

# Continuous Models (state trajectory: $\mathbb{R} \rightarrow \mathbb{R}^n$ ): Ordinary Differential Equation (ODE) formalism

$$\frac{d^n x}{dt^n} = f\left(\frac{d^{n-1}x}{dt^{n-1}}, \dots, x, u, t\right)$$

$f$  and  $x(t)$  may be vectors

$$\left\{ \begin{array}{l} x = x_0 \\ \frac{dx}{dt} = x_1 \\ \frac{dx_1}{dt} = x_2 \\ \dots \\ \frac{dx_{n-1}}{dt} = x_n = f(x_{n-1}, x_{n-2}, \dots, x_1, x_0, u, t) \end{array} \right.$$

# Euler discretisation

$$\begin{aligned} \frac{dx}{dt} &= f(x, u, t) \\ &\Downarrow \\ \frac{x(t_i + \Delta t) - x(t_i)}{\Delta t} &\approx f(x(t_i), u(t_i), t_i) \\ &\Downarrow \\ x(t_i + \Delta t) &\approx x(t_i) + \Delta t f(x(t_i), u(t_i), t_i) \end{aligned}$$

# Taylor Series Expansion

$$x(t_i + \Delta t) = x(t_i) + \frac{\Delta t}{1!} \frac{dx}{dt} \Big|_{t_i} + \frac{\Delta t^2}{2!} \frac{d^2x}{dt^2} \Big|_{t_i} + \frac{\Delta t^3}{3!} \frac{d^3x}{dt^3} \Big|_{t_i} + \dots$$

Is EXACT if expansion is not truncated

ERROR =  $O(\Delta t^{N+1})$  if truncated after term  $N$

Beyond first derivative, use difference formulas

$$\text{ERROR } \varepsilon_N \cong \text{approx}_{N+1} - \text{approx}_N$$

# Integration Methods: Euler

## Single-step

$$x_0 = \alpha_0$$

$$x_{i+1} = x_i + \Delta t f(t_i, x_i), \quad i \geq 0$$

Unsymmetrical: uses only derivative in begin point.

# Integration Methods: Modified Euler

## Single-step

$$x_0 = \alpha_0$$

$$k_1 = \Delta t f(t_i, x_i)$$

$$k_2 = \Delta t f(t_i + \Delta t, x_i + k_1)$$

$$x_{i+1} = x_i + \frac{k_1}{2} + \frac{k_2}{2}, \quad i \geq 0$$

Symmetrical: uses derivative in begin and end point.

# Integration Methods: Midpoint

## Single-step

$$x_0 = \alpha_0$$

$$k_1 = \Delta t f(t_i, x_i)$$

$$k_2 = \Delta t f\left(t_i + \frac{\Delta t}{2}, x_i + \frac{k_1}{2}\right)$$

$$x_{i+1} = x_i + k_2, \quad i \geq 0$$

Symmetrical: halfway point.

# Integration Methods: Heun

## Single-step

$$x_0 = \alpha_0$$

$$k_1 = \Delta t f(t_i, x_i)$$

$$k_2 = \Delta t f\left(t_i + \frac{2\Delta t}{3}, x_i + \frac{2k_1}{3}\right)$$

$$x_{i+1} = x_i + \frac{k_1}{4} + \frac{3k_2}{4}, \quad i \geq 0$$

# Integration Methods: Runge-Kutta Methods

**Single-step,  $q$  stages (function evaluations per step)**

$$x_{i+1} = x_i + \Delta t \phi(t_i, x_i; \Delta t)$$

$$\phi(t_i, x_i; \Delta t) = \sum_{i=1}^q \omega_i k_i$$

**Explicit method:**

$$k_i = f(t_i + \Delta t \alpha_i, x_i + \Delta t \sum_{j=1}^{i-1} \beta_{ij} k_j), \alpha_1 = 0$$



## Implicit method:

$$k_i = f\left(t_i + \Delta t \alpha_i, x_i + \Delta t \sum_{j=1}^q \beta_{ij} k_j\right)$$

- nonlinear set of equations in  $k_i$
- explicit is a special case

Order  $p$ : *exact* solution up to polynomial of order  $p$

$\Rightarrow$  can determine  $\alpha_i, \omega_i, \beta_{ij}$

For explicit method of order  $p$ , at least  $q_{min}(p)$  stages are required:

$p$	1	2	3	4	5	6	7	...
$q_{min}(p)$	1	2	3	4	6	7	9	...

# Integration Methods: Runge-Kutta 4

## Single-step

$$x_0 = \alpha_0$$

$$k_1 = \Delta t f(t_i, x_i)$$

$$k_2 = \Delta t f\left(t_i + \frac{\Delta t}{2}, x_i + \frac{k_1}{2}\right)$$

$$k_3 = \Delta t f\left(t_i + \frac{\Delta t}{2}, x_i + \frac{k_2}{2}\right)$$

$$k_4 = \Delta t f(t_i + \Delta t, x_i + k_3)$$

$$x_{i+1} = x_i + \frac{k_1}{6} + \frac{k_2}{3} + \frac{k_3}{3} + \frac{k_4}{6}, i \geq 0$$

# Integration Methods: Adams-Bashforth

Multi-step: need lower-step methods for *start-up*

## 2-step

$$x_0 = \alpha_0$$

$$x_1 = \alpha_1$$

$$x_{i+1} = x_i + \frac{\Delta t}{2} (3f(t_i, x_i) - f(t_{i-1}, x_{i-1})), \quad i \geq 1$$

## 3-step

$$x_0 = \alpha_0$$

$$x_1 = \alpha_1$$

$$x_2 = \alpha_2$$

$$x_{i+1} = x_i + \frac{\Delta t}{12} (23f(t_i, x_i) - 16f(t_{i-1}, x_{i+1}) + 5f(t_{i-2}, x_{i-2})), \quad i \geq 2$$

## 4-step

$$x_0 = \alpha_0$$

$$x_1 = \alpha_1$$

$$x_2 = \alpha_2$$

$$x_3 = \alpha_3$$

$$x_{i+1} =$$

$$x_i + \frac{\Delta t}{24} (55f(t_i, x_i) - 59f(t_{i-1}, x_{i+1}) + 37f(t_{i-2}, x_{i-2}) - 9f(t_{i-3}, x_{i-3})), \quad i \geq 3$$

