#### **Announcements**

- Oct. 27: midterm exam
  - starts at 14:20
  - [Optional] bring a A4 cheat sheet (one side only)
- Midterm 2022 is available on moodle.

## Solving linear systems III

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#### Review

- Direct methods for solving linear systems
  - Gaussian elimination with partial pivoting
  - PA = LU factorization

### Today

- *Iterative* methods for solving linear systems
  - Jacobi method
  - Gauss-Seidel method

# Solving Ax = b

- A is given, real, n x n, and b is given, real vector.
- Two types of approaches
  - Direct methods: yield exact solution in absence of roundoff error
    - Example: Gaussian elimination and its variants
  - Iterative methods: iterate in a similar fashion to what we do for solving nonlinear equations

# Iterative methods for solving Ax = b

- Starting from initial guess  $\underline{x}_0$ , generate iterates  $\underline{x}_1$ ,  $\underline{x}_2$ , ...,  $\underline{x}_k$ , hopefully converging to solution  $\underline{x}$ .
- But why not simply using a direct method?



### Why using iterative methods?

- Can be faster if the input matrix is large
  - One step of an iterative method requires only a fraction of the floating operations of a full LU factorization.
- A good approximation to the solution is already known.
- The input matrix is *sparse*.

#### Direct vs. iterative linear solvers

- Direct solvers
  - Computation is numerically stable in many relevant cases. ☺
  - Can solve economically for several right-hand sides. ☺
  - But: fill-in limits usefulness (memory). ☺
- Iterative solvers
  - Only a rough approximation to x is required. ☺
  - A good  $\underline{x}_0$  approximating  $\underline{x}$  is generally known (warm start).  $\odot$
  - Quality may depend on "right" choice of parameters.

### Typical scenarios

- Direct solvers
  - Many linear systems with the same matrix A
  - Applications that require very accurate solutions
- Iterative solvers
  - Many linear systems with "slightly changing" matrices
  - Matrix-free applications
  - Very large problems

#### Iterative methods

- Jacobi method
- Gauss-Seidel method

# Jacobi m $\epsilon \begin{bmatrix} u_0 \\ v_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

- A form of fixed-point iteration  $\begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \begin{bmatrix} \frac{5-v_0}{3} \\ \frac{5-u_0}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{3} \\ \frac{5}{2} \end{bmatrix}$
- Example: Solve the linear system

$$3u + v = 5$$
$$u + 2v = 5$$

$$\begin{bmatrix} u_2 \\ v_2 \end{bmatrix} = \begin{bmatrix} \frac{5-v_1}{3} \\ \frac{5-u_1}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-5/2}{3} \\ \frac{5-5/3}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{6} \\ \frac{5}{3} \end{bmatrix}$$

- The answer is u = 1, v = 2.
- First, we have

$$u = \frac{5 - v}{3}$$
$$5 - u$$

$$v = \frac{5 - u}{2}.$$

$$\begin{bmatrix} u_3 \\ v_3 \end{bmatrix} = \begin{bmatrix} \frac{5-5/3}{3} \\ \frac{5-5/6}{2} \end{bmatrix} = \begin{bmatrix} \frac{10}{9} \\ \frac{25}{12} \end{bmatrix}$$

•

• Start from an initial guess (0, 0)

#### Jacobi method

 Use the example again, but now the equations are given in the reverse order.

$$u + 2v = 5$$

$$3u + v = 5$$

$$\begin{bmatrix} u_0 \\ v_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

- The answer is u = 1,  $v = \begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \begin{bmatrix} 5 2v_0 \\ 5 3u_0 \end{bmatrix} = \begin{bmatrix} 5 \\ 5 \end{bmatrix}$
- We have

$$u = 5 - 2v$$
$$v = 5 - 3u$$

$$\left[\begin{array}{c} u_2 \\ v_2 \end{array}\right] = \left[\begin{array}{c} 5 - 2v_1 \\ 5 - 3u_1 \end{array}\right] = \left[\begin{array}{c} -5 \\ -10 \end{array}\right]$$

$$\begin{bmatrix} u_3 \\ v_3 \end{bmatrix} = \begin{bmatrix} 5 - 2(-10) \\ 5 - 3(-5) \end{bmatrix} = \begin{bmatrix} 25 \\ 20 \end{bmatrix}$$

Start from an initial guess (0, 0)

#### The difference?

VS.

