

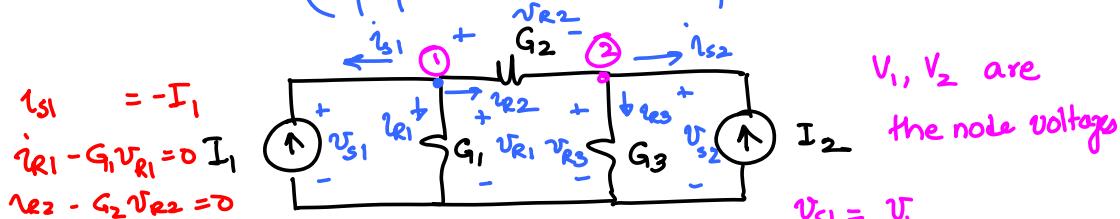
Office Hours: Tu/Th 5-6 pm KEC 409S
or by appointment

Formulation of circuit equations

- Nodal analysis (NA)

- Modified nodal analysis (MNA)

(paper on MNA posted on class website)



KCL at nodes ① & ②

$$i_{S1} + i_{R1} + i_{R2} = 0$$

$$-i_{R2} + i_{R3} + i_{S2} = 0$$

$$v_{S1} = V_1$$

$$v_{R1} = V_1$$

$$v_{R2} = V_1 - V_2$$

$$v_{R3} = V_2$$

$$v_{S2} = V_2$$

$$\underbrace{\begin{bmatrix} 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & -1 & 1 & 1 \end{bmatrix}}_A \begin{bmatrix} i_{S1} \\ i_{R1} \\ i_{R2} \\ i_{R3} \\ i_{S2} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Incidence matrix

General form $A I_b = 0$ KCL

Branch to node voltage relations

$$\begin{bmatrix} v_{S1} \\ v_{R1} \\ v_{R2} \\ v_{R3} \\ v_{S2} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & -1 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \end{bmatrix}$$

$$V_b = A^T V_n$$

BCR or BCE

$$\begin{bmatrix} i_{S1} \\ i_{R1} \\ i_{R2} \\ i_{R3} \\ i_{S2} \end{bmatrix} - \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & G_1 & 0 & 0 & 0 \\ 0 & 0 & G_2 & 0 & 0 \\ 0 & 0 & 0 & G_3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} v_{S1} \\ v_{R1} \\ v_{R2} \\ v_{R3} \\ v_{S2} \end{bmatrix} = \begin{bmatrix} I_1 \\ 0 \\ 0 \\ 0 \\ -I_2 \end{bmatrix}$$

$$I_b - G_b V_b = S$$

In general: $K_i I_b + K_v V_b = S$

(Augmented) Incidence Matrix A'

$A' = \begin{bmatrix} & 1 & 2 & 3 & \dots & j & \dots & b \\ 1 & & & & & & & \\ 2 & & & & & & & \\ 3 & & & & & & & \\ \vdots & & & & & & & \\ i & & & & & & & (+1, -1, 0) \\ \vdots & & & & & & & \\ n+1 & & & & & & & \end{bmatrix}$

includes ground → nodes i branches b (n+1) nodes b branches

$A_{ij} = +1$ if node i is the '+' terminal of branch j

$A_{ij} = -1$ if node i is the '-' terminal branch j

$A_{ij} = 0$ if node i is not connected to branch j

All columns will have 2 entries & sum up to zero

The Sparse Tableau Analysis (STA) 1969-71 by IBM research

KCL $A I_b = 0$ $\begin{bmatrix} & I_1 & \vdots & I_b \\ A & & & \end{bmatrix} = 0$