# Integration Methods: Milne

## Predictor-corrector

- Predictor

$$x_0 = \alpha_0$$

$$x_1 = \alpha_1$$

$$x_2 = \alpha_2$$

$$x_3 = \alpha_3$$

$$x_{i+1}^{(0)} = x_{i-3} + \frac{4\Delta t}{3} (2f(t_i, x_i) - f(t_{i-1}, x_{i-1}) + 2f(t_{i-2}, x_{i-2})), i \geq 3$$

- Corrector

$$x_{i+1}^{(k+1)} = x_{i-1} + \frac{\Delta t}{3} (f(t_{i+1}, x_{i+1}^{(k)}) + 4f(t_i, x_i) + f(t_{i-1}, x_{i-1})),$$

$$i \geq 2, k = 1, 2, \dots$$

# Adaptive Step-size Control

- want to attain pre-set *minimum (step-wise) accuracy*
- want *minimum computation*

Solution: use accuracy estimate to *double/halve step-size*

Obtaining accuracy estimate:

1. step doubling (e.g., RK4)
2.  $\epsilon_N \cong approx_{N+1} - approx_N$

# Adaptive Step-size Control

RK4 + RK5  $\rightarrow$  Runge-Kutta Fehlberg (embedded)

$$k_1 = \Delta t f(t_i, x_i)$$

$$k_2 = \Delta t f(t_i + a_2 \Delta t, x_i + b_{12} k_1)$$

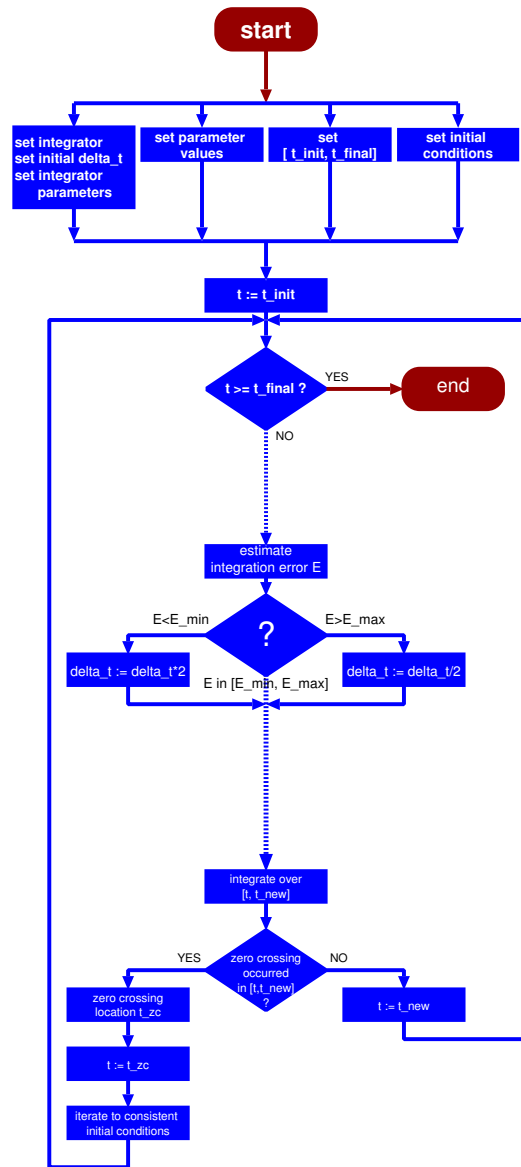
...

$$k_6 = \Delta t f(t_i + a_6 \Delta t, x_i + b_{61} k_1 + b_{62} k_2 + \dots + b_{65} k_5)$$

$$x_{i+1} = x_i + c_1 k_1 + c_2 k_2 + \dots + c_6 k_6 + O(\Delta t^6)$$

$$x_{i+1}^* = x_i + c_1^* k_1 + c_2^* k_2 + \dots + c_6^* k_6 + O(\Delta t^5)$$

$$\epsilon_{estim} = x_{i+1} - x_{i+1}^* = \sum_{i=1}^6 (c_i - c_i^*) k_i$$





# Stiff Systems

$$\begin{aligned}u' &= 998u + 1998v \\v' &= -999u - 1999v \\u(0) &= 1, v(0) = 0\end{aligned}$$

$$\begin{aligned}u &= 2y - z \\v &= -y + z\end{aligned}$$

$$\begin{aligned}u &= 2e^{-t} - e^{-1000t} \\v &= -e^{-t} + e^{-1000t}\end{aligned}$$

# Stiff Systems: solvers

$$x' = -cx$$

- Explicit: Forward Euler:  $x_{i+1} = x_i + \Delta t x'_i$   
 $x_{i+1} = (1 - c\Delta t)x_i$
- Implicit: Backward Euler:  $x_{i+1} = x_i + \Delta t x'_{i+1}$   
 $x_{i+1} = \frac{x_i}{1+c\Delta t}$   
Rosenbrock, Gear, ... methods

# Differential Algebraic Equations (DAE)

$$f\left(\frac{d^n x}{dt^n}, \frac{d^{n-1} x}{dt^{n-1}}, \dots, x, u, t\right) = 0$$
$$g(x, t) = 0$$

