# Assignment 1

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March 23, 2024

For the n by n matrix with entries  $A_{ij} = \frac{5}{i+2j-1}$ , set  $x = [1, 1, ..., 1]^T$  and b = Ax. Use the Matlab's backslash command to compute  $\bar{x}$ , the double-precision computed solution. Find the infinity norm of the forward error,  $||x - \bar{x}||_{\infty}$ , and of the backward error,  $||A\bar{x} - b||_{\infty}$ . Compute the ratio between the relative forward error and the relative backward error of Ax = b in the infinity norm and compare this ratio with the condition number of A for n = 6.

#### Solution:

```
% Set the size of the matrix
n = 6;
\% Initialize the matrix A
A = zeros(n, n);
% Populate the matrix A with the given formula
for i = 1:n
   for j = 1:n
        A(i, j) = 5 / (i + 2*j - 1);
    end
end
% Create the vector x
x = ones(n, 1);
% Compute b = Ax
b = A * x;
% Use the backslash operator to solve Ax = b
x_bar = A \setminus b;
% Compute the infinity norm of the forward error ||x - x_bar||_inf
forward_error = norm(x - x_bar, inf);
% Compute Ax_bar
Ax_bar = A * x_bar;
% Compute the infinity norm of the backward error ||Ax_bar - b||_inf
backward_error = norm(Ax_bar - b, inf);
% Compute the relative forward error
relative_forward_error = forward_error / norm(x, inf);
% Compute the relative backward error
relative_backward_error = backward_error / norm(b, inf);
% Compute the condition number of A
condition_number_A = cond(A, inf);
% Compute the ratio between the relative forward error and the relative backward error
error_ratio = relative_forward_error / relative_backward_error;
Forward error: 6.4408e-10
Backward error: 4.4409e-16
Relative forward error: 6.4408e-10
Relative backward error: 7.2504e-17
```

Ratio between relative forward error and relative backward error: 8883341.6875 Condition number of A for n = 6: 70342013.9293

Here is what the system looks like for n = 6.

$$\begin{pmatrix} \frac{5}{2} & \frac{5}{4} & \frac{5}{6} & \frac{5}{8} & \frac{1}{2} & \frac{5}{12} \\ \frac{5}{3} & 1 & \frac{5}{7} & \frac{5}{9} & \frac{5}{11} & \frac{5}{13} \\ \frac{5}{4} & \frac{5}{6} & \frac{5}{8} & \frac{1}{2} & \frac{5}{12} & \frac{5}{14} \\ 1 & \frac{5}{7} & \frac{5}{9} & \frac{5}{9} & \frac{5}{11} & \frac{5}{13} & \frac{5}{16} \\ \frac{5}{6} & \frac{5}{8} & \frac{1}{2} & \frac{1}{12} & \frac{1}{14} & \frac{5}{16} \\ \frac{5}{7} & \frac{5}{9} & \frac{5}{11} & \frac{5}{13} & \frac{1}{3} & \frac{1}{3} & \frac{5}{17} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{49}{8} \\ \frac{43024}{9009} \\ \frac{223}{9009} \\ \frac{3401}{9009} \\ \frac{341}{112} \\ \frac{6161944556856429}{11286685248} \end{pmatrix}$$

The error ratio is 8,883,341.6875 and the condition number is 70,342,013.9293. The condition number is larger, implying that the system is ill-conditioned. It is also much larger than the error ratio.

## Question 3(A)

Let A be an n by n matrix. Show that if A is symmetric and positive definite, then  $a_{ii} > 0$  for any i = 1, 2, ..., n.

#### **Solution:**

Suppose A is an n by n, symmetric and positive definite matrix. Since A is positive definite,  $x^T A x > 0$  for all non-zero x. Choose  $x = e_i$ . Then  $e_i^T A e_i > 0$ . Since A is symmetric,  $A = A^T$ . So,  $e_i^T A e_i = e_i^T A^T e_i = (A e_i)^T e_i$ . Let  $y = A e_i$ . Then  $(A e_i)_i = a_{ii}$ . So,  $e_i^T A e_i = y^T e_i = a_{ii} > 0$ , as wanted.

## Question 3(B)

Let A be an n by n matrix. Show that if A is symmetric and positive definite, then the matrix A is non-singular.

