

HW # 3 Part 2 Due Monday Nov. 7

HW # 2 Graded handed back

part 2 100 pts

→ Stamps

→ abs()

ABS()

→ Array bounds

Solution, old Solution NumEqns+1

for (i=0; i ≤ NumEqns; i++)

i < NumEqns+1 i++

<math.h>

fabs()

double fabs()

→ Convergence checks

$x^{k+1}$ ,  $x^k$

$$\left| x^{k+1}[i] - x^k[i] \right| \leq \text{ABSTOL} + \text{RELTOL} \times \text{MAX}(|x^{k+1}[i]|, |x^k[i]|)$$

Newton Loop

for (numIter ≤ MAXITER or NOT Converged) {

}

Stability of integration methods

$$\underbrace{(z^p + \alpha_1 z^{p-1} + \dots + \alpha_p)}_{P_N(z)} + \sigma \underbrace{(\beta_0 z^p + \beta_1 z^{p-1} + \dots + \beta_p)}_{P_D(z)} = 0$$

$$\sigma = - \frac{P_N(z)}{P_D(z)}$$

For large  $h \Rightarrow \sigma \rightarrow -\infty$   
 $\Rightarrow P_D(z) \rightarrow 0$

$$\Rightarrow \beta_0 z^p + \beta_1 z^{p-1} + \dots + \beta_p = 0$$

What if we set  $\beta_1, \beta_2, \dots, \beta_p = 0$

$$\Rightarrow \beta_0 z^p = 0$$

The roots are  $z=0$  of multiplicity  $p$   
 There are  $p$  roots, all at the origin  $\Rightarrow$  stability region

includes  $h = \infty$

This is the idea behind Gear's methods  
 (or backward differentiation formula)

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

$$k = 1: \quad \alpha_0 x_n + \alpha_1 x_{n-1} + h \beta_0 \dot{x}_n = 0$$

$$\text{BE method: } \alpha_0 = 1, \alpha_1 = -1, \beta_0 = -1$$

(1st order Gear's method or BDF)

Variable time steps

$h$  is no longer uniform

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

with  $\sum_{i=0}^p \beta_i = -1$  for LE calculations

Example 2nd order Gear's method (BDF)

$$\sum_{i=0}^2 \alpha_i x_{n-i} - h_n \dot{x}_n = 0$$

$$\alpha_0 x_n + \alpha_1 x_{n-1} + \alpha_2 x_{n-2} - h_n \dot{x}_n = 0$$

For a 2nd order method  $\Rightarrow$  exact for polynomials of degree  $\leq 2$

Take  $P(t) = (t - t_n)^k$   $k=0, 1, 2$

$$k=0: \quad x(t) = 1 \quad \dot{x}(t) = 0$$

$$k=1: \quad x(t) = (t - t_n) \quad \dot{x}(t) = 1$$

$$k=2: \quad x(t) = (t - t_n)^2 \quad \dot{x}(t) = 2(t - t_n)$$

For LE calculation we use exact solution

$$k=0: \quad \alpha_0 x(t_n) + \alpha_1 x(t_{n-1}) + \alpha_2 x(t_{n-2}) - h_n \dot{x}(t_n) = 0$$

$$\alpha_0 + \alpha_1 + \alpha_2 = 0$$

$$k=1: \quad \alpha_0 (0) + \alpha_1 \underbrace{(t_{n-1} - t_n)}_{(-h_n)} + \alpha_2 \underbrace{(t_{n-2} - t_n)}_{-(h_n + h_{n-1})} - h_n = 0$$

$k=2:$

$$\alpha_0 = h_n \left( \frac{1}{h_n} + \frac{1}{h_n + h_{n-1}} \right)$$

All coefficients are dependent on the timestep

In General for  $k$ -th BDF ( $1 \leq \text{order} \leq k$ )

$$\alpha_0 = h_n \sum_{i=1}^k \frac{1}{t_n - t_{n-i}}$$

$$1 \leq j \leq k: \quad \alpha_j = \frac{-h_n}{(t_n - t_{nj})} \prod_{\substack{i=1 \\ i \neq j}}^k \frac{t_n - t_{n-i}}{(t_{nj} - t_{n-i})}$$