n nodes

KVL $V_b = A^T V_n \Rightarrow V_b - A^T V_n = 0$

BCR $K_i I_b + K_v V_b = S$

Ex R: $I_b - G V_b = 0$

V: $0 I_b - V_b = V_s$

$A I_b = 0$

$V_b - A^T V_n = 0$

$K_i I_b + K_v V_b = S$

$$\begin{bmatrix} I_b \\ \hline V_b \\ \hline V_n \end{bmatrix}$$

$$\begin{array}{c}
 \text{Diagram of the matrix equation:} \\
 \left[\begin{array}{cc|cc|cc}
 & b & & b & n & \\
 & \swarrow & \searrow & & & \\
 A & & 0 & 0 & & \\
 \hline
 & - & | & - & - & \\
 0 & I & | & -A^T & & \\
 \hline
 K_i & + & K_v & | & 0 & \\
 \end{array} \right] \left[\begin{array}{c}
 I_{b1} \\ I_{b2} \\ I_{bb} \\ V_{b1} \\ \vdots \\ V_{bb} \\ V_{n1} \\ \vdots \\ V_{nn}
 \end{array} \right] = \left[\begin{array}{c}
 0 \\ 0 \\ \vdots \\ S
 \end{array} \right]
 \end{array}$$

n+2b unknowns; n+2b equations
 A has at most 2b nonzero entries
 A^T " " 2b "
 I has b nonzeros
 K_i, K_v have at most b nonzero entries

Maximum # of nonzero entries $2b + 2b + b + b + b = 7b$
 Matrix size $(n+2b) \times (n+2b)$ \Rightarrow very sparse matrix

Summary of STA :

- can be applied to any circuit
- Equations can be assembled by inspection
- STA matrix is very sparse

Derive NA from STA

$$A I_b = 0$$

$$V_b - A^T V_n = 0$$

$$K_i I_b + K_v V_b = S$$

$$\text{For NA } I_b = \bullet K_i^{-1} [S - K_v V_b]$$

K_i is invertible

$$A I_b = A K_i^{-1} [S - K_v V_b] = 0$$

$$A K_i^{-1} [S - K_v A^T V_n] = 0$$

$$\underbrace{A K_i^{-1} K_V A^T}_{\text{NA matrix}} \underbrace{V_n}_{\text{node voltages}} = \underbrace{A K_i^{-1} S}_{\text{NA source}}$$

General Form

$Ax = b$ is a system of equations
 ↑
 not the incidence matrix

$x = A^{-1} b$ A is a nonsingular matrix

Solution methods

1) Direct methods — Gaussian Elimination
 LU Decomposition

2) Indirect methods (iterative methods)
 — Gauss Jacobi/Seidel

⋮
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Gaussian Elimination (GE)

$$Ax = b$$

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$\text{row 2} = \text{row 2} - \frac{a_{21}}{a_{11}} \text{row 1}$$

$$\begin{array}{c} \begin{array}{l} \text{multiples} \\ \text{row 2} \\ \text{by } \frac{a_{21}}{a_{11}} \end{array} \\ \begin{array}{ccccc} a_{11} & a_{12} & \cdots & a_{1n} & \\ 0 & a_{22} - \frac{a_{21}}{a_{11}} a_{12} & \cdots & a_{2n} & \\ 0 & \vdots & & \vdots & \\ a_{n1} & a_{n2} - \frac{a_{n1}}{a_{11}} a_{12} & \cdots & a_{nn} & \end{array} \end{array} \quad \begin{array}{c} \begin{array}{l} \text{row 2} \\ = b_2^{(2)} \\ b_2 - \frac{a_{21}}{a_{11}} b_1 \end{array} \\ \begin{array}{ccccc} b_1 & & & & \\ b_2 - \frac{a_{21}}{a_{11}} b_1 & & & & \\ \vdots & & & & \\ b_n & & & & \end{array} \end{array}$$

$$\begin{bmatrix} a_{11} & a_{12}^{(2)} & \cdots & a_{1n}^{(2)} \\ 0 & a_{22}^{(2)} & \cdots & a_{2n}^{(2)} \\ \vdots & & & \\ 0 & a_{n2}^{(2)} & \cdots & a_{nn}^{(2)} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1^{(2)} \\ b_2^{(2)} \\ \vdots \\ b_n^{(2)} \end{bmatrix}$$

$$\begin{bmatrix} a_{11} & a_{12}^{(2)} & \cdots & a_{1n}^{(2)} \\ 0 & a_{22}^{(2)} & \cdots & a_{2n}^{(2)} \\ 0 & 0 & \ddots & \\ & & & a_{nn}^{(n)} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1^{(2)} \\ b_2^{(2)} \\ \vdots \\ b_n^{(n)} \end{bmatrix}$$

