

## Project 1

Math 458

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Github Link to All Code: [https://github.com/kartikbalodi/math458\\_project1](https://github.com/kartikbalodi/math458_project1)

### Assignment

- (1) By writing  $u_{ij}^N$  as a vector  $u^N$ ,  $u^N$  is a solution to a system of equations of the form

$$A^N u^N = b^N.$$

Find the matrices  $A^N$  and vector  $b^N$  and write these down explicitly in the case when  $N = 3$ . Notice, the matrix you get depends on how you write  $u_{ij}^N$  as a vector. Choose a way that will make it easy to code finding the product of  $A^N$  and a vector.

We write  $u_{ij}^N$  as a  $N^2 \times 1$  vector, which corresponds to  $1 \leq i \leq N, 1 \leq j \leq N$ . Thus, the matrix  $A^N$  take the form of a  $N^2 \times N^2$  array and  $b^N$  takes the form of a  $N^2 \times 1$  vector.

For  $u^3 = [u_{1,1} \ u_{2,1} \ u_{3,1} \ u_{1,2} \ u_{2,2} \ u_{3,2} \ u_{1,3} \ u_{2,3} \ u_{3,3}]^T$ , we express  $b^3$  and  $A^3$

$$b^3 = [b_{1,1} \ b_{2,1} \ b_{3,1} \ b_{1,2} \ b_{2,2} \ b_{3,2} \ b_{1,3} \ b_{2,3} \ b_{3,3}]^T = [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]^T$$

$$A^3 = \begin{bmatrix} 4 & -1 & 0 & -1 & 0 & 0 & 0 & 0 & 0 \\ -1 & 4 & -1 & 0 & -1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 4 & 0 & 0 & -1 & 0 & 0 & 0 \\ -1 & 0 & 0 & 4 & -1 & 0 & -1 & 0 & 0 \\ 0 & -1 & 0 & -1 & 4 & -1 & 0 & -1 & 0 \\ 0 & 0 & -1 & 0 & -1 & 4 & 0 & 0 & -1 \\ 0 & 0 & 0 & -1 & 0 & 0 & 4 & -1 & 0 \\ 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 & -1 \\ 0 & 0 & 0 & 0 & 0 & -1 & 0 & -1 & 4 \end{bmatrix}$$

Code:

```
N = 10; % N

% initialize b to g(xij, yij)
h = 1/(N+1);
b = zeros(N^2,1);
for i = 1:N
    for j = 1:N
        if i/(N+1) > 1/5 && i/(N+1) < 3/5 && j/(N+1) > 1/4 && j/(N+1) < 1/2
            b((j-1)*N+i,1) = -h^2; % bij = -h^2*g(xij,yij)
        end
    end
end

% filling in A
A = zeros(N^2,N^2);
for j = 1:N
    for i = 1:N
        %
        %      B I 0          4 1 0
        % A= I B I where B = 1 4 1 , I is 3x3 negative identity matrix and
        %      0 I B          0 1 4    0 is 3x3 zero matrix
        % we use this to get
        % 4ui;j - ui+1;j - ui-1;j - ui;j+1 - ui;j-1
        % fill in identity matrices on the side diagonals
        c = (j-1)*N;
        if j > 1
            A(c+i-N, c+i) = -1;
        end
        if j < N
            A(c+i+N, c+i) = -1;
        end
        % then fill in B
        A(c+i,c+i) = 4;
        if i ~= 1
            A(c+i, c+i-1) = -1;
        end
        if i ~= N
            A(c+i, c+i+1) = -1;
        end
    end
end

fmt = [repmat('%4d ', 1, size(A,2)-1), '%4d\n'];
fprintf(fmt, A. '); %transpose is important
```

- (2) Write a matlab script that generates the matrix  $A^N$  and vector  $b^N$  for any  $N$  and solves the system of equations using each of the following methods.
- Jacobi method,
  - Gauss-Seidel with successive over-relaxation with any parameter  $\omega$ , and
  - the conjugate gradient method.

Github: [https://github.com/kartikbalodi/math458\\_project1](https://github.com/kartikbalodi/math458_project1)

Jacobi:

```
%-----
% Start of Jacobi Method
%-----

% these are initializations which are used for every program
atol = 1e-3; % absolute tolerance
programTimeLimit = 300; % runtime limit in seconds
x0 = zeros(N^2, 1); % initial guess
r = norm(b - A * x0, 1); % initial residual (b-A*xk)
xk = x0; % kth iteration
iterCount = 0; % number of iterations completed

figure(1);
ax = axes();
hold(ax)
xlabel(ax, 'iterations')
ylabel(ax, 'residual')
%title(ax,'residual against time/s and iteration/count')
ax_top = axes();
hold(ax_top)
ax_top.Position = ax.Position;
ax_top.YAxis.Visible = 'off';
ax_top.XAxisLocation = 'top';
xlabel(ax_top, 'time')

tic
% repeat iterations until r <= atol or program time limit reached
while r > atol
    plot(ax,iterCount,r,'kx');
    time = toc;
    plot(ax_top,time,r,'kx');
    if time > programTimeLimit
        disp('Program time limit '+programTimeLimit+' reached');
    break
end
drawnow

% find the x in next iteration by given formula
iterNext = zeros(N^2,1);
for i = 1:N^2
    sum = 0;
    for j = 1:N^2
        if j ~= i
            sum = sum + A(i,j) * xk(j, 1);
        end
    end
    iterNext(i, 1) = (b(i,1)-sum)/A(i,i);
end
xk = iterNext;
r = norm(b - A * xk, 1); % update residual
iterCount = iterCount + 1; % increment iteration count
end

% plot one more time for the last pt i.e. first r > atol
plot(ax,iterCount,r,'kx');
time = toc;
plot(ax_top,time,r,'kx');
drawnow

disp('N = ');
disp(N);
disp('time = ');
disp(time);
disp('iterations = ');
disp(iterCount);

%hold(ax_top,'off')
%hold(ax,'off')
```