Residual Solvers

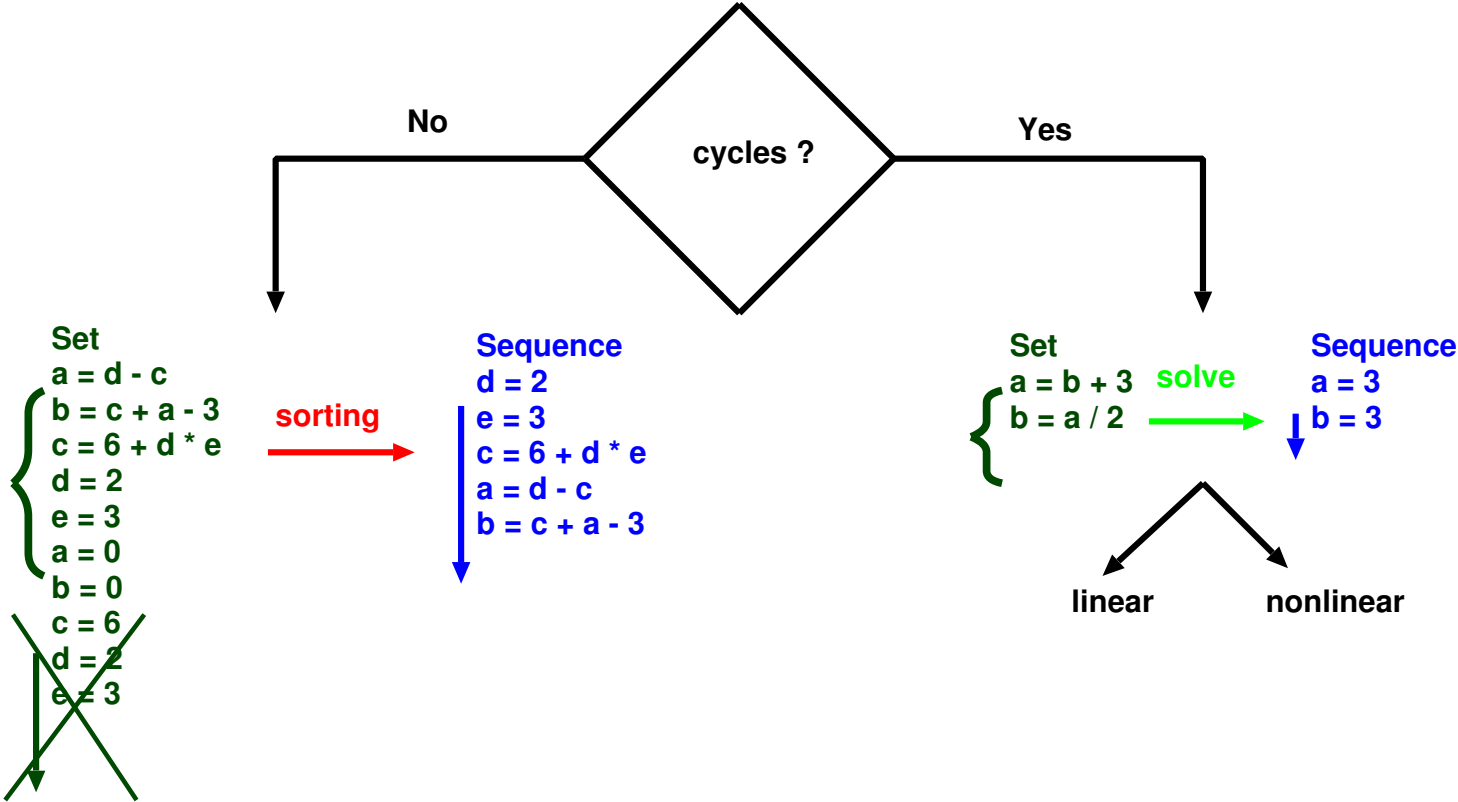
DASSL (Petzold)

<http://www.engineering.ucsb.edu/cse/software.html>

# Causal continuous-time models

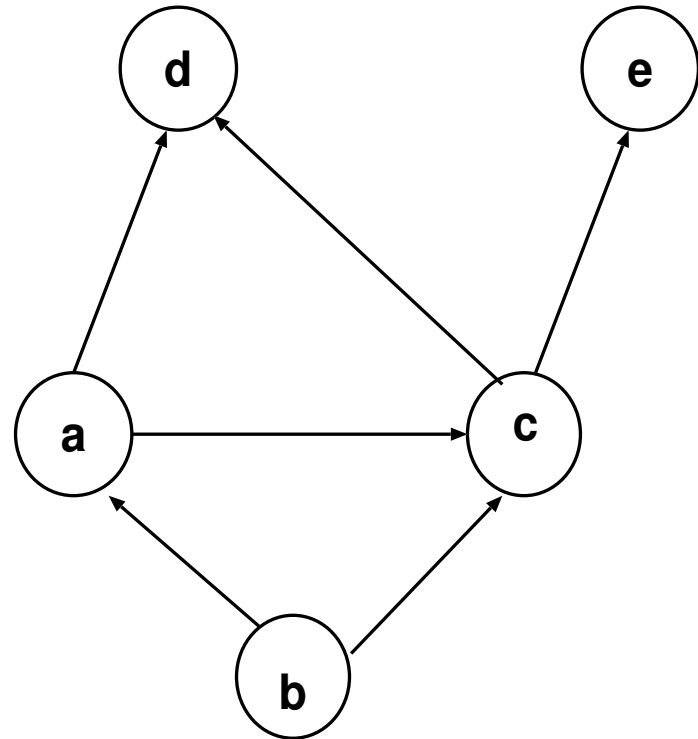
# Problems with algebraic model solving

**DAE-set  $\neq$  DAE-sequence**



# Dependency Graph

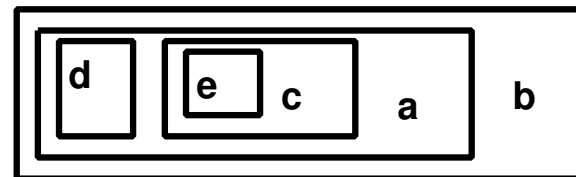
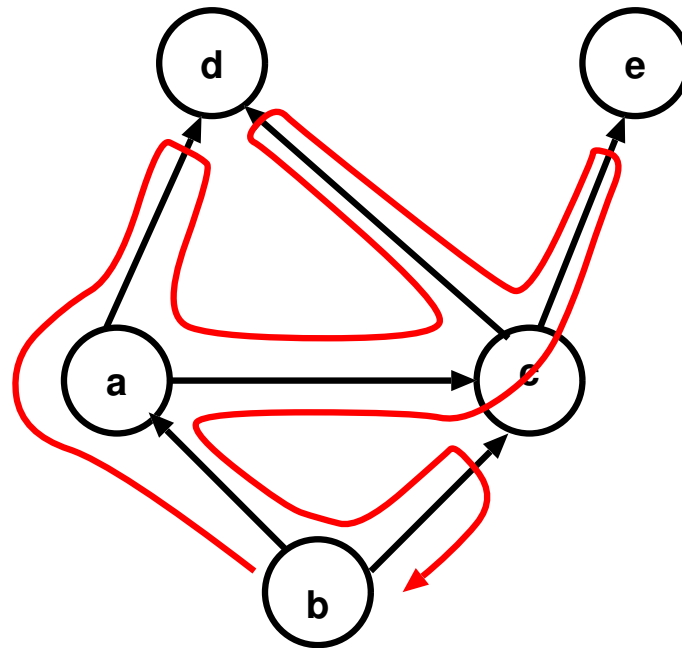
$$\left\{ \begin{array}{l} \mathbf{a = d - c} \\ \mathbf{b = c + a - 3} \\ \mathbf{c = 6 + d * e} \\ \mathbf{d = 2} \\ \mathbf{e = 3} \end{array} \right.$$



