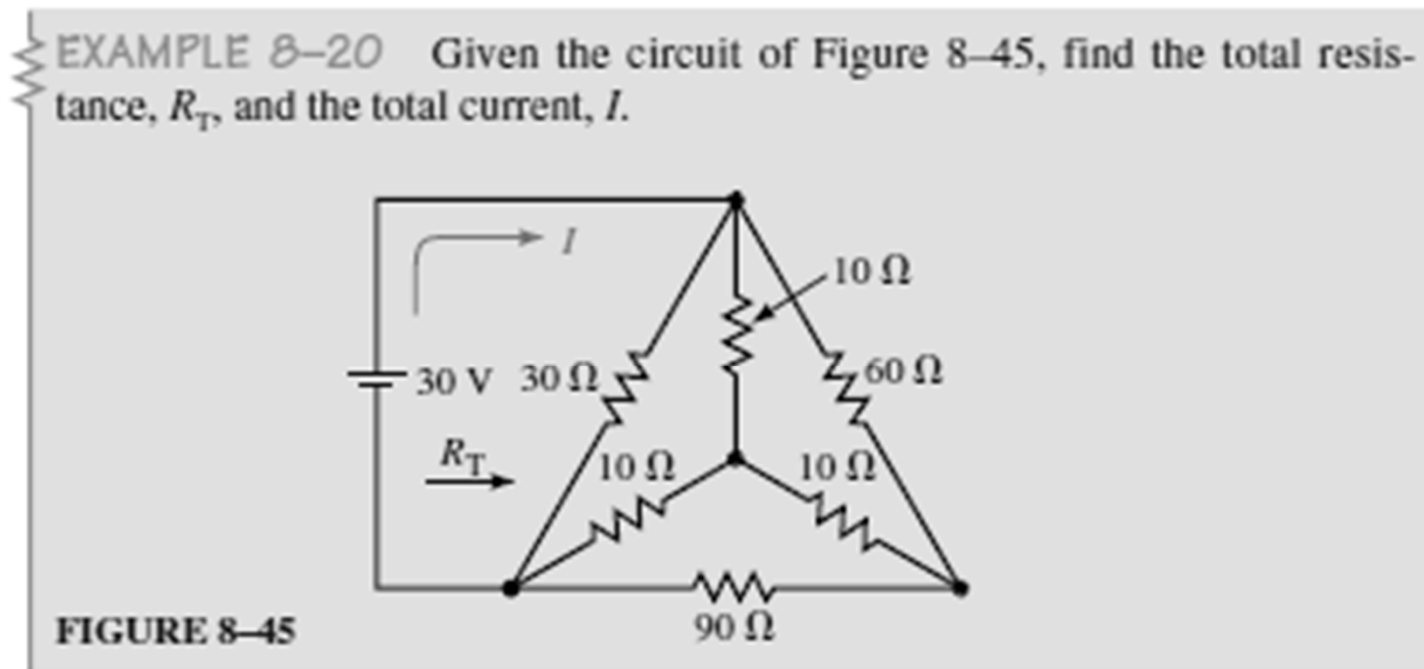


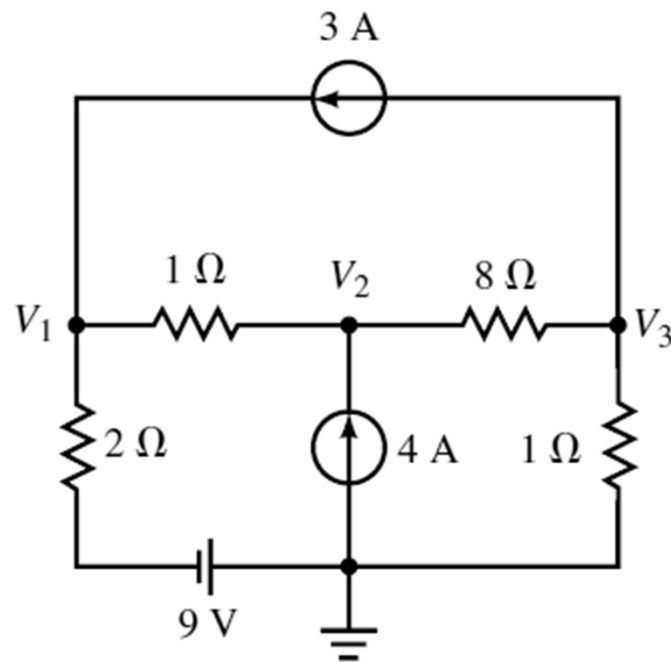
System Of Equations

Find out the current and the resistance



Find out V_1 , V_2 and V_3

FIGURE 8–38



Answers: $V_1 = 3.00\text{ V}$, $V_2 = 6.00\text{ V}$, $V_3 = -2.00\text{ V}$

Linear Systems of Equations

- Constant coefficients with several variables.
- Typically, the number of condition is the same as variables. (Condition shall still be examined)

$$E_1 : a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$E_2 : a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

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

Linear Systems of Equations

- Analytical solutions have been taught in high school and in the university.
- Huge scale equations are not easy to solve by hand.
- Try to solve these equations with computer

System Of Equations

- Direct Method
 - Gaussian Elimination with Backward Substitution
 - Pivoting Strategy
 - Matrix Factorization (LU decomposition)
- Iterative Methods

Example (I)

- How do you solve these equations

$$E_1 : x_1 + x_2 + 3x_4 = 4$$

$$E_2 : 2x_1 + x_2 - x_3 + x_4 = 1$$

$$E_3 : 3x_1 - x_2 - x_3 + 2x_4 = -3$$

$$E_4 : -x_1 + 2x_2 + 3x_3 - x_4 = 4$$

- Crammer's Rule
- 加減消去法，帶入消去法，Gauss Elimination

Example (I)

