

# Techniques for circuit simulation

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**M. B. Patil**

[www.ee.iitb.ac.in/~sequel](http://www.ee.iitb.ac.in/~sequel)

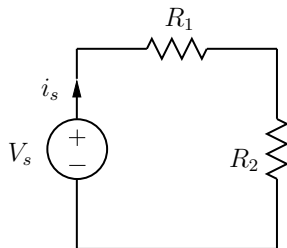
Department of Electrical Engineering  
Indian Institute of Technology Bombay

- \* Circuit simulation: introduction
- \* Nodal analysis
- \* Modified nodal analysis
- \* Sparse tableau approach
- \* Nonlinear circuits
- \* Transient (dynamic) analysis

# Circuit simulation

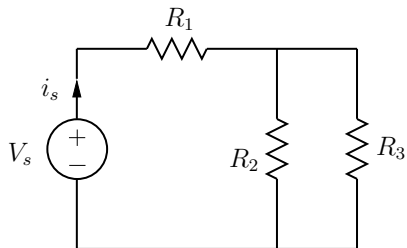
- \* DC analysis
- \* transient (time-domain) analysis
- \* AC (frequency-domain) analysis
- \* logic-level simulation
- \* mixed-signal simulation
- \* noise computation
- \* periodic steady state computation
- \* sensitivity analysis

# Why do we need circuit simulation?



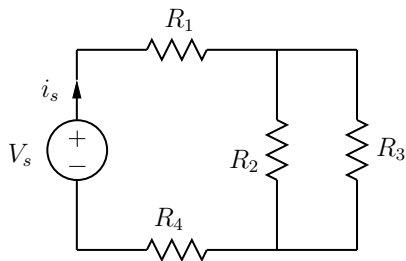
## Example 1

# Why do we need circuit simulation?



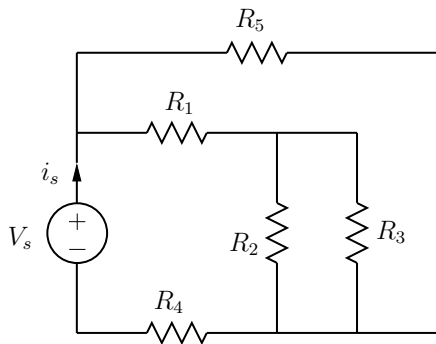
Example 1

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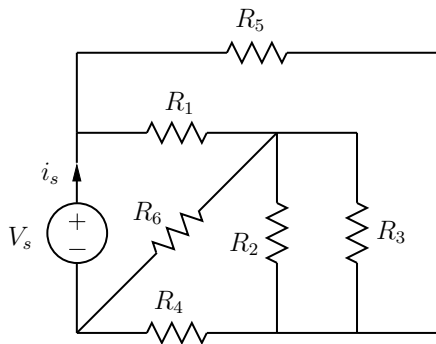
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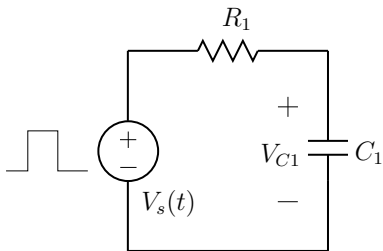
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Example 1

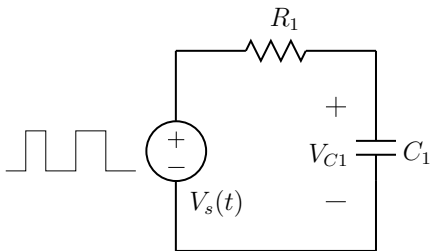


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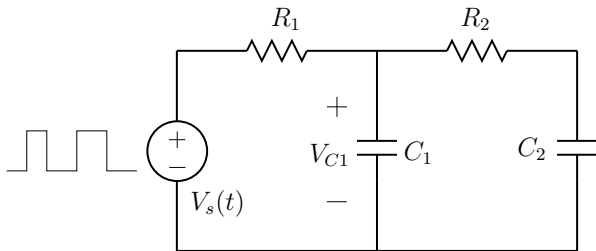
## Example 2

# Why do we need circuit simulation?



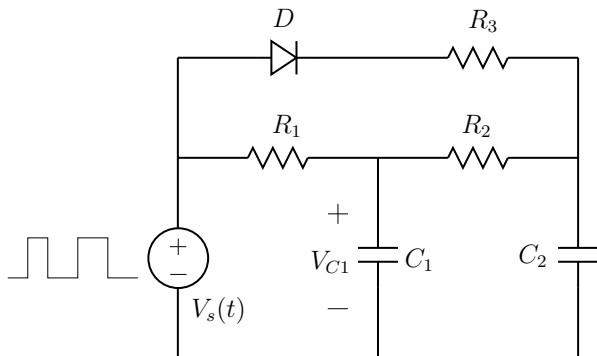
## Example 2

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Example 2

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# Circuit simulation on a computer

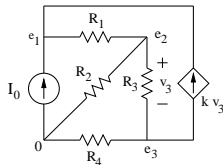
- \* Must be efficient in terms of CPU time (especially for large circuits).
- \* Must make good use of the memory available. If a matrix is sparse, it should not be stored in the  $a(i,j)$  form.
- \* The approach must be systematic. “Tricks” such as resistors in series or parallel, star-to-delta conversion, etc. will work in special cases. What we need is a *general-purpose* method that will work for *all* circuits.

- \* Circuit simulation: introduction
- \* **Nodal analysis**
- \* Modified nodal analysis
- \* Sparse tableau approach
- \* Nonlinear circuits
- \* Transient (dynamic) analysis



# Nodal Analysis of a linear circuit

- \* Take some node as the “reference node” and denote the node voltages of the remaining nodes by  $e_1$ ,  $e_2$ , etc.

