

# System of Equations:

## Direct method

## Gauss Elimination Method:

Consider  $n \times (n+1)$  matrix can be used to represent the linear system

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$\vdots$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n$$

$$A = [a_{ij}] = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

$$b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} \quad [A, b] = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} & b_n \end{bmatrix}$$

$\therefore$  given linear system can be written in the form

$$Ax = b$$

where  $x = [x_1, x_2 \dots x_n]^T$  and

array  $[A \mid b]$  is called an augmented matrix.

Solve:

$$= x_1 + x_2 + 3x_4 = 4$$

$$2x_1 + x_2 - x_3 + x_4 = 1$$

$$3x_1 - x_2 - x_3 + 2x_4 = -3$$

$$-x_1 + 2x_2 + 3x_3 - x_4 = 4$$

Augmented matrix  $[A \mid B]$  is

$$= \left[ \begin{array}{cccc|c} 1 & 1 & 0 & 3 & 4 \\ 2 & 1 & -1 & 1 & 1 \\ 3 & -1 & -1 & 2 & -3 \\ -1 & 2 & 3 & -1 & 4 \end{array} \right]$$

$$\sim \begin{bmatrix} 1 & 1 & 0 & 3 & : & 4 \\ 0 & -1 & -1 & -5 & : & -7 \\ 0 & -4 & -1 & -7 & : & -15 \\ 0 & 3 & 3 & 2 & : & 8 \end{bmatrix} \begin{array}{l} R_2 \rightarrow R_2 - 2R_1 \\ R_3 \rightarrow R_3 - 3R_1 \\ R_4 \rightarrow R_4 + R_1 \end{array}$$

$$\sim \begin{bmatrix} 1 & 1 & 0 & 3 & : & 4 \\ 0 & -1 & -1 & -5 & : & -7 \\ 0 & 0 & 3 & 13 & : & 13 \\ 0 & 0 & 0 & -13 & : & -13 \end{bmatrix} \begin{array}{l} R_3 \rightarrow R_3 - 4R_2 \\ R_4 \rightarrow R_4 + 3R_2 \end{array}$$

From E4

$$\Rightarrow -13x_4 = -13$$

$$\boxed{x_4 = 1}$$

From E3

$$\Rightarrow 3x_3 + 13x_4 = 13$$

$$\Rightarrow \boxed{x_3 = 0} \quad (\because x_4 = 1)$$

from E2

$$-x_2 - x_3 - 5x_4 = -7$$

$$\Rightarrow \boxed{x_2 = 2}$$

from  $E_1$

$$x_1 + x_2 + 3x_4 = 4$$

$$\boxed{x_1 = -1}$$

## Generalization of Gauss Elimination method:

Rewrite the augmented matrix as

$$\tilde{A} = [A, b] = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & : & a_{1,n+1} \\ a_{21} & a_{22} & \dots & a_{2n} & : & a_{2,n+1} \\ \vdots & \vdots & & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} & : & a_{n,n+1} \end{bmatrix}$$

where we assumed that  $b_i = a_{i,n+1}$   
 $i = 1$  to  $n$ .

Suppose  $a_{11} \neq 0$ , we perform

$$(1) \quad \left( E_j - \left( \frac{a_{j1}}{a_{11}} \right) E_1 \right) \rightarrow E_j$$

for each  $j = 2, 3, \dots, n$ . To eliminate  
the coeff. of  $x_1$  in each of these rows.

(2) Though entries in rows  $2, 3, \dots$  are expected to change however we use the same notation  $a_{ij}$  for convenience

(3) Then perform the operation

$$\left( E_j - \frac{a_{ji}^0}{a_{ii}^0} E_i \right) \rightarrow E_j$$

$$j = i+1, i+2, \dots, n$$

provided  $a_{ii} \neq 0$

Therefore the resulting matrix is of

the form

$$\tilde{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n}; a_{1,n+1} \\ 0 & a_{22} & \dots & a_{2n}; a_{2,n+1} \\ \vdots & & & \vdots \\ 0 & 0 & \dots & 0 & a_{nn}; a_{n,n+1} \end{bmatrix}$$

Thus the new linear system is given in the following form

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = a_{1,n+1}$$

$$a_{22}x_2 + \dots + a_{2n}x_n = a_{2,n+1}$$

...

$$a_{nn}x_n = a_{n,n+1}$$

can be solved by backward substitution.