Ex: 2nd-order BDF

$$\alpha_0 = h_n \sum_{i=1}^2 \frac{1}{t_n - t_{n-i}} = h_n \left[ \frac{1}{t_n - t_{n-1}} + \frac{1}{t_n - t_{n-2}} \right]$$

$$= h_n \left[ \frac{1}{h_n} + \frac{1}{h_n + h_{n-1}} \right]$$

$$\alpha_1 = \frac{-h_n}{(t_n - t_{n-1})} \prod_{\substack{i=1 \\ i \neq 1}}^2 \frac{t_n - t_{n-i}}{t_{n-1} - t_{n-i}}$$

$$= \frac{-h_n}{h_n} \frac{t_n - t_{n-2}}{t_{n-1} - t_{n-2}} = -\frac{h_n + h_{n-1}}{h_{n-1}}$$

For kth order BDF LE

$$C_{k+1} = \frac{1}{(k+1)!} \frac{1}{h_n^k} \prod_{i=1}^k (t_n - t_{n-i})$$

where LE =  $C_{k+1} h_n^{k+1} \mathcal{L}^{(k+1)}(t_n)$

let us consider TR

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

Non uniform  $\alpha_0 x_n + \alpha_1 x_{n-1} + h_n \beta_0 \dot{x}_n + h_n \beta_1 \dot{x}_{n-1} = 0$

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

For variable time steps:

$$x_n - x_{n-1} - \frac{h_n}{2} (\dot{x}_n + \dot{x}_{n-1}) = 0$$

same as TR with uniform h

$$LE = -\frac{h_n^3}{12} \ddot{x}(t_n)$$

## Timestep control

How do we determine the stepsize  $h_n$ ?

Goal: choose as few timepoints as possible given an error constraint

⇒ minimize # of timepoints

Assume  $E_n$  is a bound on the absolute error  
 $|LE_n| \leq E_n$

$$|LE_n| = \left| C_{k+1} h_n^{k+1} x^{(k+1)}(t_n) \right| \leq E_n$$

$$h_n \leq \left[ \frac{E_n}{|C_{k+1} x^{(k+1)}(t_n)|} \right]^{\frac{1}{k+1}}$$

$E_n$  is given,  $C_{k+1}$  is known  
 what about  $x^{(k+1)}(t_n)$ ?

In SPICE divided differences are used

$$DD_1 = \frac{x_n - x_{n-1}}{h_n} \approx \dot{x}_n$$

$$DD_2 = \frac{DD_1(t_n) - DD_1(t_{n-1})}{h_n + h_{n-1}} \approx \frac{\ddot{x}_n}{2!}$$

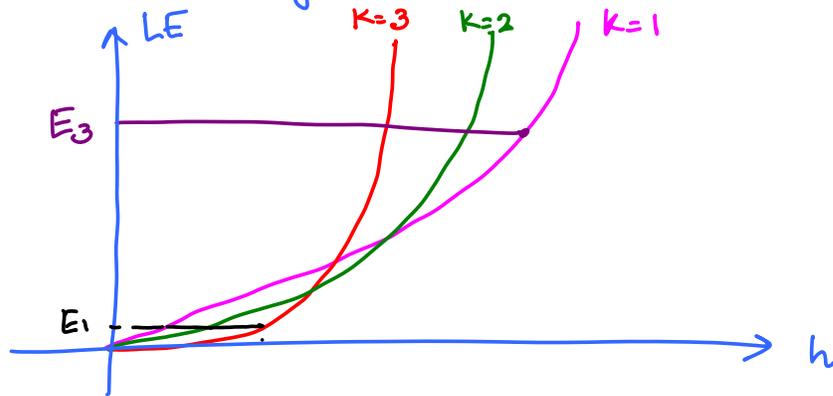
⋮

$$DD_{k+1} = \frac{DD_k(t_n) - DD_k(t_{n-k})}{\sum_{i=0}^k h_{n-i}} \approx \frac{x_n^{(k+1)}}{(k+1)!}$$