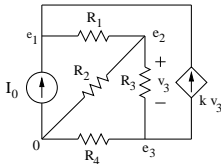


SPICE file

```
I0 1 0 1m
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R2 2 0 1.2k
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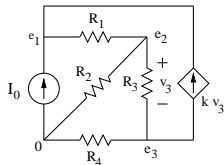


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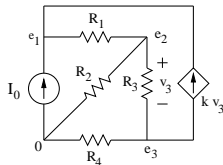


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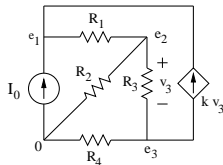


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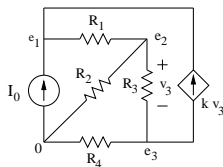


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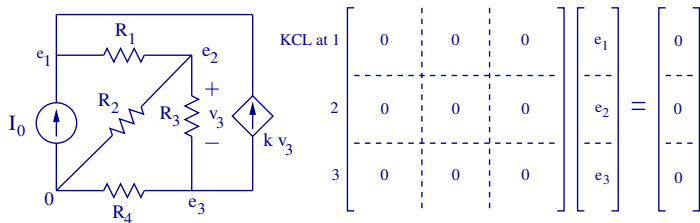
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- \* The equation assembly (also called “parsing”) can be done element-by-element, i.e., by considering one line of the circuit file at a time.
- \* The computer cannot see the entire circuit; it can, however, go through the circuit file line by line.



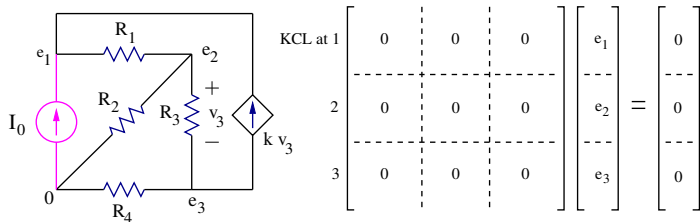
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# Step 1: Initialize

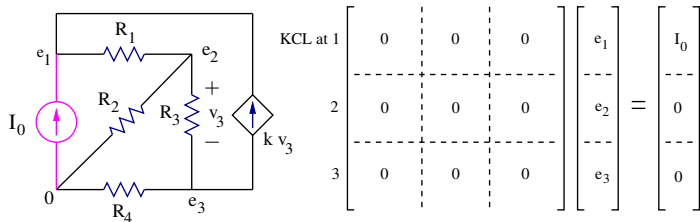


## Step 2

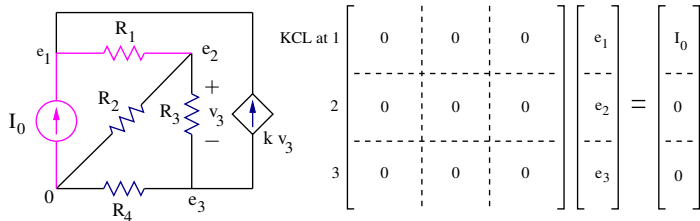




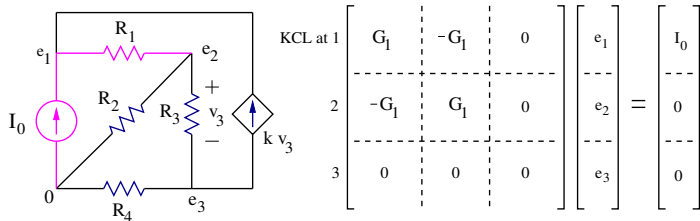
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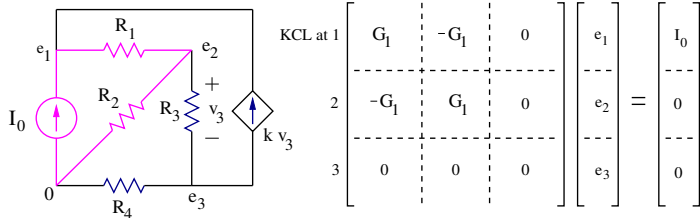
# Step 3



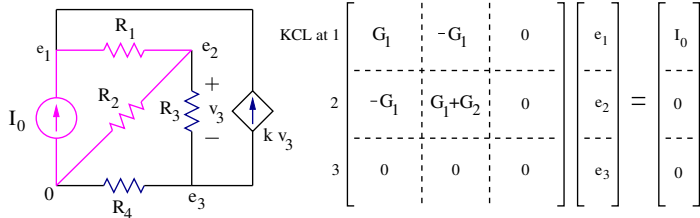
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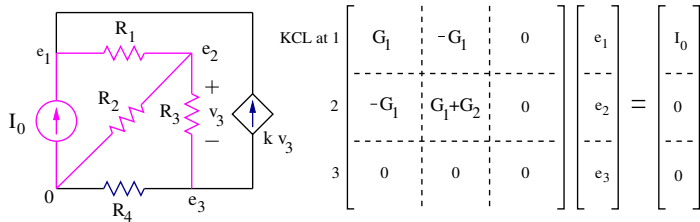
# Step 4



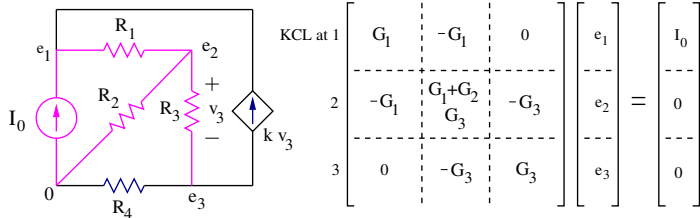
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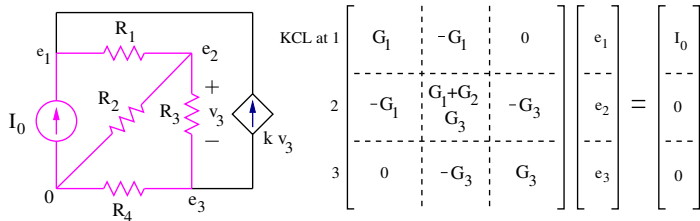
# Step 5



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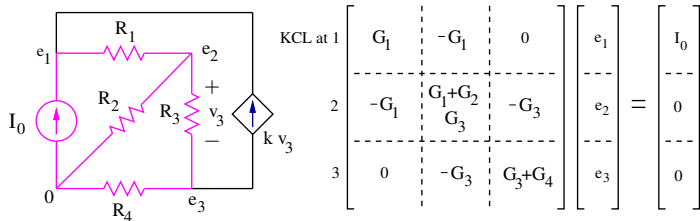


# Step 6

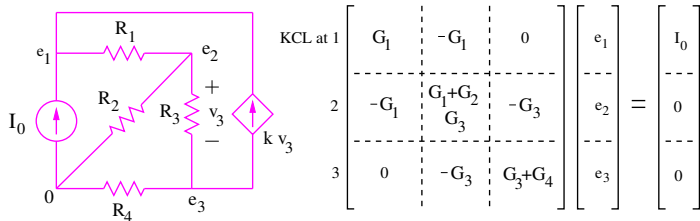




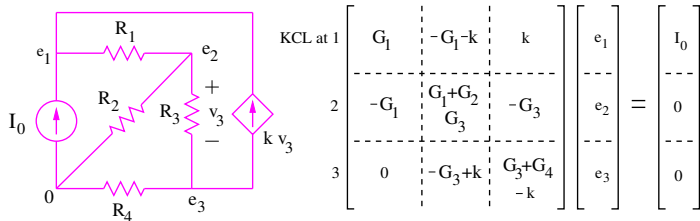
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# Step 7

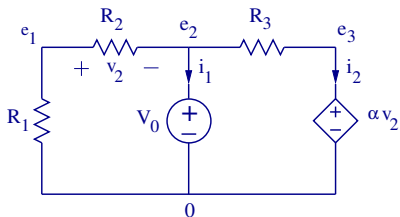


# Step 7



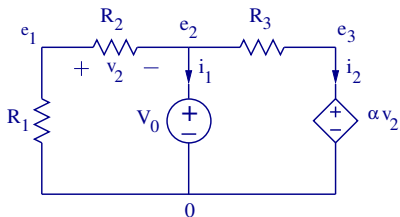
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# Modified Nodal Analysis (MNA) of a linear circuit



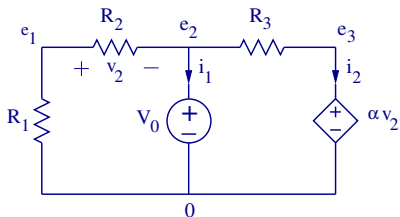
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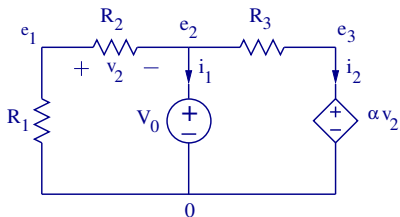
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- \* We also need to get an additional equation since the number of unknowns has gone up by 1. This equation is provided by the branch equation of the voltage source.

# Modified Nodal Analysis (MNA) of a linear circuit

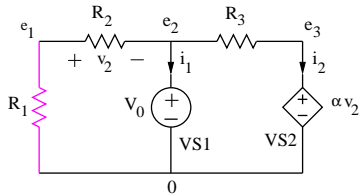


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- \* Treat the current through the voltage source as an additional unknown.
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- \* The “solution vector” now contains the voltage source currents in addition to the node voltages.



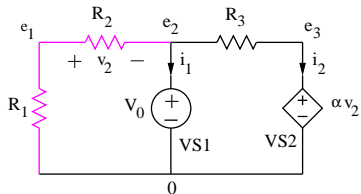


# Modified Nodal Analysis of a linear circuit



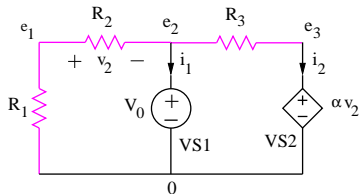
$$\begin{array}{l}
 \text{KCL at 1} \\
 \text{KCL at 2} \\
 \text{KCL at 3} \\
 \text{BCE for VS1} \\
 \text{BCE for VS2}
 \end{array}
 \begin{bmatrix}
 G_1 & & & & \\
 & & & & \\
 & & & & \\
 & & & & \\
 & & & & 
 \end{bmatrix}
 \begin{bmatrix}
 e_1 \\
 e_2 \\
 e_3 \\
 i_1 \\
 i_2
 \end{bmatrix}
 =
 \begin{bmatrix}
 \\
 \\
 \\
 \\
 \end{bmatrix}$$

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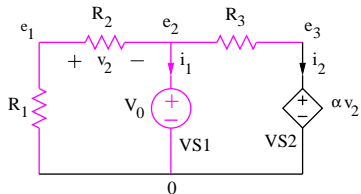
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 \begin{bmatrix}
 G_1+G_2 & -G_2 & & & \\
 -G_2 & G_2 & & & \\
 & & & & \\
 & & & & \\
 & & & & 
 \end{bmatrix}
 \begin{bmatrix}
 e_1 \\
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 \\
 \\
 \\
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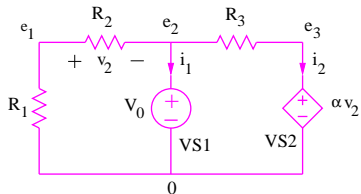
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 & -G_3 & G_3 & & \\
 & 1 & & & \\
 & & & & 
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 e_1 \\
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 \\
 \\
 \\
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 \\
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 \\
 \\
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# Sparse Tableau Analysis (STA)

- \* Variables: node voltages, branch currents, and branch voltages



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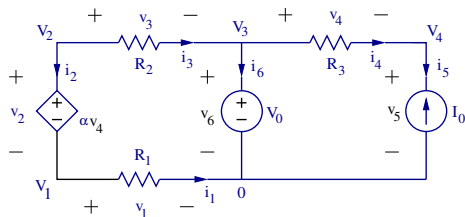
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# Sparse Tableau Analysis (STA)

- \* Variables: node voltages, branch currents, and branch voltages
- \* No need for special treatment of voltage sources or any other elements
- \* Circuit topology and element equations are decoupled.
- \* Easier to implement as compared to MNA



# STA example

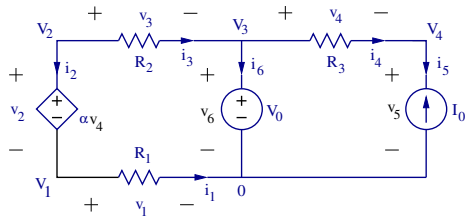


KCL

$$\left[ \begin{array}{cccccc|cccccc}
 1 & -1 & & & & & & & & & & \\
 1 & 1 & & & & & & & & & & \\
 & -1 & 1 & & & & & & & & & 0 \\
 & & -1 & 1 & & & & & & & & \\
 & & & & 1 & & & & & & & -1 \\