Gauss Jordan Method:

To find inverse of a matrix.

$$1) \quad A = \begin{bmatrix} 1 & -1 & 2 \\ 3 & 0 & 1 \\ 1 & 0 & 2 \end{bmatrix}$$

Augment matrix A with identity matrix

$$[A, I] = \left[ \begin{array}{ccc|ccc} 1 & -1 & 2 & 1 & 0 & 0 \\ 3 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 2 & 0 & 0 & 1 \end{array} \right]$$

$$R_2 \rightarrow R_2 - 3R_1$$

$$R_3 \rightarrow R_3 - 12R_1$$

$$\approx \left[ \begin{array}{ccc|ccc} 1 & -1 & 2 & : & 1 & 0 & 0 \\ 0 & 3 & -5 & : & -3 & 1 & 0 \\ 0 & \textcircled{1} & 0 & : & -1 & 0 & 1 \end{array} \right]$$

$$R_3 \rightarrow 3R_3 - R_2$$

$$\approx \left[ \begin{array}{ccc|ccc} 1 & -1 & \textcircled{2} & : & 1 & 0 & 0 \\ 0 & 3 & \textcircled{-5} & : & -3 & 1 & 0 \\ 0 & 0 & 5 & : & 0 & -1 & 3 \end{array} \right]$$

$$R_1 \rightarrow 5R_1 - 2R_3$$

$$R_2 \rightarrow R_2 + R_3$$

$$\approx \left[ \begin{array}{ccc|ccc} 5 & \textcircled{-5} & 0 & : & 5 & 2 & -6 \\ 0 & 3 & 0 & : & -3 & 0 & 3 \\ 0 & 0 & 5 & : & 0 & -1 & 3 \end{array} \right]$$

$$R_1 \rightarrow 3R_1 + 5R_2$$

$$\approx \left[ \begin{array}{ccc|ccc} 15 & 0 & 0 & : & 0 & 6 & -3 \\ 0 & 3 & 0 & : & -3 & 0 & 3 \\ 0 & 0 & 5 & : & 0 & -1 & 3 \end{array} \right]$$

$$\begin{array}{l}
 R_1 \rightarrow R_1/5 \\
 R_2 \rightarrow R_2 \cdot \frac{12}{3} \\
 R_3 \rightarrow R_3 \cdot \frac{12}{5}
 \end{array}
 \rightarrow
 \begin{bmatrix}
 1 & 0 & 0 & : & 0 & 2/5 & -1/5 \\
 0 & 1 & 0 & : & -1 & 0 & 1 \\
 0 & 0 & 1 & : & 0 & -1/5 & 3/5
 \end{bmatrix}$$

$$\therefore A^{-1} = \begin{bmatrix} 0 & 2/5 & -1/5 \\ -1 & 0 & 1 \\ 0 & -1/5 & 3/5 \end{bmatrix}$$

Iterative Methods:

Gauss Jacobi Method:

Diagonally Dominant:

We say that an  $n \times n$  matrix  $A$  is diagonally dominant if and only if

for each  $i = 1, 2, \dots, n$

$$|a_{ii}| > \sum_{\substack{j=1 \\ j \neq i}}^n |a_{ij}| \quad i = 1, 2, \dots, n$$



Eg (6)  $x_1 - 2x_2 + x_3 = 11$

$$x_1 + 2x_2 - 5x_3 = -1$$

$$-2x_1 + 7x_2 + 2x_3 = 5$$

Check the above system is diagonally dominant.

Sol:

= No. It can be written as dia. dominant

$$6x_1 - 2x_2 + x_3 = 11 \quad 6 > 2+1$$

$$-2x_1 + 7x_2 + 2x_3 = 5 \quad 7 > 2+2$$

$$x_1 + 2x_2 - 5x_3 = -1 \quad 5 > 1+2$$

Jacobi Method: (Gauss Jacobi Method)

The Jacobi iteration method is used to solve the equation  $Ax = b$ . To obtain  $x_i$  we use following algorithm provided  $a_{ii} \neq 0$ .

