Math 5301 – Numerical Analysis – Spring 2025 w/Professor Du

 $\begin{array}{c} {\rm Paul~Carmody} \\ {\rm Homework}~\#4-{\rm March}~26,~2025 \end{array}$

Assignment: Consider 1D Poisson Equation $-\Delta x = \sin(\pi x)$, over the region $(0, \pi/2)$ with boundary conditions u(0) = 0 and $u(\pi/2) = 1$. Using central difference scheme and a mesh of 128, obtain a linear system of Au = f for the problem, then the solve the system using the following methods until a relative residual of 10^{-4} is reached. For all methods below, plot the analytical solution, the numerical solution and the error distribution. Use zero vectors as your initial guess.

- (a) Jacobis's Method. Plot the rate of convergence and compareit with analysis.
- (b) Steepest Descent Method. Compare the rate of convergence with that obtained in (a).

Analytical and Numerical Solutions to Poisson's Equation

Introduction:

We will begin by describing the equation and providing an analytical solution. Then solve this equation using Jacobi's Method and the Steepest Descent Method. Comparisons will be made as to accurancy and rate of convergence.

Poison's Equation and an Analytical Solution:

As described in the assignment we will focus our attenion on this Poisson equation and initial conditions.

$$-\Delta x = \sin(\pi x)$$

 $u(0) = 0$ and $u(\pi/2) = 1$.

When we solve this problem analtyically we get

$$u'(x) = \int -\sin(\pi x)dx$$

$$= \frac{1}{\pi}\cos(\pi x) + C$$

$$u(x) = \int \left(\frac{1}{\pi}\cos(\pi x) + C\right)dx$$

$$= \frac{1}{\pi^2}\sin(\pi x) + Cx + D$$

$$u(0) = 0 = \frac{1}{\pi^2}\sin(\pi 0) + Cx + D$$

$$D = 0$$

$$u(\pi/2) = 1 = \frac{1}{\pi^2}\sin(\pi^2/2) + C(\pi/2)$$

$$C = \frac{2\left(1 - \frac{1}{\pi^2}\right)\sin(\pi^2/2)}{\pi^2} = \frac{2\left(1 - \frac{1}{9.869604064}\right)0.086022097}{9.869604064} = 1.998233797 \approx 2$$

$$u(x) \approx \frac{1}{\pi^2}\sin(\pi x) + 2x$$

Centered Difference Scheme

We will attempt to approximate the curve of the solution at particular points u_i by calculating a slope at a point by using the preceding point u_{i-1} and succeeding point u_{i+1} .

$$\frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} \approx -\sin(\pi x)$$

where $h = (\pi/2)/129 = \pi/258$ (we use N+1 as we start with the left boundary 0). This can be reduced to

$$-u_{i+1} + 2u_i - u_{i-1} = h^2 \sin(\pi x).$$

This expands to a linear function over the a matrix A and vector $u = \{u_i\}$ reflecting the left hand side and the value to our function on the right with $f = \{f_i\}$, $f_i = \sin(\pi x_i)$ or

$$Au = h^{2}f$$

$$\begin{pmatrix} 2 & -1 & 0 & 0 & \cdots & 0 \\ -1 & 2 & -1 & 0 & \cdots & 0 \\ 0 & -1 & 2 & -1 & \cdots & 0 \\ 0 & 0 & -1 & 2 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & 2 \end{pmatrix} \begin{pmatrix} u_{1} \\ u_{2} \\ u_{3} \\ u_{4} \\ \vdots \\ u_{i} \end{pmatrix} = h^{2} \begin{pmatrix} f_{1} \\ f_{2} \\ f_{3} \\ f_{4} \\ \vdots \\ f_{i} \end{pmatrix}$$