 & & & & & 1 & & & & & & 1 & -1 \\
 & & 0 & & & & 1 & & & & & -1 & 1 \\
 & & & & & & & 1 & & & & & -1 \\
 -R_1 & & & & & 1 & & & & & & & \\
 & & -R_2 & & & & 1 & & -\alpha & & & & \\
 & & & -R_3 & & & & 1 & & & & & 0 \\
 & & & & 1 & & & & & & & & \\
 & & & & & & & & & & & 1 & \\
 \end{array} \right] \begin{bmatrix} i_1 \\ i_2 \\ i_3 \\ i_4 \\ i_5 \\ i_6 \\ v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \\ v_6 \\ V_1 \\ V_2 \\ V_3 \\ V_4 \end{bmatrix} = \begin{bmatrix} \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ -I_0 \\ \\ V_0 \\ \end{bmatrix}$$



# STA example



KCL	$\begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 1 & 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & -1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & -1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & -1 & 0 & -1 & 0 \\ \hline -R_1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ -R_2 & 0 & 0 & 0 & 1 & 1 & -\alpha & 0 & 0 & 0 \\ -R_3 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ \hline \end{bmatrix}$	$\begin{bmatrix} i_1 \\ i_2 \\ i_3 \\ i_4 \\ i_5 \\ i_6 \\ v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \\ v_6 \\ V_1 \\ V_2 \\ V_3 \\ V_4 \\ \hline -I_0 \\ V_0 \end{bmatrix}$	$=$	$\begin{bmatrix} \dots \\ \dots \end{bmatrix}$
Branch Voltage Definition				
Branch Equations				

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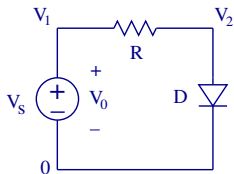
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- \* STA is generally slower than MNA, but this is not a concern for relatively small problems (including many problems in power electronics).
- \* Historically, STA was the first systematic approach used for circuit simulation (ASTAP by IBM). SPICE, based on MNA, was developed subsequently at UC Berkeley.

# Comparison of MNA and STA

- \* STA matrix is larger, but more sparse.
- \* If  $\mathbf{A}$  is an  $N \times N$  matrix, the CPU time to solve  $\mathbf{Ax} = \mathbf{b}$  is proportional to  $N^\alpha$ , where  $\alpha$  is 3 for a dense matrix and typically 1.5 to 2 for a sparse matrix.
- \* STA is generally slower than MNA, but this is not a concern for relatively small problems (including many problems in power electronics).
- \* Historically, STA was the first systematic approach used for circuit simulation (ASTAP by IBM). SPICE, based on MNA, was developed subsequently at UC Berkeley.
- \* Most of the circuit simulation programs available today are based on MNA, and many of them make use of SPICE.

- \* Circuit simulation: introduction
- \* Nodal analysis
- \* Modified nodal analysis
- \* Sparse tableau approach
- \* **Nonlinear circuits**
- \* Transient (dynamic) analysis

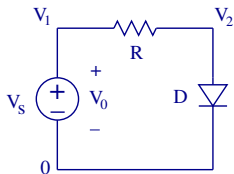
# Nonlinear circuits: Newton-Raphson method



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<sup>1</sup>Note that a circuit simulator such as SPICE will use a combination of MNA and N-R to solve this problem. Here, we will reduce it to the form  $f(x) = 0$  for simplicity.

## Nonlinear circuits: Newton-Raphson method

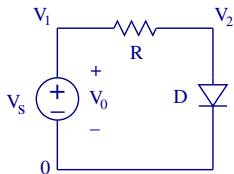


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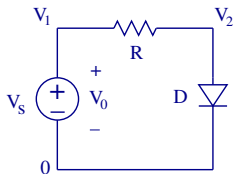
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# Nonlinear circuits: Newton-Raphson method



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Rewrite<sup>1</sup> as  $f(V_2) = 0$ . In general, consider  $f(x) = 0$ . Expand around an initial guess  $x_0$ .

$$f(x_0 + \Delta x) = f(x_0) + \Delta x f'(x_0) + \dots$$

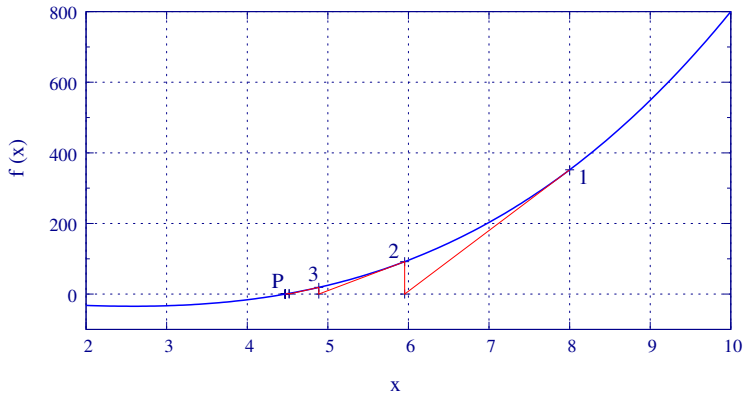
We want  $\Delta x$  such that  $f(x_0 + \Delta x) = 0$ .

$$\Delta x = -\frac{f(x_0)}{f'(x_0)}$$

---

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# Newton-Raphson method: graphical interpretation of $\Delta x = -\frac{f(x_0)}{f'(x_0)}$



Solution of  $x^3 - 20x = 0$ , with  $x = 8$  as the initial guess.

## Newton-Raphson method: convergence

$i$	$x^{(i)}$	$f(x^{(i)})$	$\Delta x^{(i)}$
1	$0.800000 \times 10^1$	$0.352 \times 10^3$	$-0.204 \times 10^1$
2	$0.595349 \times 10^1$	$0.919 \times 10^2$	$-0.106 \times 10^1$
3	$0.488846 \times 10^1$	$0.190 \times 10^2$	-0.368
4	$0.451992 \times 10^1$	$0.194 \times 10^1$	$-0.470 \times 10^{-1}$
5	$0.447288 \times 10^1$	$0.298 \times 10^{-1}$	$-0.746 \times 10^{-3}$
6	$0.447214 \times 10^1$	$0.748 \times 10^{-5}$	$-0.187 \times 10^{-6}$
7	$0.447214 \times 10^1$	$0.470 \times 10^{-12}$	$-0.117 \times 10^{-13}$

Solution of  $f(x) = x^3 - 20x = 0$ , with  $x = 8$  as the initial guess.

# Convergence of Newton-Raphson method

Consider solving  $f(x) = 0$  with the N-R method. Define

$$g(x) = x - \frac{f(x)}{f'(x)}. \quad (1)$$

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Application of Taylor's theorem to Eq. 1 yields,

$$g(x) = g(r) + g'(r)(x - r) + \frac{g''(\xi)}{2}(x - r)^2, \quad (3)$$

where  $\xi$  lies between  $x$  and  $r$ .