## Order Control

If we have a method of order  $k$   
 (BDF:  $1 \leq k \leq 6$ )

→ Choose order which gives you the largest timestep



In SPICE the default method is TR

(METHOD = GEAR      MAXORD = N (2 ≤ N ≤ 6))

Heuristic rules for timestep control

- Do not change timestep too often
- Change order only if improvement is worthwhile  
i.e. at least 2h

- Attempt change of stepsize and order only if  
 $|LE| < E_n$  (error is large)

In SPICE:  $E_n = E_A + E_R \text{MAX}(|x_n|, |x_{n-1}|)$

↑	
ABSTOL	(currents)
VNTOL	(voltages)
CHGTOL	(charges)

## Linear Multistep Methods – Stability

$$\sum_{i=0}^p \alpha_i x_{n-i} + h\beta_i \dot{x}_{n-i} = 0; \text{ Testproblem } \frac{d}{dt} x(t) = \lambda x(t), x(0) = 1 \quad (1)$$

$$\sum_{i=0}^p (\alpha_i + \beta_i h\lambda) x_{n-i} = \sum_{i=0}^p (\alpha_i + \sigma\beta_i) x_{n-i} = 0 \quad (2)$$

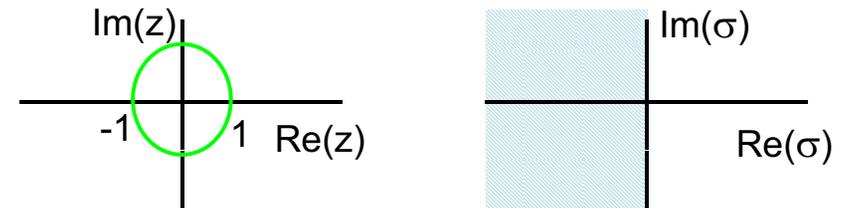
- A method is **stable** if all solutions of the associated difference equation (2) obtained by setting  $\sigma=0$  remain bounded as  $n \rightarrow \infty$
- The **region of absolute stability** of a method is the set of  $\sigma$  (complex) such that all solutions of (2) remain bounded as  $n \rightarrow \infty$
- **Note:** A method is stable if its region of absolute stability contains the origin, i.e.,  $\sigma=0$

## Stability

The region of absolute stability of a method is the set of  $\sigma$  such that all the roots of  $\sum_{i=0}^p (\alpha_i + \sigma\beta_i) z^{p-i} = 0$  are inside or on the complex unit circle, i.e.,  $|z| \leq 1$ , and the roots for which  $|z| = 1$  are of multiplicity 1

A method is **A-stable** if the region of absolute stability contains the entire left-half plane ( $\text{Re}(\sigma) < 0$ )

