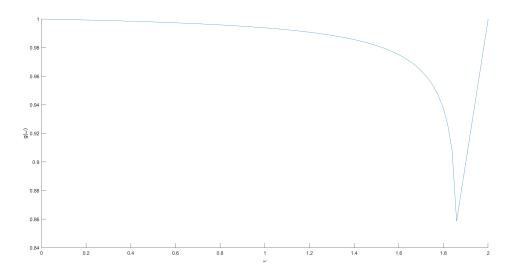
- 4.1 a. Run this program for each method and produce a plot similar to Figure 4.1.
 - b. The convergence behavior of SOR is very sensitive to the choice of ω . Try changing from the optimal ω to $\omega = 1.8$ or 1.95.
 - c. Let $g(\omega) = \rho(G(\omega))$ be the spectral radius of the iteration matrix G for a given value of ω . Write a program to produce a plot of $g(\omega)$ for $0 \le \omega \le 2$.
 - d. Try this computationally and observe that it does not work well. Explain what is wrong with this and derive the correct expression (4.24).

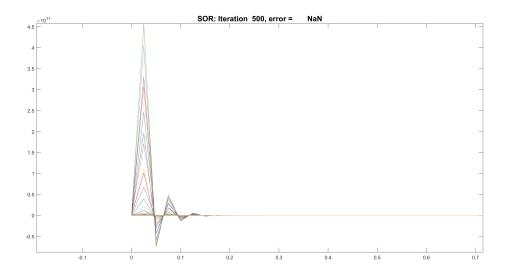
Solution.







We see by the figure below that this method clearly does not work well. Given that the true solution was supposed to be a second order polynomial, this naïve SOR method does not hold.



Now, starting from equations (2.22), we have

$$u_i^{GS} = \frac{1}{2}(u_{i-1}^{[k+1]} + u_{i+1}^{[k]} - h^2 f_i)$$

$$u_i^{[k+1]} = u_i^{[k]} + \omega(u_i^{GS} - u_i^{[k]})$$

Hence,

$$\begin{aligned} u_i^{[k+1]} &= u_i^{[k]} + \omega \big(\frac{1}{2} \big((u_{i-1}^{[k+1]} + u_{i+1}^{[k]} - h^2 f_i \big) - u_i^{[k]} \big) \big) = \\ u_i^{[k]} &- \omega u_i^{[k]} + \frac{\omega}{2} \big(u_{i-1}^{[k+1]} + u_{i+1}^{[k]} - h^2 f_i \big) = (1 - \omega) u_i^{[k]} + \frac{\omega}{2} \big(u_{i-1}^{[k+1]} + u_{i+1}^{[k]} - h^2 f_i \big) \end{aligned}$$

So. we have that

$$u_i^{[k+1]} = (1 - \omega)Du^{[k]} + \frac{\omega}{2}(Lu^{[k+1]} + Uu^{[k]} - h^2 f_i)$$
$$\omega \frac{1}{\omega}((1 - \omega)D + \frac{\omega}{2}U)u^{[k]} + \frac{\omega}{2}Lu^{[k+1]} - \frac{h^2\omega}{2}f_i$$

Therefore, we see that

$$M = \frac{1}{\omega}(D - \omega L)$$
 and $N = \frac{1}{\omega}((1 - \omega)D + \omega U)$

4.2 a. We can also define a backwards Gauss-Seidel method by setting

$$u_i^{[k+1]} = \frac{1}{2}(u_{i-1}^{[k]} + u_{i+1}^{[k+1]} - h^2 f_i), \text{ for } i = m, m-1, m-2, ..., 1$$

Show that this is a matrix splitting method of the type described in Section 4.2 with M = D - U and N = L.

b. Implement this method in *iter_bvp_Asplit.m* and observe that it converges at the same rate as forward Gauss-Siedel for this problem.

Solution.

Let M = D - U and N = L. Then, plugging these into

$$Mu_i^{[k+1]} = Nu_i^{[k]} + f$$

yields

$$(D-U)u_i^{[k+1]} = Lu_i^{[k]} + f_i$$

Hence,

$$Du_i^{[k+1]} - Uu_i^{[k+1]} = Lu_i^{[k]} + f_i$$

So,

$$Du_i^{[k+1]} = Uu_i^{[k+1]} + Lu_i^{[k]} + f_i$$

and therefore

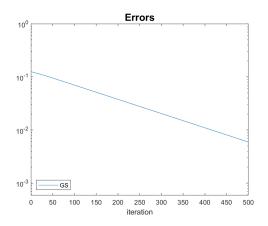
$$\frac{-2}{h^2} I u_i^{[k+1]} = \frac{-1}{h^2} (u_{i+1}^{[k+1]} + u_{i-1}^{[k]}) + f_i$$

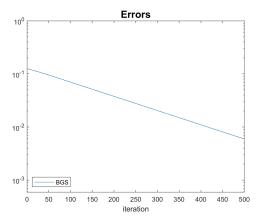
Thus,

$$u_i^{[k+1]} = \frac{1}{2}(u_{i+1}^{[k+1]} + u_{i-1}^{[k]} - h^2 f_i)$$

which is the type of splitting method desired.