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The derivative  $g'(x)$  can be obtained from Eq. 1 as,

$$g'(x) = 1 - \frac{[f'(x)]^2 - f(x)f''(x)}{[f'(x)]^2}. \quad (4)$$

# Convergence of Newton-Raphson method

Since  $f(r) = 0$ , we get  $g(r) = r$  from Eq. 1 and  $g'(r) = 0$  from Eq. 4. Substituting for  $g(r)$  and  $g'(r)$  in Eq. 3, we get,

$$g(x) = r + \frac{g''(\xi)}{2}(x - r)^2 . \quad (5)$$



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Replace  $x$  by  $x^{(n)}$  and use the fact that  $g(x^{(n)})$  is the same as  $x^{(n+1)}$  in the N-R procedure, to get

$$\left(x^{(n+1)} - r\right) = \frac{g''(\xi)}{2} \left(x^{(n)} - r\right)^2 . \quad (6)$$

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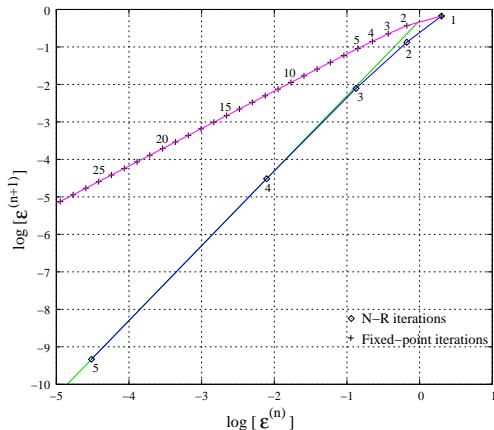
$$(x^{(n+1)} - r) = \frac{g''(\xi)}{2} (x^{(n)} - r)^2. \quad (6)$$

As  $x^{(n)}$  converges to  $r$ , so does  $\xi$ ; and we can replace  $g''(\xi)$  by  $g''(r)$ , a constant. Further, if we define  $\epsilon^{(n)} \equiv x^{(n)} - r$  (the “error” at the  $n^{\text{th}}$  N-R iteration), we can write Eq. 6 as

$$\epsilon^{(n+1)} = k [\epsilon^{(n)}]^2, \quad (7)$$

where  $k = g''(r)/2$ . Eq. 7 describes the well-known feature of “quadratic convergence” of the N-R method, i.e., the error goes down quadratically as  $x^{(n)} \rightarrow r$ .

# Convergence of Newton-Raphson method



$\log(\epsilon^{(n+1)})$  versus  $\log(\epsilon^{(n)})$  with the N-R scheme and the fixed-point iteration method for  $f(x) = x^2 - 6x + 8 = 0$ , with  $x = 0$  as the initial guess. The green line represents

$\epsilon^{(n+1)} = \frac{g''(r)}{2}(\epsilon^{(n)})^2$ . The iteration numbers are also shown for each scheme. Note the quadratic convergence of the N-R method. (Both schemes were found to converge to  $r = 2$  for the specified initial guess.)

# Newton-Raphson method for $N$ equations

Consider a system of  $N$  ODEs:

$$\begin{aligned}f_1(x_1, x_2, \dots, x_N) &= 0, \\f_2(x_1, x_2, \dots, x_N) &= 0, \\&\dots, \\f_N(x_1, x_2, \dots, x_N) &= 0.\end{aligned}$$

$$\mathbf{f} = \begin{bmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \\ \dots \\ f_N(\mathbf{x}) \end{bmatrix},$$

The correction vector  $\Delta \mathbf{x}$  can be obtained by solving

$$\mathbf{J}^{(i)} \Delta \mathbf{x}^{(i)} = -\mathbf{f}^{(i)},$$

where  $i$  is the iteration number,  $\mathbf{J}$  is the Jacobian matrix, and  $\mathbf{f}$  is the function vector.

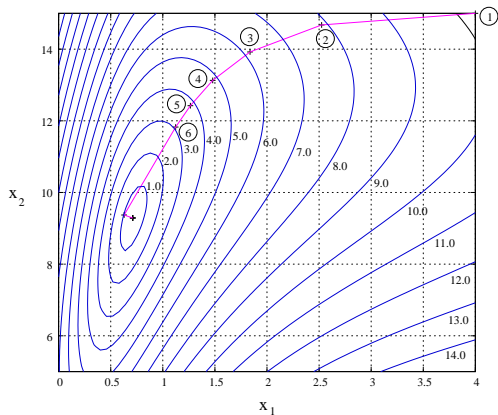
$$\mathbf{J} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_N} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_N} \\ \dots & \dots & \dots & \dots \\ \frac{\partial f_N}{\partial x_1} & \frac{\partial f_N}{\partial x_2} & \dots & \frac{\partial f_N}{\partial x_N} \end{bmatrix}.$$

# N-R method: example with two variables

$i$	$x_1^{(i)}$	$x_2^{(i)}$	$\ f\ _2$	$\Delta x_1^{(i)}$	$\Delta x_2^{(i)}$
1	$0.40000 \times 10^1$	$0.15000 \times 10^2$	$0.10241 \times 10^2$	$-0.73776 \times 10^1$	$-0.16223 \times 10^1$
2	$0.25244 \times 10^1$	$0.14675 \times 10^2$	$0.78909 \times 10^1$	$-0.34368 \times 10^1$	$-0.37631 \times 10^1$
3	$0.18371 \times 10^1$	$0.13922 \times 10^2$	$0.61523 \times 10^1$	$-0.17887 \times 10^1$	$-0.39712 \times 10^1$
4	$0.14793 \times 10^1$	$0.13128 \times 10^2$	$0.48512 \times 10^1$	$-0.10737 \times 10^1$	$-0.35342 \times 10^1$
5	$0.12646 \times 10^1$	$0.12421 \times 10^2$	$0.38481 \times 10^1$	$-0.70747$	$-0.29789 \times 10^1$
6	$0.11231 \times 10^1$	$0.11826 \times 10^2$	$0.30620 \times 10^1$	$-0.49427$	$-0.24548 \times 10^1$
7	0.62883	$0.93711 \times 10^1$	0.95091	$0.80932 \times 10^{-1}$	$-0.80932 \times 10^{-1}$
8	0.70976	$0.92902 \times 10^1$	$0.31487 \times 10^{-1}$	$0.28690 \times 10^{-2}$	$-0.28690 \times 10^{-2}$
9	0.71263	$0.92873 \times 10^1$	$0.38735 \times 10^{-4}$	$0.35381 \times 10^{-5}$	$-0.35381 \times 10^{-5}$
10	0.71263	$0.92873 \times 10^1$	$0.58855 \times 10^{-10}$	$0.53759 \times 10^{-11}$	$-0.53753 \times 10^{-11}$

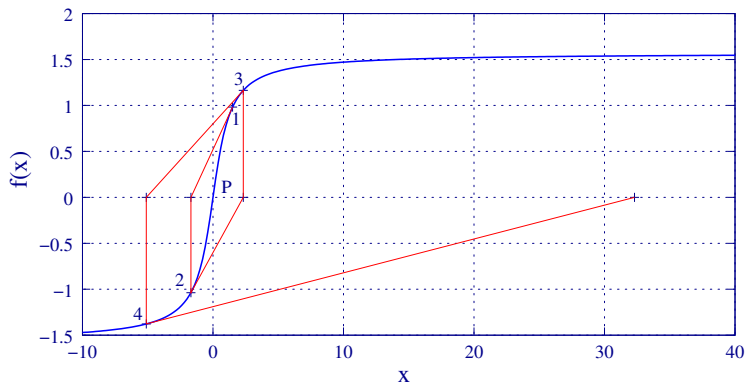
Application of the N-R method to a system of two equations, with  $f_1 \equiv x_1 + x_2 - 10 = 0$ , and  $f_2 \equiv x_2 - 15 \tan^{-1}(x_1) = 0$ . (damping was used for the first 5 iterations.)

## N-R method: example with two variables



Application of the N-R method to a system of two equations, with  $f_1 \equiv x_1 + x_2 - 10 = 0$ , and  $f_2 \equiv x_2 - 15 \tan^{-1}(x_1) = 0$ . The contours are labelled by the 2-norm,  $\|f\|_2$ . Circled integers represent the iteration numbers. (damping was used for the first 5 iterations.)

## Newton-Raphson method: convergence issues



Application of the N-R method to  $f(x) = \tan^{-1} x = 0$ , with  $x = 1.5$  as the initial guess.

Instead of

$$x^{(n+1)} = x^{(n)} + \Delta x^{(n)},$$

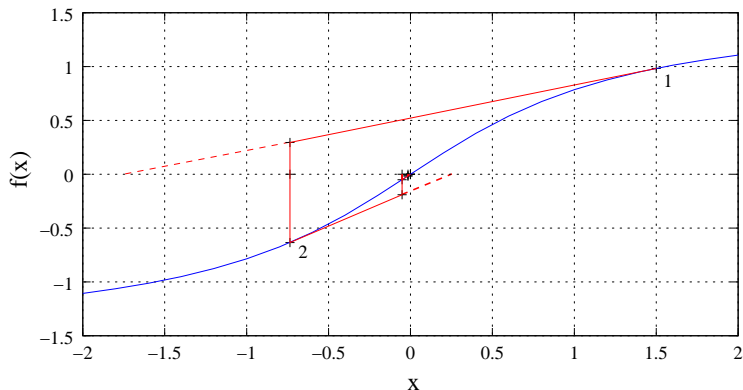
as in the standard N-R algorithm, we use

$$\begin{aligned}x^{(n+1)} &= x^{(n)} + k \Delta x^{(n)} \\ &= x^{(n)} + k \left\{ -[f'(x^{(n)})]^{-1} f(x^{(n)}) \right\},\end{aligned}$$