TR is an A-stable method



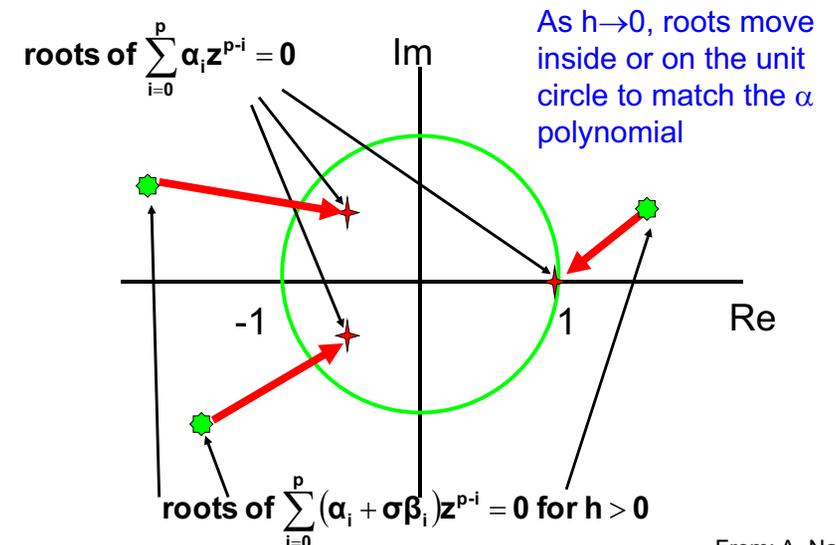
From: A. Nardi

## Stability

- Each method is associated with two polynomials of coefficients  $\alpha$  and  $\beta$ :
  - $\alpha$ : associated with past function values ( $x_{n-i}$ )
  - $\beta$ : associated with past derivative values ( $x'_{n-i}$ )
- **Stability:** roots of  $\alpha$  polynomial must satisfy  $|z| \leq 1$  and be of multiplicity 1 for  $|z|=1$
- **Absolute stability:** roots of  $(\alpha + \sigma\beta)$  polynomial must satisfy  $|z| \leq 1$  and be of multiplicity 1 for  $|z|=1$

From: A. Nardi

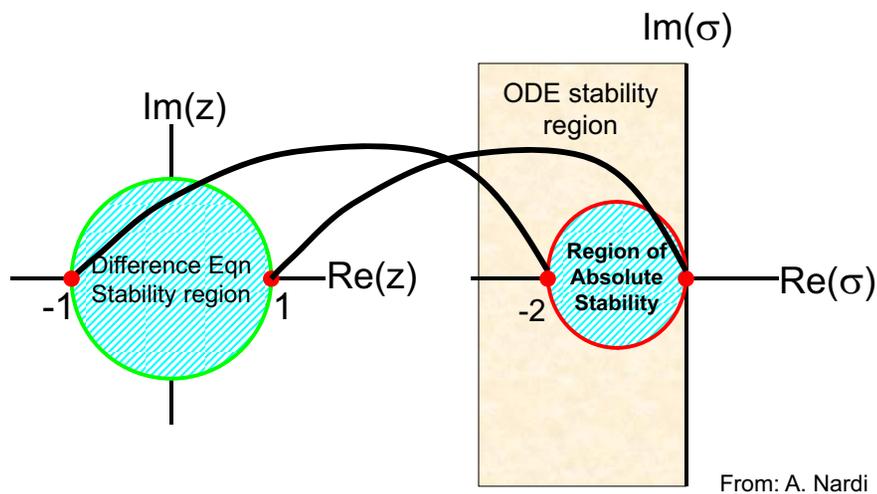
## Stability and Region of Absolute Stability



From: A. Nardi

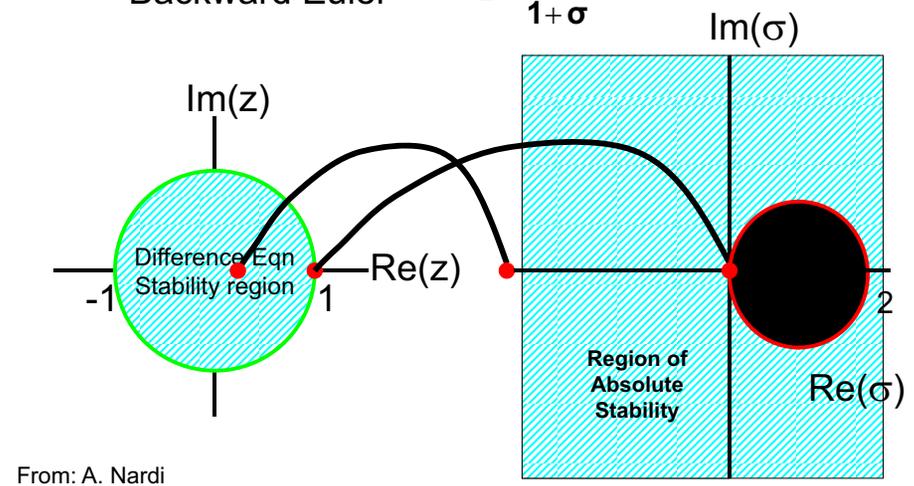
## FE Region of Absolute Stability

Forward Euler  $z = 1 + \sigma$



## BE Region of Absolute Stability

Backward Euler  $z = \frac{1}{1 + \sigma}$



## L-Stability

- An A-stable method is **L-stable** if  $\text{Re}(\lambda) < 0$  implies  $\lim_{h \rightarrow \infty} x_n = 0$  for all  $x_{n-1}$