# Problems with model re-use: sorting post-order traversal depth-first search

1. find\_root(s)
2. DFS(root)

```
DFS(node)
{
  if (not_visited(node))
  {
    mark_visited(node)
    foreach child_node of node
    {
      DFS(child_node)
    }
    print(node)
  }
}
```



## Dependency Cycle (aka Algebraic Loop)

$$\begin{cases} x = y + 16 \\ y = -x - z \\ z = 5 \end{cases}$$

can *never* be sorted due to a dependency *cycle*

aka *strong component* (every vertex in the component is reachable from every other)

$$x \rightarrow y \rightarrow x$$



May be solved implicitly

$$\begin{cases} z = 5 \\ x - y = -6 \\ x + y = -z \end{cases}$$

Implicit set of  $n$  equations in  $n$  unknowns.

- non-linear  $\rightarrow$  non-linear residual solver.
- linear  $\rightarrow$  numeric or symbolic solution.

May be solved symbolically  
(if linear and not too large)

$$x = \frac{\begin{vmatrix} -6 & -1 \\ -z & 1 \end{vmatrix}}{\begin{vmatrix} 1 & -1 \\ 1 & 1 \end{vmatrix}} = \frac{-6-z}{2}; \quad y = \frac{\begin{vmatrix} 1 & -6 \\ 1 & -z \end{vmatrix}}{\begin{vmatrix} 1 & -1 \\ 1 & 1 \end{vmatrix}} = \frac{6-z}{2}$$

$$\begin{cases} z = 5 \\ x = \frac{-6-z}{2} \\ y = \frac{6-z}{2} \end{cases}$$

# Simple Loop Detection

1. Build dependency matrix  $D$
2. Calculate transitive closure  $D^*$
3. If *True* on diagonal of  $D^*$ , a loop exists

Even with Warshall's algorithm, still  $O(n^3)$  and don't know immediately which nodes involved in the loop(s).

# Tarjan's $O(n + m)$ Loop Detection (1972)

1. Complete Depth First Search (DFS) on  $G$   
(possibly multiple DFS trees), postorder numbering

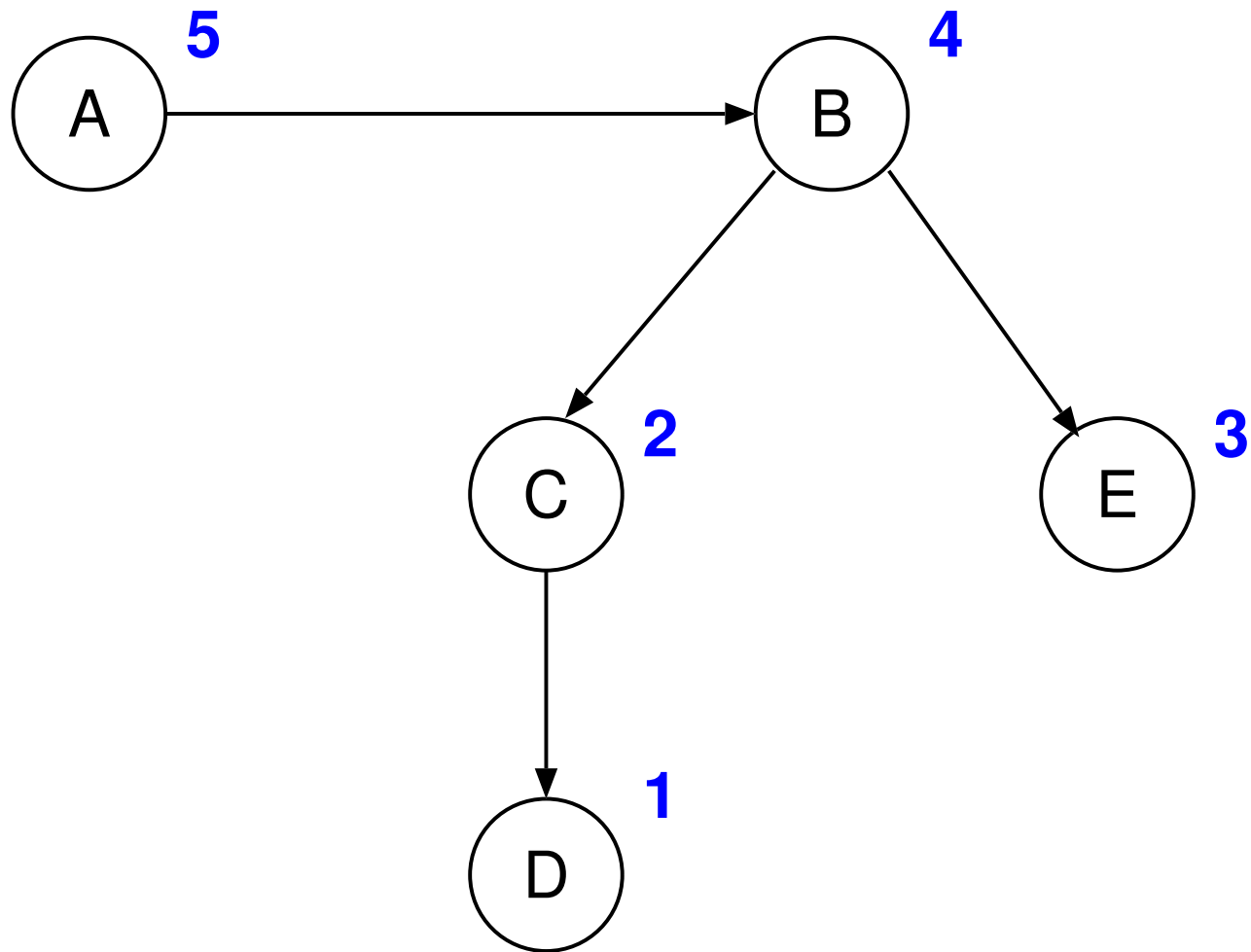
```
FOREACH  $v$  IN  $V$ 
  dfsNr[ $v$ ]  $\leftarrow$  0
FOREACH  $v$  IN  $V$ 
  IF dfsNr[ $v$ ] == 0
    DFS( $v$ )
```

2. Reverse edges in the annotated  $G \rightarrow G_R$
3. DFS on  $G_R$  starting with highest numbered  $v$   
set of vertices in each DFS tree = strong component.  
Remove strong component and repeat.