where  $k (< 1)$  is the “damping factor.”

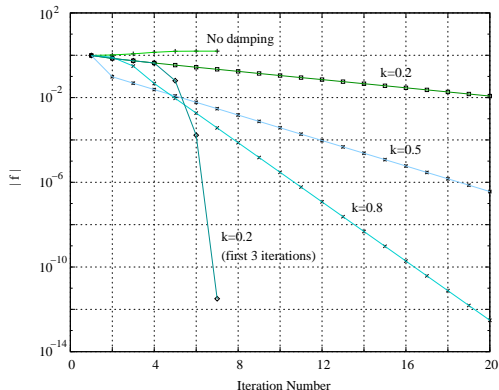


## Newton-Raphson method: use of damping



Application of the N-R method to  $f(x) = \tan^{-1} x = 0$ , with  $x = 1.5$  as the initial guess and a damping factor  $k = 0.7$ .

# Newton-Raphson method: use of damping



Application of the N-R method to  $f(x) = \tan^{-1} x = 0$ , with  $x = 1.5$  as the initial guess and different damping factors. (For the case with no damping, N-R iterations stopped due to  $\frac{df}{dx}$  becoming too small.)

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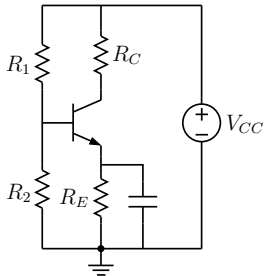
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- \* Damping is very useful in power electronic circuits since they are highly non-linear (due to switches).
- \* For transient simulation, in addition to damping, reducing the time step may also help in convergence.

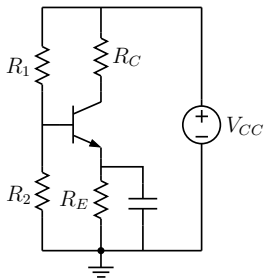
# Convergence of N-R iterations



- \* We are interested in obtaining the DC (“bias”) solution for a circuit with highly non-linear elements (e.g., BJTs).

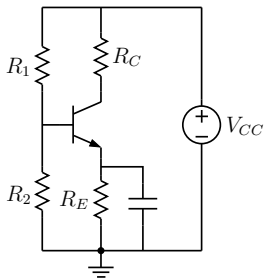


# Convergence of N-R iterations



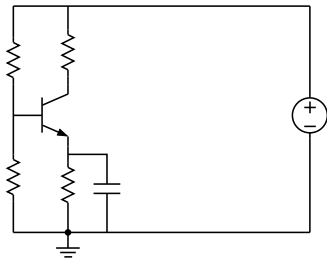
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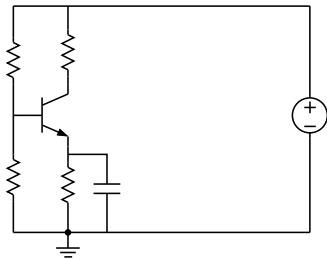
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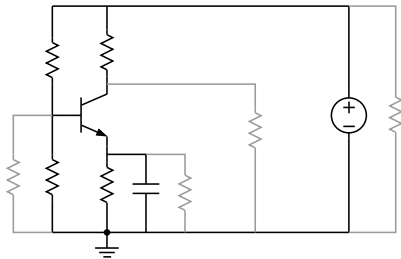
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- \* Two tricks: (a)  $g_{\min}$  stepping, (b)  $V_{CC}$  stepping.

# $g_{\min}$ stepping

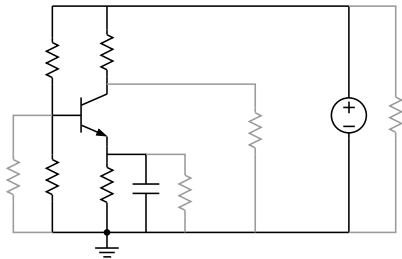




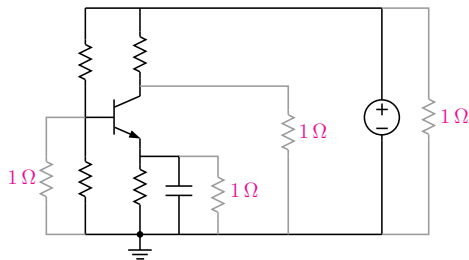
\* Connect  $R = 1/g$  between each node and ground.



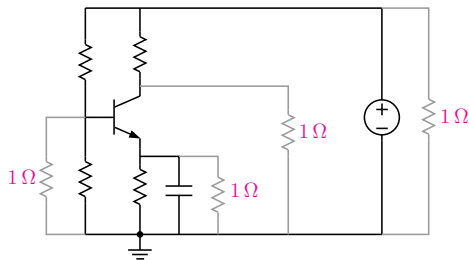
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- \* Assign a small value (say,  $1\ \Omega$ ) to each resistance, i.e., a large value to  $g$  ( $1\ \text{U}$ ).



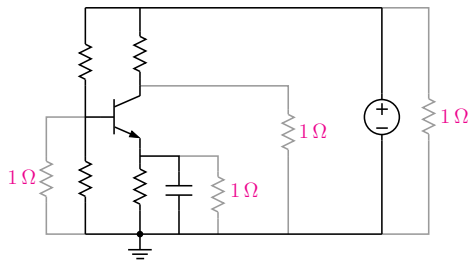
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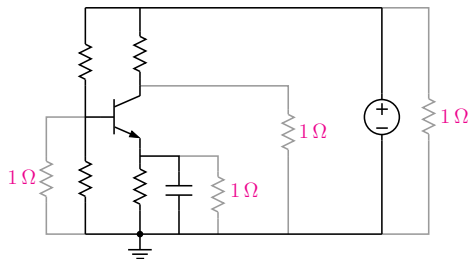


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→ easy convergence since the non-linear elements got bypassed.



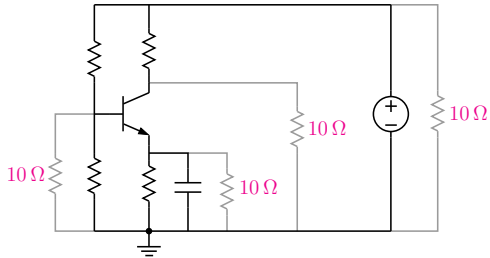
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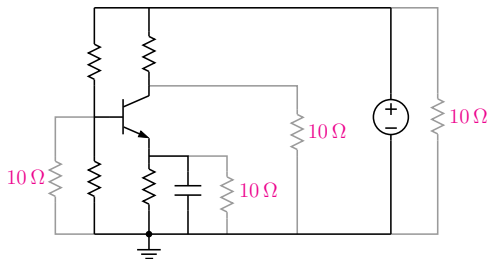


\* Increase  $R$  from, say,  $1\ \Omega$  to  $10\ \Omega$ , i.e., decrease  $g$  from  $1\ \text{U}$  to  $0.1\ \text{U}$ .

# $g_{\min}$ stepping

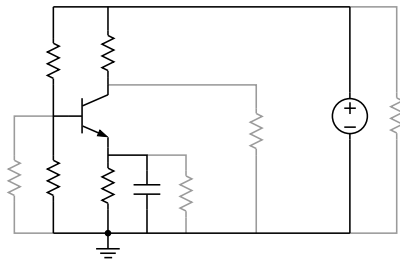


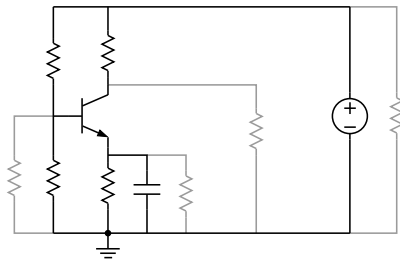
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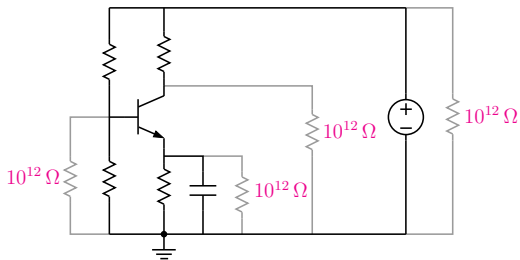
- \* Increase  $R$  from, say,  $1\ \Omega$  to  $10\ \Omega$ , i.e., decrease  $g$  from  $1\ \text{U}$  to  $0.1\ \text{U}$ .
- \* Convergence is easy since the previous solution serves as a good initial guess.

# $g_{\min}$ stepping

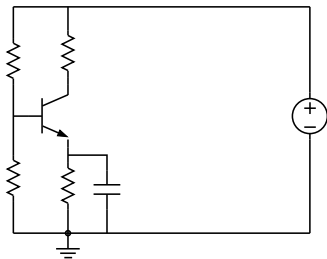




\* Keep increasing  $R$  (i.e., decreasing  $g$ ) and solve every time.

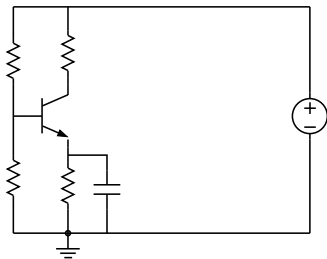


- \* Keep increasing  $R$  (i.e., decreasing  $g$ ) and solve every time.
- \* When  $g = 10^{-12} \text{ S}$ , for example,  $R = 10^{12} \Omega$ , which is as good as an open circuit.



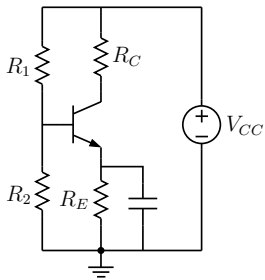
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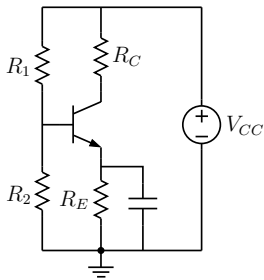
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- \* We have now got the DC solution for the original circuit.

# Voltage supply stepping



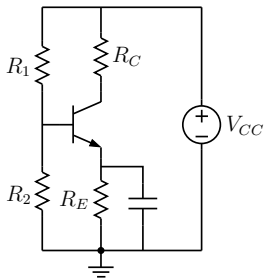
- \* When  $V_{CC} = 0\text{V}$ , the zero initial solution (all node voltages equal to  $0\text{V}$ ) is valid.

# Voltage supply stepping



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- \* Treating that as the initial guess, solve for a small value of  $V_{CC}$  (say,  $0.1\text{V}$ ). The N-R iterations are likely to converge since  $V_{CC} = 0.1\text{V}$  is a small change from  $V_{CC} = 0\text{V}$ .

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- \* Repeat.  $V_{CC} : 0\text{V} \rightarrow 0.1\text{V} \rightarrow 0.2\text{V} \rightarrow \dots \rightarrow 5\text{V}$

## N-R method: effect of scaling and precision

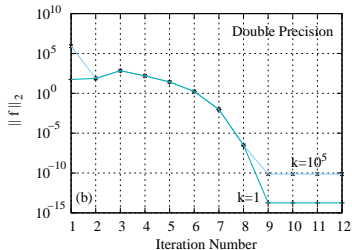
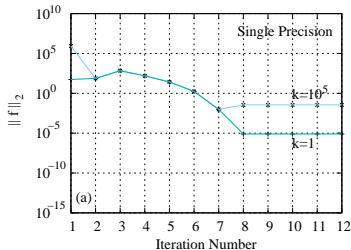
Consider the system of equations,

$$\begin{aligned}f_1(x_1, x_2) &\equiv k(x_1 + x_2 - 6\sqrt{3}) = 0, \\f_2(x_1, x_2) &\equiv 10x_1^2 - x_2^2 + 45 = 0.\end{aligned}\tag{8}$$

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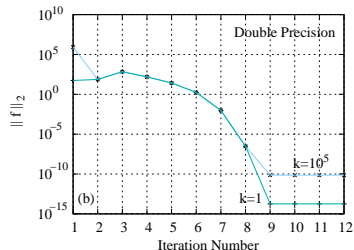
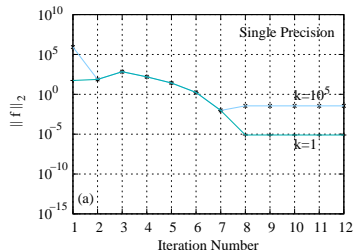


$\|f\|_2$  versus N-R iteration number for Eq. 8, with  $x_1 = x_2 = 1$  as the initial guess, (a) Single precision arithmetic, (b) Double precision arithmetic.

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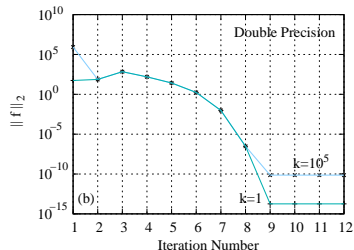
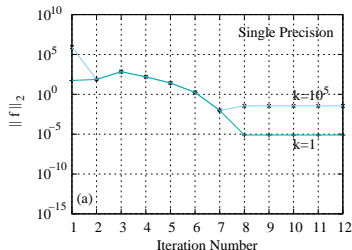