- $(E_2 - 2E_1) \rightarrow E_2, E_3 - 3E_1 \rightarrow E_3, E_4 + E_1 \rightarrow E_1$

$$E_1 : x_1 \quad + x_2 \quad \quad \quad + 3x_4 = \quad 4$$

$$E_2 : 2x_1 \quad + x_2 \quad - x_3 \quad + x_4 = \quad 1$$

$$E_3 : 3x_1 \quad - x_2 \quad - x_3 \quad + 2x_4 = \quad -3$$

$$E_4 : -x_1 \quad + 2x_2 \quad + 3x_3 \quad - x_4 = \quad 4$$

Example (I)

- $(E_3 - 4E_2) \rightarrow E_3, E_4 + 3E_2 \rightarrow E_4$

$$E_1 : x_1 + x_2 + 3x_4 = 4$$

$$E_2 : -x_2 - x_3 - 5x_4 = -7$$

$$E_3 : -4x_2 - x_3 - 7x_4 = -15$$

$$E_4 : 3x_2 + 3x_3 + 2x_4 = 8$$

Example (I)

- Generally an augmented array is used to simplify the representation of Gauss Elimination

$$\begin{array}{l}
 E_1: x_1 + x_2 + 3x_4 = 4 \\
 E_2: 2x_1 + x_2 - x_3 + x_4 = 1 \\
 E_3: 3x_1 - x_2 - x_3 + 2x_4 = -3 \\
 E_4: -x_1 + 2x_2 + 3x_3 - x_4 = 4
 \end{array}
 \left| \begin{array}{cccc|c}
 1 & 1 & 0 & 3 & 4 \\
 2 & 1 & -1 & 1 & 1 \\
 3 & -1 & -1 & 2 & -3 \\
 -1 & 2 & 3 & -1 & 4
 \end{array} \right|$$

Example (I)

- $(E_2 - 2E_1) \rightarrow E_2, E_3 - 3E_1 \rightarrow E_3, E_4 + E_1 \rightarrow E_4$

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

Example (I)

- $(E_2 - 2E_1) \rightarrow E_2, E_3 - 3E_1 \rightarrow E_3, E_4 + E_1 \rightarrow E_4$

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

$$\left[\begin{array}{cccc|c} 1 & 1 & 0 & 3 & 4 \\ 0 & -1 & -1 & -5 & -7 \\ 0 & -4 & -1 & -7 & -15 \\ 0 & 3 & 3 & 2 & 8 \end{array} \right]$$

Example (I)

- $E_3 - 4E_2 \rightarrow E_3, E_4 + 3E_2 \rightarrow E_4$

$$\left[\begin{array}{cccc|c} 1 & 1 & 0 & 3 & 4 \\ 0 & -1 & -1 & -5 & -7 \\ 0 & -4 & -1 & -7 & -15 \\ 0 & 3 & 3 & 2 & 8 \end{array} \right]$$

$$\left[\begin{array}{cccc|c} 1 & 1 & 0 & 3 & 4 \\ 0 & -1 & -1 & -5 & -7 \\ 0 & 0 & 3 & 13 & 13 \\ 0 & 0 & 0 & -13 & -13 \end{array} \right]$$

Example (I)

- Row Echelon Form

$$\begin{bmatrix} 1 & 1 & 0 & 3 & | & 4 \\ 0 & -1 & -1 & -5 & | & -7 \\ 0 & 0 & 3 & 13 & | & 13 \\ 0 & 0 & 0 & -13 & | & -13 \end{bmatrix}$$

$$\begin{array}{rclclcl}
 x_1 & + & x_2 & & + & 3x_4 & = & 4 \\
 & & -x_2 & -x_3 & - & 5x_4 & = & -7 \\
 & & & 3x_3 & + & 13x_4 & = & 13 \\
 & & & & - & 13x_4 & = & -13
 \end{array}$$

Example (I)

- Backward Substitution

$$\begin{array}{rclcl} x_1 & + x_2 & & + 3x_4 & = 4 \\ & - x_2 & - x_3 & - 5x_4 & = -7 \\ & & 3x_3 & + 13x_4 & = 13 \\ & & & - 13x_4 & = -13 \end{array}$$

$$x_4 = \frac{-13}{-13} = 1$$

$$x_3 = \frac{13 - (13x_4)}{3} = 0$$

$$x_2 = \frac{-7 - (-x_3 - 5x_4)}{(-1)} = 2$$

$$x_1 = \frac{4 - (x_2 + 0x_3 + 3x_4)}{1} = -1$$

Gauss Elimination with backward substitution

- Converting the problem to the augmented matrix

$$E_1 : a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$E_2 : a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

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

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} & b_n \end{bmatrix}$$

Gauss Elimination

- For $i = 1:n-1$

For $j = i+1 : n$

Apply $(E_j - (a_{ji}/a_{ii})E_i) \rightarrow E_j$

End

End

a_{11}	a_{12}	\dots	a_{1n}	b_1
a_{21}	a_{22}	\dots	a_{2n}	b_2
\vdots	\vdots	\vdots	\vdots	\vdots
a_{n1}	a_{n2}	\dots	a_{nn}	b_n
- Row Echelon Form

a_{11}	a_{12}	\dots	a_{1n}	$a_{1,n+1}$
0	a_{22}	\dots	a_{2n}	$a_{2,n+1}$
\vdots	\vdots	\vdots	\vdots	\vdots
0	0	\dots	a_{nn}	$a_{n,n+1}$

Backward substitution

- Check a_{nn} first.

