

Component PERCUSSION

episode II



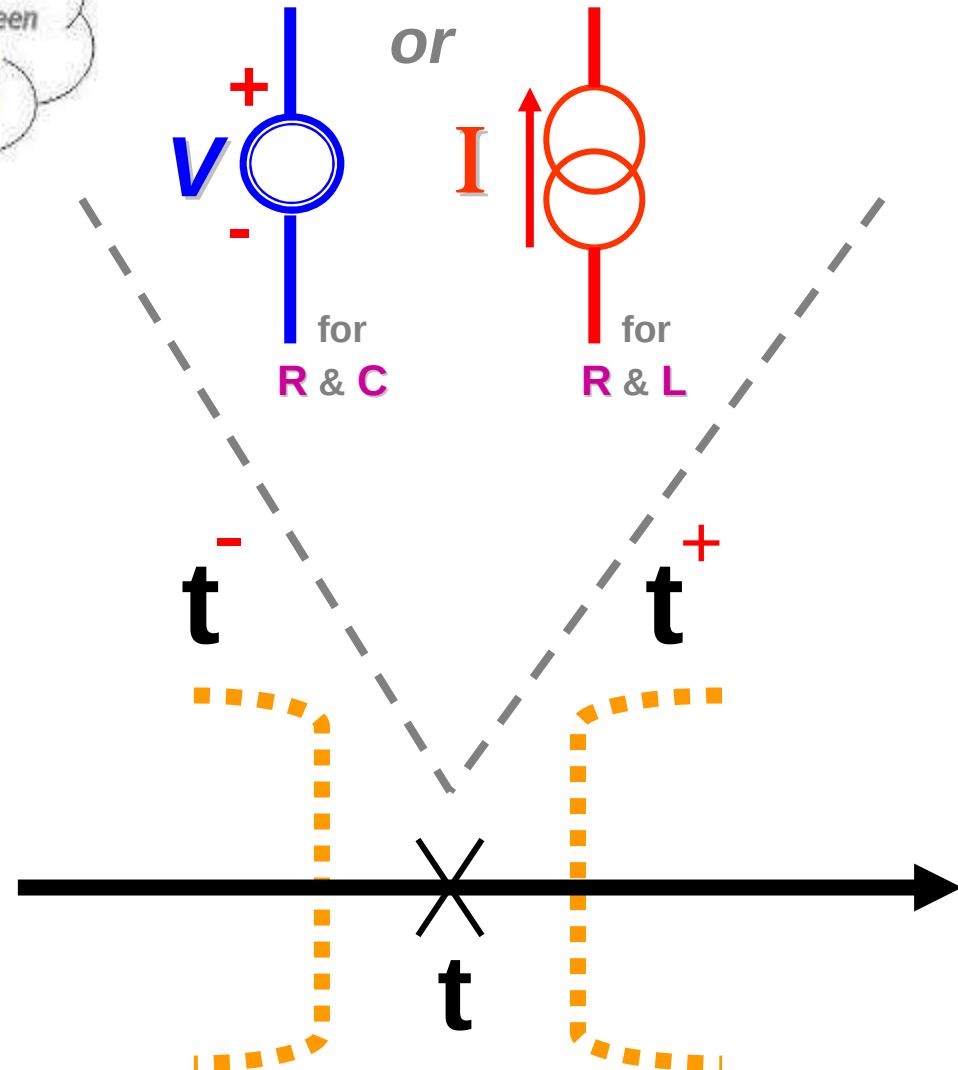


***In the previous
episode ...***



Differential equations by means of *Distributions*

percuted component

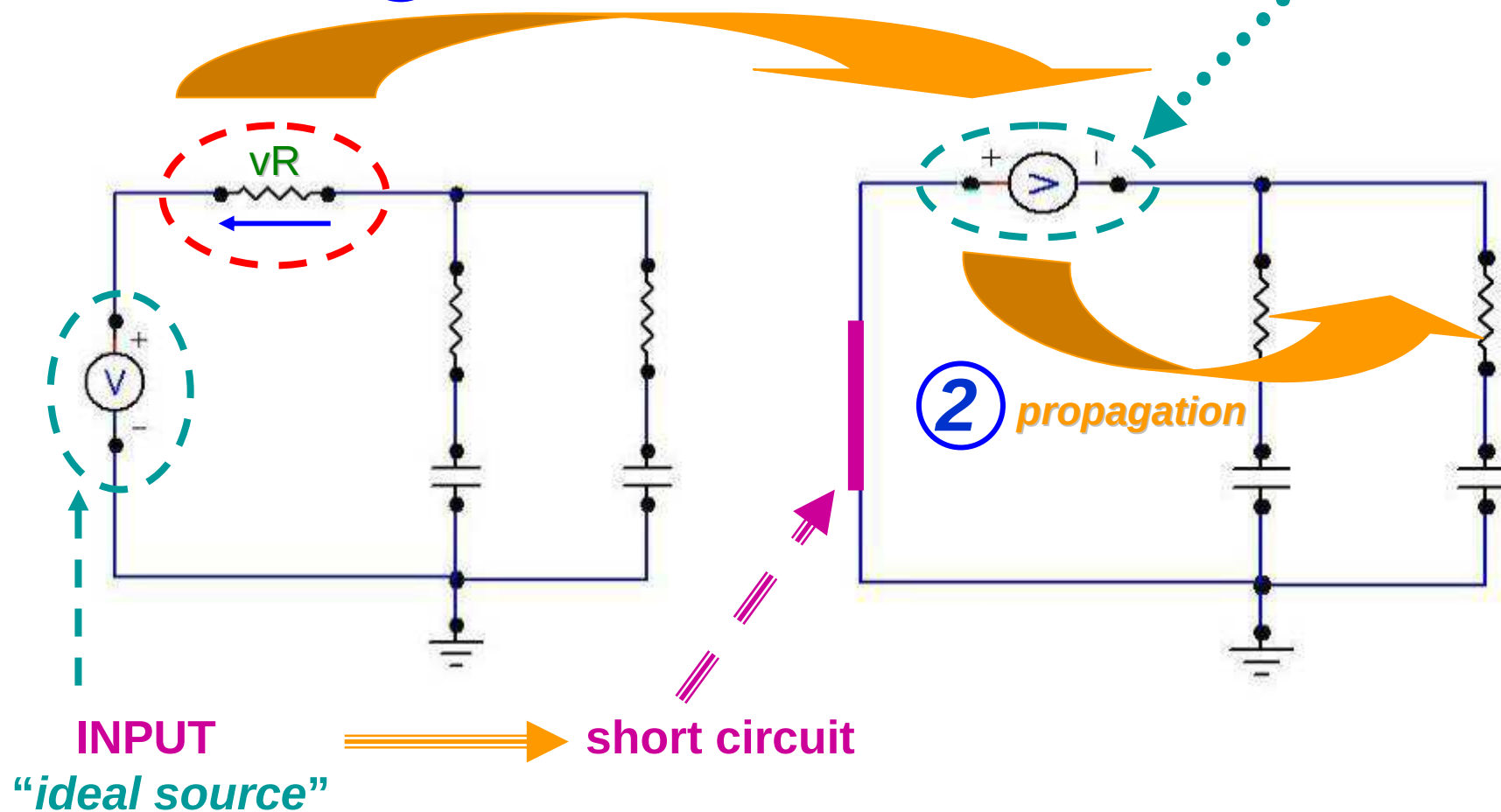


2 stages mystery . . .

