

Newton LoopoldSol $\leftarrow 0$

spClear(); Clear Rhs

ckt Load (matrix, Rhs)

spFactor (matrix)

SpSolve ("") \rightarrow Sol

$$f(x) = 0$$

$$J(x^k)(x^{k+1} - \underline{x^k}) = -f(x^k)$$

This formulation is being used

$$\underbrace{J(x^k)}_{\text{Jacobian}} \underline{x^{k+1}} = \boxed{\underline{J(x^k)} \underline{x^k} - f(x^k)}$$

$$|x^{k+1} - x^k| < \epsilon_A + \epsilon_R \max(1|x^{k+1}|, |x^k|)$$

$\uparrow 10^{-6}$ $\uparrow 10^{-3}$

Copy Sol to oldSol

$$\text{oldSol}[i] = \text{sol}[i]$$

for ($i = 0; i \leq \text{NumEqn}; i++$)

$$\text{Rhs}[i] = 0.0;$$

Linear equation: $f_1 = x_1 + x_2 + x_3 = 0$

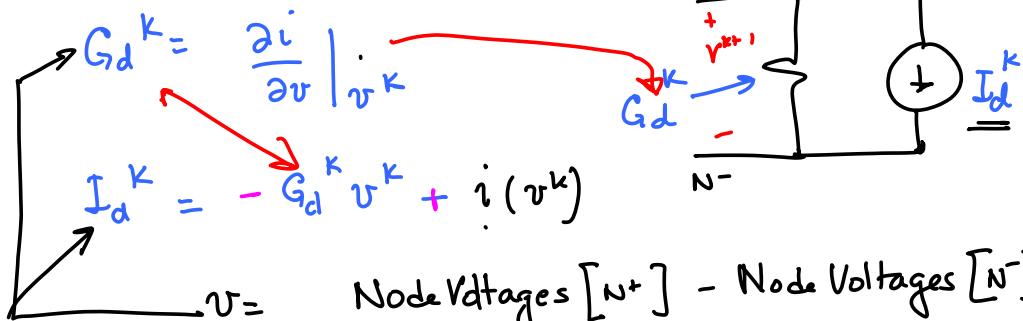
$$\frac{\partial f_1}{\partial x_i} = [1 \ 1 \ 1]$$

$$[1 \ 1 \ 1] \begin{bmatrix} x_1^{k+1} \\ x_2^{k+1} \\ x_3^{k+1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1^k \\ x_2^k \\ x_3^k \end{bmatrix} - \begin{bmatrix} x_1^k + x_2^k + x_3^k \end{bmatrix} = 0$$

Diode example:

$$i = I_s (e^{\frac{v}{V_{th}}} - 1) \quad \text{Node Voltages}$$

$$\frac{\partial i}{\partial v} = \frac{I_s}{V_{th}} e^{\frac{v}{V_{th}}}$$

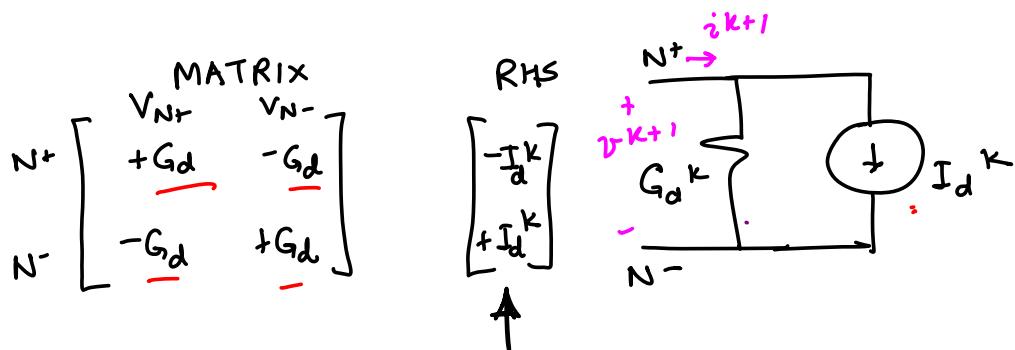


Companion model

$$J(x^k)$$

$$J(x^k)x^k - f(x^k)$$

$$\dot{i}^{k+1} = \underline{i(v^k)} + \frac{\partial i}{\partial v} \Big|_{v^k} (v^{k+1} - v^k) \\ \frac{\partial i}{\partial v} \Big|_{v^k} v^{k+1} + \boxed{i(v^k) - \frac{\partial i}{\partial v} \Big|_{v^k}}$$