• 
$$3u + v = 5$$

• 
$$u + 2v = 5$$

$$A = \begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix}$$

• 
$$u + 2v = 5$$

• 
$$3u + v = 5$$

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix}$$

The Jacobi method converges. ©

The Jacobi method diverges. 🕾

### Strictly diagonally dominant

• **Definition**. The  $n \times n$  matrix  $A = (a_{ij})$  is **strictly diagonally dominant** if, for each  $1 \le i \le n$ ,  $|a_{ii}| > \sum_{i \neq i} |a_{ii}|.$ 

Which matrix is strictly diagonally dominant?

$$A = \begin{bmatrix} 3 & 1 & -1 \\ 2 & -5 & 2 \\ 1 & 6 & 8 \end{bmatrix} \qquad B = \begin{bmatrix} 3 & 2 & 6 \\ 1 & 8 & 1 \\ 9 & 2 & -2 \end{bmatrix}$$

$$B = \begin{bmatrix} 3 & 2 & 6 \\ 1 & 8 & 1 \\ 9 & 2 & -2 \end{bmatrix}$$



**Theorem**. If the  $n \times n$  matrix A is strictly diagonally dominant, then for every vector  $\underline{b}$  and every starting guess, the Jacobi method applied to  $A\underline{x} = \underline{b}$  converges to the solution.

### The previous example

• 
$$3u + v = 5$$

• 
$$u + 2v = 5$$

• 
$$u + 2v = 5$$

• 
$$3u + v = 5$$

$$A = \begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix}$$

strictly diagonally dominant

The Jacobi method converges. ©

**NOT** strictly diagonally dominant

The Jacobi method diverges. 😂

## A form of fixed-point iteration $\underline{x} = g(\underline{x})$

• Start with 
$$A\underline{x} = \underline{b}$$
  
 $(D + L + U)\underline{x} = \underline{b}$   
 $D\underline{x} = \underline{b} - (L + U)\underline{x}$   
 $\underline{x} = D^{-1}(\underline{b} - (L + U)\underline{x})$   
 $D = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -5 & 0 \\ 0 & 0 & 8 \end{bmatrix} L = \begin{bmatrix} 0 & 0 & 0 \\ 2 & 0 & 0 \\ 1 & 6 & 0 \end{bmatrix} U = \begin{bmatrix} 0 & 1 & -1 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}$ 

Jacobi Method

$$x_0 = \text{initial vector}$$
  
 $x_{k+1} = D^{-1}(b - (L + U)x_k) \text{ for } k = 0,1,2,...$ 

### Example: Jacobi method

Jacobi method

$$x_0$$
 = initial vector  $x_{k+1} = D^{-1}(b - (L + U)x_k)$  for  $k = 0,1,2,...$ 

The previous example

$$-3u + v = 5$$

$$-u + 2v = 5$$

$$D = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}, L = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, U = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, b = \begin{bmatrix} 5 \\ 5 \end{bmatrix}$$