$$u = h^{2}A^{-1}f$$

remembering the boundary conditions. Since A is tri-diagonal we can use several methods and compare the cost and efficiency. The MatLab code for the Central Difference Method is

```
clc; clear; close all;
  figure; hold on;
  function [x, dsc] = centered\_difference\_scheme(mesh)
6
       h = 1/mesh;
       x = 0:h: pi/2;
      N = length(x) - 2;
9
      % Construct finite difference matrix A
       A = (1/h^2) * (diag(-2*ones(N,1)) + diag(ones(N-1,1),1) + diag(ones(N-1,1),1)
12
          -1,1),-1));
       b = \sin(pi*x(2:end-1));
13
14
      % Solve the linear system A*u = b
      \%u = A \setminus reshape(b, [], 1);
16
       u = A \setminus b.
17
18
       % Include boundary values u(0) = 0, u(pi/2) = 1)
19
       dsc = [0; u; 1];
20
21
  end
22
  [x, u_full] = centered_difference_scheme(128);
24
25
  %calculate the analytic solution
```

```
% analytic = (-1/pi^2)*sin(pi*x);

28 analytic = 1/pi^2*sin(pi*x)+2/pi*(1-1/pi^2)*x;

29 % analytic = (-1/pi^2)*sin(pi*x)+2*x;

30

31 % Plot the solutions

32 plot(x, u_full, '-b', 'DisplayName', sprintf('h = %.3f', 1/128))

33 plot(x, analytic, '-r', 'DisplayName', 'analytic')

34 xlabel('x');

35 ylabel('u(x)');

36 title('Poisons Equation on [0,pi/2]: Solution using 128');

37 legend;
```

Jacobi's Iteration Method

Since A is a tri-diagonal matrix it can be divided into three separate matrices that add up. Let D be zero everywhere except the diagonal where it will hold the values of A_{ii} (namely all 2s). By applying this iteratively we get

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

Here is the MatLab code used to generate the graphs that follow:

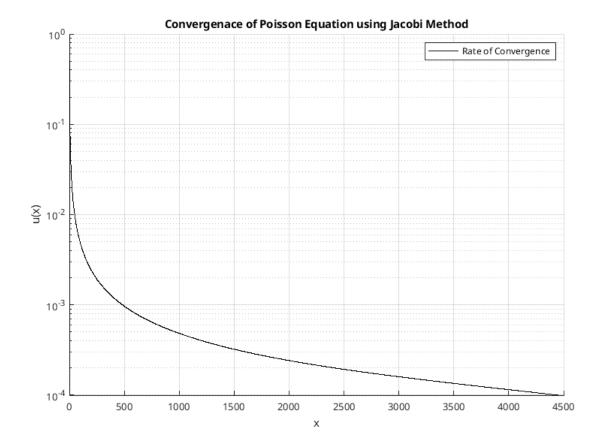
```
clc; clear; close all;
  figure; hold on;
  function [conv, cnt] = jacobi_method(A, b, h, tol, max_iter)
      % Solves Ax = b using Jacobi's iterative method
6
      % Inputs:
      %
                     - Coefficient matrix (NxN)
          Α
      %
                    - Right-hand side vector (Nx1)
           b
                    - Convergence tolerance
10
           max_iter - Maximum number of iterations
      % Output:
12
           x - Solution vector (Nx1)
13
14
      N = length(b);
                                     % Number of equations
15
       x = b;\%zeros(N, 1);
                                        % Initial guess (zero vector)
                                     % Store previous iteration
       x_{old} = x;
       h2=h^2;
18
       conv_loc = 1:10000;
19
       conv=conv_loc;
20
       cnt = 0;
21
       for k=1:max_iter
23
           for i=2:N-1
24
               x(i) = 1/2*(x_old(i-1) + x_old(i+1) - h2*b(i));
25
           end
26
27
           % Check for convergence
28
            if norm(x - x_old, 2) < tol
29
            conv_loc(k) = norm(x-x_old, Inf);
30
            cnt = k;
31
32
           if norm(x - x_old, Inf) < tol
33
```

```
fprintf('Converged in %d iterations.\n', k);
34
               conv = conv_loc;
35
               return;
36
           end
           x_{old} = x; % Update solution
      end
39
40
       fprintf('Max iterations reached without convergence.\n');
41
      conv = conv_loc;
42
  end
43
  % Define problem parameters
44
  N = 128;
                              % Number of internal grid points
  h = (pi/2) / (N+1);
                              % Grid spacing
  x = linspace(h, pi/2-h, N); % Grid points
                            % Right—hand side vector
  f = h^2 * sin(pi * x);
48
49
  % Construct tridiagonal matrix A
  A = 2 * eye(N) - diag(ones(N-1,1),1) - diag(ones(N-1,1),-1);
51
  % Modify last element of f to include boundary condition u(pi/2) = 1
  f(1) = 0; f(N) = 1;
54
55
  % Solve using Jacobi method
56
  tol = 1e-4; % Convergence tolerance
57
  max_iter = 10000; % Maximum iterations
59
  [u, cnt] = jacobi_method(A, f, h, tol, max_iter);
60
61
  output = u(1:cnt);
62
  grid on;
63
  yscale log;
  horizontal = 1:length(output); %linspace(1, cnt, 1);
  plot (horizontal, output, '-k', 'DisplayName', 'Rate of Convergence');
  xlabel('x'); ylabel('u(x)');
67
  title ('Convergenace of Poisson Equation using Jacobi Method');
68
  legend;
```