## Gauss-Seidel with SOR:

```
%Stopping Conditions: a tolerance limit and max number of
iterations
%before quitting
tolerance = 1e-12;
max_iterations = 100000;
%LDU Decomposition of A
%Isolates lower diagonal (below the main diagonal)
L = -tril(A, -1);
%Isolates main diagonal
D = diag(diag(A));
%Isolates upper diagonal (above the main diagonal)
U = -triu(A, 1);
%Creates initial solution array of zeros
x = zeros(length(A));
%Stored in reference since it is used repeatedly
%A in the equation  $Ax_i = b$  where  $x_i$  represents the current
%estimation of x
ref = (D - omega * L);
%Repeats until stopping criterion of iterations is met
for i = 1:max_iterations
    %Solving the iteration step
    x =
mldivide(ref, ((1-omega)*D+omega*U)*x_init)+omega*mldivide(ref
,b);
    %Ends the loop if stopping criterion of tolerance is
    %met
    %Tolerance is the max. acceptable distance between x
    %and x_init
    %(previous estimate of x)
    if norm(x - x_init) < tolerance
        break;
    end
    %x_init stores previous value of x
    x_init = x;
end
```

Conjugate Gradient:

%-----

% Start of Conjugate Gradient Method

%-----

```
% Conjugate Gradient Method
% Takes in matrix A, matrix b, and solves for our solution x
function x = conjGradMethod(A, b, x)
    % use a variable for tolerance to easily change it
    tolerance = 1e-3;
    figure(1);
    ax = axes();
    hold(ax);
    xlabel(ax, 'iterations');
    ylabel (ax, 'residual');
    iterCount = 0;
    tic
    % initialize r to -1/2 (gradient)
    r = b - A*x;
    % initialize p
    p = r;
    % n iterations
    while (norm(r) >= tolerance)
        norm1 = norm(r);
        plot(ax, iterCount, norm1, 'kx');
        time = toc;
        title(time);
        drawnow()
        % calculate Ap and p' * Ap to avoid multiple calculations
        Ap = A * p;
        pAp = p' * Ap;
        x = x + (r' * r) / (pAp) * p;
        r = r - (p' * r) / (pAp) * Ap;
        % return if norm is less than tolerance already
        if (norm(r) < tolerance)
            % print out our answer
            disp(x);
            return;
        end
        % calculate our new p
        p = r - (r' * Ap) / (pAp) * p;
        iterCount = iterCount + 1;
    end
end
```

- (3) Experiment with different values of the relaxation parameter  $\omega$  to find an optimal choice for the SOR method. Describe how you did this and how you decided what choice was optimal.

```
%Set value of omega, can change for tests
omega = 0.01:0.01:1.99;
x = zeros(N^2, 1);

least_time = 10000000;
best_omega = 0;
for i = 1:length(omega)
    tic;
    x = SORMethod(A,b,x, omega(i));
    time = toc;
    if time < least_time
        best_omega = omega(i);
        least_time = time;
    end
end

function x = SORMethod(A,b,x_init, omega)
    %Stopping Conditions: a tolerance limit and max number of iterations
    %before quitting
    tolerance = 1e-12;
    max_iterations = 100000;

    %LDU Decomposition of A
    %Isolates lower diagonal (below the main diagonal)
    L = -tril(A, -1);
    %Isolates main diagonal
    D = diag(diag(A));
    %Isolates upper diagonal (above the main diagonal)
    U = -triu(A, 1);

    %Creates initial solution array of zeros
    x = zeros(length(A));

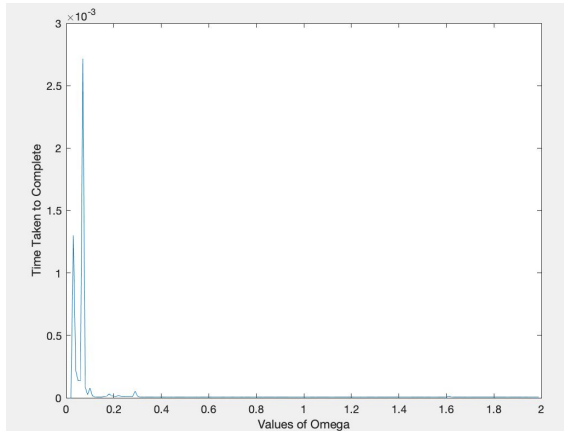
    %Stored in reference since it is used repeatedly
    %A in the equation Ax_i = b where x_i represents the current
    %estimation of x
    ref = (D - omega * L);

    %Repeats until stopping criterion of iterations is met
    for i = 1:max_iterations
        %Solving the iteration step
        x = mldivide(ref,((1-omega)*D+omega*U)*x_init+omega*mldivide(ref,b));

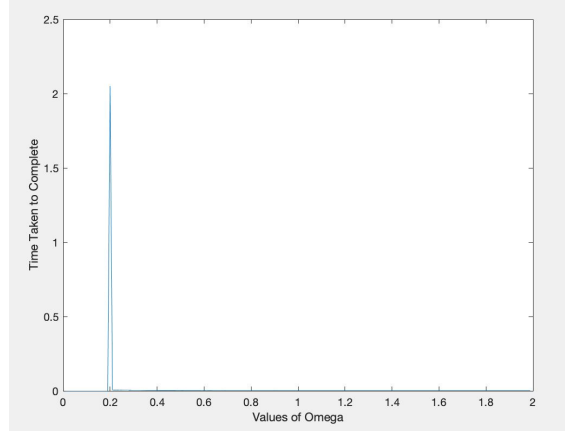
        %Ends the loop if stopping criterion of tolerance is met
        %Tolerance is the max. acceptable distance between x and x_init
        % (previous estimate of x)
        if norm(x - x_init) < tolerance
            break;
        end
        %x_init stores previous value of x
        x_init = x;
    end
end
```