$$\begin{array}{ccccc}
 a_{11} & a_{12} & \dots & a_{1n} & a_{1,n+1} \\
 0 & a_{22} & \dots & a_{2n} & a_{2,n+1} \\
 \vdots & \vdots & \vdots & \vdots & \vdots \\
 0 & 0 & \dots & a_{nn} & a_{n,n+1}
 \end{array}$$

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

$$x_n = \frac{a_{n,n+1}}{a_{n,n}}$$

$$x_{n-1} = \frac{a_{n-1,n+1} - a_{n-1,n}x_n}{a_{n-1,n-1}}$$

$$x_{n-2} = \frac{a_{n-2,n+1} - (a_{n-2,n}x_n + a_{n-2,n-1}x_{n-1})}{a_{n-2,n-2}}$$

\vdots

$$x_1 = \frac{a_{1,n+1} - \sum_{j=2}^n a_{1j}x_j}{a_{11}}$$

Example (II)

- Consider the linear system

$$E_1 : x_1 - x_2 + 2x_3 - x_4 = 4$$

$$E_2 : 2x_1 - 2x_2 + 3x_3 - 3x_4 = -7$$

$$E_3 : x_1 + x_2 + x_3 = -15$$

$$E_4 : x_1 - x_2 + 4x_3 + 3x_4 = 8$$

Example (II)

- After eliminating x_1 , the pivot element of row 2 is 0.

$$\begin{bmatrix} 1 & -1 & 2 & -1 & -8 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 2 & -1 & 1 & 6 \\ 0 & 0 & 2 & 4 & 12 \end{bmatrix}$$

Example (II)

- interchange Row 2 and Row 3, then carry on Gaussian Elimination

$$\begin{bmatrix} 1 & -1 & 2 & -1 & -8 \\ 0 & 2 & -1 & 1 & 6 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 0 & 2 & 4 & 12 \end{bmatrix}$$

Example (II)

- $E_4 + 2E_3 \rightarrow E_4$

$$\begin{bmatrix} 1 & -1 & 2 & -1 & -8 \\ 0 & 2 & -1 & 1 & 6 \\ 0 & 0 & -1 & -1 & -4 \\ 0 & 0 & 0 & 2 & 4 \end{bmatrix}$$

Gaussian Elimination with Pivoting and Backward Substitution

- To solve the nxn linear system

$$E_1 : a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$E_2 : a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

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

- INPUT: number of unknowns, the augmented matrix
- OUTPUT: x_1, \dots, x_n or other message

Gaussian Elimination with Pivoting and Backward Substitution

- For $i = 1:n-1$ (row index)
 - {Find the column index of the 1st non-zero term:p}
 - {Interchange i-th row to p-th row}
 - {if first n column elements are all 0, no unique sol}
 - For $j = i+1:n$
 - $m_{ji} = a_{ji}/a_{ii};$
 - $E_j = E_j - m_{ji} * E_i$
 - end
- End

Gaussian Elimination with Pivoting and Backward Substitution

- If $a_{nn}=0$, {no unique solution}

- $x_n = a_{n,n+1}/a_{n,n}$

- For $i = 1:n-1$

for $j = 0:i-1$

$$x_{n-i} = x_{n-i} + x_{n-j} a_{n-i,n-j}$$

end

$$x_{n-i} = [a_{n-i,n+1} - x_{n-i}] / a_{n-i,n-i};$$

- End

- Output x_1-x_n ;

Computation Complexity

- Arithmetic only takes places at calculating the ratio to the pivoting element, subtraction each row , and backward substitution
- And addition/subtraction and multiplication/division are separated counted as they takes different time to perform calculation

Computation Complexity (Row Echelon Form)

- At the step $E_j - m_{ji} E_i$
 - M_{ji} : $n-i$ division
 - $m_{ji} E_i$ $(n-i)(n-i+1)$ multiplications
 - $E_j - m_{ji} E_i$: $(n-i)(n-i+1)$ subtractions
- Because $I = 1:n$
 - Multiplication/Division : $\frac{2n^3 + 3n^2 - 5n}{6}$
 - Addition / Subtraction: $\frac{n^3 - n}{3}$

Computation Complexity (Backward Substitution)

- Backward substitution

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

- $a_{ij}x_j$: (n-i) multiplication
- Sigma: n-i-1 addition
- And $i = 1:n-1$

- Multiplication/Division :

$$\frac{n^2 + n}{2}$$

- Addition / Subtraction:

$$\frac{n^2 - n}{2}$$

Total Number of computation

- Multiplication/Division :

$$\frac{2n^3 + 3n^2 - 5n}{6} + \frac{n^2 + n}{2} = \frac{n^3}{3} + n^2 - \frac{n}{3}$$

- Addition / Subtraction

$$\frac{n^3 - n}{3} + \frac{n^2 - n}{2} = \frac{n^3}{3} + \frac{n^2}{2} - \frac{5n}{6}$$

Gaussian Elimination with Backward Substitution

- A very basic approach to find the roots of a linear system of equations
- The computation complexity is proportional to n^3

A case where the round-off error becomes severe

- Explore the root of the equation using 4 digit arithmetic

$$E_1 : \quad 0.003000x_1 \quad + 59.14x_2 \quad = 59.17$$

$$E_2 : \quad 5.291x_1 \quad - 6.130x_2 \quad = 46.78$$

$$m_{11} = \frac{5.291}{0.003000} = 1763.66\dots$$

A case where the round-off error becomes severe

- $E_2 - 1764 * E_1 \rightarrow E_2$

$$E_1 : \quad 0.003000x_1 \quad + 59.14x_2 \quad = 59.17$$

$$E_2 : \quad \quad \quad -104300x_2 \quad = -104400$$

$$\Rightarrow x_2 \approx \frac{104400}{104300} = 1.001 \qquad x_1 \approx \frac{59.17 - (59.14) * 1.001}{0.003000} = -10.00$$