① PERCUSSION

"ideal source"
but for a null time !!!

$f(vR, \text{filter coeffs})$





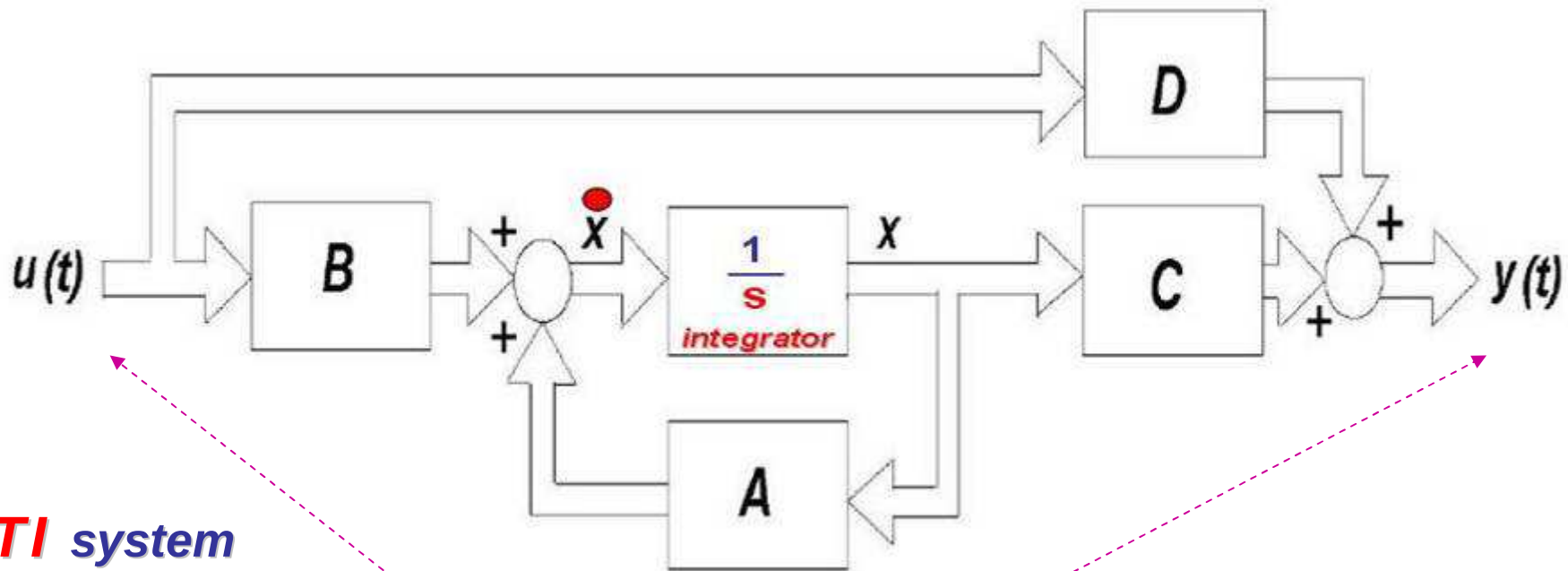
Semi-analytical recursive algorithms for convolution calculations
W. Janke G. Blakiewicz
Department of Electronics, Technical University of Gdansk, Poland
IEE Proc. Circuits Devices Syst., Vol. 142 , No. 2, April 1995

STATE MODEL

MATRICES & VECTORS

$$\dot{\mathbf{x}}(t) = \mathbf{A} \mathbf{x}(t) + \mathbf{B} \mathbf{u}(t)$$

$$\mathbf{y}(t) = \mathbf{C} \mathbf{x}(t) + \mathbf{D} \mathbf{u}(t)$$



LTI system
block diagram

multiple **inputs** / multiple **outputs** (**MIMO**)



Rudolf (Rudy) Emil Kálmán

born May 19, 1930

awarded in 2009 by

President Barack Obama with the
National Medal of Science

$$\begin{aligned}\dot{\mathbf{x}}(t) &= \mathbf{A} \mathbf{x}(t) + \mathbf{B} \mathbf{u}(t) \\ \mathbf{y}(t) &= \mathbf{C} \mathbf{x}(t) + \mathbf{D} \mathbf{u}(t)\end{aligned}$$

In the 60's, based on **Kálmán** research,
the **State Model** approach has been fully extended
and used, for examples, on the **Apollo** and **Polaris** projects

For a given system, the **transfer function** output / input is unique
but multiple set of **A, B, C, D** matrices can be expressed
possibly corresponding to different **orders** (and thus different number of **poles** !!!)

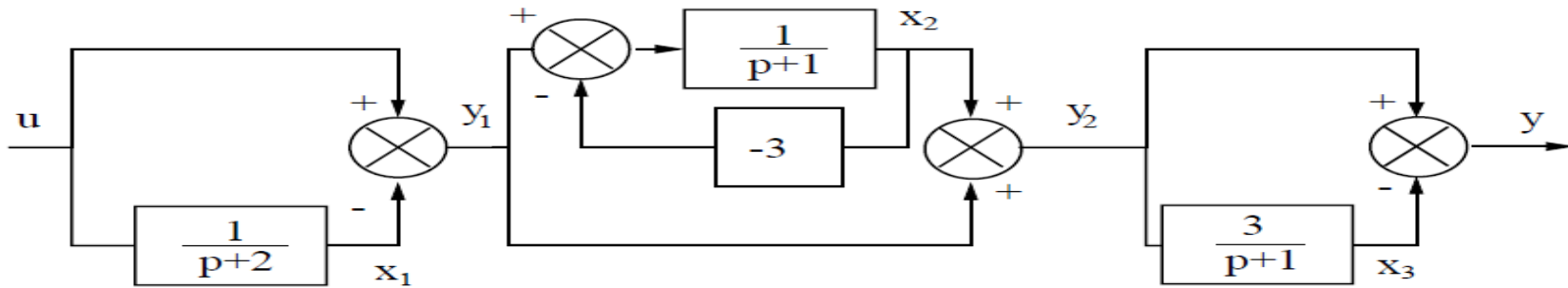
But, the **minimal order** only corresponds to the **A matrices** that enforce both
Controllable and **Observable** **Kálmán criteria**