#### Iterative methods

- Jacobi method
- Gauss-Seidel method

#### Gauss-Seidel method

- Uses the most recently updated values of the knowns, even if the updating occurs in the current step!
- The previous exa $\begin{bmatrix} u_0 \\ v_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

$$3u + v = 5$$

$$u + 2v = 5$$

$$u = \frac{5 - v}{3}$$

 $v = \frac{5-u}{2}$ 

$$\begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \begin{bmatrix} \frac{5-v_0}{3} \\ \frac{5-u_1}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-0}{3} \\ \frac{5-5/3}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{3} \\ \frac{5}{3} \end{bmatrix}$$

$$\begin{bmatrix} u_2 \\ v_2 \end{bmatrix} = \begin{bmatrix} \frac{5-v_1}{3} \\ \frac{5-u_2}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-5/3}{3} \\ \frac{5-10/9}{2} \end{bmatrix} = \begin{bmatrix} \frac{10}{9} \\ \frac{35}{18} \end{bmatrix}$$

$$\begin{bmatrix} u_3 \\ v_3 \end{bmatrix} = \begin{bmatrix} \frac{5-v_2}{3} \\ \frac{5-u_3}{3} \end{bmatrix} = \begin{bmatrix} \frac{5-35/18}{3} \\ \frac{5-55/54}{3} \end{bmatrix} = \begin{bmatrix} \frac{55}{54} \\ \frac{215}{109} \end{bmatrix}$$

### A form of fixed-point iteration x = g(x)

• Start with  $A\underline{x} = \underline{b}$  $(L+D+U)\underline{x} = \underline{b}$   $(L+D)\underline{x}_{k+1} = \underline{b} - U\underline{x}_{k}$   $D\underline{x}_{k+1} = \underline{b} - U\underline{x}_{k} - L\underline{x}_{k+1}$   $\underline{x}_{k+1} = D^{-1}(\underline{b} - U\underline{x}_{k} - L\underline{x}_{k+1})$   $\underline{g}(x)$ 

$$A = \begin{bmatrix} 3 & 0 & 0 \\ 2 & 5 & 2 \\ 1 & 6 & 8 \end{bmatrix}$$

$$D = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -5 & 0 \\ 0 & 0 & 8 \end{bmatrix} \quad L = \begin{bmatrix} 0 & 0 & 0 \\ 2 & 0 & 0 \\ 1 & 6 & 0 \end{bmatrix}$$

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

Gauss-Seidel Method

$$x_0 = \text{initial vector}$$
  
 $x_{k+1} = D^{-1}(b - Ux_k - Lx_{k+1}) \text{ for } k = 0,1,2,...$ 

#### Gauss-Seidel Method

$$x_0 = \text{initial vector}$$
  $x_{k+1} = D^{-1}(b - Ux_k - Lx_{k+1}) \text{ for } k = 0,1,2,...$ 

• Example  $\begin{bmatrix} 3 & 1 & -1 \\ 2 & 4 & 1 \\ 1 & 2 & 5 \end{bmatrix} \begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \\ 1 \end{bmatrix}$ 

$$\begin{bmatrix} u_{k+1} \\ v_{k+1} \\ w_{k+1} \end{bmatrix} = \begin{bmatrix} 1/3 & 0 & 0 \\ 0 & 1/4 & 0 \\ 0 & 0 & 1/5 \end{bmatrix} \begin{pmatrix} \begin{bmatrix} 4 \\ 1 \\ 1 \end{bmatrix} - \begin{bmatrix} 0 & 1 & -1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} u_k \\ v_k \\ w_k \end{bmatrix} - \begin{bmatrix} 0 & 0 & 0 \\ 2 & 0 & 0 \\ -1 & 2 & 0 \end{bmatrix} \begin{bmatrix} u_{k+1} \\ v_{k+1} \\ w_{k+1} \end{bmatrix} )$$

$$= \begin{bmatrix} 1/3 & 0 & 0 \\ 0 & 1/4 & 0 \\ 0 & 0 & 1/5 \end{bmatrix} \left( \begin{bmatrix} 4 - v_k + w_k \\ 1 - w_k - 2u_{k+1} \\ 1 + u_{k+1} - 2v_{k+1} \end{bmatrix} \right)$$

$$3u + v - w = 4 \implies u = \frac{1}{3}(4 - v + w) - u + 2v + 5w = 1 \implies w = \frac{1}{5}(1 + u - 2v)$$