## Set of Algebraic Eqns, no Loops

$$\left\{ \begin{array}{l} a = b^2 + 3 \\ b = \sin(c \times e) \\ c = \sqrt{d - 4.5} \\ d = \pi/2 \\ e = u() \end{array} \right.$$

# Sorting, no Loops



# Sorting Result

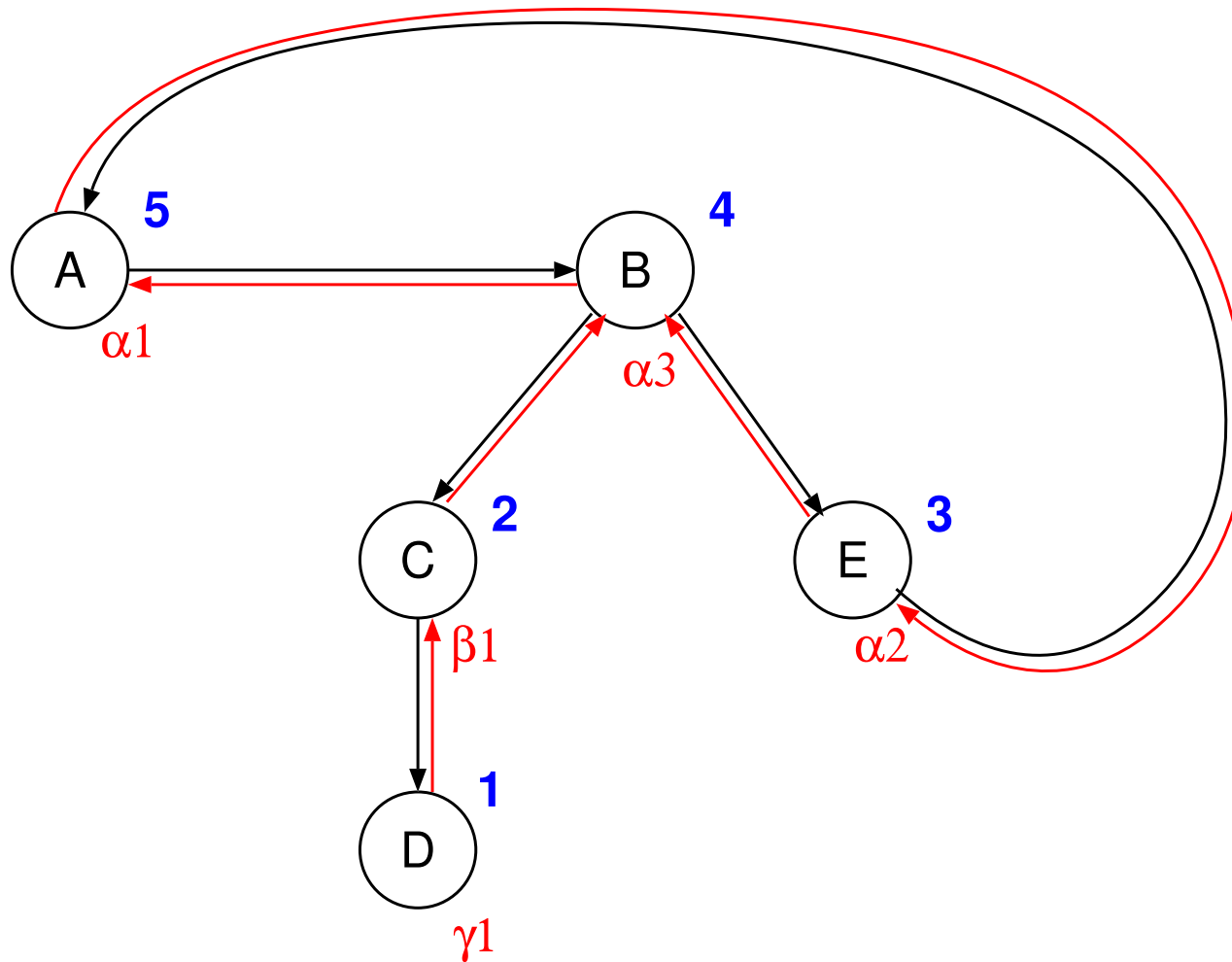
$$\left[ \begin{array}{l} d = \pi/2 \\ e = u() \\ c = \sqrt{d - 4.5} \\ b = \sin(c \times e) \\ a = b^2 + 3 \end{array} \right.$$

# Algebraic Loop (Cycle) Detection

$$\left\{ \begin{array}{l} a = b^2 + 3 \\ b = \sin(c \times e) \\ c = \sqrt{d - 4.5} \\ d = \pi/2 \\ e = a^2 + u() \end{array} \right.$$



# Algebraic Loop (Cycle) Detection



## Algebraic Loop (Cycle) Detection Result

$$\left[ \begin{array}{l} d = \pi/2 \\ c = \sqrt{d-4.5} \\ \left\{ \begin{array}{l} b = \sin(c \times e) \\ a = b^2 + 3 \\ e = a^2 + u() \end{array} \right. \end{array} \right. ; \left[ \begin{array}{l} d = \pi/2 \\ c = \sqrt{d-4.5} \\ \left\{ \begin{array}{l} b - \sin(c \times e) = 0 \\ a - b^2 - 3 = 0 \\ a^2 - e + u() = 0 \end{array} \right. \end{array} \right.$$

# Continuous System Simulation Languages (CSSLs)

- Analog Computers
- block oriented vs. equation based
- the CSi 1968 CSSL standard
- CSSL-IV, ACSL, ADSIM/RT, ...