This is an upper triangular system

Solve for

$$x_n = \frac{b_n^{(n)}}{a_{nn}^{(n)}}$$

$$x_{n-1} = \frac{b_{n-1}^{(n-1)} - a_{n-1,n}^{(n-1)} x_n}{a_{n-1,n-1}^{(n-1)}}$$

$$x_1 = \frac{b_1 - \sum_{j=2}^n a_{1j} x_j}{a_{11}}$$

This process is called backward substitution

Overall GE: 1) triangularization

$A \rightarrow$ Upper triangular matrix U

2) backward substitution/solve

Triangularization (assume $A[i][i] \neq 0$)

For $i = 1$ to n { each row $\alpha = 1/A[i][i]$

L division For $j = i+1$ to n { each row below row i
 $m = \alpha * A[j][i]$

$(n-i)$ mult

For $k = i+1$ to n { traverse row j

$(n-i)(n-i)$ mult

$(n-i)(n-i)$ addition

$$A[j][k] = A[j][k] - m * A[i][k]$$

$$(n-i) \text{ mult } b[j] = b[j] - m * b[i]$$

$$\text{Total \# of mult/div} = 1 + \frac{(n-i) + (n-i)(n-i)}{(n-i)(n-i+1)} + \underbrace{(n-i)}_{\text{RHS}}$$

$$\sum_{i=1}^n (n-i)(n-i+1) + (n-i) + 1$$

$$\frac{n^3 - n}{3} \quad \frac{n^2 - n}{2} \quad \sim \frac{n^3}{3}$$

Backward substitution $\sim n^2/2$

What about solving $Ax = b_A$
 $= b_B$
 \vdots
 $= b_X$

$$A \rightarrow \begin{bmatrix} U \\ O \end{bmatrix} \begin{bmatrix} x \end{bmatrix} = RHS$$

LU Factorization (or Decomposition)

$$A \rightarrow \begin{bmatrix} & \\ & U \\ & \end{bmatrix} \begin{bmatrix} x \\ \vdots \\ 1 \end{bmatrix} = b$$

$$\begin{aligned} b_1^{(1)} &= b_1 \\ b_2^{(2)} &= b_2 - \frac{a_{21}}{a_{11}} b_1^{(1)} \\ &\vdots \\ b_n^{(n)} &= b_n - \frac{a_{n1}}{a_{11}} b_1^{(1)} - \frac{a_{n2}}{a_{22}} b_2^{(2)} \cdots - \frac{a_{nn}}{a_{nn}} b_n^{(n)} \end{aligned}$$

$$\left[\begin{array}{cccc} 1 & & & x_n \\ l_{21} & 1 & & \\ l_{31} & l_{32} & 1 & \\ \vdots & & & \\ l_{n1} & l_{n2} & \cdots & l_{n,n-1} & 1 \end{array} \right] \left[\begin{array}{c} b_1^{(1)} \\ b_2^{(2)} \\ \vdots \\ b_n^{(n)} \end{array} \right] = \left[\begin{array}{c} b_1 \\ b_2 \\ \vdots \\ b_n \end{array} \right]$$