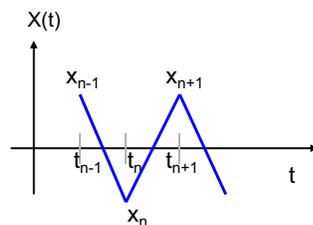
Consider TR  $\text{Re}(\lambda) < 0$ :  $x_n - x_{n-1} - \frac{h}{2}(\dot{x}_n + \dot{x}_{n-1}) = 0$

$$\left(1 - \frac{\lambda h}{2}\right)x_n - \left(1 + \frac{\lambda h}{2}\right)x_{n-1} = 0$$

$$\Rightarrow x_n = \frac{\left(1 + \frac{\lambda h}{2}\right)}{\left(1 - \frac{\lambda h}{2}\right)} x_{n-1} \Rightarrow \lim_{h \rightarrow \infty} x_n = -x_{n-1}$$

i.e., there is ringing

TR is not an L-stable method



## Finding the Region of Absolute Stability

$$(1 + \sigma\beta_0)z^p + (\alpha_1 + \sigma\beta_1)z^{p-1} + \dots + (\alpha_p + \sigma\beta_p) = 0$$

For what values of  $\sigma$  do all the  $p$  roots of this polynomial satisfy the stability condition?

$$z^p + \alpha_1 z^{p-1} + \dots + \alpha_p + \sigma(\beta_0 z^p + \beta_1 z^{p-1} + \dots + \beta_p) = 0$$

Methods

- Choose  $\sigma$ , compute roots, test, repeat for all  $\sigma$  (BAD)
- Solve for  $\sigma = -P_N(z)/P_D(z)$

$$P_N(z) = z^p + \alpha_1 z^{p-1} + \dots + \alpha_p$$

$$P_D(z) = \beta_0 z^p + \beta_1 z^{p-1} + \dots + \beta_p$$

$$\text{Consider } S = \left\{ \sigma \mid \sigma = -\frac{P_N(z)}{P_D(z)}, |z| \leq 1 \right\} \text{ and let } z \text{ vary in } |z| \leq 1$$

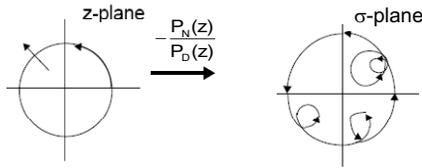
May get same  $\sigma$  values for two or more different  $z$ 's, one with  $|z| \leq 1$  and one with  $|z| > 1$

$\Rightarrow S$  is a superset of the region of absolute stability

## Boundary Locus $\Gamma_\sigma$

Boundary locus  $\Gamma_\sigma$  is the contour  $-\frac{P_N(z)}{P_D(z)}$  when  $|z|=1$ , i.e.,  $z = e^{i\theta}$ ,

$0 \leq \theta \leq 2\pi$ . It is a map from the z-plane to the  $\sigma$ -plane



### Basic Results from Theory of Complex Variables

- Mapping is conformal, i.e., angle preserving
- $\Gamma_\sigma$  separates the  $\sigma$ -plane into disjoint sets. In each set, the number of roots outside the unit circle is constant
- The boundary of the stability region is a subset of  $\Gamma_\sigma$

## Region of Absolute Stability

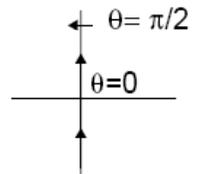
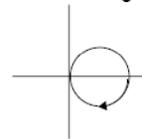
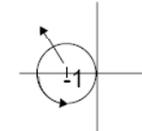
Move counterclockwise along  $|z|=1$ , then you see **one more root to the outside of the unit circle** for those  $\sigma$  to the right than for those  $\sigma$  to the left when traversing  $\Gamma_\sigma$