# CSSL Requirements

- Easy Model Description (equation based, block oriented)
- Integrator control:
  - select integrator
  - (initial) step size
  - error control
  - variable initialisation
  - parameter setting
- Documentation of model and experiments
- Structured: model vs. experiments (re-use)

# Model Description

$$DX = \text{INTEG}(F - B * X - A * DX, DX0)$$

$$X = \text{INTEG}(DX, X0)$$

**or**

$$DX' = F - B * X - A * DX$$

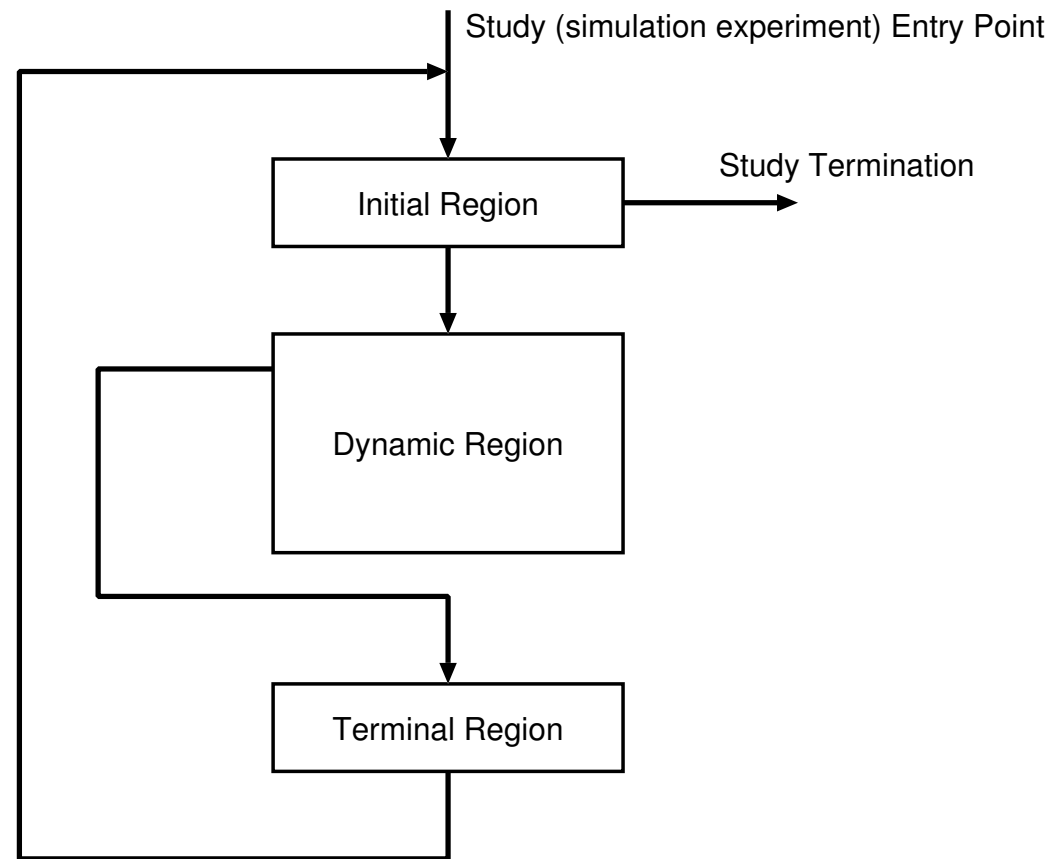
$$X' = DX$$

Initial Conditions at  $t = 0$ :