$$x_i = \sum_{\substack{j=1 \\ j \neq i}}^n - \frac{a_{ij}}{a_{ii}} x_j + \frac{b_i}{a_{ii}}$$

$$i=1, 2, \dots, n$$

$$a_{11} x_1 + a_{12} x_2 + a_{13} x_3 = b_1$$

$$a_{21} x_1 + a_{22} x_2 + a_{23} x_3 = b_2$$

$$a_{31} x_1 + a_{32} x_2 + a_{33} x_3 = b_3$$

(1) Convert the system as diagonally dominant

$$(2) \quad x_1 = \frac{b_1 - a_{12} x_2 - a_{13} x_3}{a_{11}}$$

For each  $k \geq 1$ , generate the components  $x_i^{(k)}$  of  $x^{(k)}$  from

$$x_i^{(k-1)}$$

by

$$x_i^{(k)} = \frac{1}{a_{ii}} \left[ \sum_{\substack{j=1 \\ i \neq j}}^n -a_{ij} x_j^{(k-1)} + b_i \right]$$

Remark:

Convert  $Ax = b$  into an equivalent

System

$$x = Tx + C \quad \text{for some}$$

matrix  $T$  and vector  $C$

Guess  $x^{(0)}$  initial vector. Then

sequence of approximate soln vector  
is given by

$$x^{(k)} = Tx^{(k-1)} + C$$

for each  $k=1, 2, \dots, n$ . This is

similar like of the fixed point  
method

Solve:  
=

$$10x_1 - x_2 + 2x_3 = 6$$

$$-x_1 + 11x_2 - x_3 + 3x_4 = 25$$

$$2x_1 - x_2 + 10x_3 - x_4 = -11$$

$$3x_2 - x_3 + 8x_4 = 15$$

Use Jacobi method to find approximate solutions with initial guess  $x^{(0)} = (0, 0, 0, 0)$

until  $\frac{\|x^k - x^{k-1}\|}{\|x^k\|} < 10^{-3}$ .

At least 3 digits accuracy

Sol:

(1) Problem is diagonally dominant

$$(2) \quad x_1 = \frac{x_2 - 2x_3 + 6}{10}$$

$$x_2 = \frac{x_1 + x_3 - 3x_4 + 25}{11}$$

$$x_3 = \frac{-2x_1 + x_2 + x_4 - 11}{10}$$

$$x_4 = \frac{-3x_2 + x_3 + 15}{8}$$

$$(3) \quad x^{(0)} = \begin{matrix} x_1 & x_2 & x_3 & x_4 \\ (0, & 0, & 0, & 0) \end{matrix}$$

$$x_1^{(1)} = \frac{6}{10} = 0.6$$

$$x_2^{(1)} = \frac{25}{11} = 2.2727$$

$$x_3^{(1)} = \frac{-11}{10} = -1.1$$

$$x_4^{(1)} = \frac{15}{8} = 1.875$$

$$x_1^{(2)} = \frac{2.2727 - 2(-1.1) + 6}{10}$$

$$= 1.04727$$

$$x_2^{(2)} = \frac{0.6 + (-1.1) - 3(1.875) + 25}{11}$$

$$= 1.715909$$

$$x_3^{(2)} = -0.80523$$

$$x_4^{(2)} = 0.8852$$

$i$	$x_1$	$x_2$	$x_3$	$x_4$
0	0	0	0	0
1	0.6	2.2727	-1.1	1.875
2	1.04727	1.715909	-0.80523	0.8852
3	0.9326	2.053	-1.0493	1.1309
4	1.0152	1.9537	-0.9681	0.9739
5	0.9890	2.0114	-1.0103	1.0214
6	1.0032	1.9922	-0.9945	0.9944
7	0.9982	2.0023	-1.0020	1.0036
8	1.0006	1.9987	-0.9990	0.9985
9	0.9987	2.0004	-1.0004	1.0006
10	1.0001	1.9998	-0.9998	0.9998

$$\text{tolerance} = \frac{\|x^{(10)} - x^{(9)}\|_{\infty}}{\|x^{(10)}\|_{\infty}}$$

$$= \frac{8 \times 10^{-4}}{1.9998} < 10^{-3}$$

$$\|x^{(10)}\|_{\infty} = \max \{ \|x_1\|, \|x_2\|, \|x_3\|, \|x_4\| \}$$

Remark:

Jacobi method can be written as

$$x^{(k)} = \tau x^{(k-1)} + c$$

by splitting  $A$  into its diagonal and off diagonal form,

$$A = \begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & a_{nn} \end{bmatrix} - \begin{bmatrix} 0 & \dots & 0 \\ a_{21} & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \dots & \dots & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & a_{12} & \dots & a_{1n} \\ 0 & 0 & \dots & a_{2n} \\ \vdots & & \ddots & 0 & a_{n-1,n} \\ 0 & \dots & 0 & \dots & 0 \end{bmatrix}$$

$$= D - L - U$$

$$\therefore Ax = b$$

$$\Rightarrow (D - L - U)x = b$$

$$\Rightarrow Dx = (L + U)x + b$$

Suppose  $D^{-1}$  exists  $a_{ii} \neq 0$  then

$$x = D^{-1}(L + U)x + D^{-1}b$$

For  $k = 1, 2, \dots, n$ ,

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

$$\Rightarrow x^{(k)} = T x^{(k-1)} + c \quad \text{when}$$

$$T = D^{-1}(L + U) \quad \text{and} \quad c = D^{-1}b$$



## Gauss Seidel Method:

The components of  $x^{(k-1)}$  are used to compute all the components of  $x^{(k)}$  of  $x^{(k)}$ . But if the components  $x_1^{(k)}, \dots, x_{i-1}^{(k)}$  of  $x^{(k)}$  have already been computed and are expected to be better approximation to the actual

Solution.

$$x_i^{(k)} = \frac{1}{a_{ii}} \left[ - \sum_{j=1}^{i-1} a_{ij} x_j^{(k)} - \sum_{j=i+1}^n a_{ij} x_j^{(k-1)} + b_i \right]$$

$$i = 1, 2, \dots, n$$

This modification in Gauss-Jacobi method is called as Gauss-Seidel method.

Eg: Same problem as in previous method

(1) Diagonally dominant

$$(2) \quad x_1 = 6 + x_2 - 2x_3 \mid 10$$

$$x_2 = 25 + x_1 + x_3 - 3x_4 \mid 11$$

$$x_3 = -11 - 2x_1 + x_2 + x_4 \mid 10$$

$$x_4 = 15 - 3x_2 + x_3 \mid 8$$

$$(3) \quad x^{(0)} = (0, 0, 0, 0)$$

$$(4) \quad x_1^{(1)} = 6 + 0 - 0 \mid 10 = 0.6$$

$$x_2^{(1)} = 25 + 0.6 + 0 - 0 \mid 11 = 2.3272$$

$$x_3^{(1)} = \frac{-11 - 2(0.6) + 2.3272 + 0}{10}$$

$$= -0.98728$$

$$x_4^{(1)} = \frac{15 - 3(2.3272) + (-0.98728)}{8} = 0.87889$$

$$x_1^{(k)} = 6 + x_2^{(k-1)} - 2x_3^{(k-1)} \quad | 10$$

$$x_2^{(k)} = 25 + x_1^{(k)} + x_3^{(k-1)} - 3x_4^{(k-1)} \quad | 11$$

$$x_3^{(k)} = -11 - 2x_1^{(k)} + x_2^{(k)} + x_4^{(k-1)} \quad | 10$$

$$x_4^{(k)} = 15 - 3x_2^{(k)} + x_3^{(k)} \quad | 8$$

it	$x_1$	$x_2$	$x_3$	$x_4$
0	0	0	0	0
1	0.6	2.3272	-0.9823	0.87889
2	1.030	2.037	-1.014	0.9844
3	1.0065	2.0036	-1.0025	0.9985
4	1.0009	2.0003	-1.0003	0.9999
5	1.0001	2.0000	-1.0000	1.0000

$$\text{ToA} = \frac{\|x^5 - x^4\|_\infty}{\|x^5\|_\infty} = \frac{0.0008}{2} = 0.0004 < 10^{-3}$$

Note:

Jacobi method required twice as many as iterations for the same accuracy than Seidel method. It means Seidel method converges quickly than Jacobi method.

Iterative method to find Eigenvalues:

Power method:

\* A standard iterative procedure for computing approximate values of the eigen-values of an  $n \times n$  matrix is called the power method.

\* This method applies to any  $n \times n$  matrix that has a dominant eigen values.