$$2u + 4v + w = 1 \Rightarrow v = \frac{1}{4}(1 - w - 2u)$$

### Jacobi vs. Gauss-Seidel

$$\begin{bmatrix} u_0 \\ v_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \begin{bmatrix} \frac{5-v_0}{3} \\ \frac{5-u_0}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-0}{3} \\ \frac{5-0}{2} \end{bmatrix}$$

$$\begin{bmatrix} u_2 \\ v_2 \end{bmatrix} = \begin{bmatrix} \frac{5-v_1}{3} \\ \frac{5-u_1}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-5/2}{3} \\ \frac{5-5/3}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{6} \\ \frac{5}{3} \end{bmatrix}$$

$$\left[\begin{array}{c} u_3 \\ v_3 \end{array}\right] = \left[\begin{array}{c} \frac{5-5/3}{3} \\ \frac{5-5/6}{2} \end{array}\right] = \left[\begin{array}{c} \frac{10}{9} \\ \frac{25}{12} \end{array}\right]$$

$$\underline{x}_{k+1} = D^{-1}(\underline{b} - (L+U)\underline{x}_k)$$

$$\left[\begin{array}{c} u_0 \\ v_0 \end{array}\right] = \left[\begin{array}{c} 0 \\ 0 \end{array}\right]$$

$$\begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \begin{bmatrix} \frac{5-v_0}{3} \\ \frac{5-u_0}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-0}{3} \\ \frac{5-0}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{3} \\ \frac{5}{2} \end{bmatrix} \qquad \begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \begin{bmatrix} \frac{5-v_0}{3} \\ \frac{5-u_1}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-0}{3} \\ \frac{5-5/3}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{3} \\ \frac{5}{3} \end{bmatrix}$$

$$\begin{bmatrix} u_2 \\ v_2 \end{bmatrix} = \begin{bmatrix} \frac{5-v_1}{3} \\ \frac{5-u_1}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-5/2}{3} \\ \frac{5-5/3}{2} \end{bmatrix} = \begin{bmatrix} \frac{5}{6} \\ \frac{5}{3} \end{bmatrix} \qquad \begin{bmatrix} u_2 \\ v_2 \end{bmatrix} = \begin{bmatrix} \frac{5-v_1}{3} \\ \frac{5-u_2}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-5/3}{3} \\ \frac{5-10/9}{2} \end{bmatrix} = \begin{bmatrix} \frac{10}{9} \\ \frac{35}{18} \end{bmatrix}$$

$$\begin{bmatrix} u_3 \\ v_3 \end{bmatrix} = \begin{bmatrix} \frac{5-v_2}{3} \\ \frac{5-u_3}{2} \end{bmatrix} = \begin{bmatrix} \frac{5-35/18}{3} \\ \frac{5-55/54}{2} \end{bmatrix} = \begin{bmatrix} \frac{55}{54} \\ \frac{215}{108} \end{bmatrix}$$

$$\underline{x}_{k+1} = D^{-1}(\underline{b} - U\underline{x}_k - L\underline{x}_{k+1})$$

#### Next week

- Solving nonlinear systems
- Will use your code that solves a linear system in the next assignment!

### 程式練習

And, please upload your program on moodle2.

 Please use Jacobi, Gauss-Seidel or SOR to solve the system of six equations in six unknowns:

$$\begin{bmatrix} 3 & -1 & 0 & 0 & 0 & \frac{1}{2} \\ -1 & 3 & -1 & 0 & \frac{1}{2} & 0 \\ 0 & -1 & 3 & -1 & 0 & 0 \\ 0 & 0 & -1 & 3 & -1 & 0 \\ 0 & \frac{1}{2} & 0 & -1 & 3 & -1 \\ \frac{1}{2} & 0 & 0 & 0 & -1 & 3 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \\ u_5 \\ u_6 \end{bmatrix} = \begin{bmatrix} \frac{5}{2} \\ \frac{3}{2} \\ 1 \\ 1 \\ \frac{3}{2} \\ \frac{5}{2} \end{bmatrix}$$