It also exists, for singular systems, a “**generalized state representation**”

$$\begin{aligned}\mathbf{E} \dot{\mathbf{X}} &= \mathbf{A} \mathbf{X} + \mathbf{B} \mathbf{U} \\ \mathbf{Y} &= \mathbf{C} \mathbf{X} + \mathbf{D} \mathbf{U}\end{aligned}$$

E is called the “**derivative**” matrix

for **non-singular** systems **E = Identity**



system “**external**” VIEW

the **transfer function** expression is **unique**

1 pole: -2

$$\frac{Y(p)}{U(p)} = \frac{p-1}{p+2}$$

pole-zero cancellations
has occurred

The **output / input differential equation**

2 poles: -2 , -1

$$\ddot{y} + 3\dot{y} + 2y = \ddot{u} - u$$

system “**internal**” VIEW

state equation

matrix **A**

3 poles: -2 , -1 , 2

$$\dot{x} = \begin{bmatrix} -2 & 0 & 0 \\ -1 & 2 & 0 \\ -3 & 3 & -1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix} u$$

one of the possible “**not minimal**” state equation

1 pole: -2

$$\dot{x} = [-2] x + [1] u$$

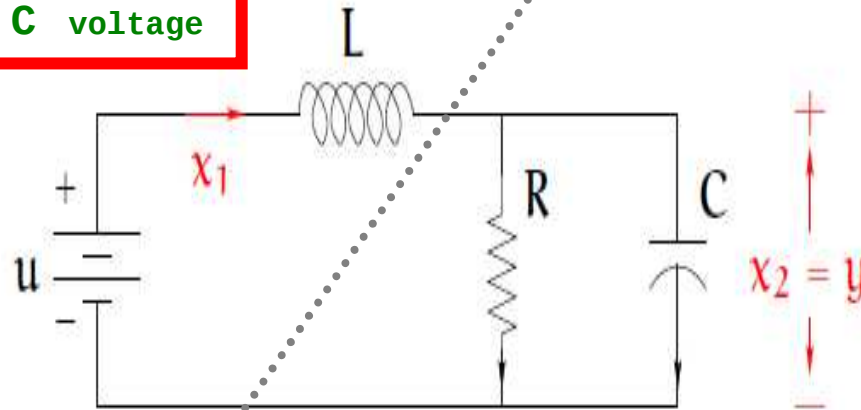
“**minimal**” state equation

Equivalent **STATE** equations

STATES can be:

L current, **C** voltage, LOOP current,
NODE voltage, **L** flux, **C** charge, etc ...

L current
C voltage

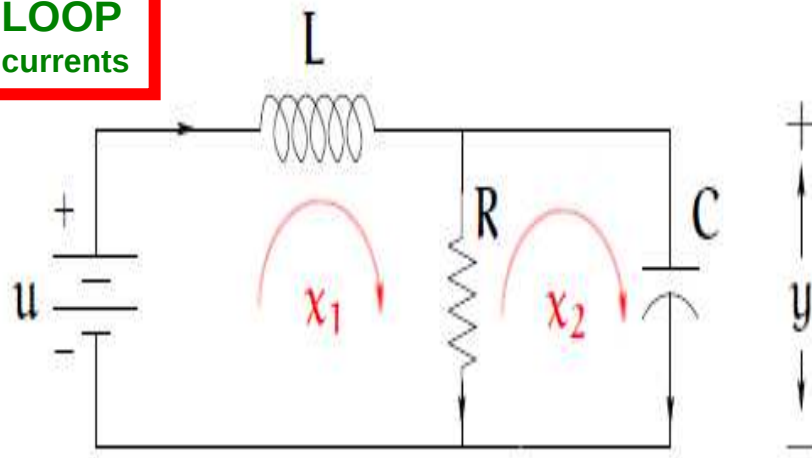


$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$$

$$y = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$R = 1\Omega \quad L = 1H \quad C = 1F$$

LOOP
currents



$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u$$

$$y = \begin{bmatrix} 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

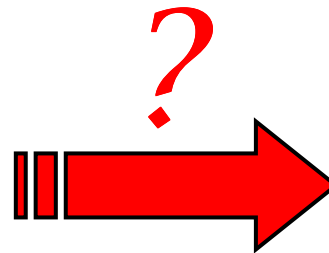
STATE

equivalence transformation

$$Z = T X \iff X = T^{-1} Z$$

Algebraic Equivalence

$$\begin{aligned} \dot{X} &= A X + B U \\ Y &= C X + D U \end{aligned}$$



$$\begin{aligned} \dot{Z} &= \tilde{A} Z + \tilde{B} U \\ Y &= \tilde{C} Z + \tilde{D} U \end{aligned}$$

$$\tilde{A} = T A T^{-1}$$

$$\tilde{B} = T B$$

$$\tilde{C} = C T^{-1}$$

$$\tilde{D} = D$$

STATE MODEL

STATE equation

OUTPUT equation

$$\dot{x}(t) = Ax(t) + Bu(t)$$

$$y(t) = Cx(t) + Du(t)$$

WITH

vector of signals
having derivative
due to inductance
or capacitance

A
B
C
D

X

U

Y

state MATRIX

input MATRIX

output MATRIX

direct transmission MATRIX

(or feedthrough , feedforward)

state VECTOR

input (or control) VECTOR

output VECTOR

This formalism really highlights the **Linear Time Invariant** terminology

It is mostly used by the “**automatism**” world (“**control command**”)

$$\dot{\mathbf{x}}(t) = \mathbf{A} \mathbf{x}(t) + \mathbf{B} \mathbf{u}(t)$$

$$\mathbf{y}(t) = \mathbf{C} \mathbf{x}(t) + \mathbf{D} \mathbf{u}(t)$$

DIMENSIONS

MATRICES

VECTORS

Variable	Dimension	Name
s		Number of states
i		Number of inputs
o		Number of outputs
A	$s \times s$ matrix	State matrix
B	$s \times i$ matrix	Input matrix
C	$o \times s$ matrix	Output matrix
D	$o \times i$ matrix	Direct transmission matrix
$\mathbf{x}$	s vector	State vector
$\mathbf{u}$	i vector	Input vector
$\mathbf{y}$	o vector	Output vector

STATE equation

$$\dot{X} = A X + B U$$

not easy to compute (searching X involves a “convolution”)

order of matrix A is **nearly** related to the number of *Inductances* and *Capacitances*

When *order* $\neq$ *number energy-storing elements* the network is called “degenerate network”

The *order* of complexity of any network equals the total number of *energy-storing elements* minus the *number of all-capacitor “loops”* and the *number of all-inductor “cut-sets”*