\* It means  $|\lambda|$  is greater than the absolute values of the other eigen values

\* Power method helps to find the dominant eigen values.

Thm:

Suppose  $A$  is  $n \times n$  a real symmetric matrix. Let  $x \neq 0$  be any real vector with  $n$  components. Let

$$y = Ax, \quad m_0 = x^T x, \quad M_1 = x^T y$$

$$m_2 = y^T y$$

Then  $\rho = \frac{m_1}{m_0}$  is an approximation for an eigenvalue  $\lambda$  of  $A$ .

Remark: <sup>app.</sup>  $\rightarrow$  <sup>exact</sup>

24  $\hat{q} = \lambda - \epsilon$  so that  $\epsilon$  is

the error of  $\hat{q}$  then

$$|\epsilon| \leq \sqrt{\frac{m_2}{m_0} - \hat{q}^2}$$

Power Method:

(1) Start from any initial vector  
 $x_0 \neq 0$  with  $n$  components

$$(2) \quad x_1 = Ax_0$$

$$x_2 = Ax_1$$

$\vdots$

$$x_n = Ax_{n-1}$$

Remark:

$$Ax = \begin{bmatrix} a_{ij} \end{bmatrix}_{n \times n} \begin{bmatrix} x_i \end{bmatrix}_{n \times 1} \\ = \begin{bmatrix} \lambda_i \end{bmatrix}_{n \times 1}$$

$$= \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_n \end{bmatrix}$$

$$= \lambda_1 \begin{bmatrix} 1 \\ \lambda_2 / \lambda_1 \\ \vdots \\ \lambda_n / \lambda_1 \end{bmatrix}$$

if  $\lambda_1 > \lambda_2 > \lambda_3 > \dots > \lambda_n$

Note We have to scale with the largest (in magnitude only) value of the vector.

1) Use power method to find the dominant eigen value of  $A = \begin{bmatrix} 5 & 4 \\ 1 & 2 \end{bmatrix}$

Sol. Initial guess  $x_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \neq 0$

$$(5-\lambda)(2-\lambda) - 4 = 0$$

$$\lambda^2 - 7\lambda + 6 = 0$$

$$\begin{array}{r|l} -7 & 6 \\ \hline -1 & \\ -6 & \end{array}$$

$$x_1 = Ax_0 = \begin{bmatrix} 5 & 4 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 5 \\ 1 \end{bmatrix} = \textcircled{5} \begin{bmatrix} 1 \\ 0.2 \end{bmatrix}$$

$$x_2 = Ax_1 = \begin{bmatrix} 5.8 \\ 1.4 \end{bmatrix}$$

$$= 5.8 \begin{bmatrix} 1 \\ 0.2423 \end{bmatrix}$$

$$x_3 = Ax_2$$

$$= 5.9652 \begin{bmatrix} 1 \\ 0.2485 \end{bmatrix}$$

$$x_4 = Ax_3$$

$$= 5.994 \begin{bmatrix} 1 \\ 0.24974 \end{bmatrix}$$

$$x_5 = Ax_4$$

$$= 5.99896 \begin{bmatrix} 1 \\ 0.24995 \end{bmatrix}$$

Dominant eigenvalue  
Corresponds

EV



2)

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

$$(1) \quad x_0 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \neq 0$$

$$\begin{aligned} x_1 &= Ax_0 \\ &= \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ -0.5 \\ 0 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} x_2 &= Ax_1 \\ &= \begin{bmatrix} 2.5 \\ -2 \\ 0.5 \end{bmatrix} = 2.5 \begin{bmatrix} 1 \\ -0.8 \\ 0.2 \end{bmatrix} \end{aligned}$$

$$x_3 = 2.8 \begin{bmatrix} 1 \\ -1 \\ 0.43 \end{bmatrix}$$

$$x_4 = 3.43 \begin{bmatrix} 0.87 \\ -1 \\ 0.54 \end{bmatrix}$$

$$x_5 = 3.41 \begin{bmatrix} 0.8 \\ \textcircled{-1} \\ 0.61 \end{bmatrix}$$

$$x_6 = 3.41 \begin{bmatrix} 0.74 \\ \textcircled{-1} \\ 0.64 \end{bmatrix}$$

Restriction:

- \* It gives only dominant eigenvalue
- \* If A has more than one dominant eigen value then power method may not converges.