**F.E.:**  $q(z = e^{i\theta}) = e^{i\theta} - 1$

**T.R.:**  $x_n = x_{n-1} + \frac{q}{2}(x_{n-1} + x_n)$

$$\Rightarrow q = \frac{2(z-1)}{z+1} = \frac{2(e^{i\theta}-1)}{e^{i\theta}+1}$$

**B.E.:**  $q(z = e^{i\theta}) = 1 - \frac{1}{e^{i\theta}} = 1 - e^{-i\theta}$

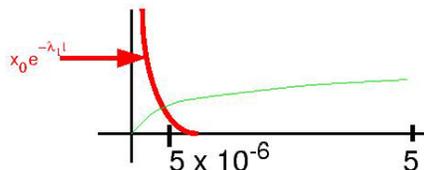


From: A. Sangiovanni-Vincentelli

## Large Timestep Issues Stiff Problems

$$\begin{cases} \frac{dx(t)}{dt} = -\lambda_1(x - s(t)) + \frac{ds(t)}{dt} & \text{where } s(t) = 1 - e^{-\lambda_2 t} \\ x(0) = x_0 \end{cases} \quad \lambda_1 = 10^6, \lambda_2 = 1$$

Exact solution:  $x(t) = x_0 e^{-\lambda_1 t} + 1 - e^{-\lambda_2 t}$



For  $t \geq 5 \cdot 10^{-6}$   $x_0 e^{-\lambda_1 t} \approx 0$   
 For  $t \geq 5$   $1 - e^{-\lambda_2 t} \approx 1$

Interval of interest is  $[0, 5]$

Uniform step size (for accuracy)

$$\Rightarrow \Delta t \leq 10^{-6}$$

$$\Rightarrow 5 \times 10^6 \text{ steps !!!}$$

From: A. Nardi

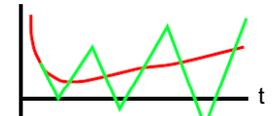
## One Possible Strategy – Variable Time Steps

- Take 5 steps of size  $10^{-6}$  for accuracy during initial phase and then 5 steps of size 1
- With FE cannot use  $h > 2 \times 10^{-6}$  ( $\lambda = 10^6$ )

### Stiff problem:

1. Natural time constants
2. Input time constants
3. Interval of interest

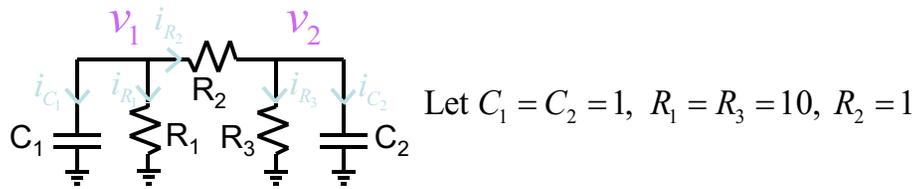
If these are widely separated, then the problem is **stiff**



From: A. Nardi

# Application Example

## On-chip Signal Transmission– 2x2 example

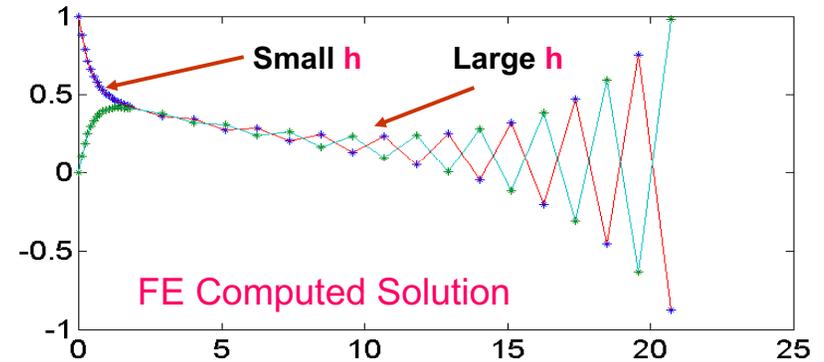


Time Constants 1/0.1 sec and 1/2.1 sec

$$\frac{dx}{dt} = \underbrace{\begin{bmatrix} -1.1 & 1.0 \\ 1.0 & -1.1 \end{bmatrix}}_A x$$

