Homework 7

Due date: March 13, 2025

1. Note that

$$||A||_{\infty} = \max\{|3| + |2| + |4|, |2| + |0| + |2|, |4| + |2| + |3|\} = \max\{9, 4, 9\} = 9.$$

Furthermore, we have that

$$||A||_1 = \max\{|3| + |2| + |4|, |2| + |0| + |2|, |4| + |2| + |3|\} = \max\{9, 4, 9\} = 9.$$

Finally, we compute the eigenvalues of A in the following way.

$$\det(A - \lambda I) = \det\begin{pmatrix} 3 - \lambda & 2 & 4 \\ 2 & -\lambda & 2 \\ 4 & 2 & 3 - \lambda \end{pmatrix} = (3 - \lambda)(-\lambda(3 - \lambda) - 4) - 2(2(3 - \lambda) - 8) + 4(4 + 4\lambda)$$
$$= -\lambda^3 + 6\lambda^2 + 15\lambda + 8 = -(\lambda + 1)^2(\lambda - 8).$$

Therefore, the eigenvalues are given by the roots of $-(\lambda + 1)^2(\lambda - 8)$, which are $\lambda = -1$ and $\lambda = 8$. Since A is symmetric, we have that

$$||A||_2 = \rho(A) = \max\{|8|, |-1|\} = 8.$$

2. (a) Proof. Let $M_1 = \max_{||x|| \neq 0} \frac{||Ax||}{||x||}$ and $M_2 = ||A|| = \max_{||x|| = 1} ||Ax||$. It suffices to show that $M_1 = M_2$. Note that if $||x|| \neq 0$, then by properties of norms, we have

$$\frac{||Ax||}{||x||} = \left| \left| A \frac{x}{||x||} \right| \right|.$$

Furthermore¹, since $\left|\left|\frac{x}{||x||}\right|\right| = \frac{1}{||x||}||x|| = 1$ we must have that $M_1 \leq M_2$ by definition of M_2 . Furthermore, if we fix an arbitrary vector, x, such that ||x|| = 1. Then we must have

$$||Ax|| = \left| \left| A\frac{x}{1} \right| \right| = \left| \left| A\frac{x}{||x||} \right| \right| = \frac{||Ax||}{||x||}$$

by definition of M_1 , we must have that $M_2 \leq M_1$, and therefore, $M_1 = M_2$, which was the desired result.

(b) *Proof.* Note that by part (a), we have that for any vector $x \neq 0$,

$$||A|| = \max_{y \neq 0} \frac{||Ay||}{||y||} \ge \frac{||Ax||}{||x||}.$$

Multiplying by ||x|| gives that $||Ax|| \le ||A|| ||x||$ for all $x \ne 0$. Furthermore, note that the inequality also holds when x = 0, since Ax = 0. Therefore,

$$||Ax|| \le ||A||||x||$$

for all x. We then have that

$$||AB|| = \max_{||x||=1||} ||ABx|| \leq \max_{||x||=1} ||A||||Bx|| = ||A||\max_{||x||=1} ||Bx|| = ||A||||B||$$

which is the desired result.

¹It should be noted that $\frac{x}{||x||}$ is intended to mean $\frac{1}{||x||}x$, to agree with multiplying x by a scalar.

(c) *Proof.* By repetedly applying part (b), we have that for all $k \in \mathbb{N}$

$$||A^k|| = ||A \cdot A^{k-1}|| \le ||A|| ||A^{k-1}|| = ||A|| ||A \cdot A^{k-2}|| \le ||A||^2 ||A^{k-2}|| \le \dots ||A||^k.$$

(d) *Proof.* Note that by part (b), we have

$$||A||||A^{-1}|| \ge ||AA^{-1}|| = ||I|| = \max_{||x||=1} ||Ix|| = \max_{||x||=1} ||x|| = 1$$

which is the desired result.

3.

4. For the Gauss-Seidel method, we get the following after each iteration:

```
Iteration 1: x[0] = 0.5000000, x[1] = 2.833333, x[2] = -1.083333
```

Iteration 2:
$$x[0] = 1.916667, x[1] = 2.944444, x[2] = -1.027778$$

Iteration 3:
$$x[0] = 1.972222, x[1] = 2.981481, x[2] = -1.009259$$

Iteration 4:
$$x[0] = 1.990741, x[1] = 2.993827, x[2] = -1.003086$$

Iteration 5:
$$x[0] = 1.996914, x[2] = 2.997942, x[3] = -1.001029.$$

Leaving the final solution as [1.996914, 2.997924, -1.001029]

Here is the code that was used:

```
import numpy as np
```

```
n = len(b)

x = np.zeros(n) if x0 is None else np.array(x0, dtype=float)
```

 $def gauss_seidel(A, b, x0=None, iterations=5)$:

```
\begin{array}{lll} & \text{for iteration in range} \, (1\,, \, \, \text{iterations} \, + \, 1) \colon \\ & \text{print} \, (\, f'' \, \text{Iteration} \, \, \{ \, \text{iteration} \, \} \colon '' \, ) \\ & \text{for i in range} \, (n) \colon \\ & \text{sum1} \, = \, \text{sum} \, (A[\, i \, ][\, j \, ] \, * \, x[\, j \, ] \, \, \text{for j in range} \, (\, i \, ) \, ) \\ & \text{sum2} \, = \, \text{sum} \, (A[\, i \, ][\, j \, ] \, * \, x[\, j \, ] \, \, \text{for j in range} \, (\, i \, + \, 1 \, , \, n) \, ) \\ & x[\, i \, ] \, = \, (\, b[\, i \, ] \, - \, \, \text{sum1} \, - \, \, \text{sum2} \, ) \, / \, A[\, i \, ][\, i \, ] \\ & print \, (\, f'' \, x[\, \{ \, i \, \} ] \, = \, \{ x[\, i \, ] \colon . \, 6 \, f \, \} \, " \, ) \\ & print \, (\, ) \end{array}
```

return x

A = np.array([[2, -1, 0], [-1, 3, -1], [0, -1, 2]], dtype=float) b = np.array([1, 8, -5], dtype=float)
$$x0 = [0, 0, 0]$$

```
solution = gauss_seidel(A, b, x0, iterations=5)
print("Final solution after 5 iterations:", solution)
```