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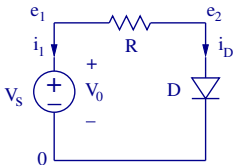


$\|f\|_2$  versus N-R iteration number for Eq. 8, with  $x_1 = x_2 = 1$  as the initial guess, (a) Single precision arithmetic, (b) Double precision arithmetic.

- \* If  $k$  is made larger, the norm saturates at a higher value.
- \* Precision has a significant effect on the lowest achievable norm.



# Non-linear circuit analysis



MNA equations:

$$i_1 + G(e_1 - e_2) = 0,$$

$$G(e_2 - e_1) + i_D(e_2) = 0,$$

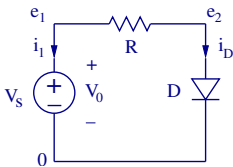
$$e_1 = V_0,$$

where

$$i_D(e_2) = I_{s0} [\exp(e_2/V_T) - 1].$$

- \* The circuit equations can be assembled using the MNA or STA approach.

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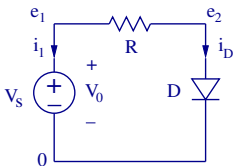
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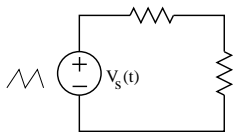
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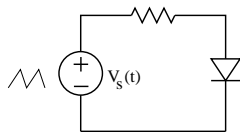
- \* The circuit equations can be assembled using the MNA or STA approach.
- \* Since the equations are non-linear, the N-R method is used to solve them.
- \* More expensive than a linear circuit of the same size, since several (typically 3 to 5) N-R iterations are involved, each requiring the solution of  $\mathbf{J}\Delta\mathbf{x} = -\mathbf{f}$ .

- \* Circuit simulation: introduction
- \* Nodal analysis
- \* Modified nodal analysis
- \* Sparse tableau approach
- \* Nonlinear circuits
- \* Transient (dynamic) analysis

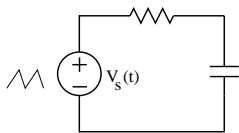
# Transient (dynamic) analysis



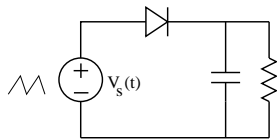
(a)



(b)

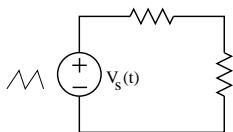


(c)

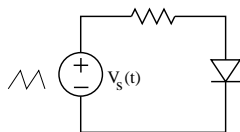


(d)

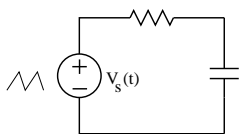
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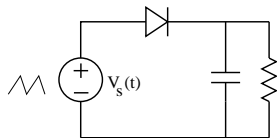
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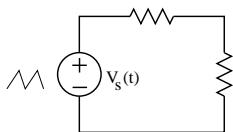
(c)



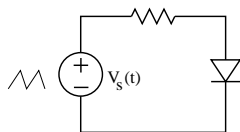
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- \* In (a) and (b), we can use the techniques seen earlier. At a given time  $t$ , we simply need to replace the source with a DC source with voltage =  $V_s(t)$ .

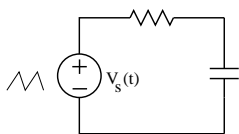
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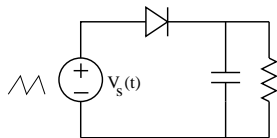
(a)



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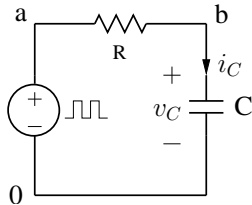
(c)



(d)

- \* In (a) and (b), we can use the techniques seen earlier. At a given time  $t$ , we simply need to replace the source with a DC source with voltage =  $V_s(t)$ .
- \* In (c) and (d), the situation is very different due to the presence of a capacitor which involves time derivatives.

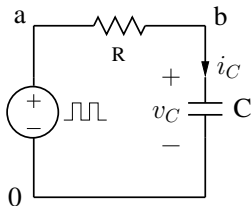
# Transient analysis



- \* The capacitor current,  $i_C = C \frac{dv_C}{dt}$ , cannot be written in terms of the instantaneous node voltages or branch voltages since its value depends on the past behaviour of  $v_C$ .

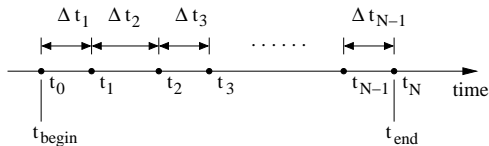


# Transient analysis



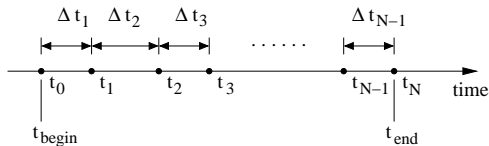
- \* The capacitor current,  $i_C = C \frac{dv_C}{dt}$ , cannot be written in terms of the instantaneous node voltages or branch voltages since its value depends on the past behaviour of  $v_C$ .
- \* We need some way of approximating the derivative in terms of the past behaviour of  $v_C$ .

# Discretization of time



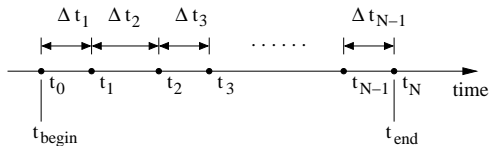
- \* Discretization of time is required since numerical solution can only be obtained at a finite number of points.

# Discretization of time



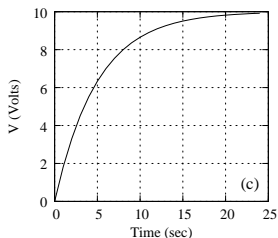
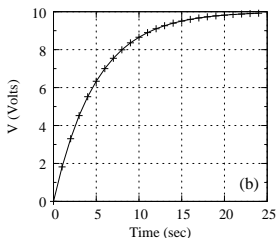
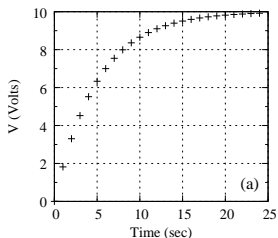
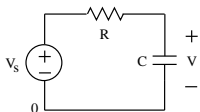
- \* Discretization of time is required since numerical solution can only be obtained at a finite number of points.
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# Discretization of time



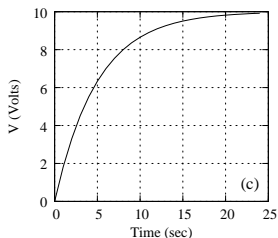
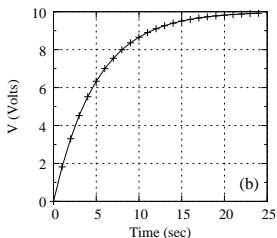
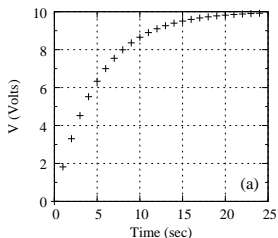
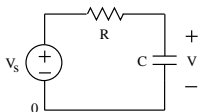
- \* Discretization of time is required since numerical solution can only be obtained at a finite number of points.
- \* The time steps ( $\Delta t_i$ ) may not be uniform.
- \* Generally, the time steps are computed *dynamically*, not *a priori*.

# Discretization of time



(a) Typical simulator output.

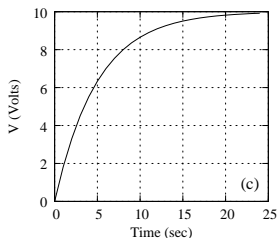
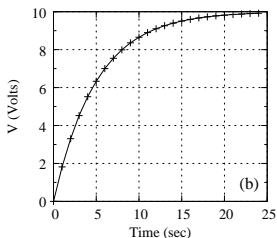
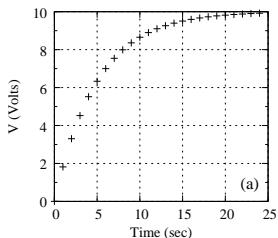
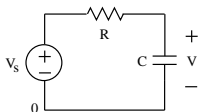
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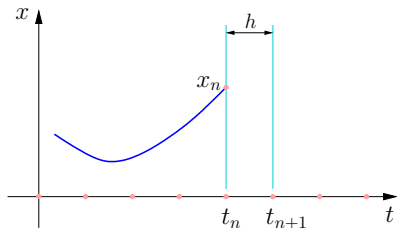
(b) After connecting the output points with line segments.

# Discretization of time



- (a) Typical simulator output.
- (b) After connecting the output points with line segments.