$$Ax = b$$

$$\underline{L} \circled{U} x = b$$



$\sim n^2/2$

$\sim n^2/2$

$$Ly = b.$$

$$Ux = y$$

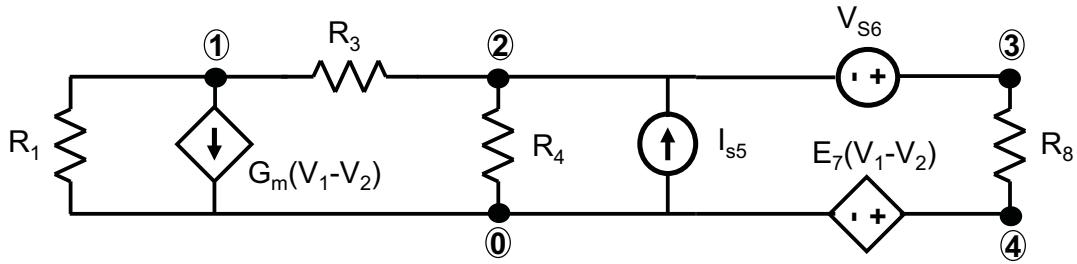
Forward Substitution

Backward Substitution

One time operation LU Factorization $\sim n^3/3$

Repeated Forward/Back solves for different
RHS $\sim n^2$

MNA Example



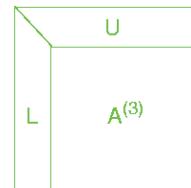
$$\left[\begin{array}{ccc|cc} \frac{1}{R_1} + G_m + \frac{1}{R_3} & -\left(G_m + \frac{1}{R_3}\right) & 0 & 0 & 0 \\ -\frac{1}{R_3} & \frac{1}{R_3} + \frac{1}{R_4} & 0 & 0 & -1 \\ 0 & 0 & \frac{1}{R_8} & -\frac{1}{R_8} & 1 \\ 0 & 0 & -\frac{1}{R_8} & \frac{1}{R_8} & 0 \\ \hline 0 & -1 & 1 & 0 & 0 \\ E_7 & -E_7 & 0 & -1 & 0 \end{array} \right] \begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ i_6 \\ i_7 \end{bmatrix} = \begin{bmatrix} 0 \\ I_{s5} \\ 0 \\ 0 \\ V_{S6} \\ 0 \end{bmatrix}$$

From: A. Nardi

Pivoting for Accuracy

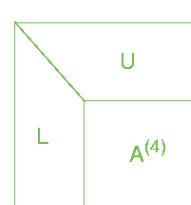
Example 1: After two steps of G.E. MNA matrix becomes:

$$\left[\begin{array}{cccc|cc} x & x & 0 & 0 & 0 & 0 \\ 0 & x & 0 & 0 & x & 0 \\ 0 & 0 & \frac{1}{R} & -\frac{1}{R} & 1 & 0 \\ 0 & 0 & -\frac{1}{R} & \frac{1}{R} & 0 & 1 \\ 0 & 0 & x & 0 & x & 0 \\ 0 & 0 & 0 & x & x & 0 \end{array} \right]$$



$$\left[\begin{array}{cccc|cc} x & x & 0 & 0 & 0 & 0 \\ 0 & x & 0 & 0 & x & 0 \\ 0 & 0 & \frac{1}{R} & -\frac{1}{R} & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & x & x & 0 \\ 0 & 0 & 0 & x & x & 0 \end{array} \right]$$

$$I_{14} = \frac{a_{14}^{(4)}}{a_{44}^{(4)}} = \frac{0}{0} = \infty !!!$$



From: A. Sangiovanni-Vincentelli

Solution:

Interchange rows and/or columns to bring non-zero element into position (k,k):

$$\begin{bmatrix} 0 & 1 & 1 \\ x & x & 0 \\ x & x & 0 \end{bmatrix} \xrightarrow{\text{swap rows}} \begin{bmatrix} x & x & 0 \\ 0 & 1 & 1 \\ x & x & 0 \end{bmatrix}$$

Example 2:

$$\begin{bmatrix} 1.25 \times 10^{-4} & 1.25 \\ 12.5 & 12.5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 6.25 \\ 75 \end{bmatrix}$$

solution to 5 digit accuracy

$$x_1 = 1.0001$$

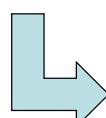
$$x_2 = 5.0000$$

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From: A. Sangiovanni-Vincentelli

Problem with Small Pivots

$$\begin{array}{l} \left[\begin{array}{cc|c} 1.25 \cdot 10^{-4} & 1.25 & 6.25 \\ 12.5 & 12.5 & 75 \end{array} \right] \\ \text{GE} \\ \left[\begin{array}{cc|c} 1.25 \cdot 10^{-4} & 1.25 & 6.25 \\ 0 & 12.5 - 1.25 \cdot 10^5 & 75 - 6.25 \cdot 10^5 \end{array} \right] \end{array}$$



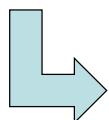
$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{5\text{ digits}} = \begin{bmatrix} 1.0001 \\ 4.9999 \end{bmatrix}$$