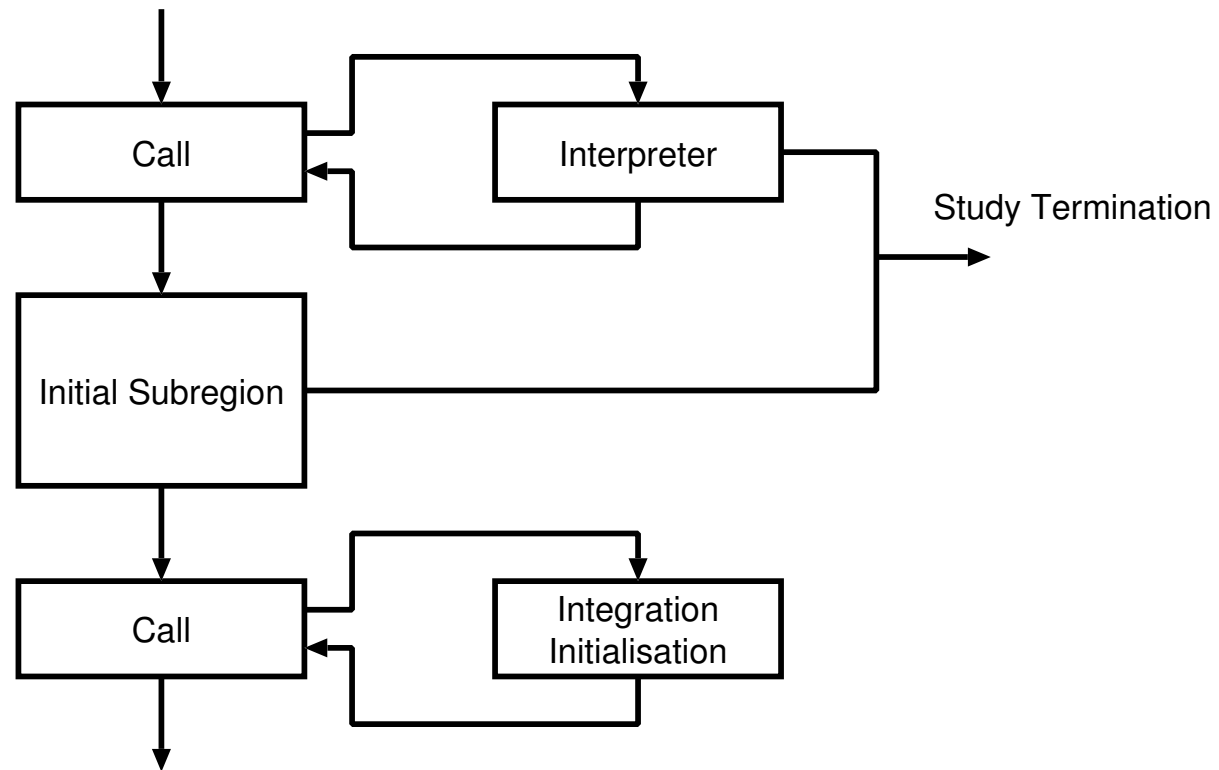
$$X = X0$$

$$DX = DX0$$

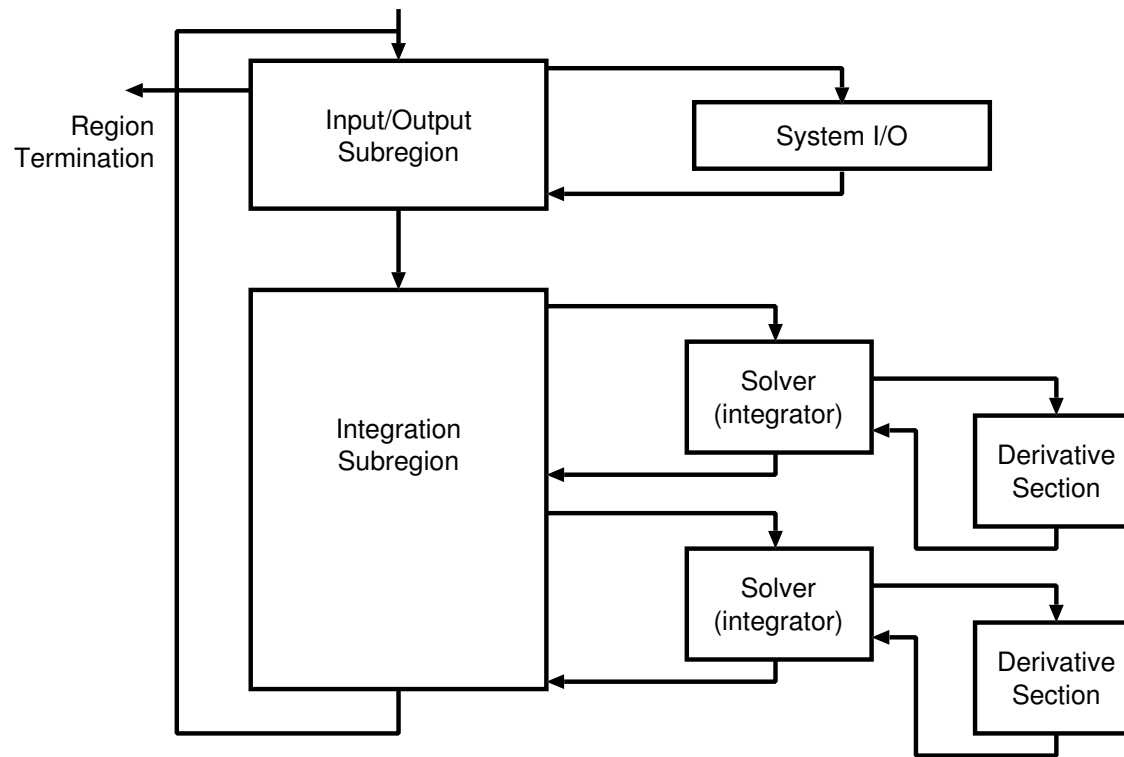
# “CSSL study” structure



# “CSSL initial region” structure

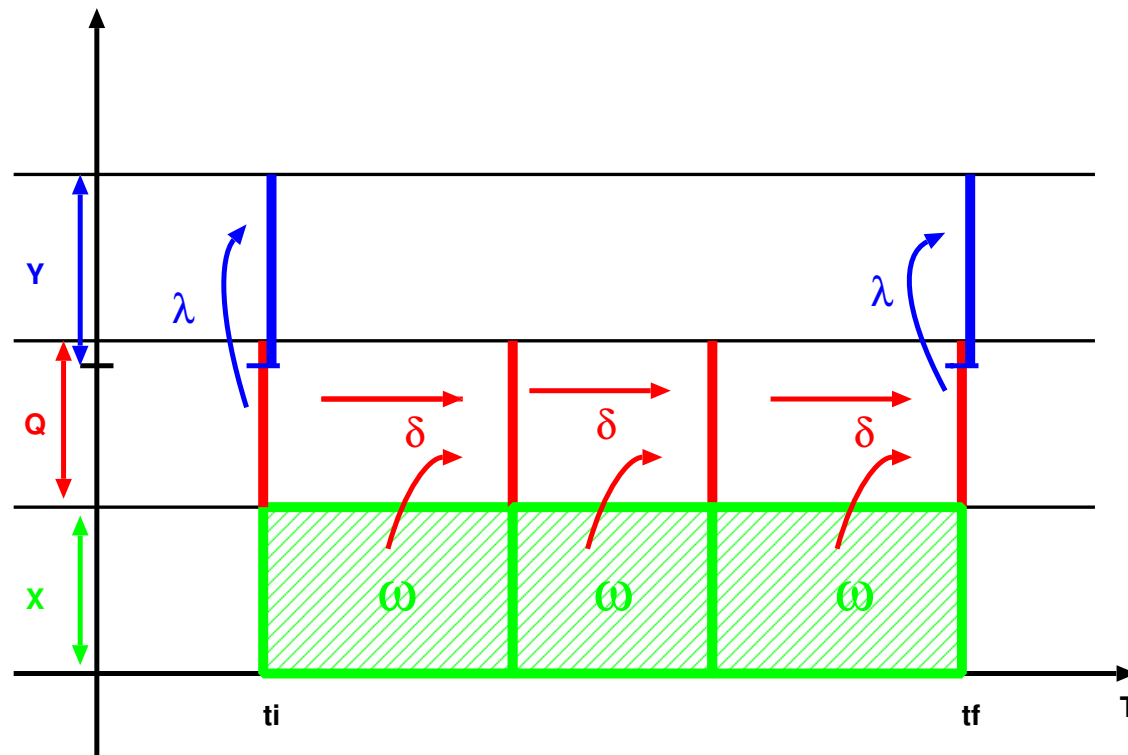


# “CSSL dynamic region” structure

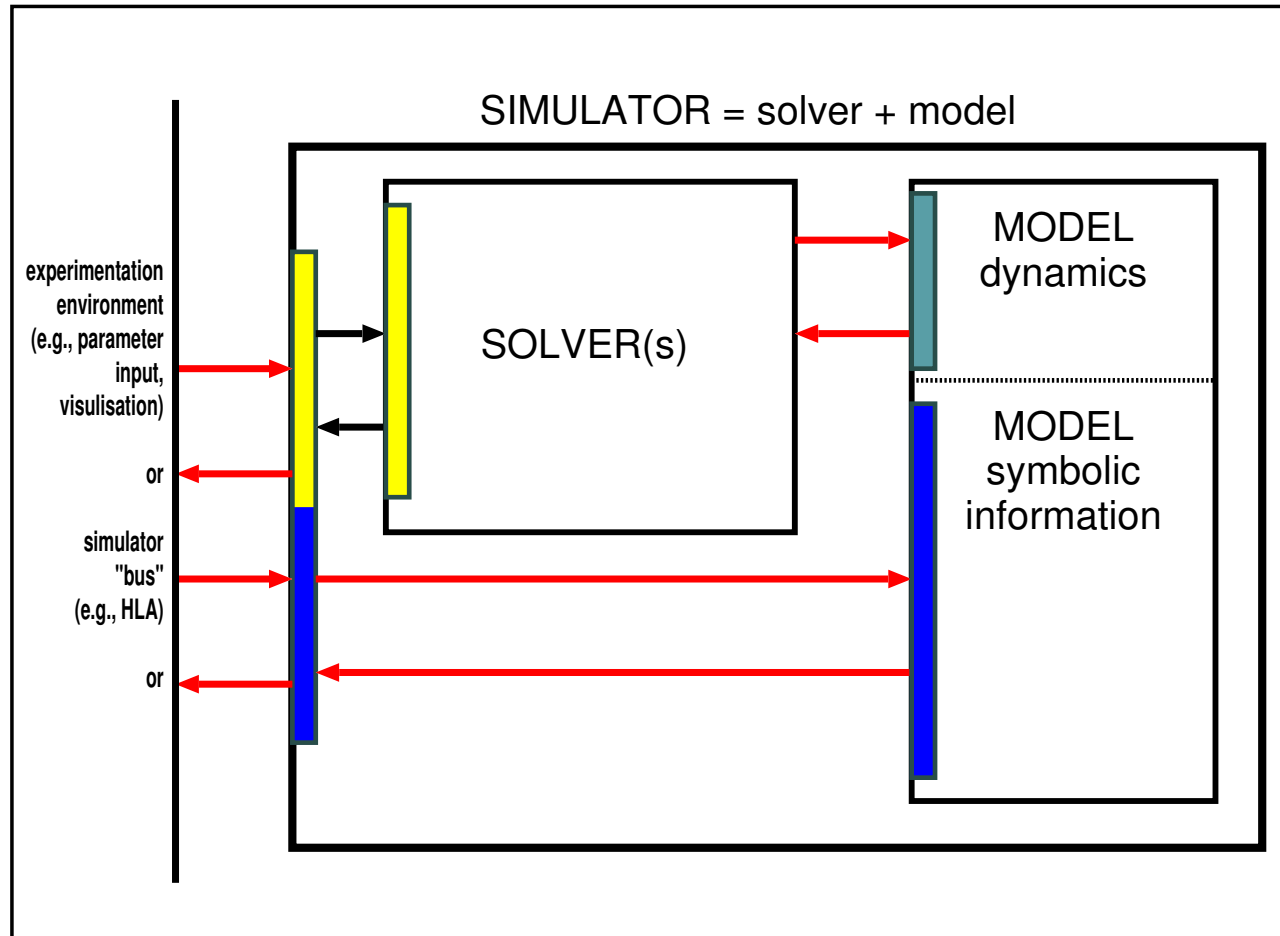




# General, state-based simulation kernel



# Model-solver Architecture



# MSL-EXEC Model Representation

```

#include <math.h>
#include <assert.h>
#include "MSLE.h"
#include "MSLEExternal.h"
#include "MSLU.h"
#include "Circle.h"

#define _t_ IndepVarValues[0]
#define _x_out_ OutputVarValues[0]
#define _y_out_ OutputVarValues[1]
#define _x_ DerStateVarValues[0]
#define _y_ DerStateVarValues[1]
#define _D_x_ Derivatives[0]
#define _D_y_ Derivatives[1]

CircleClass :: CircleClass(StringType name_arg)
{
    set_name(name_arg);
    set_description("Circle test.");
    set_class_name("CircleClass");

    set_no_indep_vars(1);
    set_indep_var(0, new MSLEIndepVarClass("t", "s"));

    set_no_output_vars(2);
    set_output_var(0, new MSLEOutputVarClass("x_out", "", 0));
    set_output_var(1, new MSLEOutputVarClass("y_out", "", 0));

