#### Exercise 5.1: The Jacobi Method

### 5.1.1 Introduction to the Problem

In this problem, we wish to solve the linear system of equations given by

$$\begin{cases}
7x_1 - 3x_2 &= 4 \\
-3x_1 + 9x_2 + x_3 &= -6 \\
x_2 + 3x_3 - x_4 &= 3 \\
-x_3 + 10x_4 - 4x_5 &= 7 \\
-4x_4 + 6x_5 &= 2
\end{cases}$$
(5.1.1)

using the Jacobi Method. To do this, we first reconstruct the given system in Equation 5.1.1 as the equation

$$Ax = b (5.1.2)$$

where

$$A = \begin{bmatrix} 7 & -3 & 0 & 0 & 0 \\ -3 & 9 & 1 & 0 & 0 \\ 0 & 1 & 3 & -1 & 0 \\ 0 & 0 & -1 & 10 & -4 \\ 0 & 0 & 0 & -4 & 6 \end{bmatrix}, x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix}, \text{ and } b = \begin{bmatrix} 4 \\ -6 \\ 3 \\ 7 \\ 2 \end{bmatrix}$$

In implementing the method, we use the  $l_{\infty}$ -norm relative jump based stopping criterion with  $\epsilon = 10^{-8}$ .

# 5.1.2 The Fixed Point Iteration for Linear Systems

Now, we solve the linear system in Equation 5.1.2 by obtaining its corresponding fixed-point iteration given by

$$x^{(k+1)} = T_I x^{(k)} + c_I, \qquad k > 0 (5.1.3)$$

where  $T_J$  and  $c_J$  are the iteration matrix and translation matrix, respectively. We note that this iteration converges to a unique solution from any initial estimate  $x^{(0)}$  provided that the spectral radius of T, denoted as  $\rho(T)$ , does not exceed 1.

We note a property that for any matrix A, we can split the matrix as A = M - N where M is a non-singular matrix and N is any matrix of the right size. From this, we can then obtain Equation 5.1.3 from Equation 5.1.2 as follows:

$$Ax = b$$

$$(M - N)x = b$$

$$Mx = Nx + b$$

$$x = M^{-1}(Nx + b)$$

$$x = M^{-1}Nx + M^{-1}b$$

The final equation suggests that we take  $T_J = M^{-1}N$  and  $c_J = M^{-1}b$  as long as  $\rho(T) < 1$ . The problem now is on finding M and N. For this, we use the Jacobi Method.

### 5.1.3 The Jacobi Method

We first deconstruct the matrix A as A = D - L - U where D is a diagonal matrix consisting of the diagonal entries of A while L and U are the lower and upper triangular matrices, respectively, with zeros as its diagonal entries and that satisfies the equation. That is,

By properties of matrix, we obtain that the inverse of D is given by

$$D^{-1} = \begin{bmatrix} \frac{1}{7} & 0 & 0 & 0 & 0\\ 0 & \frac{1}{9} & 0 & 0 & 0\\ 0 & 0 & \frac{1}{3} & 0 & 0\\ 0 & 0 & 0 & \frac{1}{10} & 0\\ 0 & 0 & 0 & 0 & \frac{1}{6} \end{bmatrix}$$

This implies that  $M = D \Rightarrow M^{-1} = D^{-1}$  and N = L + U. Consequently, we have

$$T_J = D^{-1}(L+U) (5.1.4)$$

$$c_J = D^{-1}b (5.1.5)$$

From Equation 5.1.4, we have

$$T_{j} = \begin{bmatrix} \frac{1}{7} & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{9} & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{10} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{6} \end{bmatrix} \begin{pmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 3 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 4 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 3 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 4 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{7} & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{9} & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{6} \end{bmatrix} \begin{bmatrix} 0 & 3 & 0 & 0 & 0 \\ 3 & 0 & -1 & 0 & 0 \\ 0 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 4 \\ 0 & 0 & 0 & 4 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & \frac{3}{7} & 0 & 0 & 0 \\ \frac{1}{3} & 0 & -\frac{1}{9} & 0 & 0 \\ 0 & -\frac{1}{3} & 0 & \frac{1}{3} & 0 \\ 0 & 0 & \frac{1}{10} & 0 & \frac{2}{5} \\ 0 & 0 & 0 & \frac{2}{3} & 0 \end{bmatrix}$$

From Equation 5.1.5, we have

$$c_{j} = \begin{bmatrix} \frac{1}{7} & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{9} & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{10} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{6} \end{bmatrix} \begin{bmatrix} 4 \\ -6 \\ 3 \\ 7 \\ 2 \end{bmatrix} = \begin{bmatrix} \frac{4}{7} \\ -\frac{2}{3} \\ 1 \\ \frac{7}{10} \\ \frac{1}{3} \end{bmatrix}$$