#### **Solution:**

Suppose A is an n by n, symmetric and positive definite matrix.

Assume, for contradiction, that A is singular. Then det(A) = 0. Since A is symmetric, it has real eigenvalues. Let  $\lambda$  be an eigenvalue of A with corresponding eigenvector  $v \neq 0$ . Thus,  $Av = \lambda v$ .

Furthermore, since A is positive definite, we have  $v^T A v = v^T \lambda v = \lambda ||v||^2$ . If A is singular, there must be at least one  $\lambda = 0$ , so  $v^T A v = v^T \lambda v = 0$ , contradicting  $v^T A v > 0$  for all v. Therefore, A must be non-singular.

## Question 3(C)

Show that if A is a strictly diagonally dominant matrix, then (-A) is a strictly diagonally dominant matrix.

#### **Solution:**

Suppose A is a strictly diagonally dominant matrix with entries  $(a_{ii})$ . By definition then,  $|a_{ii}| > \sum_{j \neq i} |a_{ij}|$ . Consider -A with entries  $(-a_{ii})$ . Then  $|-a_{ii}| = |a_{ii}| > \sum_{j \neq i} |a_{ij}| = \sum_{j \neq i} |-a_{ij}|$ . This is the definition of a strictly diagonally dominant matrix, therefore -A is strictly diagonally dominant.

Apply the Jacobi method to approximate in Matlab the solution of the sparse system for Ax = b, with three correct decimal places (forward error in the infinity norm) for n = 100. The correct solution is  $[1, -1, 1, -1, ..., 1, -1]^T$ . Give the number of iterations needed. The system is

$$\begin{bmatrix} 2 & 1 & 0 & \cdots & 0 & 0 \\ 1 & 2 & 1 & \cdots & 0 & 0 \\ 0 & 1 & 2 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 2 & 1 \\ 0 & 0 & 0 & \cdots & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ -1 \end{bmatrix}$$

## Solution:

```
% Define the size of the system
n = 100;
% Create the sparse matrix A using diag function
A = diag(ones(1, n-1), -1) + 2*diag(ones(1, n), 0) + diag(ones(1, n-1), 1);
% Create the right-hand side vector b
b = [1; zeros(n-2, 1); -1];
% Initialize the solution vector x with zeros
x = zeros(n, 1);
% Set the tolerance for convergence
tolerance = 1e-6;
% Set the maximum number of iterations
maxIterations = 100000;
% Perform Jacobi iteration
for iter = 1:maxIterations
   xNew = zeros(n, 1);
   for i = 1:n
        xNew(i) = (b(i) - A(i, :)*x + A(i, i)*x(i)) / A(i, i);
   end
   % Check for convergence
    if norm(xNew - x, inf) < tolerance</pre>
        break;
    end
   % Update solution for the next iteration
   x = xNew;
end
% Display the result
disp(['Approximate solution: ', num2str(x', '%.3f')]);
disp(['Number of iterations needed: ', num2str(iter)]);
Approximate solution: 1.000-1.000 1.000-1.000 1.000-1.000 1.000-0.999 0.999-0.999 0.999-0.999 0.999-0.999
Number of iterations needed: 13276
```

Solve the following problem Ax = b as below by carrying out the conjugate gradient method by hand:

$$\begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

## Solution:

Following the algorithm, we start with  $x^{(0)} = \vec{0}$  and  $d^{(0)} = \pi^{(0)} = \vec{b}$  and stop when  $\pi(k) = \vec{0}$ . For the first iteration, we have k = 1:

$$\alpha_{1} = \frac{(\pi^{(0)})^{T}\pi^{(0)}}{(d^{(0)})^{T}Ad^{(0)}}$$

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

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

$$= \frac{1}{2}$$

$$x^{(1)} = x^{(0)} + \alpha_{1} \cdot d^{(0)}$$

$$= \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \frac{1}{2} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{2} \\ 0 \end{bmatrix}$$

$$\pi^{(1)} = \pi^{(0)} - \alpha_{1} \cdot A \cdot d^{(0)}$$

$$= \begin{bmatrix} 1 \\ 0 \end{bmatrix} - \begin{bmatrix} 1 \\ -\frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix}$$