    set_no_der_state_vars(2);
    set_der_state_var(0, new MSLEDerStateVarClass("x", "", 0.1));
    set_der_state_var(1, new MSLEDerStateVarClass("y", "", 0.1));
}

```

```
set_no_indep_var_values(1);
GetIndepVar(0)->LinkValue(this, MSLE_INDEP_VAR, 0);

set_no_output_var_values(2);
GetOutputVar(0)->LinkValue(this, MSLE_OUTPUT_VAR, 0);
GetOutputVar(1)->LinkValue(this, MSLE_OUTPUT_VAR, 1);

set_no_der_state_var_values(2);
GetDerStateVar(0)->LinkValue(this, MSLE_DER_STATE_VAR, 0);
GetDerStateVar(1)->LinkValue(this, MSLE_DER_STATE_VAR, 1);
GetDerStateVar(0)->LinkInitialValue(this, 0);
GetDerStateVar(1)->LinkInitialValue(this, 1);
GetDerStateVar(0)->LinkDerivative(this, 0);
GetDerStateVar(1)->LinkDerivative(this, 1);

Reset();
}
```

```
void CircleClass :: ComputeOutput(void)
{
    _x_out_ = _x_;
    _y_out_ = _y_;
}

void CircleClass :: ComputeInitial(void)
{
}

void CircleClass :: ComputeState(void)
{
    _D_x_ = _y_;
    _D_y_ = -_x_;
}

void CircleClass :: ComputeTerminal(void)
{
}

#undef _t_
#undef _x_out_
#undef _y_out_
#undef _x_
#undef _y_
```

# MSL-EXEC simulator demo

# GUI