It is obviuos by the figures below that the the rate of convergence for the forward and the backward Gauss-Siedel methods is the same.





```
1 % Solve 1d BVP obtained by discretizing u''(x) = f(x)
2 % with Dirichlet boundary condtions u(0) = alpha, u(1) = beta
3 % Discretized using centered difference.
5 % Compare matrix splitting methods: Jacobi, Gauss-Seidel, and SOR
7 % From http://www.amath.washington.edu/¬rjl/fdmbook/ (2007)
9 %4.1.a. We iterate through each of the different techinques and
         create plots to mirror illustration 4.1
12 method = 'SOR'; % 'Jacobi' or 'GS' or 'SOR'
13 nplot = 10;
                       % iterate u is plotted every nplot iterations
14 maxiter = 500;
                        % number of iterations to take
16 \text{ m} = 39;
17 \text{ ax} = 0;
18 \text{ bx} = 1;
19 \text{ alpha} = 0;
20 beta = 0;
21 f = @(x) \text{ ones(size(x))}; % f(x) = 1
h = (bx-ax) / (m+1);
x = linspace(ax, bx, m+2)';
26 % determine exact solution to linear system by setting up
27 % system Au = f and solving with backslash:
28 e = ones(m+2,1);
30 A = 1/h^2 * spdiags([e -2*e e], [-1 0 1], m+2, m+2);
A(1,1:2) = [1 0];
32 A (m+2, m+1:m+2) = [0 1];
33 \text{ rhs} = f(x);
34 \text{ rhs}(1) = \text{alpha};
35 rhs(m+2) = beta;
36 ustar = A \cdot rhs;
37
38
39 % Decompose A = DA - LA - UA:
40 DA = diag(diag(A)); % diagonal part of A
                        % strictly lower triangular part of A (negated)
41 LA = DA - tril(A);
                        % strictly upper triangular part of A (negated)
42 UA = DA - triu(A);
44 % set up iteration matrix:
45 switch method
     case 'Jacobi'
46
       M = DA;
47
        N = LA + UA;
48
     case 'GS'
49
        M = DA - LA;
50
        N = UA;
52 % Added this case to impliment the backwards GS method
    case 'BGS'
```

```
M = DA - UA;
          N = LA;
      case 'SOR'
56
57
          % use optimal omega for this problem:
58
   %4.1.b. iterating the SOR method through different omega values ...
59
       commented
         out below
60
61
62
          omega = 2 / (1 + \sin(pi*h));
63
          %omega = 1.8;
          %omega = 1.95;
64
          M = 1/omega * (DA - omega*LA);
65
          N = 1/\text{omega} * ((1-\text{omega})*DA + \text{omega*UA});
66
    end
68
70 u = zeros(size(x)); % initial guess for iteration
72 figure(1)
73 clf
74 plot(x,ustar, 'r')
75 hold on
76 plot(x,u)
78 error = nan(maxiter+1,1);
  error(1) = max(abs(u-ustar));
80
81
   % Iteration:
83
  %Bad iteration method for the SOR method(commented out for use of ...
       actual code)
86 %To run the bad iteration, uncomment lines 90-92 and comment out lines
  %94-96
88
   %for iter=1:maxiter
       uGS = (DA - LA) \setminus (UA*u + rhs);
        u = u + omega * (uGS - u);
93
   for iter=1:maxiter
94
      u = M \setminus (N*u + rhs);
95
96
      error(iter+1) = max(abs(u-ustar));
97
98
       if mod(iter, nplot) == 0
          % plot u every nplot iterations
99
          plot(x, u)
100
         title(sprintf('%s: Iteration %4i, error = %9.3e',...
101
                 method,iter,error(iter+1)),'FontSize',15)
102
103
          drawnow
          pause(.1)
      end
105
```

```
106 end
107
108 % plot errors vs. iteration:
109 figure(2)
110 semilogy(error)
111 axis([0 maxiter min(error)/10 1])
112 title('Errors','FontSize',15)
113 xlabel('iteration')
114 hold on
% legend(\{'GS','BGS'\},'Location','southwest') used as a legend for the
116 %comparison grpahs in the command window.
117
118 % compute spectral radius of iteration matrix G:
_{119} G = M \setminus N;
120 rhoG = max(abs(eig(full(G))));
121 disp(sprintf('Spectral radius of G for %s is %5f', method, rhoG))
```

```
1 %Stephanie Klumpe
2 %4.1.c.
3 %Spectral radius iterated over omega values between 0 and 2
4 clear;
5 close all;
6 clc;
7 \text{ nplot} = 10;
                        % iterate u is plotted every nplot iterations
8 maxiter = 500;
                        % number of iterations to take
10 \text{ m} = 39;
11 \text{ ax} = 0;
12 \text{ bx} = 1;
13 alpha = 0;
14 beta = 0;
15 f = Q(x) ones(size(x)); % f(x) = 1
17 h = (bx-ax) / (m+1);
18 x = linspace(ax, bx, m+2)';
20 % determine exact solution to linear system by setting up
21 % system Au = f and solving with backslash:
22 e = ones(m+2,1);
23
A = 1/h^2 * spdiags([e -2*e e], [-1 0 1], m+2, m+2);
25 A(1,1:2) = [1 0];
26 A(m+2, m+1: m+2) = [0 1];
27 rhs = f(x);
28 rhs(1) = alpha;
29 rhs (m+2) = beta;
30 ustar = A \cdot rhs;
33 % Decompose A = DA - LA - UA:
34 DA = diag(diag(A)); % diagonal part of A
35 LA = DA - tril(A); % strictly lower triangular part of A (negated)
36 UA = DA - triu(A);
                        % strictly upper triangular part of A (negated)
omega=linspace(1e-16,2);
  for i=1:length(omega)
       M = 1/omega(i) * (DA - omega(i)*LA);
      N = 1/\text{omega}(i) * ((1-\text{omega}(i))*DA + \text{omega}(i)*UA);
40
      u = zeros(size(x)); % initial guess for iteration
41
       G = M \setminus N;
42
       rhoG(i) = max(abs(eig(full(G))));
44 end
45 figure
46 axis([0 2 0.84 1])
47 xlabel('\omega')
48 ylabel('g(\omega)')
49 hold on;
50 plot(omega, rhoG);
51 hold off;
```