Integration Methods

FE, BE, TR (one step methods)

$$\text{FE: } x_n = x_{n-1} + h \dot{x}_{n-1}$$

$$\text{BE: } x_n = x_{n-1} + h \dot{x}_n$$

$$\text{TR: } x_n = x_{n-1} + \frac{h}{2} (\dot{x}_n + \dot{x}_{n-1})$$

Linear multistep method (p -step method)

$$\sum_{i=0}^p \alpha_i x_{n-i} + h \sum_{i=0}^p \beta_i \dot{x}_{n-i} = 0$$

$|LTE|$ bounded by LE

$$\begin{aligned} LE_n &= \sum_{i=0}^p \alpha_i x(t_{n-i}) + h \sum_{i=0}^p \beta_i \dot{x}(t_{n-i}) \\ &= \sum_{i=0}^p \alpha_i x(t_{n-ih}) + h \sum_{i=0}^p \beta_i \dot{x}(t_{n-ih}) \end{aligned}$$

Once α_i, β_i are selected i.e., a particular integration method then only 'h' can be used to control the error

$$LE_n = E[x(t), h]$$

$$\begin{aligned} E[x, h] &= E[x_0] + E^{(1)}[x_0]h + E^{(2)}[x_0] \frac{h^2}{2} \\ &\quad + \dots + E^{(k+1)}[x_0] \frac{h^{k+1}}{(k+1)!} + O(h^{k+2}) \end{aligned}$$

$$E[x, h] = \sum_{i=0}^p \alpha_i x(t_{n-ih}) + h \sum_{i=0}^p \beta_i \dot{x}(t_{n-ih})$$

$$\begin{aligned} E^{(1)}[x, h] &= \sum_{i=0}^p \alpha_i \dot{x}(t_{n-ih})(-i) + \sum_{i=0}^p \beta_i \ddot{x}(t_{n-ih})(i) \\ &\quad + h \sum_{i=0}^p \beta_i \dddot{x}(t_{n-ih})(-i) \end{aligned}$$

$$E^{(1)}[x_0] = \sum_{i=0}^p \alpha_i \dot{x}(t_n)(-i) + \sum_{i=0}^p \beta_i \ddot{x}(t_n)$$

$$E[x_0] = \sum_{i=0}^p \alpha_i x(t_n)$$

A multistep method is said to be of order k if the method is exact for polynomials of degree $\leq k$

$$x(t) = c, t, t^2, \dots, t^k$$

$$x^{(k+1)}(t) = 0 \quad \text{for a polynomial of degree } k$$

Exactness constraints

$$\left\{ \begin{array}{l} 0 = E[x, 0] \Rightarrow \sum_{i=0}^p \alpha_i = 0 \\ 0 = E^{(1)}[x, 0] \Rightarrow \sum_{i=0}^p [\alpha_i(-i) + \beta_i] = 0 \\ \vdots \\ 0 = E^{(k)}[x, 0] \Rightarrow \sum_{i=0}^p [\alpha_i(-i)^k + k\beta_i(-i)^{k-1}] = 0 \\ 0 \neq E^{(k+1)}[x, 0] \Rightarrow \sum_{i=0}^p [\alpha_i(-i)^{k+1} + (k+1)\beta_i(-i)^k] \neq 0 \end{array} \right.$$

Example BE $x_n = x_{n-1} + h \dot{x}_n$

$$x_n - x_{n-1} - h \dot{x}_n = 0$$

1-step $\alpha_0 = 1, \alpha_1 = -1, \beta_0 = -1, \beta_1 = 0$

$$E[x, 0] = \sum_{i=0}^p \alpha_i = \alpha_0 + \alpha_1 = 1 + -1 = 0$$

$$\begin{aligned} E^{(1)}[x, 0] &= \sum_{i=0}^1 \alpha_i(-i) + \beta_i = \alpha_0(0) + \alpha_1(-1) + \beta_0 \\ &\quad + \beta_1 \\ &= (-1)(-1) - 1 = 0 \end{aligned}$$

$$\begin{aligned} E^{(2)}[x, 0] &= \sum_{i=0}^1 \alpha_i(-i)^2 + 2\beta_i(-i) \\ &= \alpha_0(0)^2 + \alpha_1(-1)^2 + 2\beta_0(0) + 2\beta_1(-1) \\ &= 1 + 1 + 2(-1) + 2(0) = -1 \neq 0 \end{aligned}$$