- (c) After removing the output points (but retaining the segments), the waveform looks continuous, but this is an illusion!

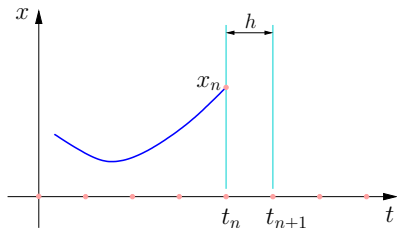
## Transient simulation: Forward Euler method



- \* Consider  $\frac{dx}{dt} = f(t, x)$ . We have the solution at  $t_n$  and want to obtain  $x(t_{n+1})$ .

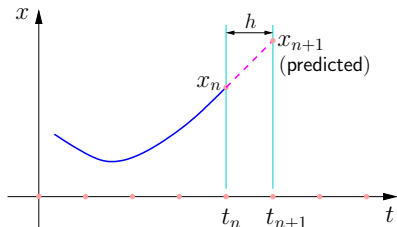


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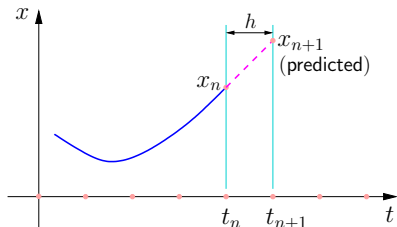
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- \* Compute the slope at  $t_n$ :  $\left. \frac{dx}{dt} \right|_{t=t_n} = f(t_n, x_n)$ .

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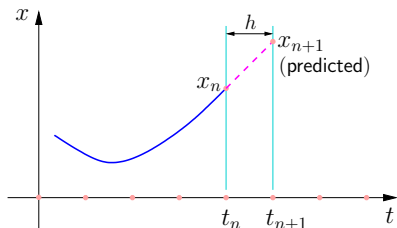
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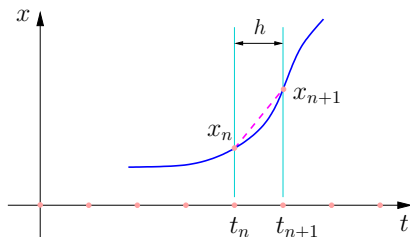
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- \*  $\frac{x_{n+1} - x_n}{t_{n+1} - t_n} \approx f(t_n, x_n)$

# Transient simulation: Forward Euler method



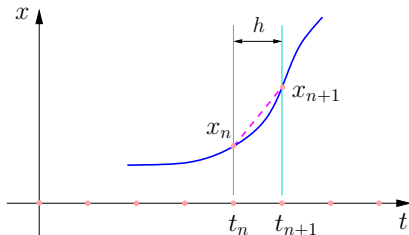
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- \*  $\frac{x_{n+1} - x_n}{t_{n+1} - t_n} \approx f(t_n, x_n) \rightarrow x_{n+1} = x_n + h f(t_n, x_n)$ .

# Transient analysis: a quick look



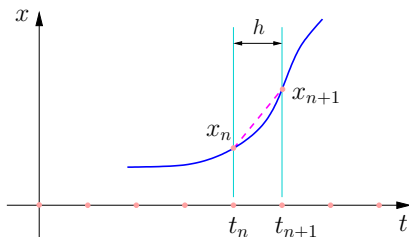
Method	Approximation for $\frac{dx}{dt} = f(t, x)$
Forward Euler	$\frac{x_{n+1} - x_n}{h} = f(t_n, x_n)$

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# Transient analysis: a quick look



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Backward Euler	$\frac{x_{n+1} - x_n}{h} = f(t_{n+1}, x_{n+1})$
Trapezoidal	$\frac{x_{n+1} - x_n}{h} = \frac{1}{2} [f(t_n, x_n) + f(t_{n+1}, x_{n+1})]$

## Application to $\dot{x} = -x$ , with $x(0) = 1$

$$FE : \quad \frac{x_{n+1} - x_n}{h} = f(t_n, x_n) = -x_n$$

$$BE : \quad \frac{x_{n+1} - x_n}{h} = f(t_{n+1}, x_{n+1}) = -x_{n+1}$$

$$TRZ : \quad \frac{x_{n+1} - x_n}{h} = \frac{1}{2} [f(t_n, x_n) + f(t_{n+1}, x_{n+1})] = -\frac{1}{2} (x_n + x_{n+1})$$

Simple manipulation yields the following approximations:

$$FE : \quad x_{n+1} = x_n (1 - h)$$

$$BE : \quad x_{n+1} = x_n \frac{1}{1 + h}$$

$$TRZ : \quad x_{n+1} = x_n \frac{1 - h/2}{1 + h/2}$$



## Application to $\dot{x} = -x$ , with $x(0) = 1$

The exact solution is  $\hat{x}(t) = e^{-t}$ . Expanding around  $t_n$ , we get,

$$\hat{x}_{n+1} = \hat{x}_n + h \frac{d\hat{x}}{dt} + \cdots = \hat{x}_n + h(-e^{-t_n}) + \cdots = \hat{x}_n(1 - h + h^2/2 - h^3/6 + \cdots).$$

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Compare with

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\* If  $h \ll 1$ , the three approximations are equivalent, as we would expect.

## Application to $\dot{x} = -x$ , with $x(0) = 1$

The exact solution is  $\hat{x}(t) = e^{-t}$ . Expanding around  $t_n$ , we get,

$$\hat{x}_{n+1} = \hat{x}_n + h \frac{d\hat{x}}{dt} + \dots = \hat{x}_n + h(-e^{-t_n}) + \dots = \hat{x}_n(1 - h + h^2/2 - h^3/6 + \dots).$$

Compare with

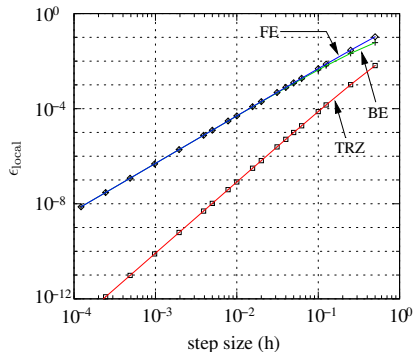
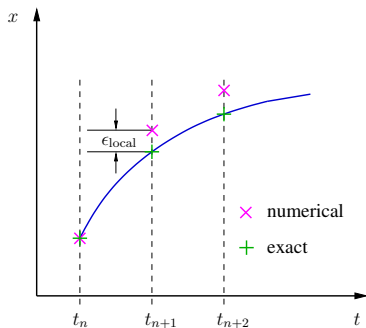
$$FE : \quad x_{n+1} = x_n(1 - h)$$

$$BE : \quad x_{n+1} = x_n \frac{1}{1 + h} = x_n(1 - h + h^2 + \dots)$$

$$TRZ : \quad x_{n+1} = x_n \frac{1 - h/2}{1 + h/2} = x_n(1 - h + h^2/2 - h^3/4 + \dots)$$

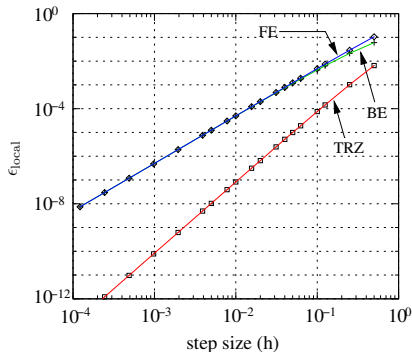
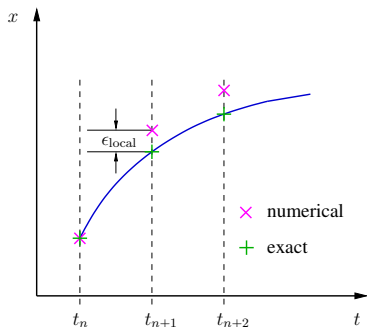
- \* If  $h \ll 1$ , the three approximations are equivalent, as we would expect.
- \* If the starting point  $x(t_n)$  is the same, the “error” (difference between the exact and numerical solutions) is  $O(h^2)$  for FE and BE, and  $O(h^3)$  for TRZ.

# Application to $\dot{x} = -x$ , with $x(0) = 1$



- \* The *local* error is the error made in a *single* step, assuming that the starting point is exact. In this case, starting from the exact value,  $x(0) = 1$ , the difference  $|x(h) - \hat{x}(h)|$  has been computed.

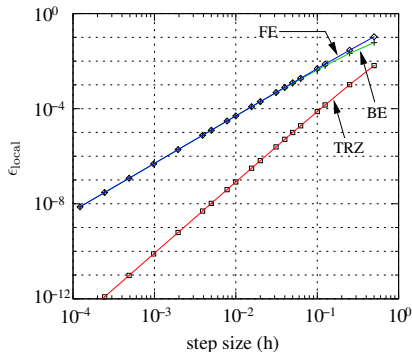
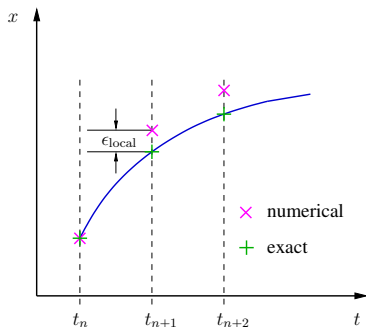
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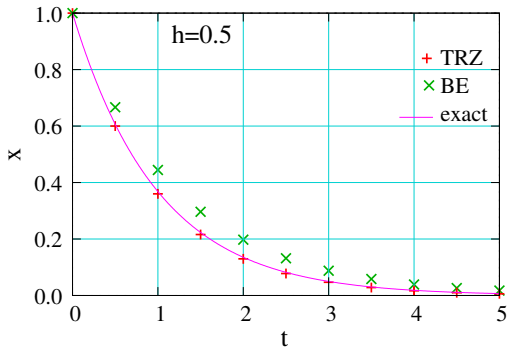


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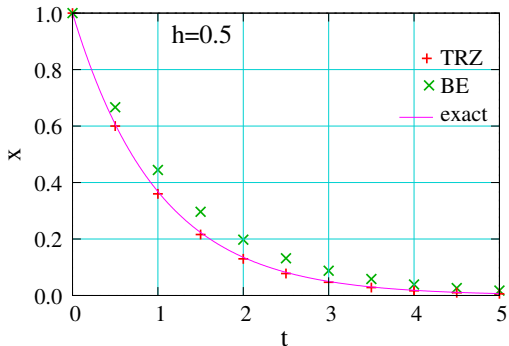


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- \* If  $h \rightarrow h/10$ , the error decreases by a factor of  $10^2$  for the FE and BE methods, and by  $10^3$  for the TRZ method.
- \* The TRZ method is therefore said to be more *accurate* than FE or BE.

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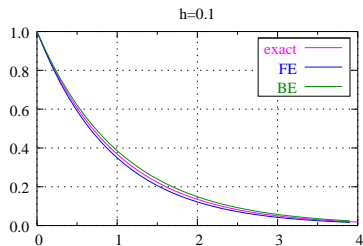


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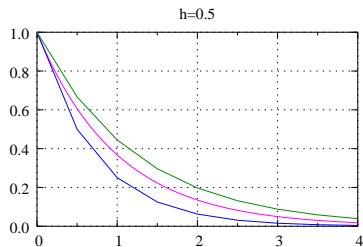
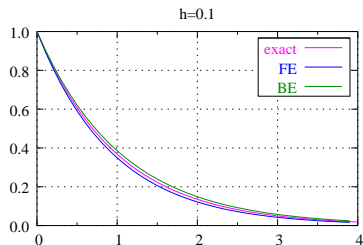


\* The higher accuracy of the TRZ method allows larger time steps.

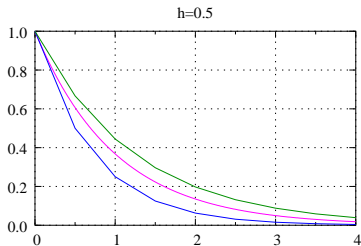
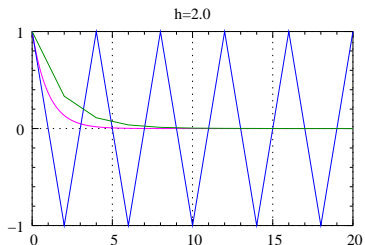
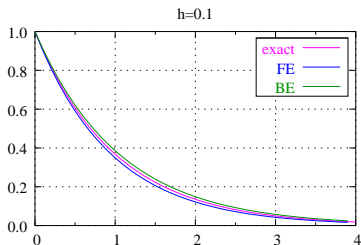
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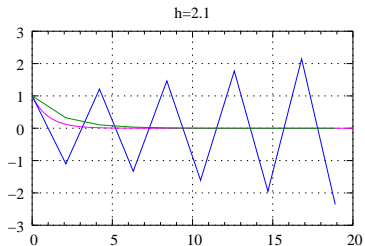
