

## 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.

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

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

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

Furthermore<sup>1</sup>, since  $\left\| \frac{x}{\|x\|} \right\| = \frac{1}{\|x\|} \|x\| = 1$  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

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

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  $x \neq 0$ ,

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

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

$$\|Ax\| \leq \|A\| \|x\|$$

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. ■

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<sup>1</sup>It 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 repeatedly applying part (b), we have that for all  $k \in \mathbb{N}$

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

■

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

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

which is the desired result. ■

3. *Proof. Step 1.*  $\|A\|_1 \leq \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.$

Fix  $\mathbf{x}$  with  $\|\mathbf{x}\|_1 = 1$ . ( $\Rightarrow \sum_{j=1}^n |x_j| = 1$ )

$$\begin{aligned} \|A\mathbf{x}\|_1 &= \sum_{i=1}^n \left| \sum_{j=1}^n a_{ij} x_j \right| \\ &\leq \sum_{i=1}^n \sum_{j=1}^n |a_{ij} x_j| \\ &\leq \sum_{j=1}^n |x_j| \sum_{i=1}^n |a_{ij}| \\ &\leq \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|. \end{aligned}$$

Taking the maximum over all  $\mathbf{x}$  with  $\|\mathbf{x}\|_1 = 1$ , we get:

$$\|A\|_1 = \max_{\|\mathbf{x}\|_1=1} \|A\mathbf{x}\|_1 \leq \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.$$

**Step 2.**  $\|A\|_1 \geq \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.$

There exists some  $p$  such that:

$$\sum_{i=1}^n |a_{ip}| = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.$$

Define  $\mathbf{x} = [x_j]$  as:

$$x_j = \begin{cases} 1 & j = p, \\ 0 & \text{else} \end{cases}$$

( $\|\mathbf{x}\|_1 = 1$ . and  $a_{ij}x_j = |a_{ip}|$ .)

$$\begin{aligned}\|A\mathbf{x}\|_1 &= \sum_{i=1}^n \left| \sum_{j=1}^n a_{ij}x_j \right| \\ &= \sum_{i=1}^n |a_{ip}| \\ &= \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.\end{aligned}$$

Thus,

$$\|A\|_1 = \max_{\|\mathbf{x}\|_1=1} \|A\mathbf{x}\|_1 \geq \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.$$

Since both inequalities hold, we conclude:

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}|.$$

■

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

def gauss_seidel(A, b, x0=None, iterations=5):
    n = len(b)
    x = np.zeros(n) if x0 is None else np.array(x0, dtype=float)

    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(i))
            sum2 = sum(A[i][j] * x[j] for j in range(i + 1, n))
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        x[i] = (b[i] - sum1 - sum2) / A[i][i]
        print(f"x[{i}] = {x[i]:.6f}")
    print()

    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:  $x[0] = 0.500000, x[1] = 2.666667, x[2] = -2.500000$

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
```

```

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

    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 = jacobi_method(A, b, x0, iterations=5)

```

```
print("Final solution after 5 iterations:", 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

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

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

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

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

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

Since  $A$  is strictly diagonally dominant, we have that

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

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

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

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<sup>2</sup>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  $\|D^{-1}(L+U)\|_{\infty} = \|2I(-\frac{1}{2}I)\|_{\infty} = \|-I\|_{\infty} = 1$  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.

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

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

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. ■