From this we generate the following graphs.

grid on;

70



Steepest Descent Method

This technique for approximating our solution to the Poisson equation is to make each iteration in the direction of greatest change. That is,

$$\nabla \phi(u_{k-1}) = Au_{k-1} - f \equiv -r_{k-1}$$

where $\phi: \mathbb{R}^m \to \mathbb{R}$ of the form

$$\phi(u) = \frac{1}{2}u^T A u - u^T f$$

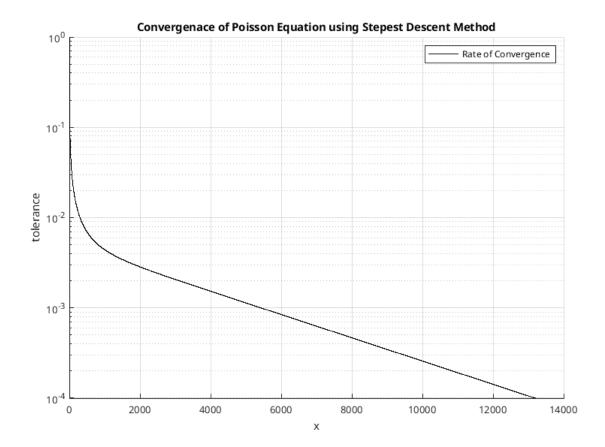
which is a quadratic function in u and can be mapped with local extrema either as a top, a bowl or a saddle point, all based on the eigenvalues of A (negative, positive, or neither, respectively). Thus, when A is SPD we can expect r_k to progress ever closer towards the extremum.

This is the source code

```
clc; clear; close all;
  figure; hold on;
3
  function [conv, cnt] = steepest_descent(A, f, tol, max_iter)
      % Solves Ax = b using steepest descent method
      % Inputs:
      %
          A
                    - Coefficient matrix (NxN)
                    - Right-hand side vector (Nx1)
      %
          b
                    - Convergence tolerance
10
           max_iter - Maximum number of iterations
11
      % Output:
12
          x - Solution vector (Nx1)
13
      N = length(f);
                                    % Number of equations
15
```

```
u = zeros(N, 1);
                                    % Initial guess (zero vector)
16
       u_old = u;
17
      conv = 1:10000;
18
       for k=1:max_iter
20
           r_{old} = f - A*u_{old};
21
           conv(k) = norm(r_old);
22
           cnt = k;
23
24
           if norm(r_old) < tol
25
               fprintf('Converged in %d iterations.\n', k);
26
               return;
           end
28
           alpha_old = (r_old.' * r_old)/(r_old.' * A * r_old);
29
           u = u_old + alpha_old*r_old;
30
           u_old = u;
31
      end
32
       fprintf('Max iterations reached without convergence.\n');
  end
  % Define problem parameters
36
  N = 128;
                             % Number of internal grid points
  h = (pi/2) / (N+1);
                             % Grid spacing
  x = linspace(h, pi/2-h, N); % Grid points
                            % Right-hand side vector
  f = h^2 * sin(pi * x);
  % Construct tridiagonal matrix A
  A = 2 * eye(N) - diag(ones(N-1,1),1) - diag(ones(N-1,1),-1);
44
  % Modify last element of f to include boundary condition u(pi/2) = 1
45
  f(1) = 0; f(N) = 1;
  % Solve using steepest descent method
  tol = 1e-4; % Convergence tolerance
49
  max_iter = 100000; % Maximum iterations
50
51
  [conv, size] = steepest_descent(A, f, tol, max_iter);
52
  output = conv(1: size);
  grid on;
55
  yscale log;
56
  horizontal = 1:length(output); %linspace(1, cnt, 1);
  plot (horizontal, output, '-k', 'DisplayName', 'Rate of Convergence');
  xlabel('x'); ylabel('tolerance');
  title ('Convergenace of Poisson Equation using Stepest Descent Method');
  legend;
```

From this we generate the following graphs.



Analyzing the various solutions.

It should be noted that the Jacobi method took 4,452 iterations before being completed and the Steepest Descent Method took 13,454.

The two plots use a logorithmic scale. A standard scale didn't emphasize the decay rate effectively as the initial iterations showed the great change, so much so that the following improved accuracy couldn't be seen. The logarithmic scale more effictively shows the improvement. Each iteration provides very little but necessary improvement.