For the Jacobi method, we got the following:

```
Iteration 1: x[0] = 0.500000, x[1] = 2.666667, x[2] = -2.5000000

Iteration 2: x[0] = 1.833333, x[1] = 2.000000, x[2] = -1.166667

Iteration 3: x[0] = 1.500000, x[1] = 2.888889, x[2] = -1.500000

Iteration 4: x[0] = 1.944444, x[1] = 2.666667, x[2] = -1.055556

Iteration 5: x[0] = 1.833333, x[1] = 2.962963, x[2] = -1.166667
```

Leaving final solution [1.833333, 2.962963, -1.166667]

import numpy as np

```
n = len(b)
x = np.zeros(n) if x0 is None else np.array(x0, dtype=float)
x_new = np.copy(x)

for iteration in range(1, iterations + 1):
    print(f"Iteration {iteration}:")
    for i in range(n):
        sum1 = sum(A[i][j] * x[j] for j in range(n) if j != i)
        x_new[i] = (b[i] - sum1) / A[i][i]
        print(f"x[{i}] = {x_new[i]:.6f}")
    x[:] = x_new
    print()
```

 $def jacobi_method(A, b, x0=None, iterations=5)$:

return x

$$A = \text{np.array} ([[2, -1, 0], [-1, 3, -1], [0, -1, 2]], \text{ dtype=float})$$

$$b = \text{np.array} ([1, 8, -5], \text{ dtype=float})$$

$$x0 = [0, 0, 0]$$

$$\text{solution} = \text{jacobi_method}(A, b, x0, \text{iterations} = 5)$$

$$\text{print} ("Final solution after 5 iterations} :", \text{solution})$$

5.

6. Proof. First, assume that A is strictly diagonally dominant. Namely, suppose that

$$|a_{ii}| > \sum_{j \neq i} |a_{ij}|$$

for all $1 \leq i \leq n$. Then, we must have that

$$||D^{-1}(L+U)||_{\infty} \le ||D^{-1}||_{\infty} ||L+U||_{\infty} = ||D^{-1}||_{\infty} ||D-A||_{\infty}$$

where we used the fact that A = D - (L + U). Furthermore, since $D = \text{diag}(a_{ii})^2$, we must have that

$$D^{-1} = \operatorname{diag}\left(\frac{1}{a_{ii}}\right).$$

Therefore, since we have that

$$L + U = \begin{cases} a_{ij} & \text{if } i \neq j \\ 0 & \text{else} \end{cases},$$

we must have

$$[D^{-1}(L+U)]_{ij} = \begin{cases} \frac{a_{ij}}{a_{ii}} & \text{if } i \neq j \\ 0 & \text{else} \end{cases}.$$

Therefore, we have that

$$||D^{-1}(L+U)||_{\infty} = \max_{1 \le i \le n} \sum_{j=1}^{n} \left| [D^{-1}(L+U)]_{ij} \right| = \max_{1 \le i \le n} \sum_{j \ne i} \left| \frac{a_{ij}}{a_{ii}} \right| = \max_{1 \le i \le n} \frac{1}{|a_{ii}|} \sum_{j \ne i} |a_{ij}|.$$

Since A is strictly diagonally dominant, we have that

$$||D^{-1}(L+U)||_{\infty} = \max_{1 \le i \le n} \frac{1}{|a_{ii}|} \sum_{i \ne i} |a_{ij}| < \max_{1 \le i \le n} \frac{|a_{ii}|}{|a_{ii}|} = 1.$$

Therefore, $||D^{-1}(L+U)||_{\infty} < 1$.

Suppose that $||D^{-1}(L+U)||_{\infty} < 1$. As discussed before, we have that

$$||D^{-1}(L+U)||_{\infty} = \max_{1 \le i \le n} \frac{1}{|a_{ii}|} \sum_{j \ne i} |a_{ij}|.$$

Therefore, for all $1 \leq i \leq n$, we have that

$$\frac{1}{|a_{ii}|} \sum_{j \neq i} |a_{ij}| \le \max_{1 \le i \le n} \frac{1}{|a_{ii}|} \sum_{j \neq i} |a_{ij}| < 1.$$

Thus, multiplying by a_{ii} gives that for all $1 \leq i \leq n$,

$$\sum_{j\neq i} |a_{ij}| < |a_{ii}|.$$

Therefore, A is strictly diagonally dominant, which completes the proof.

 $^{^2}$ I am going to make this assumption, since the theorem is not true without it. For example, if A=I, then clearly A is strictly diagonally dominant. However, if we let $D=\frac{1}{2}I$, then we must have $L+U=D-A=-\frac{1}{2}I$. This may be achieved with $L=U=-\frac{1}{4}I$. However, this means $||D^{-1}(L+U)||_{\infty}=||2I\left(-\frac{1}{2}I\right)||_{\infty}=||-I||_{\infty}=1$ which is obviously not less than 1. I specify this because we only assumed that D was diagonal, L is lower triangular, and U is upper triangular. However, what I have written is a stronger condition.