From: A. Nardi

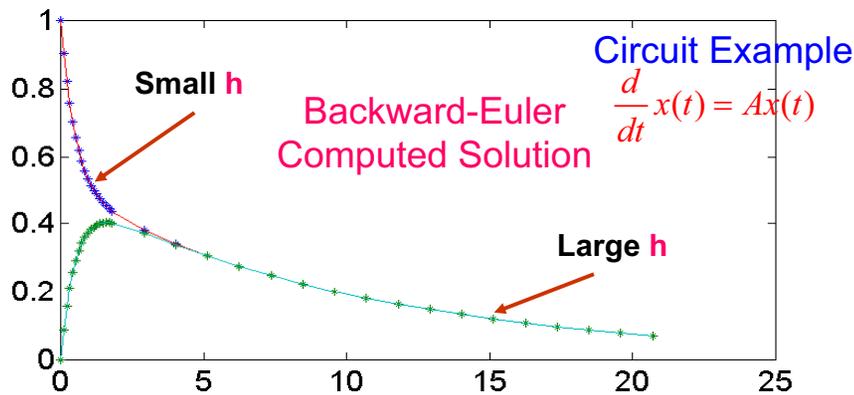
## FE on Two Time-constant Circuit



Forward-Euler is accurate for small timesteps, but goes unstable when timestep is increased

From: A. Nardi

## BE on Two Time-constant Circuit



With BE can use small timesteps for fast dynamics and then switch to large timesteps for the slow decay

From: A. Nardi

## Summary of Stiff Problems

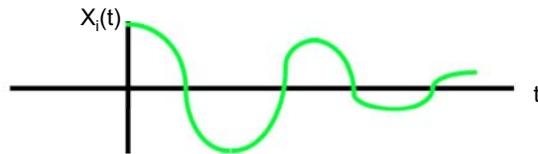
- The analysis of stiff circuits requires the use of variable step sizes
- Not all the linear multistep methods can be efficiently used to integrate stiff equations
- To be able to choose  $h$  based only on accuracy considerations, the region of absolute stability should allow a large  $h$  for large time constants, without being constrained by the small time constants
- A-stable methods satisfy this requirement

From: A. Nardi

## Requirements of Stiffly Stable Integration Methods

$$\dot{x}_i(t) = \lambda_i x_i(t) \quad i = 1, 2, \dots, n, \lambda_i = a_i + j b_i$$

$$x_i(t) = C_i e^{a_i t} \cos b_i t$$

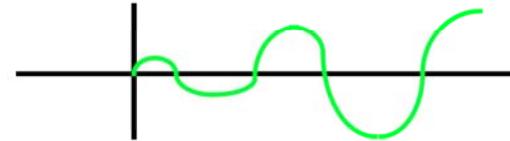


For accuracy want at least 8 points per cycle

$$\Rightarrow h \leq \frac{1}{8} \left( \frac{2\pi}{b_i} \right) \text{ or } h b_i \leq \frac{\pi}{4} \Rightarrow \text{Im}(\sigma) \leq \frac{\pi}{4}$$

## Requirements of Stiffly Stable Integration Methods

Want a region of absolute stability which gives a stable algorithm for initial transient



$$\Rightarrow a_i h > 0, a_i h < \mu$$

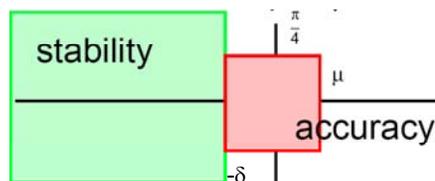
$$\text{Re}(\sigma) = a_i h$$

$$\text{Require: } 0 \leq \text{Re}(\sigma) \leq \mu$$

## Requirements of Stiffly Stable Integration Methods

Require a small h to capture fast transient

$$\lambda_{\text{fast}} = a_{\text{fast}} + j b_{\text{fast}} \Rightarrow \text{Take: } \delta > \frac{5}{a_{\text{fast}}}, a_{\text{fast}} > 0$$