- correct answer should be $x_1 = 10, x_2 = 1$

Reconsider Example (III)

- What if the two equations are interchange prior to the elimination performs?

$$E_1 : \quad 5.291x_1 \quad - 6.130x_2 \quad = 46.78$$

$$E_2 : \quad 0.003x_1 \quad + 59.14x_2 \quad = 59.17$$

$$m_{21} = \frac{0.003000}{5.291} = 0.0005670\dots$$

Reconsider Example (III)

- $E_2 - 0.0005670 * E_1 \rightarrow E_2$

$$E_1 : \quad 5.291x_1 \quad - 6.130x_2 \quad = 46.78$$

$$E_2 : \quad \quad \quad + 59.14x_2 \quad = 59.14$$

$$\Rightarrow x_2 \approx \frac{59.14}{59.14} = 1$$

$$x_1 \approx \frac{59.17 - (59.14) * 1.00}{0.003000} = 10.00$$

- correct answer should be $x_1 = 10, x_2 = 1$

Maximal Column Pivoting

- Before performing Gaussian elimination, the maximum pivot element of each column will be interchanged to the upper row.
- Maximum column pivoting can “SOMEWHAT” avoid round-off error in Gaussian elimination.
- aka. Partial Pivoting

Gaussian Elimination

- For $i = 1:n-1$ (row index)
 - {Find the column index of the 1st non-zero term:p}
 - {Interchange i-th row to p-th row}
 - {if first n column elements are all 0, no unique sol}
 - For $j = i+1:n$
 - $m_{ji} = a_{ji}/a_{ii};$
 - $E_j = E_j - m_{ji} * E_i$
 - end
- End

Gaussian Elimination with Max Column Pivoting

- For $i = 1:n-1$ (row index)
 - {from i -th row to n -th row,
find the max abs(coefficient) of i -th COLUMN, P }
 - {Interchange i -th row to p -th row}
 - {if all i -th column elements are 0, no unique sol}
 - For $j = i+1:n$
 - $m_{ji} = a_{ji}/a_{ii};$
 - $E_j = E_j - m_{ji} * E_i$
 - end
- End

Example (III)

- What if original E_1 is multiplied by 10000

$$E_1 : \quad 30.00x_1 \quad + 591400x_2 \quad = 59.1700$$

$$E_2 : \quad 5.291x_1 \quad - 6.130x_2 \quad = 46.78$$

- The equations satisfy maximum column pivoting criteria.

$$\Rightarrow x_2 \approx \frac{104400}{104300} = 1.001 \quad x_1 \approx \frac{591700 - (591400) * 1.001}{30.00} = -10.00$$

Scaled Column Pivoting

- Finding out the maximum coefficient of each row.
- Scale pivoting column with the maximum coefficient
- Use maximum column pivoting strategy according to scaled pivot column elements

Scaled Column Pivoting

- $a_{11}/\max(E_1) = 30/591400 = 0.5073 \times 10^{-4}$
- $a_{21}/\max(E_2) = 5.291/6.130 = 0.8631$
- E_1 and E_2 shall be interchanged

$$E_1 : \quad 30.00x_1 \quad + 591400x_2 \quad = 59.1700$$

$$E_2 : \quad 5.291x_1 \quad - 6.130x_2 \quad = 46.78$$

Gauss Elimination with Maximal Column Pivoting

- For $i = 1:n-1$ (row index)
 - {from i -th row to n -th row,
find the max abs(coefficient) of i -th COLUMN, P }
 - {Interchange i -th row to p -th row}
 - {if first n column elements are all 0, no unique sol}
 - For $j = i+1:n$
 - $m_{ji} = a_{ji}/a_{ii};$
 - $E_j = E_j - m_{ji} * E_i$
 - end
- End

Gauss Elimination with Scaled Column

- Find max row elements of each row, $s(i)$
- For $I = 1:n-1$ (row index)
 - {from i -th row to n -th row,
find the max $\text{abs}(\text{coefficient}/s(i))$ of i -th COLUMN, P }
 - {Interchange i -th row to p -th row}
 - {if all i -th column elements are 0, no unique sol}
 - For $j = i+1:n$
 - $m_{ji} = a_{ji}/a_{ii};$
 - $E_j = E_j - m_{ji} * E_i$
 - end
- End

Computation Complexity

- For each row, there should be $n-1$ comparisons
 - Total $n(n-1)$ comparisons
- To determine the interchange of 1st column
 - n divisions and $n-1$ comparisons
- To determine interchange of k -th column
 - $n-k+1$ divisions and $n-k$ comparisons

Computation Complexity

- Divisions

$$\sum_{k=1}^{n-1} (n - k + 1) = \frac{n(n+1)}{2} - 1$$

- Comparisons

$$n(n-1) + \sum_{k=1}^{n-1} (n - k) = \frac{3}{2}n(n-1)$$

- Still way smaller than Gauss Elimination $O(n^3)$

Iterative Methods

Similar to finding a root for nonlinear
equations

Iterative Methods

- These methods are mostly similar to those used to solve non-linear equation of single variable.
- What can be used to solve a system of non-linear equations can also be used to solve linear equations.

Iterative Methods

- Pick up an initial guess
- Use a rational iteration process to approach the true answer
- Rational “Iteration Relationship” and “Stop Criteria” will be needed

Iterative Methods

- Fixed Point Method
 - Linear and non-linear system
- Newton's Method
 - Non-linear system
- Conjugate Gradient
 - Symmetric and positive definite system
- Steepest Descent
 - Linear and non-linear

When to stop an iteration?