\Rightarrow BE is a first order method

$$\underline{TR} : \quad x_n = x_{n-1} + \frac{h}{2} [x_n + x_{n-1}]$$

1-step method

$$x_n - x_{n-1} - h \frac{1}{2} \dot{x}_n - h \frac{1}{2} \dot{x}_{n-1} = 0$$

$$\alpha_0 = 1, \quad \alpha_1 = -1, \quad \beta_0 = -\frac{1}{2}, \quad \beta_1 = -\frac{1}{2}$$

$$E[x, 0] = 0, \quad E^{(1)}[x, 0] = 0, \quad E^{(2)}[x, 0] = 0$$

$$E^{(3)}[x, 0] \neq 0$$

\Rightarrow TR is a 2nd order method

Synthesis of linear multistep integration methods

p step method of order K $\alpha_0 = 1$

$$\sum_{i=0}^p \alpha_i x_{n-i} + h \sum_{i=0}^p \beta_i \dot{x}_{n-i}$$

$\underbrace{\alpha_0, \alpha_1, \dots, \alpha_p}_{p \text{ coeff}} \quad \underbrace{\beta_0, \beta_1, \dots, \beta_p}_{p+1 \text{ coeff}}$

Need to determine $p + p+1 = 2p+1$ coeff

Exactness constraints $K+1$ equations

$2p+1$ unknowns & $K+1$ equations

$$2p+1 \geq K+1 \Rightarrow p \geq \frac{K}{2}$$

TR! $K=2, p=1$

Algorithm for choosing coeff

1) Choose order of method - K (accuracy)

2) Choose number of steps - p ($\geq \frac{K}{2}$)

3) Write the $(K+1)$ exactness constraints

4) For $p > \frac{K}{2}$ $(2p+1) - (K+1) = 2p-K$

coeff can be assigned arbitrarily (or some other crit)

Example

Suppose $p = k$ then k coeff
need additional constraints

Take $\beta_1, \beta_2, \dots, \beta_p = 0$

Then the integration method is

$$\sum_{i=0}^k \alpha_i x_{n-i} + h \beta_0 \dot{x}_n = 0 \quad (\alpha_0 = 1)$$

This is the k -th order backward differentiation formula (BDF) also referred to as Gear's methods

Example

$$p = k = 2$$

2nd-order Gear's method

$$\alpha_0 = 1, \alpha_1, \alpha_2, \beta_0$$

$$E[x, 0] = \sum_{i=0}^2 \alpha_i = 0 \quad = 1 + \alpha_1 + \alpha_2 = 0$$

$$E^{(1)}[x, 0] = \sum_{i=0}^2 \alpha_i (-i) + \beta_0 \quad \alpha_1 (-1) + \alpha_2 (-2) + \beta_0 = 0$$

$$E^{(2)}[x, 0] = \sum_{i=0}^2 \alpha_i (-i)^2 + 2\beta_0 (-i)^0 \\ \alpha_1 (-1)^2 + \alpha_2 (-2)^2 = 0$$

$$\alpha_1 = -\frac{4}{3}, \alpha_2 = \frac{1}{3}, \beta_0 = -\frac{2}{3}$$

So the integration method is

$$x_n - \frac{4}{3}x_{n-1} + \frac{1}{3}x_{n-2} - \frac{2}{3}h \dot{x}_n = 0$$

$$x_n = \frac{4}{3}x_{n-1} - \frac{1}{3}x_{n-2} + \frac{2}{3}h \dot{x}_n$$

$$\dot{x}_n = \frac{1}{h} \left[\frac{3}{2} \underline{\underline{x}_n} - \underbrace{2x_{n-1} + \frac{1}{2}x_{n-2}} \right]$$

$$= \alpha x_n + \beta$$

Local Error for a k th order method

$$E[x, 0] = 0 = E^{(1)}[x, 0] - \dots = E^{(k)}[x, 0]$$

$$E^{(k+1)}[x, 0] \neq 0$$

$$LE_n = E[x, h] = E[x_0] + E^{(1)}[x_0] \overset{\circ}{h} + E^{(2)}[x_0] \frac{h^2}{2} \\ + \dots + E^{(k+1)}[x_0] \frac{h^{k+1}}{(k+1)!} + O(h^{k+2})$$

$$LE \approx E^{(k+1)}[x_0] \frac{h^{k+1}}{(k+1)!}$$

$$= \frac{h^{k+1}}{(k+1)!} \sum_{i=0}^p \left[\alpha_i (-i)^{k+1} + (k+1) \beta_i (-i)^k \right] x^{(k+1)}(t_n) \\ = C_{k+1} \frac{h^{k+1}}{(k+1)!} x^{(k+1)}(t_n) \\ C_{k+1} = \frac{1}{(k+1)!} \sum_{i=0}^p \left[\alpha_i (-i)^{k+1} + (k+1) \beta_i (-i)^k \right]$$