OUTPUT equation

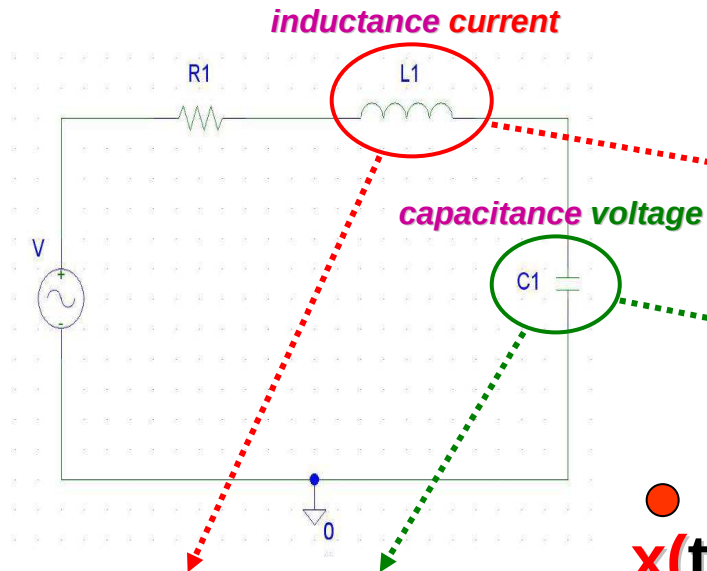
$$Y = C X + D U$$

trivial to compute as soon as X is known

able to return any *voltage* or *current* node similarly to **Kirchhoff / Ohm**

order of *vector* Y depends on the number of *expected observed values*

STATE MODEL



$$\dot{x}(t) = A x(t) + B u(t)$$

$$y(t) = C x(t) + D u(t)$$

$$L1 \frac{d iL1(t)}{d t} = - R1 iL1(t) - vC1(t) + u(t)$$

$$C1 \frac{d vC1(t)}{d t} = iL1(t)$$

2 impedance components

state dimension 2

$$\begin{pmatrix} \frac{d iL1(t)}{d t} \\ \frac{d vC1(t)}{d t} \end{pmatrix} = \begin{pmatrix} -\frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{pmatrix} \begin{pmatrix} iL1(t) \\ vC1(t) \end{pmatrix} + \begin{pmatrix} \frac{1}{L1} \\ 0 \end{pmatrix} u(t)$$

Now, let's suppose that
for the $y(t)$ output vector
we want to observe $vR1(t)$

from the
schematic

$$vR1(t) = R1 \, iL1(t)$$

$$vR1(t) = \begin{bmatrix} R1 \\ 0 \end{bmatrix} \begin{bmatrix} iL1(t) \\ vC1(t) \end{bmatrix} + \begin{bmatrix} 0 \end{bmatrix} u(t)$$

$y(t) \quad C \quad x(t) \quad D \quad u(t)$

The resulting 4 matrices $A \, B \, C \, D$ are *intrinsically characteristic* of our *circuit*

$$A = \begin{bmatrix} -\frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{bmatrix} \quad B = \begin{bmatrix} \frac{1}{L1} \\ 0 \end{bmatrix} \quad C = \begin{bmatrix} R1 \\ 0 \end{bmatrix} \quad D = \begin{bmatrix} 0 \end{bmatrix}$$

the **STATE MODEL**

is giving us the opportunity to

answer to the **frightening question** :

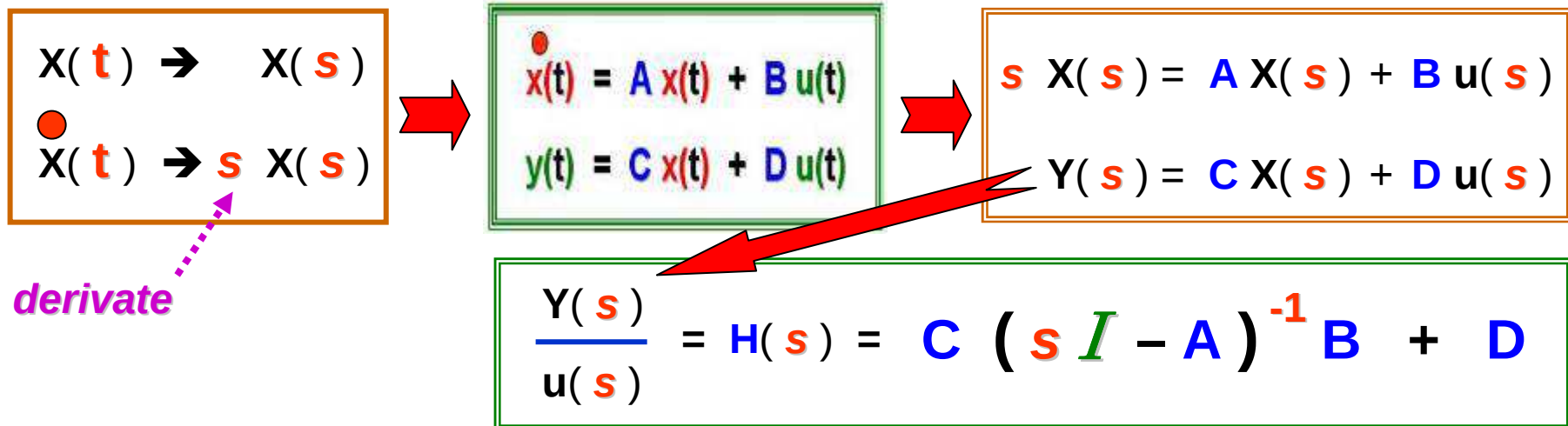
What is the rational behind

transfer function denominator ???



We claimed that the **4** matrices **A B C D** are **characteristic** of the **circuit**
 does that means that they are **independent** of the “**processing domain**” ???

In other words, are they **meaningful** as well, for example, to the **Laplace “S” domain**



$$H(s) = \frac{C [\text{cof}(s I - A)]^T B + D Q(s)}{Q(s)}$$

with $Q(s) = \det(s I - A)$ **Characteristic Polynomial** of matrix **A**

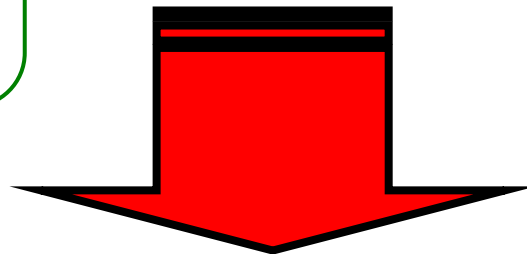
This implies that the **zeros** of this **Characteristic Polynomial** of matrix **A**

$$Q(s) = \det(sI - A) \quad \text{are also the POLES of } H(s)$$

and so ... the **POLES** of **H(s)** are from the expression of **Q(s)**
the **Eigen values** of matrix **A** (“valeurs propres” in french)

$$A = \begin{pmatrix} -\frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{pmatrix}$$