- For the case of a single variable
(e.g. $x - \cos x = 0$)
 - Criterion the absolute error: $|x_n - x_{n+1}| < \epsilon$
 - Criterion the relative error: $\frac{|x_n - x_{n+1}|}{|x_{n+1}|} < \epsilon = 10^{-10}$
 - $|x|$ denotes the distance/length for a 1D vector
- For the case of multiple variables, such as a vector $[x_1, x_2, x_3, \dots]$
 - Distance/Length shall be defined for a vector:
Vector Norm

Definition of the Vector Norm

- p-norm of a vector \vec{v} is defined as

$$|\vec{v}|_p = \left(\sum_{i=1}^N |v_i|^p \right)^{\frac{1}{p}}$$

The norm of a vector represents the length of the vector in a specific geometry representation.

Vector Norm

- For a vector \mathbf{x}

$$\mathbf{x} = [x_1, x_2, \dots, x_n]$$

- The L2 norm (Euclidean norm) is defined

$$\|\mathbf{x}\|_2 = (|x_1|^2 + |x_2|^2 + \dots + |x_n|^2)^{\frac{1}{2}}$$

- The L_∞ norm is defined

$$\begin{aligned} \|\mathbf{x}\|_\infty &= \lim_{k \rightarrow \infty} (|x_1|^k + |x_2|^k + \dots + |x_n|^k)^{\frac{1}{k}} \\ &= \max(|x_i|) \end{aligned}$$

Example

- $\vec{v} = [1, -2, 3, 0, 4]$
 - $|\vec{v}|_2 = \sqrt{1 + 4 + 9 + 0 + 16} = \sqrt{30}$
 - $|\vec{v}|_\infty = \max(|v_i|) = 4$
- $\vec{u} = [2, -9, 3, 6, 5]$
 - $|\vec{u}|_2 = \sqrt{2 + 81 + 9 + 36 + 25} = \sqrt{153}$
 - $|\vec{u}|_\infty = \max(|v_i|) = 9$
- $|\vec{u} - \vec{v}|_\infty = |[1, -7, 0, 6, 1]|_\infty = 7$

Norm and stop criteria

- L2 norm is a common measurement of the error or residual length for the stop criteria in iterative method
- And in the following discussion only L2 Norm will be used

- Stop criteria:

$$\|\mathbf{Ax} - \mathbf{b}\| < \varepsilon$$

$$\|\mathbf{x}^{(k+1)} - \mathbf{x}^{(k)}\| < \varepsilon$$

Jacobi Method

Example

- Solve the following equation using Jacobi Method

$$3x + 2y - z = 1$$

$$x - 3y + 2z = 5$$

$$2x + y - 3z = -4$$

Jacobi Method

- For a system of linear equations

$$\mathbf{Ax}=\mathbf{b}$$

- The simplest fixed point counterpart is the following form

$$\mathbf{x} = \mathbf{T}\mathbf{x} + \mathbf{c}$$

$$\mathbf{x}^{(k+1)} = \mathbf{T}\mathbf{x}^{(k)} + \mathbf{c}$$

Example

- Solve the following equation using Jacobi Method

$$E_1 : 10x_1 - x_2 + 2x_3 = 6$$

$$E_2 : -x_1 + 11x_2 - x_3 + 3x_4 = 25$$

$$E_3 : 2x_1 - x_2 + 10x_3 - x_4 = -11$$

$$E_4 : 3x_2 - x_3 + 8x_4 = 15$$

Example

- Fixed point form

$$\begin{aligned}x_1 &= 1 - 1/10x_2 - 1/5x_3 \\x_2 &= 1/11x_1 + 1/11x_3 - 3/11x_4 + 25/11 \\x_3 &= 6 - 1/5x_1 - 1/10x_2 + 1/10x_4 - 11/10 \\x_4 &= 1 - 3/8x_2 + 1/8x_3 + 15/8\end{aligned}$$

- Start with $[0,0,0,0]$

Example

- 1st Iteration

$$\begin{array}{rclcl}
 0.6000 = & 1/10(0) & -1/5(0) & & +3/5 \\
 2.2727 = & 1/11(0) & +1/11(0) & -3/11(0) & 25/11 \\
 -1.1000 = & -1/5(0) & +1/10(0) & +1/10(0) & -11/10 \\
 1.8750 = & -3/8(0) & +1/8(0) & & 15/8
 \end{array}$$

- 2nd Iteration

$$\begin{array}{rclcl}
 1.0473 = & 1/10(2.2727) & -1/5(-1.1) & & +3/5 \\
 1.7159 = & 1/11(0.6) & +1/11(-1.1) & -3/11(1.875) & 25/11 \\
 -0.8052 = & -1/5(0.6) & +1/10(2.2727) & +1/10(1.875) & -11/10 \\
 0.8852 = & -3/8(2.2727) & +1/8(-1.1) & & 15/8
 \end{array}$$

Example

k	0	1	2	3	4	5
$x_1^{(k)}$	0.000	0.6000	1.0473	0.9326	1.0152	0.9890
$x_2^{(k)}$	0.0000	2.2727	1.7159	2.053	1.9537	2.0114
$x_3^{(k)}$	0.0000	-1.1000	-0.8052	-1.0493	-0.9681	-1.0103
$x_4^{(k)}$	0.0000	1.8750	0.8852	1.1309	0.9739	1.0214

6	7	8	9	10
1.0032	0.9981	1.0006	0.9997	1.0001
1.9922	2.0023	1.9987	2.0004	1.9998
-0.9945	-1.0020	-0.9990	-1.0004	-0.9998
0.9944	1.0036	0.9989	1.0006	0.9998