BE: $x_n = x_{n-1} + h \dot{x}_n$
 $x_n - x_{n-1} - h \dot{x}_n = 0$

$$\alpha_0 = 1, \quad \alpha_1 = -1, \quad \beta_0 = -1, \quad \beta_1 = 0$$

$$C_2 = \frac{1}{2!} \sum_{i=0}^1 \left[\alpha_i (-i)^2 + 2 \beta_i (-i) \right] \\ = \frac{1}{2} \left[\alpha_0(0)^2 + \alpha_1(-1)^2 + 2 \beta_0(0) + 2 \beta_1(-1) \right]$$

$$= -\frac{1}{2}$$

$$LE \text{ for BE: } -\frac{1}{2} h^2 \ddot{x}(t_n)$$

$$LE \text{ for TR: } -\frac{1}{12} h^3 \dddot{x}(t_n)$$

Two integration methods of order k

Compare leading coefficient of LE
for error

LE for Gear's second order method

$$= \frac{1}{6} h^3 \ddot{x}(t_n)$$

More on Local Error

LMS integration method (p -step method)

$$\textcircled{1} \quad \sum_{i=0}^p \alpha_i x_{n-i} + h \sum_{i=0}^b \beta_i \dot{x}_{n-i} = 0$$

$$\text{LE} = \frac{1}{(k+1)!} \left[\sum_{i=0}^p \alpha_i (-i)^{k+1} + (k+1) \beta_i (-i)^k \right] h^{k+1} \dot{x}(t_n)$$

New integration method as

$$\textcircled{2} \quad \sum_{i=0}^b \alpha'_i x_{n-i} + h \sum_{i=0}^b \beta'_i \dot{x}_{n-i} = 0$$

$$\alpha'_i = \frac{\alpha_i}{\eta} \quad ; \quad \beta'_i = \frac{\beta_i}{\eta}$$

This scaling would change the LE
wrongly suggesting that a simple scaling will
make the method more accurate

Need a normalization such that LE is
the same regardless of whether $\textcircled{1}$ or $\textcircled{2}$
is used

Normalization: $\sum_{i=0}^p \beta_i = -1$

BE: $x_n - x_{n-1} - h \dot{x}_n = 0 \quad \beta_0 = -1, \beta_1 = 0$

TR: $x_n - x_{n-1} - \frac{h}{2} \dot{x}_n - \frac{h}{2} \dot{x}_{n-1} = 0; \quad \beta_0 = -\frac{1}{2}, \beta_1 = -\frac{1}{2}$

One could have written the TR method as:

$$\frac{x_n}{2} - \frac{x_{n-1}}{2} - \frac{h}{4} \dot{x}_n - \frac{h}{4} \dot{x}_{n-1} = 0$$

$$\beta_0 = -\frac{1}{4}, \quad \beta_1 = -\frac{1}{4}$$

Implicit Integration Methods

$$\dot{x} = f(x)$$

$$\sum_{i=0}^p [\alpha_i x_{n-i} + h \beta_i \dot{x}_{n-i}] = 0 \quad \beta_0 \neq 0$$

$$\underline{\alpha_0 x_n} + \sum_{i=1}^p \underline{\alpha_i x_{n-i}} + \underline{h \beta_0 \dot{x}_n} + h \sum_{i=1}^p \underline{\beta_i \dot{x}_{n-i}} = 0$$

$$\underline{\alpha_0 x_n} + \underline{h \beta_0 \dot{x}_n} + \sum_{i=1}^p \underline{\alpha_i x_{n-i}} + \underline{h \beta_i \dot{x}_{n-i}} = 0$$

$$f(x_n) \quad \quad \quad f(x_{n-i})$$

$$\alpha_0 x_n + h \beta_0 f(x_n) + \underbrace{\text{value calculated from previous time points}}_b = 0$$

$$F(x_n) = \alpha_0 x_n + h \beta_0 f(x_n) + b = 0$$

This is a nonlinear equation in x_n

⇒ solve using Newton's method

$$J(x_n^k) (x_n^{k+1} - x_n^k) = -F(x_n^k)$$