From: A. Nardi

Problem with Small Pivots

$$\begin{array}{l}
 \left[\begin{array}{cc} 1.25 \cdot 10^{-4} & 1.25 \\ 12.5 & 12.5 \end{array} \right] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 6.25 \\ 75 \end{bmatrix} \\
 \text{GE} \\
 \left[\begin{array}{cc} 1.25 \cdot 10^{-4} & 1.25 \\ 0 & 12.5 - 1.25 \cdot 10^5 \end{array} \right] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 6.25 \\ 75 - 6.25 \cdot 10^5 \end{bmatrix} \\
 \text{Rounded to 3 digits} \\
 \left[\begin{array}{cc} 1.25 \cdot 10^{-4} & 1.25 \\ 0 & -1.25 \cdot 10^5 \end{array} \right] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 6.25 \\ -6.25 \cdot 10^5 \end{bmatrix}
 \end{array}$$

From: A. Nardi

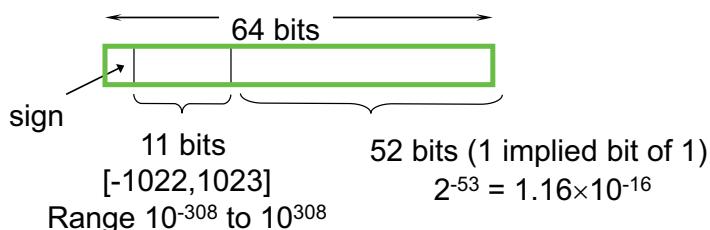


$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{3\text{ digits}} = \begin{bmatrix} 0 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{5\text{ digits}} = \begin{bmatrix} 1.0001 \\ 4.9999 \end{bmatrix}$$

Floating Point Arithmetic

Double precision number (IEEE Representation)



Machine precision: $\epsilon \approx 10^{-16}$

$\Rightarrow 1 + \epsilon = 1$

Avoid sum/subtraction of large and tiny numbers

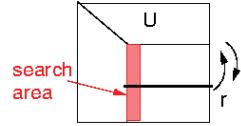
\Rightarrow Avoid big multipliers

\Rightarrow Do NOT use equalities in floating point arithmetic

Pivoting Strategies

1. Partial Pivoting: (row interchange only)

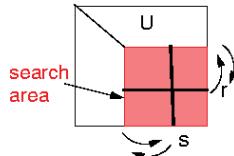
choose r as the smallest integer such that
 $|a_{rk}^{(k)}| = \max_{j=k, \dots, n} |a_{jk}^{(k)}|$



2. Complete Pivoting

(row and column interchange)

Choose r and s as the smallest integers such that
 $|a_{rk}^{(k)}| = \max_{\substack{i=k, \dots, n \\ j=k, \dots, n}} |a_{ij}^{(k)}|$



3. Threshold Pivoting

a. Apply partial pivoting only if $|a_{kk}^{(k)}| < \varepsilon_p |a_{rk}^{(k)}|$

b. Apply complete pivoting only if $|a_{kk}^{(k)}| < \varepsilon_p |a_{rs}^{(k)}|$

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From: A. Sangiovanni-Vincentelli

Partial Pivoting for Small Pivots

swap

$$\begin{bmatrix} 1.25 \cdot 10^{-4} & 1.25 \\ 12.5 & 12.5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 6.25 \\ 75 \end{bmatrix}$$

GE

$$\begin{bmatrix} 12.5 & 12.5 \\ 1.25 \cdot 10^{-4} & 1.25 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 75 \\ 6.25 \end{bmatrix}$$

Rounded to 3 digits

$$\begin{bmatrix} 12.5 & 12.5 \\ 0 & 1.25 - 12.5 \cdot 10^{-5} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 75 \\ 6.25 - 75 \cdot 10^{-5} \end{bmatrix}$$

$\xrightarrow{\hspace{1cm}}$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{3\text{digits}} = \begin{bmatrix} 1 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{5\text{digits}} = \begin{bmatrix} 1.0001 \\ 4.9999 \end{bmatrix}$$

From: A. Nardi