Gauss Seidel Method

Jacobi Method

- Iteration of Jacobi Method

$$\mathbf{x}^{(k+1)} = \mathbf{T}\mathbf{x}^{(k)} + \mathbf{c}$$

- The explicit form

$$x_1^{(k+1)} = T_{12}x_2^{(k)} + T_{13}x_3^{(k)} + \dots + c_1$$

$$x_2^{(k+1)} = T_{21}x_1^{(k)} + T_{23}x_3^{(k)} + \dots + c_2$$

\vdots

$$x_n^{(k+1)} = T_{n1}x_1^{(k)} + T_{n2}x_2^{(k)} + \dots + c_n$$

Gauss Seidel Method

- Conjecture: The $k+1$ -th result shall be more accurate than the result of k -th iteration

$$x_1^{(k+1)} = T_{12}x_2^{(k)} + T_{13}x_3^{(k)} + \dots + c_1$$

$$x_2^{(k+1)} = T_{21}x_1^{(k+1)} + T_{23}x_3^{(k)} + \dots + c_2$$

$$x_3^{(k+1)} = T_{31}x_1^{(k+1)} + T_{32}x_2^{(k+1)} + T_{34}x_4^{(k)} + \dots + c_3$$

\vdots

$$x_n^{(k+1)} = T_{n1}x_1^{(k+1)} + T_{n2}x_2^{(k+1)} + \dots + T_{n,n-1}x_{n-1}^{(k+1)} + c_n$$

Example

- Solve the following equation using Gauss Seidel Method

$$E_1 : 10x_1 - x_2 + 2x_3 = 6$$

$$E_2 : -x_1 + 11x_2 - x_3 + 3x_4 = 25$$

$$E_3 : 2x_1 - x_2 + 10x_3 - x_4 = -11$$

$$E_4 : \quad 3x_2 - x_3 + 8x_4 = 15$$

Example

- Fixed point form

$$\begin{aligned}x_1 &= && 1/10x_2 && -1/5x_3 && +3/5 \\x_2 &= 1/11x_1 && && +1/11x_3 && -3/11x_4 && 25/11 \\x_3 &= -1/5x_1 && +1/10x_2 && && +1/10x_4 && -11/10 \\x_4 &= && -3/8x_2 && +1/8x_3 && && 15/8\end{aligned}$$

- Start with $[0,0,0,0]$

Example

- 1st Iteration

$$\begin{array}{rcllcl}
 0.6 = & & 1/10(0) & & -1/5(0) & & +3/5 \\
 2.3272 = & 1/11(0.6) & & & +1/11(0) & -3/11(0) & 25/11 \\
 -0.9873 = & -1/5(0.6) & +1/10(2.3272) & & & +1/10(0) & -11/10 \\
 0.8789 = & & -3/8(2.3272) & +1/8(-0.9873) & & & 15/8
 \end{array}$$

- 2nd Iteration

$$\begin{array}{rcllcl}
 1.030 = & & 1/10(2.372) & & -1/5(-0.9873) & & +3/5 \\
 2.037 = & 1/11(1.030) & & & +1/11(-0.9873) & -3/11(0.8789) & 25/11 \\
 -1.014 = & -1/5(1.030) & +1/10(2.037) & & & +1/10(0.8789) & -11/10 \\
 0.9844 = & & -3/8(2.037) & +1/8(-1.014) & & & 15/8
 \end{array}$$

Example

k	0	1	2	3	4	5
$x_1^{(k)}$	0.0000	0.6000	1.030	1.0065	1.0009	1.0001
$x_2^{(k)}$	0.0000	2.3272	2.037	2.0036	2.0003	2.0000
$x_3^{(k)}$	0.0000	-0.9873	-1.014	-1.0025	-1.0003	-1.0000
$x_4^{(k)}$	0.0000	0.8789	0.9844	0.9983	0.9999	1.0000

Geometric Illustration of the 2D Fixed point method

- Consider a two variable linear system

$$E_1 : \quad 5x_1 \quad - x_2 \quad = 6$$

$$E_2 : \quad x_1 \quad - 3x_2 \quad = 4$$

- The fixed point counterpart is

$$f_1(x_2) = x_1 = \quad \quad \quad 1/5x_2 \quad + 6/5$$

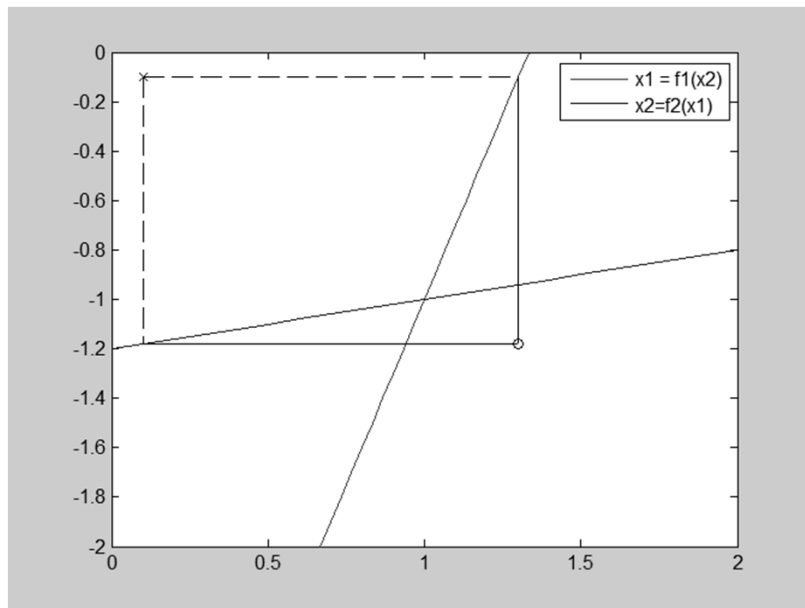
$$f_2(x_1) = x_2 = \quad 1/3x_1 \quad \quad \quad - 4/3$$