Therefore, from Equation 5.1.3, we have the fixed-point iteration

$$x^{(k+1)} = \begin{bmatrix} 0 & \frac{3}{7} & 0 & 0 & 0\\ \frac{1}{3} & 0 & -\frac{1}{9} & 0 & 0\\ 0 & -\frac{1}{3} & 0 & \frac{1}{3} & 0\\ 0 & 0 & \frac{1}{10} & 0 & \frac{2}{5}\\ 0 & 0 & 0 & \frac{2}{3} & 0 \end{bmatrix} x^{(k)} + \begin{bmatrix} \frac{4}{7}\\ -\frac{2}{3}\\ 1\\ \frac{7}{10}\\ \frac{1}{3} \end{bmatrix}, \qquad k > 0$$

## 5.1.4 Jacobi Iteration Implementation

Given our obtained matrix for  $T_J$  in the previous section, we obtain using MATLAB that  $\rho(T_J) \approx 0.5563 < 1$ . This implies that for any initial estimate  $x^{(0)}$ , the iteration converges to a unique root. For such case, we use the zero vector  $[0,0,0,0,0]^T$  as our initial guess. Implementing the iteration in MATLAB, we obtain the result in Figure 5.1.1.

n			x^(n)			rel jump
0	0.000000	0.000000	0.000000	0.000000	0.000000	NaN
1	0.571429	-0.666667	1.000000	0.700000	0.333333	Inf
2	0.285714	-0.587302	1.455556	0.933333	0.800000	4.666667E-01
3	0.319728	-0.733157	1.506878	1.165556	0.955556	1.595420E-01
4	0.257218	-0.727522	1.632904	1.232910	1.110370	1.027388E-01
5	0.259634	-0.762361	1.653477	1.307439	1.155273	4.564169E-02
6	0.244702	-0.763842	1.689933	1.327457	1.204959	3.004920E-02
7	0.244068	-0.772870	1.697100	1.350977	1.218305	1.391763E-02
8	0.240199	-0.773877	1.707949	1.357032	1.233985	9.239240E-03
9	0.239767	-0.776373	1.710303	1.364389	1.238021	4.307438E-03
10	0.238697	-0.776778	1.713587	1.366239	1.242926	2.867672E-03
11	0.238524	-0.777499	1.714339	1.368529	1.244159	1.336516E-03
12	0.238215	-0.777641	1.715343	1.369098	1.245686	8.906198E-04
13	0.238154	-0.777855	1.715579	1.369809	1.246065	4.145617E-04
14	0.238062	-0.777902	1.715888	1.369984	1.246539	2.763364E-04
15	0.238042	-0.777967	1.715962	1.370204	1.246656	1.284977E-04
16	0.238014	-0.777982	1.716057	1.370259	1.246803	8.566141E-05
17	0.238008	-0.778002	1.716080	1.370327	1.246839	3.980693E-05
18	0.237999	-0.778006	1.716110	1.370344	1.246885	2.653760E-05
19	0.237997	-0.778012	1.716117	1.370365	1.246896	1.232725E-05
20	0.237995	-0.778014	1.716126	1.370370	1.246910	8.218133E-06
21	0.237994	-0.778016	1.716128	1.370377	1.246913	3.816638E-06
22	0.237993	-0.778016	1.716131	1.370378	1.246918	2.544422E-06
23	0.237993	-0.778017	1.716131	1.370380	1.246919	1.181524E-06
24	0.237993	-0.778017	1.716132	1.370381	1.246920	7.876821E-07
25	0.237993	-0.778017	1.716133	1.370381	1.246920	3.657412E-07
26	0.237993	-0.778017	1.716133	1.370381	1.246921	2.438274E-07
27	0.237993	-0.778017	1.716133	1.370382	1.246921	1.132110E-07
28	0.237993	-0.778017	1.716133	1.370382	1.246921	7.547399E-08
29	0.237993	-0.778017	1.716133	1.370382	1.246921	3.504242E-08
30	0.237993	-0.778017	1.716133	1.370382	1.246921	2.336161E-08
31	0.237993	-0.778017	1.716133	1.370382	1.246921	1.084662E-08
32	0.237993	-0.778017	1.716133	1.370382	1.246921	7.231080E-09

Figure 5.1.1: Result of Implementation of Jacobi Iteration in MATLAB.

Results show that after 32 iterations, we obtain a good estimate that satisfies the defined stopping criterion.

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The estimate obtained after 32 iterations is x = [0.237993, -0.778017, 1.716133, 1.370382, 1.246921]
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To determine its accuracy, we solve for its relative  $l_{\infty}$ -norm error. Since the exact solution is unknown, we use the formula

$$\|e\|_{\infty} = \frac{\|Ax^* - b\|_{\infty}}{\|b\|_{\infty}}$$
 (5.1.6)

Solving this in MATLAB, we obtain that  $||e||_{\infty} \approx 8.23085 \times 10^{-9}$ .

From these results, we conclude that the convergence of the method is not that fast considering that it took 32 iterations to obtain a good estimate. This is to be expected since the Jacobi Method follows a linear rate of convergence as reflection of the fixed-point iteration. The calculated  $l_2$ -norm of the iteration matrix is  $||T_J|| = 0.667$ . This value is the upper bound for the asymptotic error constant. The relatively small value of  $||T_J||$  verifies the slow convergence of the method. Despite this, the obtained value for  $||e||_{\infty}$ , being very small, implies that our estimate is quite accurate.