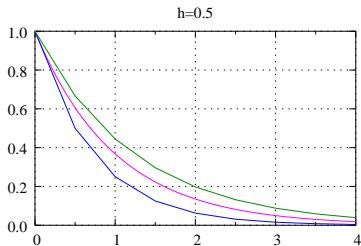
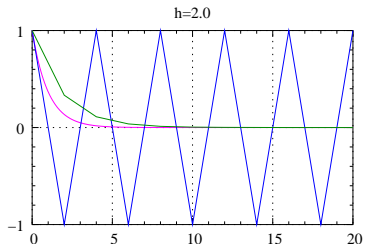
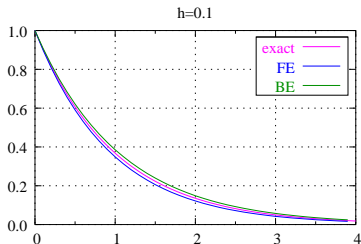
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- \* The issue of stability rules out many other methods as well.

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- \* However, the FE method is not useful because it can be unstable in some cases.



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- \* The stability constraints significantly reduce the choices available for circuit simulation. BE, Gear (order 2), and Trapezoidal methods are commonly used.

## Equivalent circuit for a capacitor

With Backward Euler method, we get

$$\frac{v_C^{n+1} - v_C^n}{h} = \frac{1}{C} i_C^{n+1}.$$

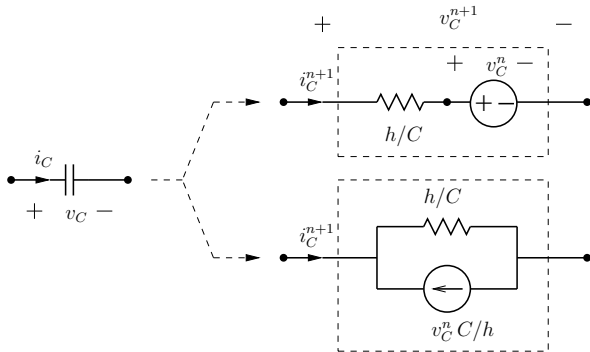
$$\text{i.e., } v_C^{n+1} = \frac{h}{C} i_C^{n+1} + v_C^n \quad \text{OR} \quad i_C^{n+1} = \frac{C}{h} v_C^{n+1} - \frac{C}{h} v_C^n.$$

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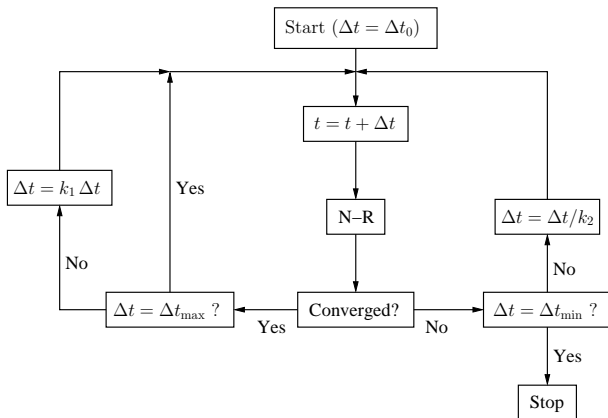
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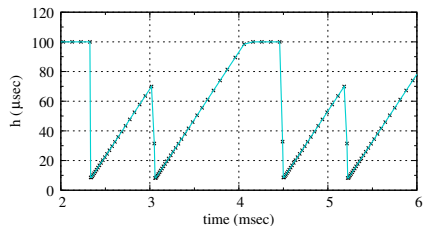
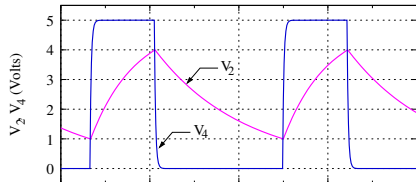
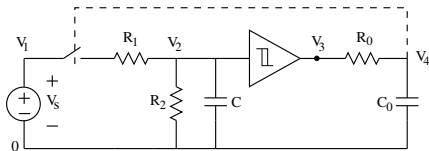
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  - $\Delta t$  too small  $\Rightarrow$  large simulation time.

# Automatic time step selection



# Automatic time step selection: example

Automatic time step selection  
based on convergence of N-R  
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- \* very difficult to judge except for simple problems
- \* In practice, reduce  $\Delta t$  by a factor of 2 and see if the results are different.
- \* Usually, the user would have some idea of the time scale, For example,
  - (a) Buck converter:  $\Delta t = T_c/50$  may be appropriate.
  - (b) Half-wave rectifier:  $\Delta t = T/50$  may be appropriate.Such a rule of thumb provides a good starting point.

# Steady-State Waveform (SSW) Analysis

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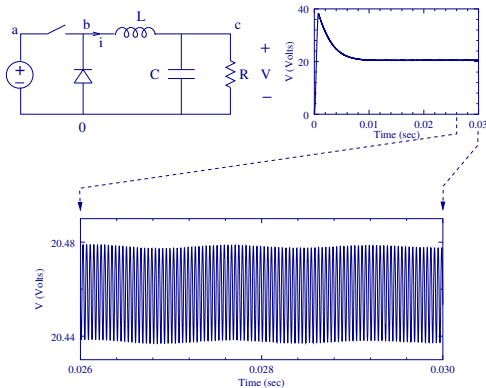
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- \* Total time for which transient simulation needs to be performed to reach the steady state is not known *a priori*; need to rely on a trial-and-error approach.
- \* It is much faster to obtain the steady-state information *directly* where a nonlinear problem in the state variables is solved.

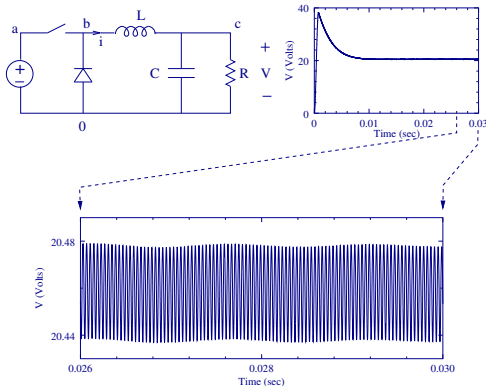
# SSW Analysis: Buck Converter



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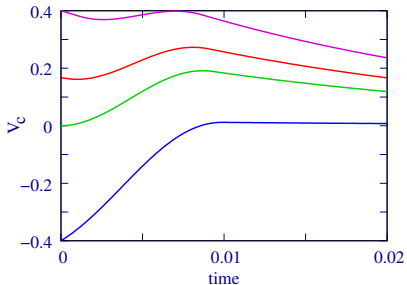
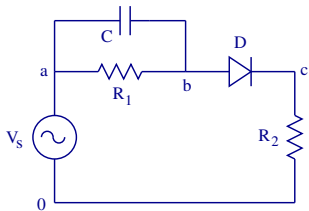


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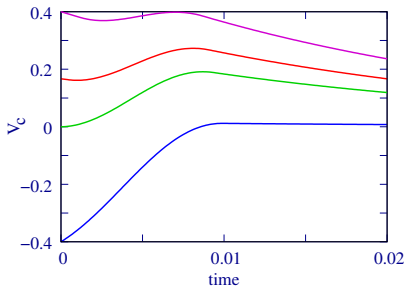
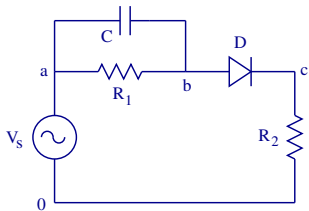
- \* A large number of cycles are required if transient simulation is used. (Note that, for this example, the steady state is not quite reached as indicated by the small amplitude variation.)
- \* If a component value ( $L$  or  $C$ ) is changed, we would not know how long to simulate to attain steady state. This is cumbersome.

# SSW Analysis: Basic idea



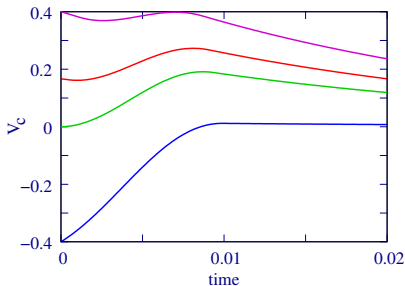
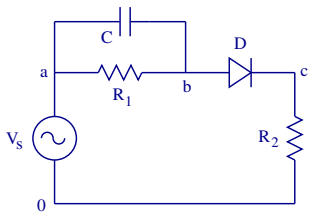
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- \* Start with an initial guess for the state variable(s) (the capacitor voltage here).
- \* Integrate for one cycle. Is  $V_c(T) = V_c(0)$ ?
- \* If yes (red curve), we have obtained the SSW solution; if not, we need to compute a better initial guess (in an *outer* Newton-Raphson loop) and repeat [7].

Example	$N_{\text{trns}}$	$N_{\text{SSW}}$
Buck Converter	750	4
Boost Converter	625	3
Cúk Converter	1250	3
1- $\phi$ half-wave rectifier	150	3
1- $\phi$ half-controlled bridge converter	110	4
3- $\phi$ diode bridge rectifier	200	4
Induction motor	125	17

- \* Note the dramatic reduction in computational effort for the SSW method as compared to transient analysis.

# References

- [1] L. O. Chua and P. M. Lin, *Computer-Aided Analysis of Electronic Circuits*, Englewood Cliffs: Prentice-Hall, 1976.
- [2] W. J. McCalla, *Fundamentals of Computer-Aided Circuit Simulation*, Boston: Kluwer Academic Publishers, 1987.
- [3] R. Raghuram, *Computer Simulation of Electronic Circuits*, New Delhi: Wiley Eastern, 1989.
- [4] K. S. Kundert, *The Designer's Guide to SPICE and SPECTRE*, Boston: Kluwer Academic Publishers, 1995.
- [5] M. B. Patil, V. Ramanarayanan, and V. T. Ranganathan, *Simulation of Power Electronic Circuits*, to be published.
- [6] C. D. Hachtel and R. K. Brayton and F. G. Gustavson, "The sparse tableau approach to network analysis and design," *IEEE Trans. CT*, vol. 18, pp. 101-113, 1971.
- [7] F. R. Colon and T. N. Trick, "Fast periodic steady-state analysis for large-signal electronic circuits," *IEEE J. Solid-State Circuits*, vol. 8, pp. 260-269, 1973.
- [8] C. F. Gerald and P. O. Whitley, *Applied Numerical Analysis*, Delhi: Pearson Education India, 1999.