Geometric Illustration of the 2D Fixed point method

- Jacobi Iteration

$$x_1^{(k+1)} = \frac{1}{5}x_2^{(k)} - 1.6/5$$

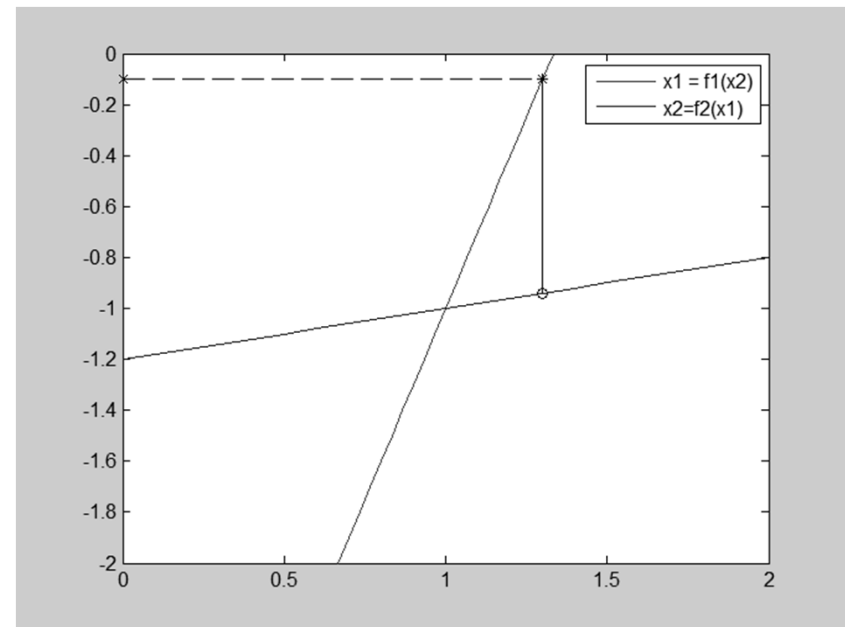
$$x_2^{(k+1)} = \frac{1}{3}x_1^{(k)} - 4/3$$



- Gauss Seidel Iteration

$$x_1^{(k+1)} = \frac{1}{5}x_2^{(k)} - 1.6/5$$

$$x_2^{(k+1)} = \frac{1}{3}x_1^{(k+1)} - 4/3$$



Geometric Illustration of the 2D Fixed point method

- What if E1 and E2 interchanges?

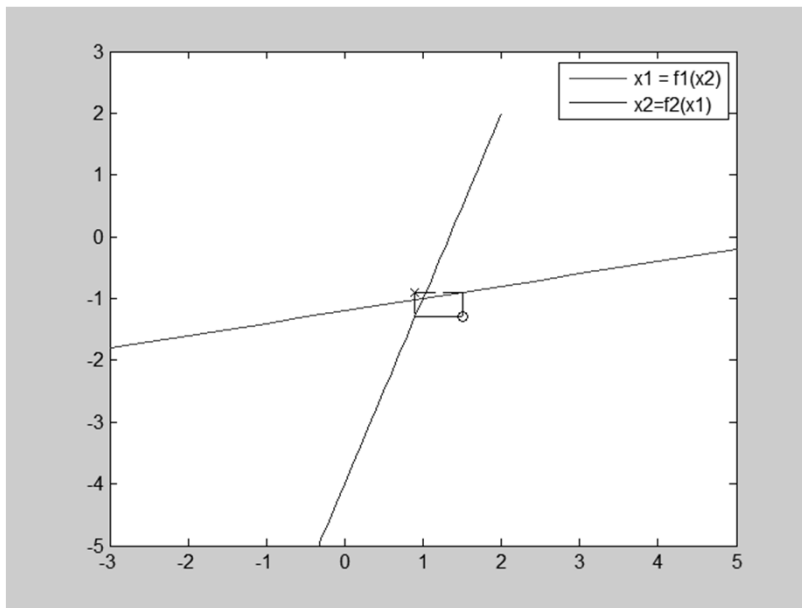
$$\begin{array}{lcl} E_1 : & x_1 & - 3x_2 = 4 \\ E_2 : & 5x_1 & - x_2 = 6 \end{array} \longrightarrow \begin{array}{lcl} & x_1 = & 3x_2 + 4 \\ & x_2 = & 5x_1 - 6 \end{array}$$