Eigen values



$$-\frac{R1}{2L1} - \frac{\sqrt{C1^2 R1^2 - 4C1L1}}{2C1L1}$$

$$-\frac{R1}{2L1} + \frac{\sqrt{C1^2 R1^2 - 4C1L1}}{2C1L1}$$

POLES of the **Transfer Function** **H(s)**

$$A = \begin{pmatrix} -\frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{pmatrix}$$

Characteristic Polynomial of matrix **A**

$$Q(s) = \det(sI - A)$$

$$Q(s) = s^2 + \frac{R1}{L1}s + \frac{1}{C1L1}$$

please note that such **theoretical denominator** expression

is **literally** $(s - p1) * (s - p2)$

and thus has an **unit** s^2 coefficient

$$= \frac{C1 L1 s^2 + C1 R1 s + 1}{C1 L1}$$

If you remember, this is effectively **coherent** with what we have found for the "**denominators**" when we used the **Kirchhoff laws** to compute **Laplace Transfer Functions**

$$\begin{aligned} \frac{vr1}{V} &= \frac{C1 R1 s}{C1 L1 s^2 + C1 R1 s + 1} \\ \frac{il1}{V} &= \frac{C1 s}{C1 L1 s^2 + C1 R1 s + 1} \\ \frac{vc1}{V} &= \frac{1}{C1 L1 s^2 + C1 R1 s + 1} \end{aligned}$$

So, what have we learned ???

The **polynomial** that, *for a given circuit*, sounds “*strangely*” reappearing on the **denominator** of its various **output Transfer Functions** is getting a **justification** based on the

Characteristic Polynomial of the **STATE** matrix **A**

which is **independent** of the **observed OUTPUT**

vr1 →

il1 →

vc1 →

$$\begin{array}{l} \frac{vr1}{V} = \frac{C1 R1 s}{C1 L1 s^2 + C1 R1 s + 1} \\ \frac{il1}{V} = \frac{C1 s}{C1 L1 s^2 + C1 R1 s + 1} \\ \frac{vc1}{V} = \frac{1}{C1 L1 s^2 + C1 R1 s + 1} \end{array}$$

←

←

←



So The expression “**Characteristic Polynomial**” is therefore taking its full sense since it truly **characterizes** the **INTRINSIC STATE** of the circuit

For the fun, let's verify that the **State Model vR1(s) Transfer Function**

is effectively
the same than
the one found
with **Kirchhoff**

$$A = \begin{bmatrix} -\frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{bmatrix} \quad B = \begin{bmatrix} \frac{1}{L1} \\ 0 \end{bmatrix} \quad C = \begin{bmatrix} R1 \\ 0 \end{bmatrix} \quad D = \begin{bmatrix} 0 \end{bmatrix}$$

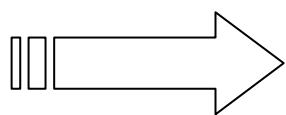
$$\frac{vR1(s)}{u(s)} = \frac{C [\text{cof}(sI - A)]^T B + D Q(s)}{Q(s)}$$

$$Q(s) = s^2 + \frac{R1s}{L1} + \frac{1}{C1L1}$$

$$[\text{cof}(sI - A)]^T =$$

transpose of the MATRIX of cofactors

$$\begin{bmatrix} s & -\frac{1}{L1} \\ \frac{1}{C1} & \frac{L1s + R1}{L1} \end{bmatrix}$$



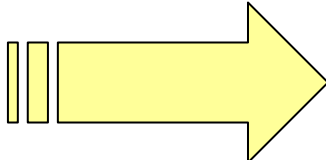
$$\frac{vR1(s)}{u(s)} =$$

$$\frac{\frac{R1s}{L1}}{C1L1s^2 + C1R1s + 1} = \frac{C1R1s}{C1L1s^2 + C1R1s + 1}$$

QED !!!