$$\beta_{1} = \frac{(\pi^{(1)})^{T}\pi^{(1)}}{(d^{(0)})^{T}Ad^{(0)}}$$

$$= \frac{[0 \quad \frac{1}{2}] \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix}}{[1 \quad 0] \begin{bmatrix} 1 \\ 0 \end{bmatrix}}$$

$$= \frac{1}{4}$$

$$d^{(1)} = \pi^{(1)} + \beta_{1} \cdot d^{(0)}$$

$$= \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix} + \frac{1}{4} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{4} \\ \frac{1}{2} \end{bmatrix}$$

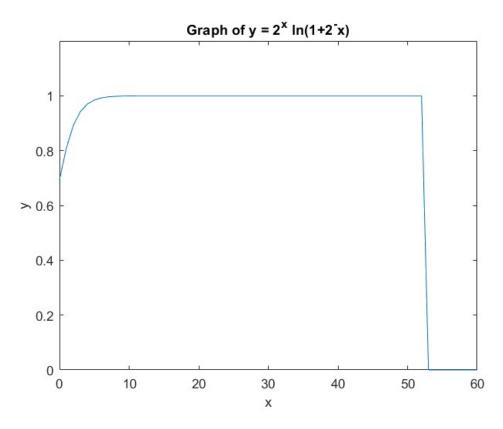
For the second iteration, we have k=2:

$$\begin{split} \alpha_2 &= \frac{(\pi^{(1)})^T \pi^{(1)}}{(d^{(1)})^T A d^{(1)}} \\ &= \frac{\begin{bmatrix} 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix}}{\begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}} \\ &= \frac{\frac{1}{4}}{\begin{bmatrix} 0 & \frac{3}{4} \end{bmatrix} \begin{bmatrix} \frac{1}{4} \\ \frac{1}{2} \end{bmatrix}} \\ &= \frac{2}{3} \\ x^{(2)} &= x^{(1)} + \alpha_2 \cdot d^{(1)} \\ &= \begin{bmatrix} \frac{1}{2} \\ 0 \end{bmatrix} + \frac{2}{3} \begin{bmatrix} \frac{1}{4} \\ \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \end{bmatrix} \\ \pi^{(2)} &= \pi^{(1)} - \alpha_2 \cdot A \cdot d^{(1)} \\ &= \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix} - \frac{2}{3} \begin{bmatrix} 0 \\ \frac{3}{4} \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix} - \begin{bmatrix} 0 \\ \frac{1}{2} \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} \end{split}$$

Since  $\pi^{(2)} = \vec{0}$ ,  $x^{(2)} = \begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \end{bmatrix}$  is the approximate (and exact) solution to Ax = b.

Use Matlab to plot a graph of the function  $y = 2^x \ln(1 + 2^{-x})$  for x going from 0 to 60. What is the value of  $\lim_{x\to\infty} y$ ? Can you explain the graph you obtained?

## **Solution:**



It is clear that the function approaches 1 as x goes to infinity, and reduces to zero after around x = 52. The limit of  $y = 2^x \ln(1 + 2^{-x})$  as  $x \to \infty$  is given by:

$$\lim_{x \to \infty} 2^x \ln(1+2^{-x}) = \lim_{x \to \infty} \frac{\ln(1+2^{-x})}{2^{-x}}$$

$$= \lim_{x \to \infty} \frac{\ln(1+2^{-x})}{2^{-x}}$$

$$\stackrel{\text{L'H}}{=} \lim_{x \to \infty} \frac{-\ln(2)2^{-x}}{-\ln(2)2^{-x}(1+2^{-x})}$$

$$= \lim_{x \to \infty} \frac{1}{1+2^{-x}}$$

$$= \frac{1}{1+0}$$

$$= 1$$

As above, the value of the limit is 1. The graph I obtained shows that y approaches 1 until about 52, and then drops to zero. There is a numerical error issue since the graph uses a limited precision to plot the graph, causing errors when  $2^{-x}$  is close enough to zero, which gives  $\ln(1+0) = \ln(1) = 0$ , multiplied by a very large number  $2^x$ . This can be avoided by using higher precision or arbitrary precision to compute the graph after