- Linear algebra says, Two interchanged rows shall not affect the answer

Geometric Illustration of the 2D Fixed point method

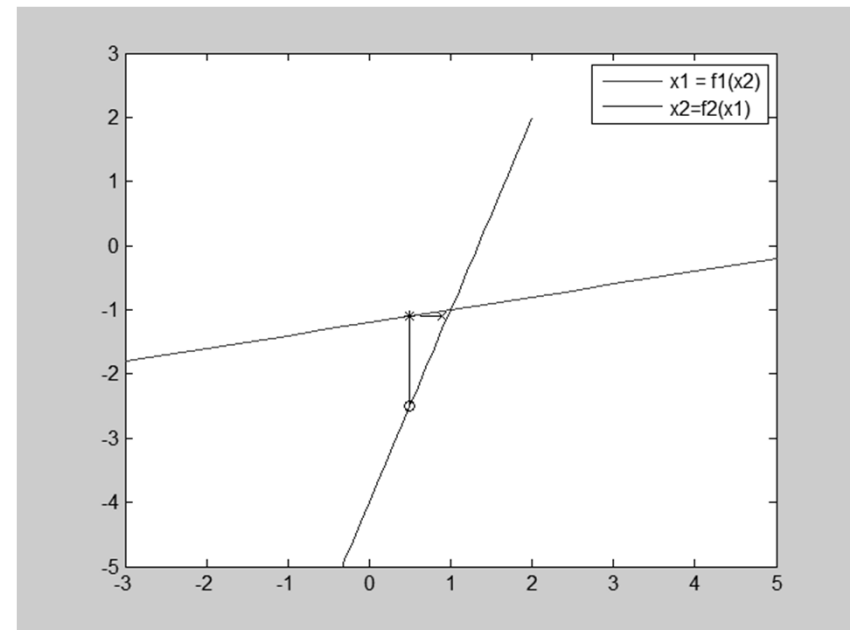
- Jacobi Iteration

$$\begin{aligned} x_1^{(k+1)} &= 3x_2^{(k)} - 1.4 \\ x_2^{(k+1)} &= 5x_1^{(k)} - 0.6 \end{aligned}$$



- Gauss Seidel Iteration

$$\begin{aligned} x_1^{(k+1)} &= 3x_2^{(k)} - 1.4 \\ x_2^{(k+1)} &= 5x_1^{(k+1)} - 0.6 \end{aligned}$$



Condition of Convergence

- Assume that \mathbf{p}_0 is the correct fixed point of the following fixed point problem

$$\mathbf{x}^{(k+1)} = \mathbf{T}\mathbf{x}^{(k)} + \mathbf{c}$$

- The distance between an approximated point \mathbf{p} and \mathbf{p}_0 is

$$\left\| \mathbf{p}^{(k+1)} - \mathbf{p}_0^{(k+1)} \right\| = \left\| \mathbf{T}(\mathbf{p}^{(k)} - \mathbf{p}_0^{(k)}) \right\|$$

$$\left\| \mathbf{p}^{(k+1)} - \mathbf{p}_0^{(k+1)} \right\| \leq \|\mathbf{T}\| \left\| \mathbf{p}^{(k)} - \mathbf{p}_0^{(k)} \right\|$$

Condition of Convergence

- A convergent fixed point algorithm shall generate a series of approximated point approaching to the true answer

$$\lim_{k \rightarrow \infty} \left\| \mathbf{p}^{(k+1)} - \mathbf{p}_0^{(k+1)} \right\| = \lim_{k \rightarrow \infty} \left\| \mathbf{T}(\mathbf{p}^{(k)} - \mathbf{p}_0^{(k)}) \right\| = 0$$

- L2-Norm of T shall be smaller than 1

Condition of Convergence

- For any \mathbf{x} defined by

$$\mathbf{x}^{(k+1)} = \mathbf{T}\mathbf{x}^{(k)} + \mathbf{c}$$

- Converges to a unique solution if and only if the maximum eigen value of \mathbf{T} is smaller than 1

Conditional Convergence

- For any eigen value λ of T , $1 - \lambda$ is an eigen value of $(I - T)$. Because no eigen value is greater than 1, that means $1 - \lambda$ cannot be zero leading that $I - T$ is non-singular.
- So $(I - T)^{-1}$ exists
- Original equation is

$$(I - T)\mathbf{x} = \mathbf{c} \qquad \mathbf{x} = (I - T)^{-1}\mathbf{c}$$

Conditional Convergence

- Assume a matrix S_m is

$$S_m = I + T + T^2 + \dots + T^m$$

- Multiply both side by $(I+T)$

$$(I - T)S_m = (I - T)(I + T + T^2 + \dots + T^m) = I - T^{m+1}$$

- Because max eigen value of is smaller than 1

$$\lim_{m \rightarrow \infty} T^m = 0$$

- therefore

$$\lim_{m \rightarrow \infty} S_m = I + T + T^2 + \dots + T^m + \dots = (I - T)^{-1}$$

Conditional Convergence

- The expansion of the iteration is

$$\begin{aligned}\mathbf{x}^{(k)} &= \mathbf{T}\mathbf{x}^{(k-1)} + \mathbf{c} = \mathbf{T}(\mathbf{T}\mathbf{x}^{(k-2)} + \mathbf{c}) + \mathbf{c} \\ &= \mathbf{T}^2\mathbf{x}^{(k-2)} + (\mathbf{T} + \mathbf{I})\mathbf{c} = \mathbf{T}^2(\mathbf{T}\mathbf{x}^{(k-3)} + \mathbf{c}) + (\mathbf{T} + \mathbf{I})\mathbf{c} \\ &= \mathbf{T}^3\mathbf{x}^{(k-3)} + (\mathbf{T}^2 + \mathbf{T} + \mathbf{I})\mathbf{c} \\ &\vdots \\ &= \mathbf{T}^k\mathbf{x}^{(0)} + (\mathbf{T}^{k-1} + \dots + \mathbf{T}^2 + \mathbf{T} + \mathbf{I})\mathbf{c}\end{aligned}$$

- Taking the limit

$$\begin{aligned}\lim_{m \rightarrow \infty} \mathbf{x}^{(k)} &= \lim_{m \rightarrow \infty} \mathbf{T}^k\mathbf{x}^{(0)} + \lim_{m \rightarrow \infty} (\mathbf{T}^{k-1} + \dots + \mathbf{T} + \mathbf{I})\mathbf{c} \\ &= (\mathbf{I} - \mathbf{T})^{-1}\mathbf{c}\end{aligned}$$