Above is the code used to calculate the optimal omega value for various values of N (generating the matrix from part (1)). We found that the optimal omega values increase w.r.t. N, as shown in the following graphs:

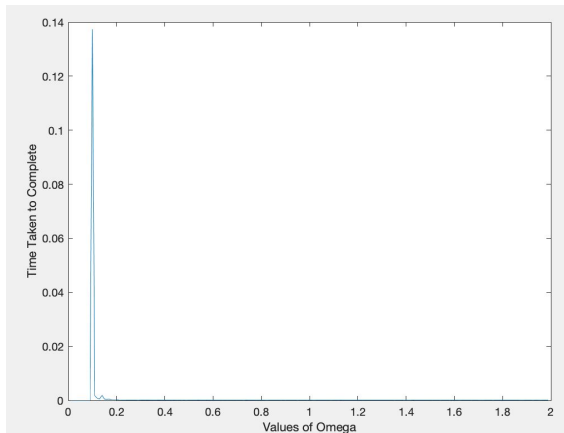
N = 3 (Best Omega = 0.990)



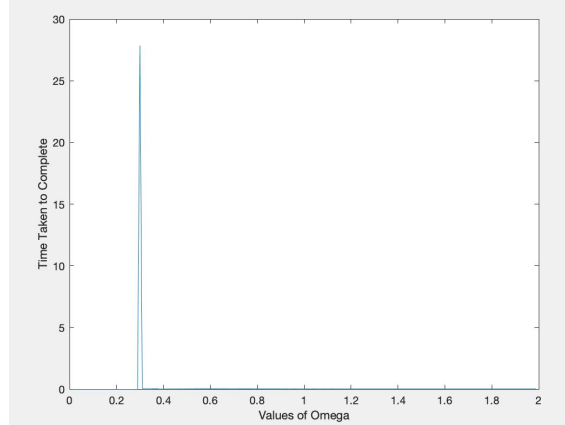
N = 20 (Best Omega = 1.830)



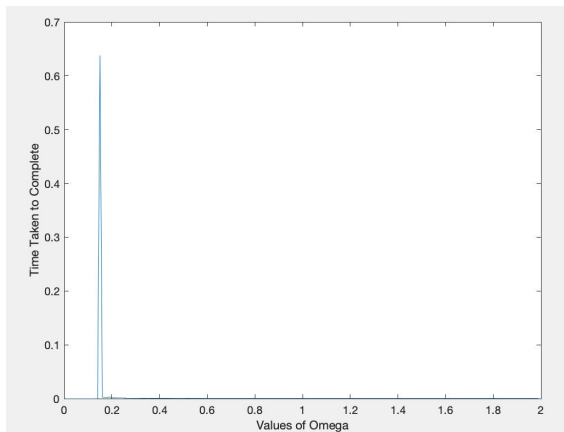
N = 10 (Best Omega = 1.410)



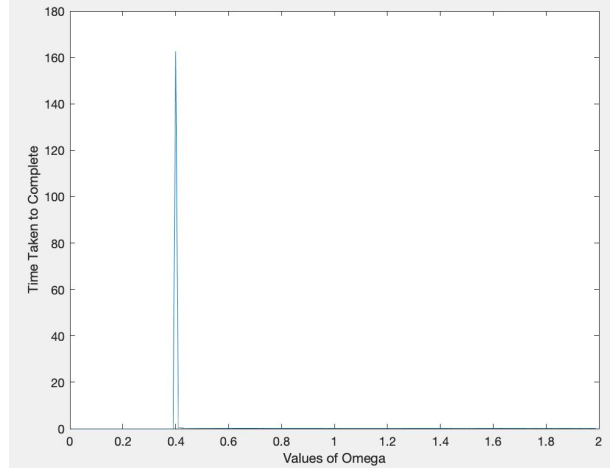
N = 30 (Best Omega = 1.930)



N = 15 (Best Omega = 1.680)



N = 40 (Best Omega = 1.940)



- (4) Experiment with the three methods of solving the equations (using the optimal choice of  $\omega$ ). Use as large an  $N$  as you can. Produce pictures similar to Figure 7.5 on page 185 of the text but remember it is not just the *number* of iterations that is important but the efficiency of the algorithm, so you might also try plotting the residual norm against time. (The `tic` and `toc` Matlab commands may be helpful for this.)