STATE MODEL



*expression(**t**)*  ***x(t)***

$$\dot{x} = ax + bu \Rightarrow \dot{x}(t) - ax(t) = bu(t)$$

$$e^{-at}(\dot{x}(t) - ax(t)) = e^{-at}(bu(t))$$

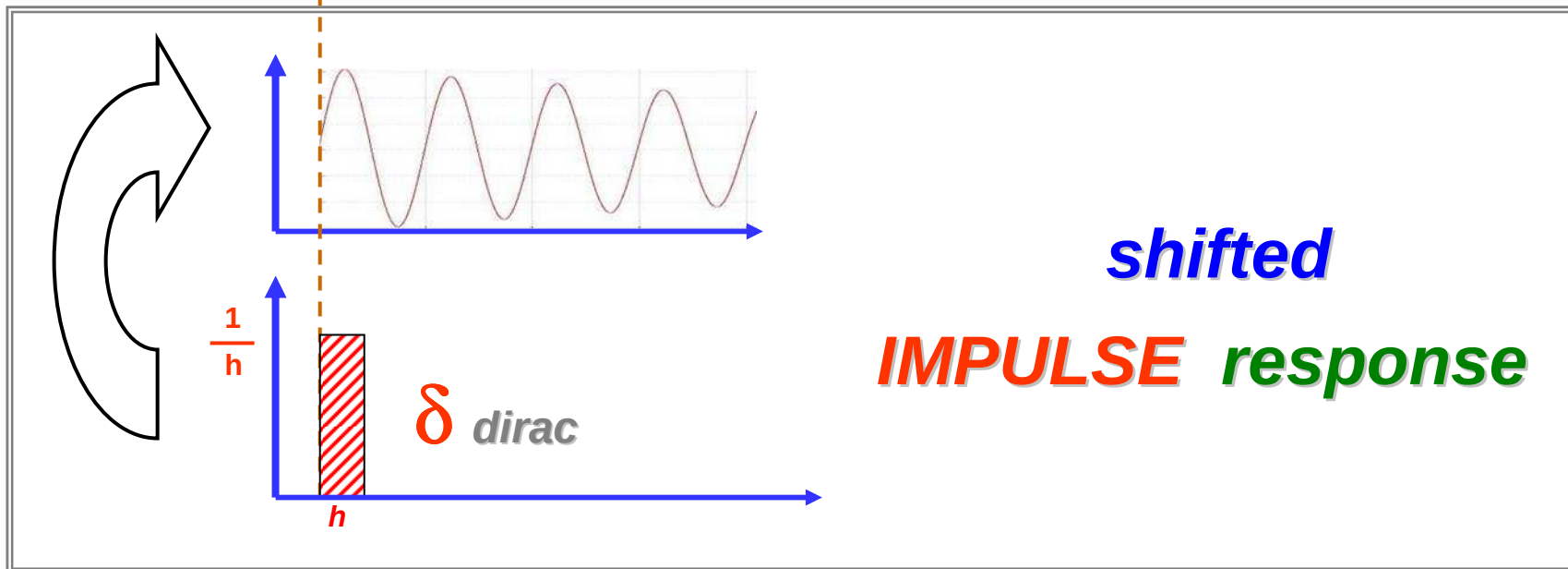
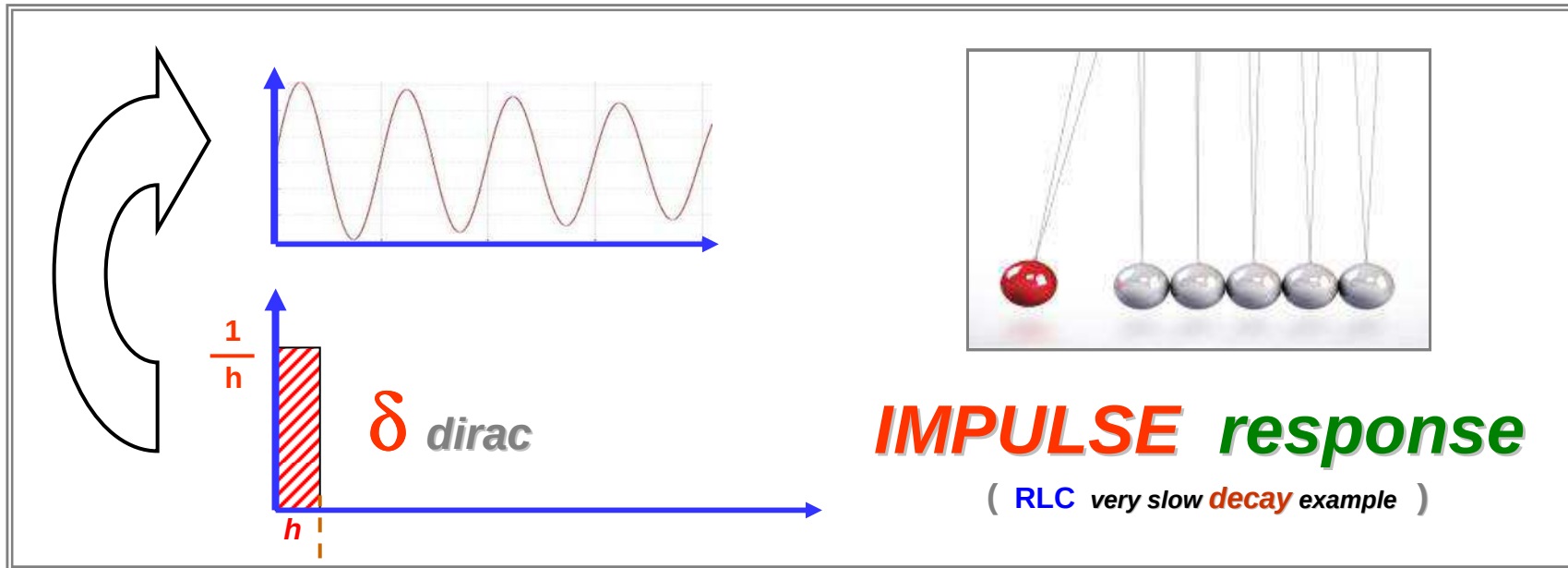
$$e^{-at}\dot{x} - e^{-at}ax = \frac{d}{dt}(e^{-at}x(t)) \Rightarrow \frac{d}{dt}(e^{-at}x(t)) = e^{-at}bu$$

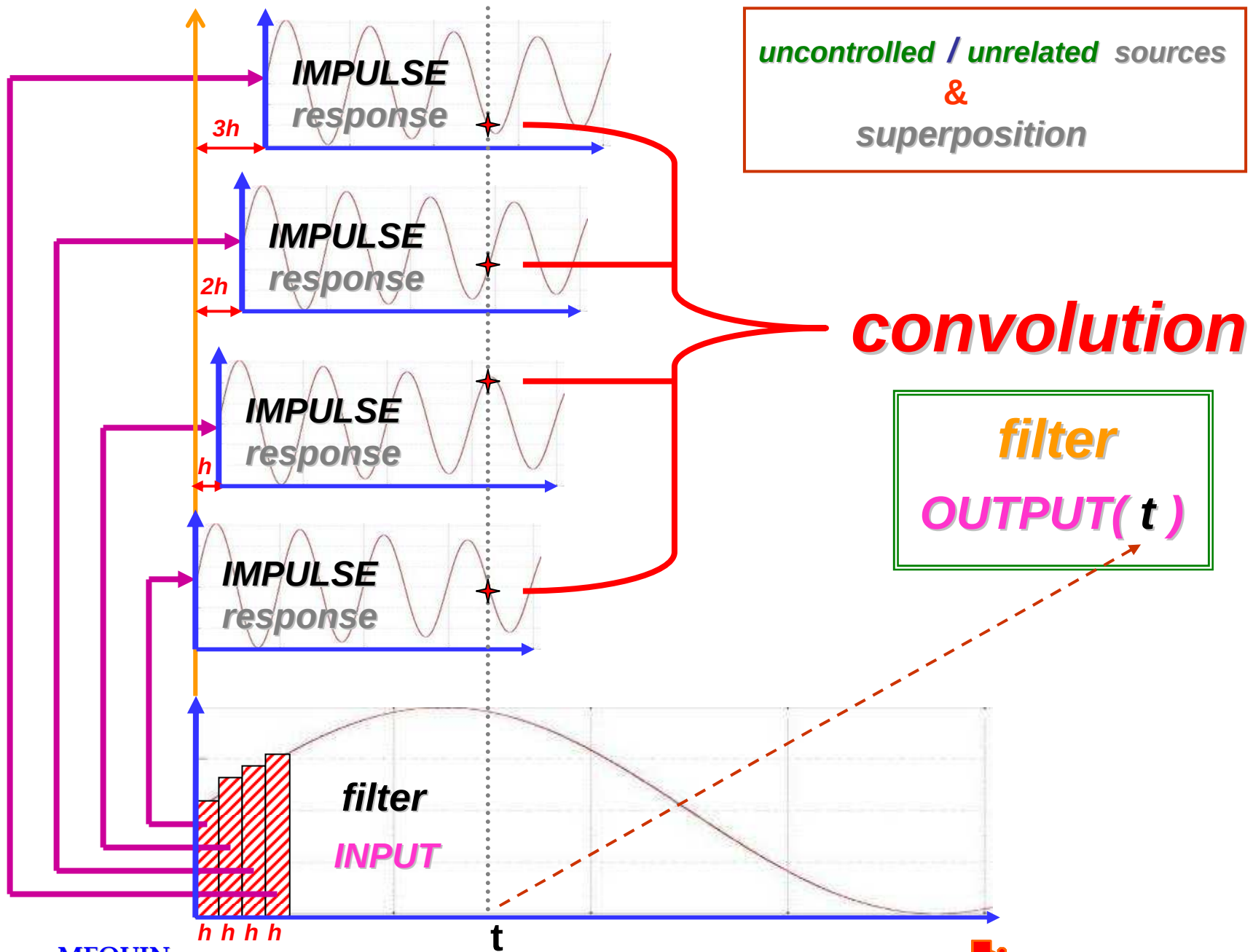
$$\int_0^t \frac{d}{d\tau}(e^{-a\tau}x(\tau)) d\tau = \int_0^t e^{-a\tau}bu(\tau) d\tau \Rightarrow e^{-at}x(t) - x(0) = \int_0^t e^{-a\tau}bu(\tau) d\tau$$