Remarks:

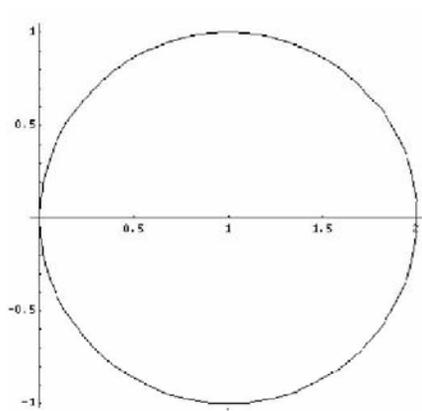
- There is a region  $\text{Re}(\sigma) < -\delta$  that is absolutely stable
- For  $0 < \text{Im}(\sigma) < \pi/4$  the region is of absolute stability and the algorithm is accurate
- For  $0 < \text{Re}(\sigma) < \mu$  the region is stable and the algorithm is accurate

## Backward Differentiation Formula - BDF (Gear Methods)

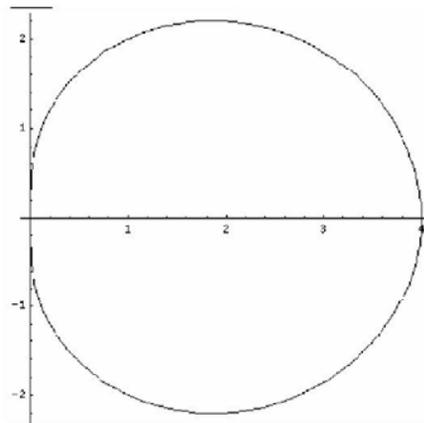
$$\sum_{i=0}^k \alpha_i x_{n-i} + h \beta_0 \dot{x}_n = 0 \quad \text{where } \beta_0 \neq 0$$

- Gear's first order method is BE
- It can be shown that:
  - Gear's methods up to order 6 are stiffly stable and are well-suited for stiff ODEs
  - Gear's methods of order higher than 6 are not stiffly stable
- Less stringent than A-stable

## Gear's Method Region of Absolute Stability (outside the closed curve)



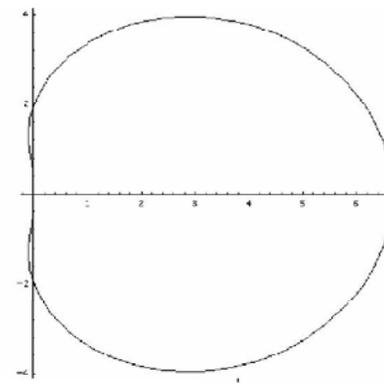
k=1



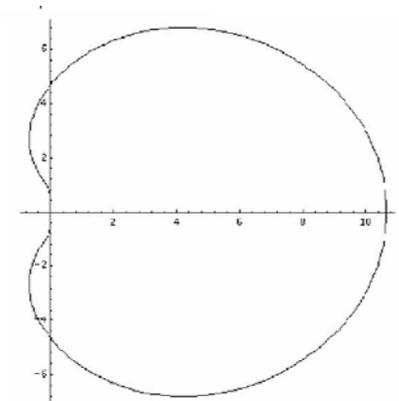
k=2

From: A. Sangiovanni-Vincentelli

## Gear's Method Region of Absolute Stability (outside the closed curve)



k=3



k=4

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## Observations on Stiff Stability

- **FE:** timestep is limited by stability and not by accuracy
- **BE:** A-stable, any timestep could be used
- **TR:** most accurate A-stable multistep method
- **Gear:** stiffly stable method (up to order 6)
  
- **The analysis of stiff circuits requires the use of variable timestep**

From: A. Nardi