Absolute tolerance was set to  $1e-3$ , and  $N$  was measured and tabulated till time taken exceeded 150s.

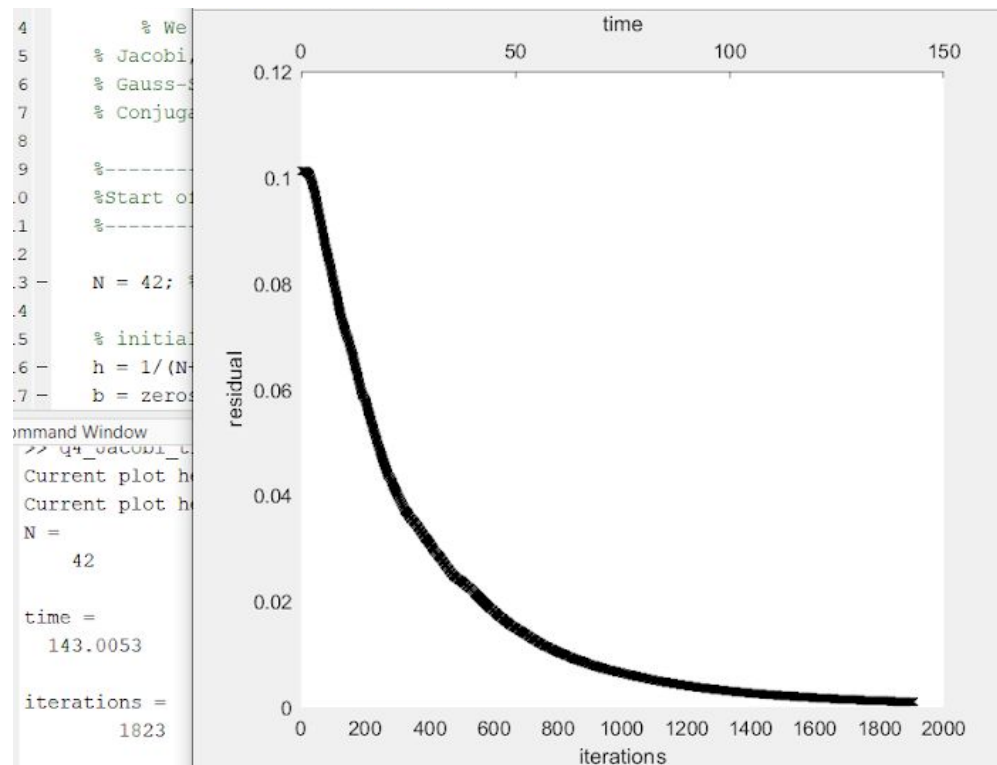
All plots were plotted as residuals against time(top x-axis) and iteration count (bottom x-axis).

### Jacobi Analysis:

N	Time Taken	Iteration Count
10	3.4230 seconds	118
20	11.6114 seconds	427
30	36.8058 seconds	946
40	105.5501 seconds	1643
<b>42</b>	<b>143.0053 seconds</b>	<b>1823</b>
43	166.9965 seconds	1879

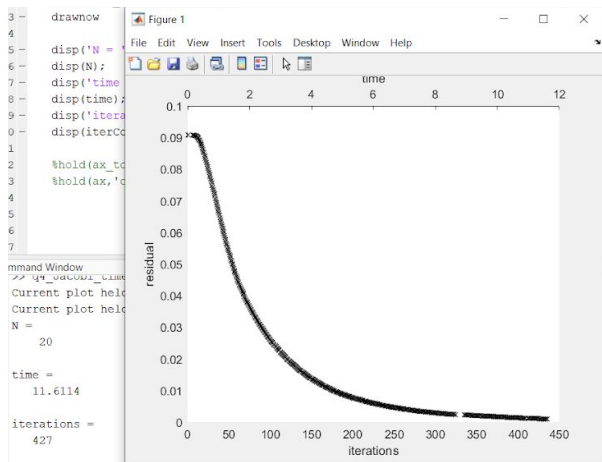
Highest  $N$  possible for Jacobi method is  $N=42$ .

Plot:

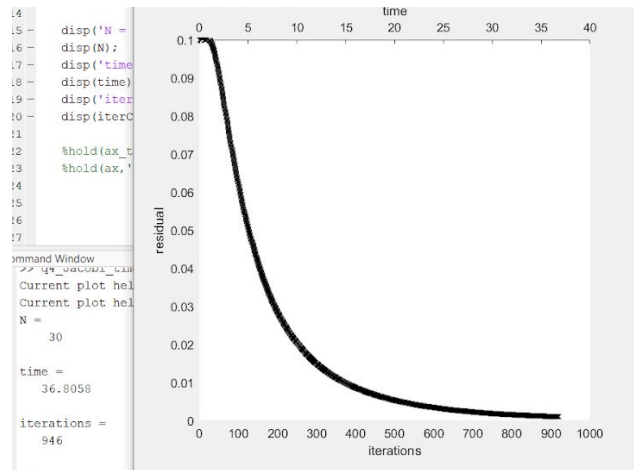




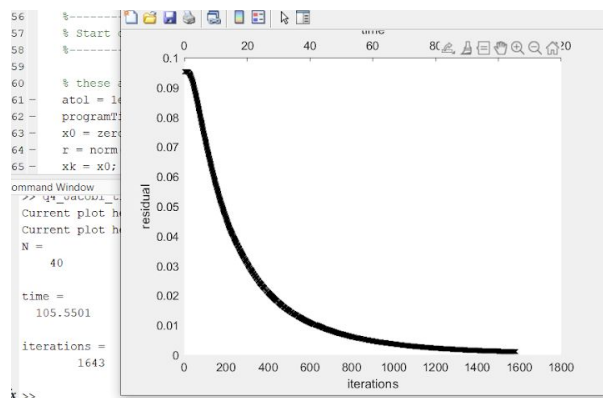
Plot for N=20, Jacobi:



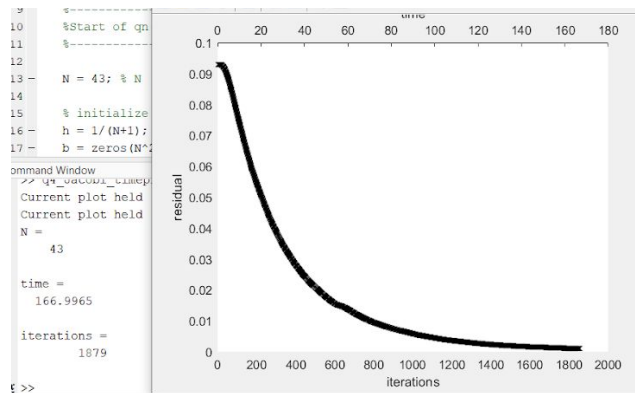
Plot for N=30, Jacobi:



Plot for N=40, Jacobi:



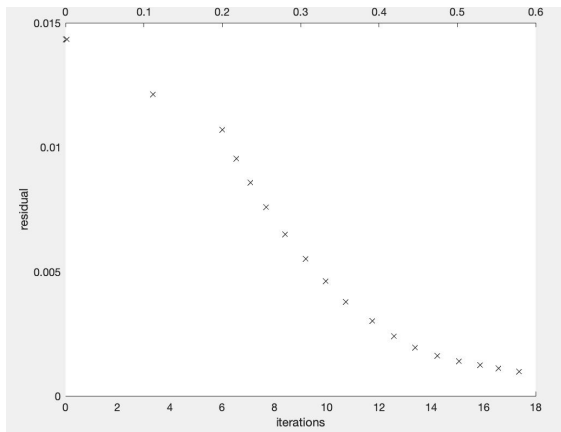
Plot for N=43, Jacobi:



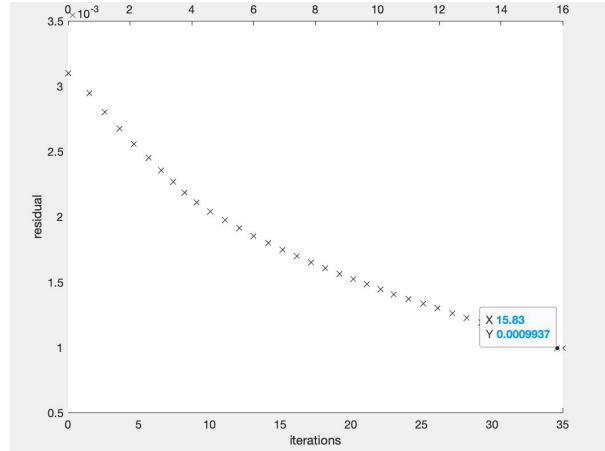
## Gauss-Seidel:

N	Time Taken	Iteration Count
20	0.579 seconds	18
40	0.9904 seconds	25
60	3.038 seconds	31
100	15.83 seconds	35
<b>125</b>	<b>69.464 seconds</b>	<b>35</b>
130	205.894 seconds	36

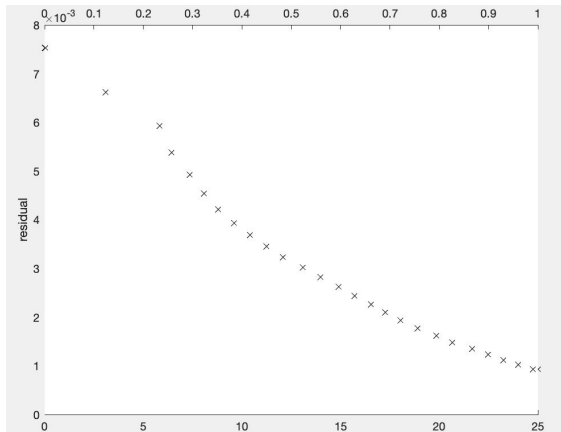
Plot for N = 20:



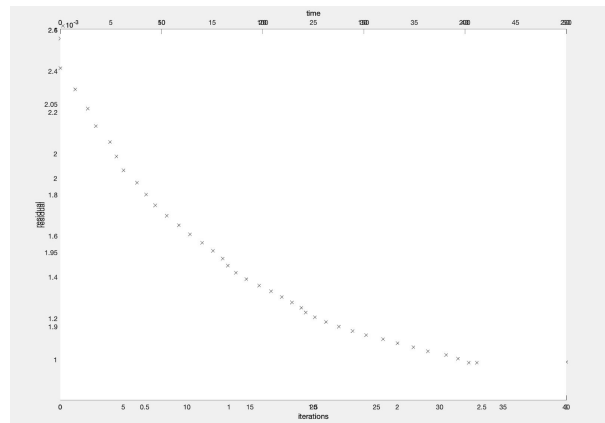
Plot for N = 100:



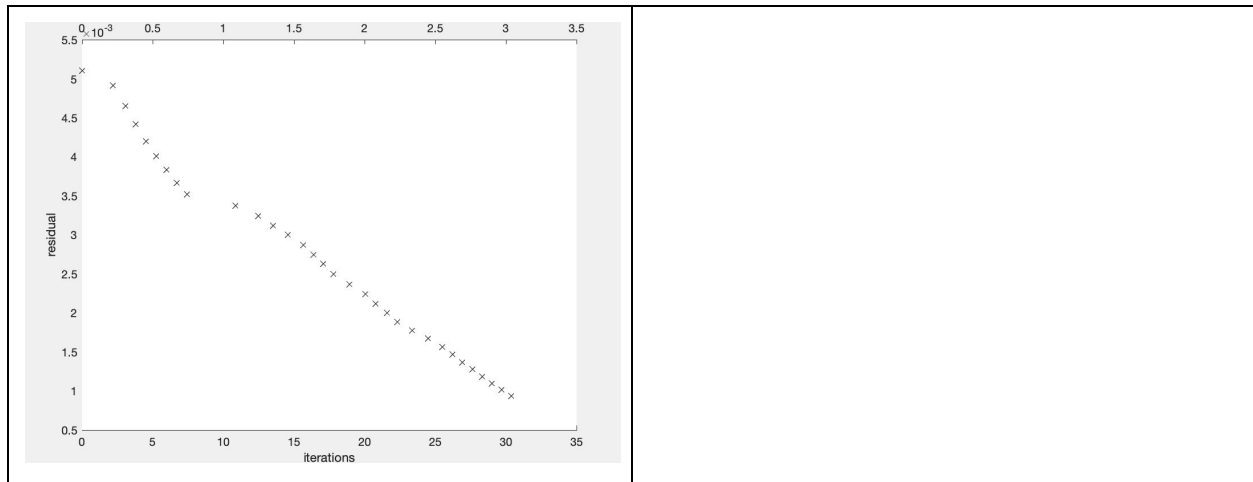
Plot for N = 40:



Plot for N = 130:



Plot for N = 60:



At 200, Matlab runs out of memory:

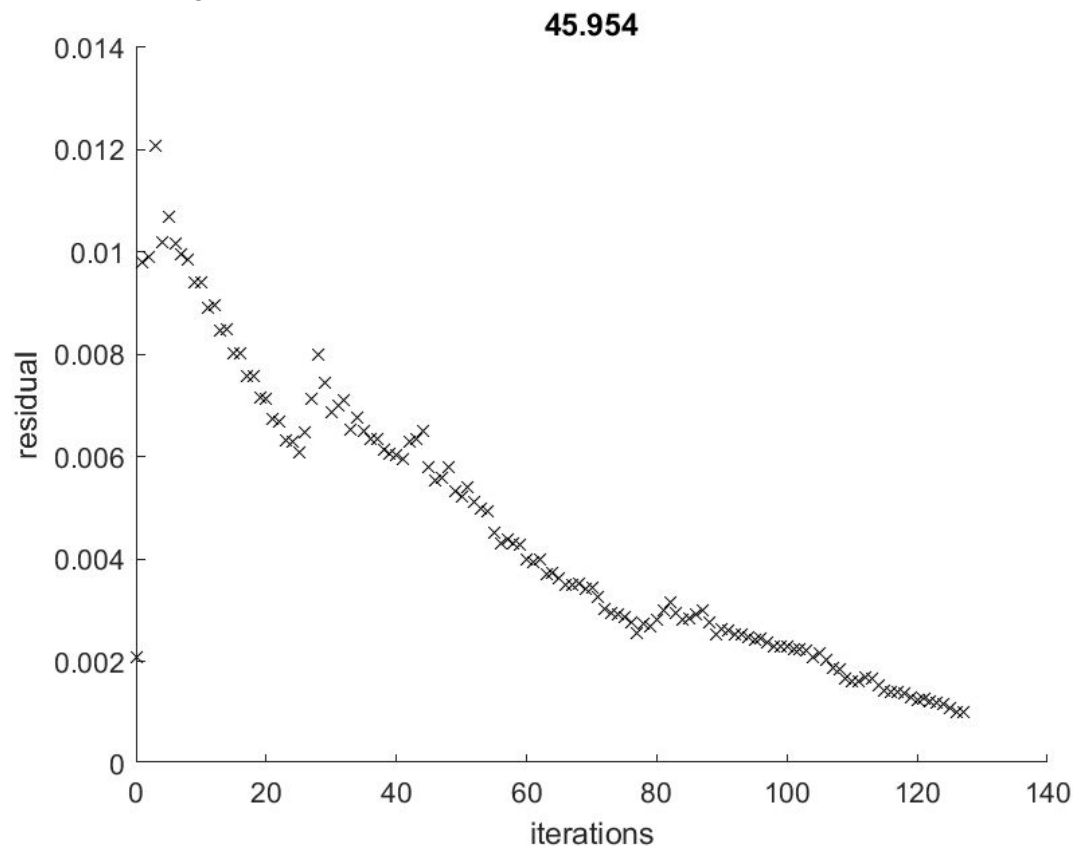
The screenshot displays the MATLAB interface with three main windows:

- Documentation Window:** Titled "Resolve 'Out of Memory' Errors", it explains that MATLAB is a 64-bit application and can run out of memory. It provides a code example: `A = rand(1e6,1e6);` and an error message: "Error using rand. Requested 1000000x1000000 (7450.6GB) array exceeds maximum array size preference. Create MATLAB to become unresponsive." It also includes a "More information" link.
- Command Window:** Shows the execution of the script `q4_gssor.m`. It displays the error message: "Error in q4\_gssor (line 68) U = -triu(A, 1);". It also includes a "More information" link.
- Workspace Window:** Lists variables in the workspace, including `A` (40000x40000 double), `ax` (1x1 Axes), `ax_top` (1x1 Axes), `b` (40000x1 double), `c` (39800), `d` (2025x1 double), `D` (40000x40000 double), `h` (0.0050), `i` (200), `iter_nums` (200x200 double), `iterationcount` (190), `iterNext` (1764x1 double), `j` (200), `L` (40000x40000 double), `max_iterations` (100000), `N` (200), `nextit` (2025x1 double), `num_iter` (0), `num_iterations` (1), `omega` (1.9300), `optit` (94), `optOmega` (1.9000), `programTime` (300), `r` (0.0016), `ref` (9x9 double), `s` ("Total time:"), `sum` (1.2436e-04), `sumTerm` (1.0363e-04), `time` (0), `timeLimit` (30), `tolerance` (1.0000e-03), `U` (2025x2025 double), and `x` (9x9 double).

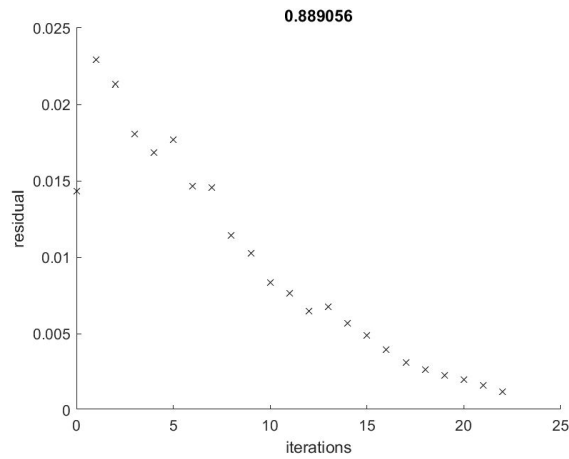
### Conjugate Gradient Analysis:

N	Time Taken	Iteration Count
20	.889 seconds	22
50	2.284 seconds	50
100	6.208 seconds	93
120	12.002 seconds	110
130	15.493 seconds	114
145	34.872 seconds	126
<b>149</b>	<b>45.954 seconds</b>	<b>127</b>
150	350.591 seconds	131

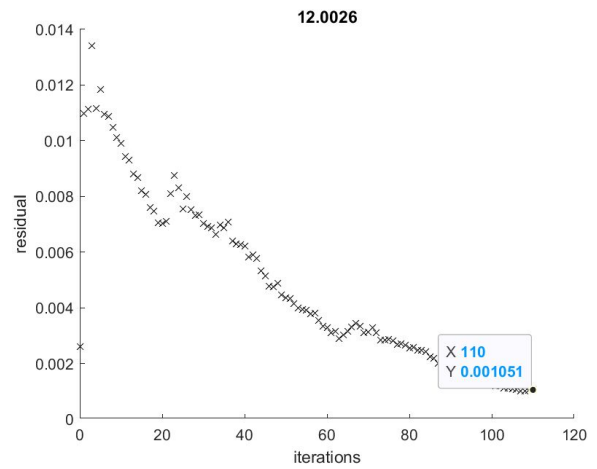
Plot for winning time (N = 149):



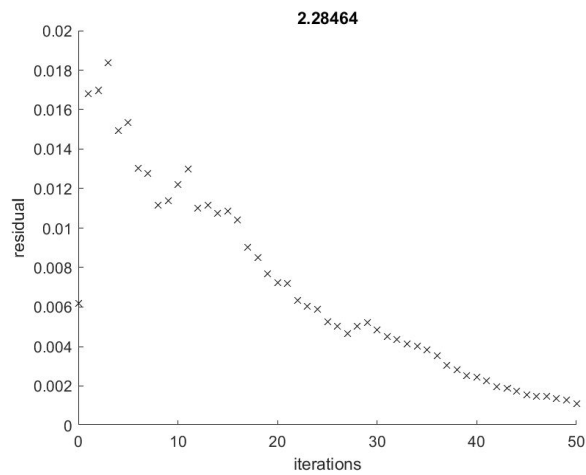
Plot for N = 20 (CGM):