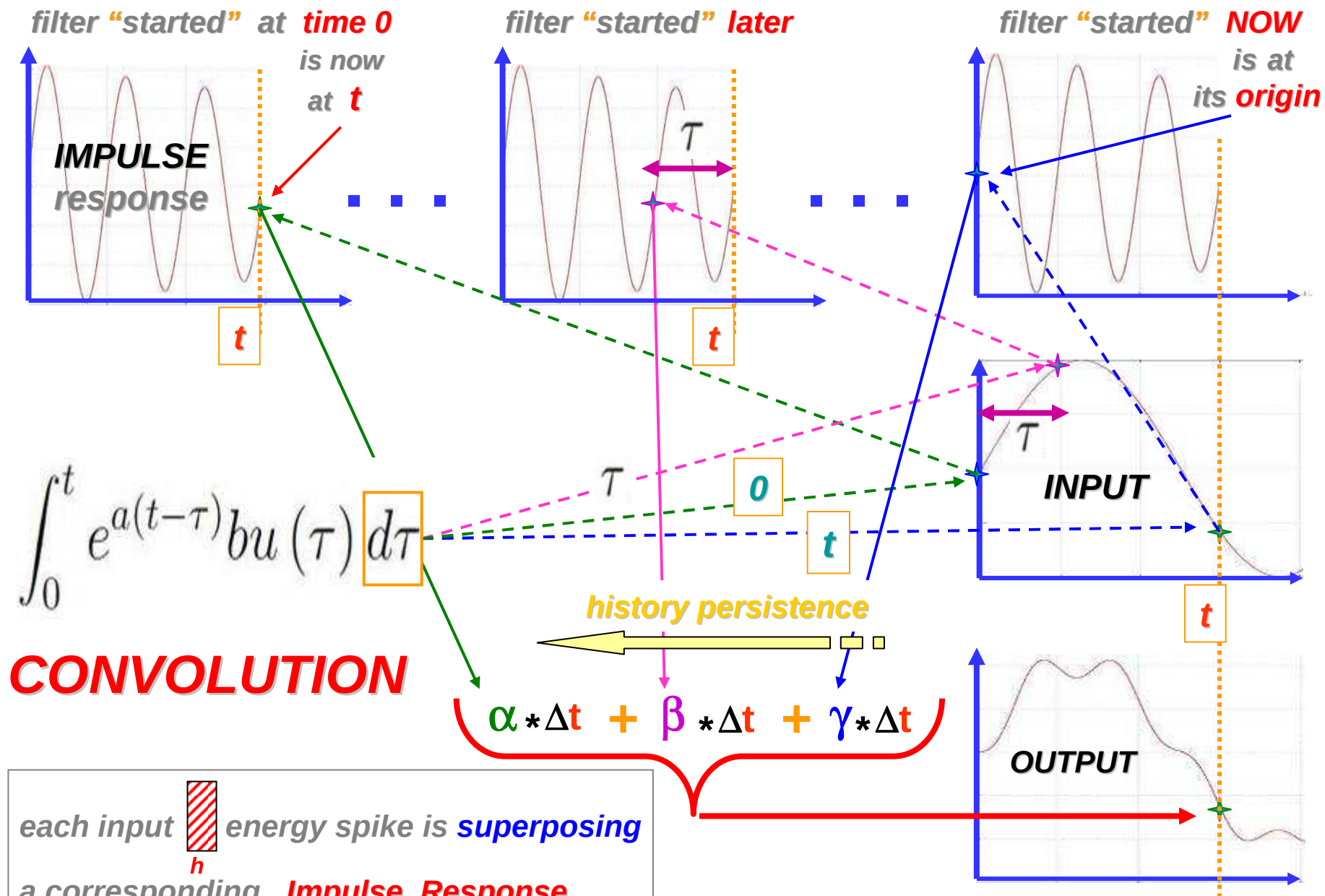
$$x(t) = e^{at}x(0) + \int_0^t e^{a(t-\tau)}bu(\tau) d\tau$$

convolution

integral







Unfortunately,

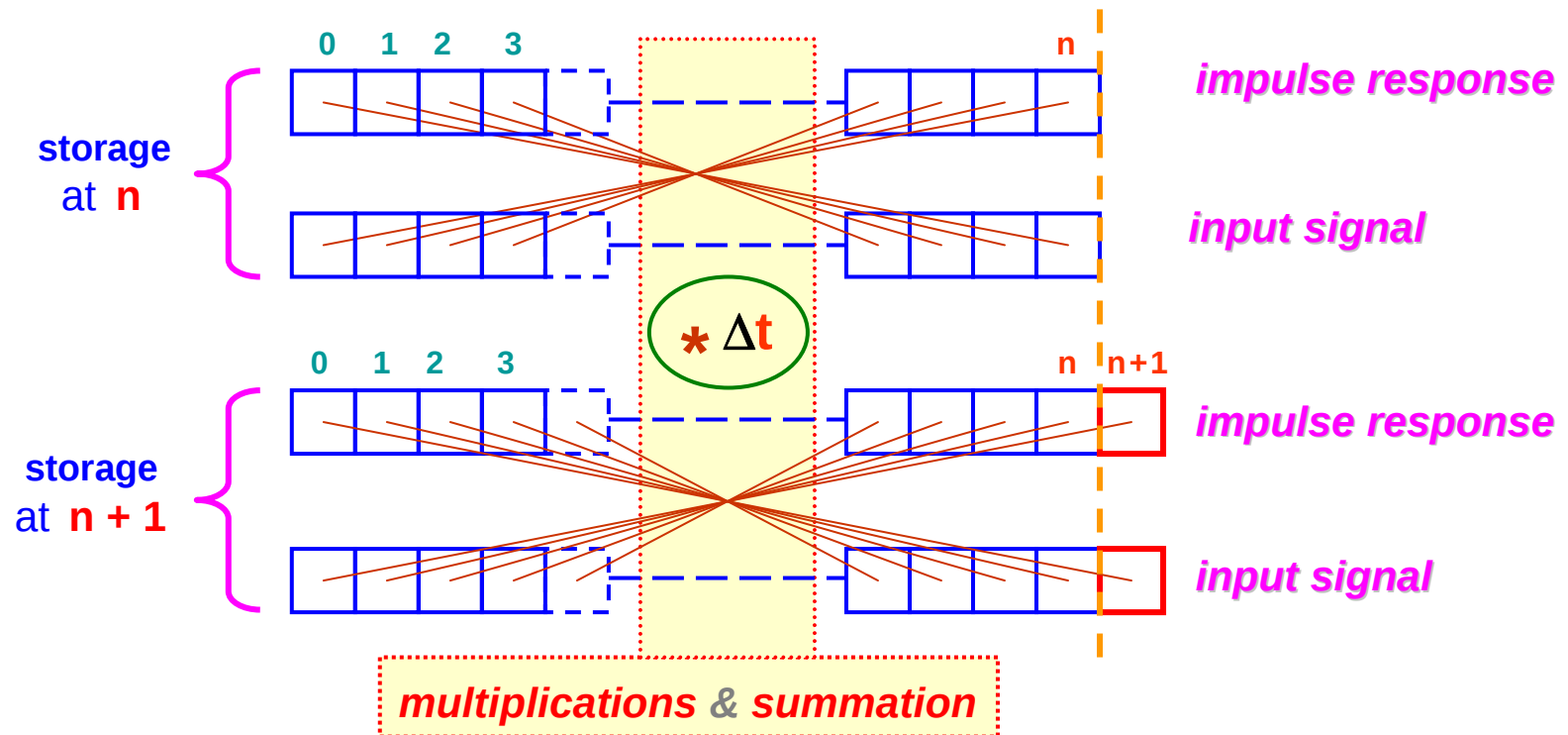
brut force convolution numerical usage is unlikely