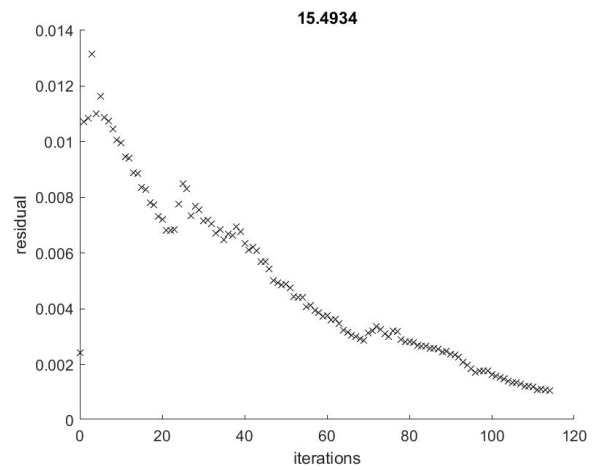
Plot for N = 120 (CGM):



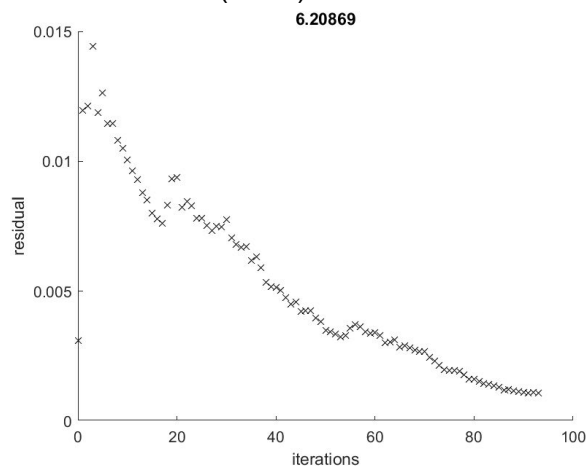
Plot for N = 50 (CGM):



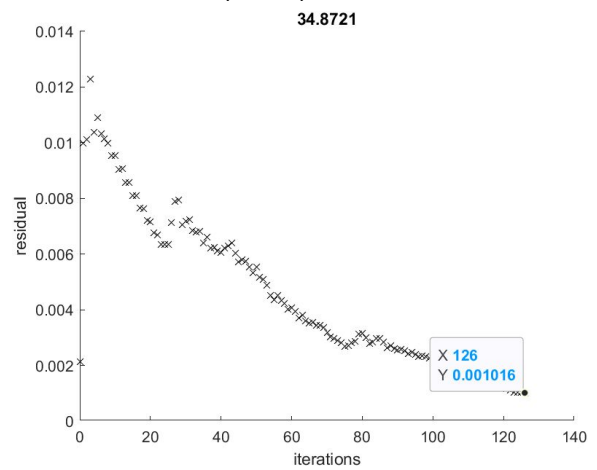
Plot for N = 130 (CGM):



Plot for N = 100 (CGM):



Plot for N = 145 (CGM):



- (5) Use your best performing algorithm with a large value of  $N$  and plot your solution.

Conjugate gradient method is far more superior in comparison to Gauss-Seidel with SOR, which in turn is far better than Jacobi.

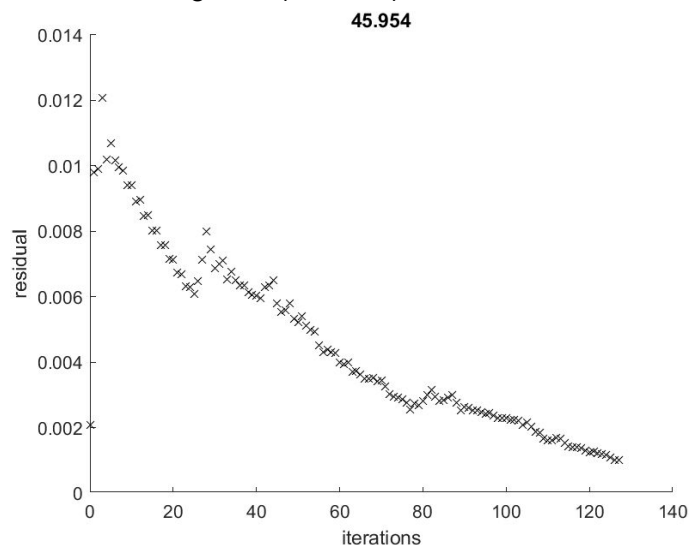
Something interesting to note is the large change in time between  $N=149$  and  $N=150$ , despite the iterations being approximately equal. Our theory is that matlab may have ran out of the space on the stack needed to perform the computations, which would require lengthy reshuffling of space. This would significantly increase the time of the computations, and would help explain the steep incline of time taken. This is further corroborated by the fact that each of the consecutive  $N$  values (151, 152, ...) were also consistent with the large jump in time taken.

If this is the case, our maximum  $N$  value could have been notably larger, judging from the relatively short time that  $N=149$  took.

Our plot is below. As you can see, the number of iterations (127) was relatively small compared to the other two methods for their maxed out  $N$ .

### Conjugate Gradient Method Solution:

Plot for winning time ( $N = 149$ ):



- (6) For each method, tell me who it was in your group who took primary responsibility for coding it up. Also, briefly describe the contributions of each of you to the rest of the project.

Kartik:

Code for part (1), Jacobi + plots & time data for Jacobi. Corresponding sections of writeup.

Saket:

Code for part (3), Gauss-Seidel + plots & time data for Gauss-Seidel. Corresponding sections of writeup.

Heather:

Code for conjugate gradient + plots & time data for conjugate gradient. Corresponding sections of writeup, part(5).