unknown **input signal function $u(t)$** prevents **direct integration**



sample based discrete numerical evaluation is very resource consuming



STATE MODEL



$expression(t) \Rightarrow x(t)$



$x(t) \Rightarrow x(t+h)$

STATE MODEL

“time domain” evolution

STATE matrix **A**

$$x(t) = e^{A(t-t_0)} x(t_0) + \int_{t_0}^t e^{A(t-\tau)} B u(\tau) d\tau$$

new
STATE
at **t**

u(t) = 0
“free” mode
since STATE **t₀**

effective contribution from **u(t)**
“forced” mode
since STATE **t₀**

$$e^x = \sum_{k=0}^{\infty} \frac{1}{k!} x^k = 1 + x + \frac{1}{2}x^2 + \frac{1}{6}x^3 + \dots$$



$$e^{At} = \sum_{k=0}^{\infty} \frac{t^k A^k}{k!} = I + At + \frac{t^2 A^2}{2!} + \dots + \frac{t^k A^k}{k!} + \dots$$

The **Transition Matrix** e^{At} can be calculated

by searching **individual** matrix **coefficient** **Inverse Laplace Transform**
OR **globally** through **analytic or approximate** **matricial methods**

STATE MODEL



$\text{expression}(t) \Rightarrow x(t)$



$x(t) \Rightarrow x(t + h)$



$\text{expression}(t) \Rightarrow e^{At}$

example: calculation based on **Inverse Laplace Transform** method

$$e^{At} = \text{Inverse Laplace Transform} \{ K^{-1} \} \quad \text{with} \quad K = sI - A$$

Characteristic Matrix
of matrix **A**

$$A = \begin{bmatrix} -\frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{bmatrix} \Rightarrow K = \begin{bmatrix} L1s + R1 & 1 \\ -\frac{1}{C1} & s \end{bmatrix} \Rightarrow K^{-1} = \begin{bmatrix} \frac{C1L1s}{C1L1s^2 + C1R1s + 1} & \frac{C1}{C1L1s^2 + C1R1s + 1} \\ \frac{L1}{C1L1s^2 + C1R1s + 1} & \frac{C1L1s + C1R1}{C1L1s^2 + C1R1s + 1} \end{bmatrix}$$

$$\Rightarrow e^{At} = \begin{bmatrix} \text{ILT} \left(\frac{s C1L1}{s C1R1 + s^2 C1L1 + 1} \right) & \text{ILT} \left(\frac{C1}{s C1R1 + s^2 C1L1 + 1} \right) \\ \text{ILT} \left(\frac{L1}{s C1R1 + s^2 C1L1 + 1} \right) & \text{ILT} \left(\frac{C1R1 + s C1L1}{s C1R1 + s^2 C1L1 + 1} \right) \end{bmatrix} = \begin{bmatrix} \text{ILT} \left(\frac{s}{(s-p1)(s-p2)} \right) & \text{ILT} \left(\frac{1}{(s-p1)(s-p2)L1} \right) \\ \text{ILT} \left(\frac{1}{(s-p1)(s-p2)C1} \right) & \text{ILT} \left(\frac{R1 + sL1}{(s-p1)(s-p2)L1} \right) \end{bmatrix}$$

transformed matrix is a symbolic “**time domain**” expression of e^{At}

MATRIX decomposition

similarity matrix Φ

$$A = M \Phi M^{-1}$$

$$f(A) = f(M \Phi M^{-1})$$

$$f(A) = M f(\Phi) M^{-1}$$

diagonalization

$$\begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix}$$

diagonal

A may not be diagonalizable (defective matrix)

JORDAN

$$\begin{bmatrix} \lambda_1 & 1 & 0 & 0 & 0 & 0 & 0 \\ & \lambda_1 & 0 & 0 & 0 & 0 & 0 \\ & & \ddots & \ddots & \ddots & \ddots & \ddots \\ & & & \lambda_1 & 1 & 0 & 0 \\ & & & & \lambda_1 & 0 & 0 \\ & 0 & & & & \lambda_2 & 1 & 0 \\ & & & & & & \lambda_2 & 0 \\ & & & & & & & \ddots & \ddots & \ddots \\ & & & & & & & & \lambda_3 & 1 & 0 \\ & & & & & & & & & \lambda_3 & 0 \end{bmatrix}$$

bi-diagonal

triangularization

Schur form

$$\begin{bmatrix} \lambda_1 & t_{12} & t_{13} \\ 0 & \lambda_2 & t_{23} \\ 0 & 0 & \lambda_3 \end{bmatrix}$$

upper-triangular

strictly upper-triangular (nilpotent property)

example: e^{At} calculation based on **Jordan**

(even if this **A** is directly **diagonalizable**)

$$A = \begin{bmatrix} \frac{R1}{L1} & -\frac{1}{L1} \\ \frac{1}{C1} & 0 \end{bmatrix}$$

Jordan =

$$\begin{bmatrix} \frac{\sqrt{C1^2 R1^2 - 4 C1 L1} + C1 R1}{2 C1 L1} & 0 \\ 0 & \frac{\sqrt{C1^2 R1^2 - 4 C1 L1} - C1 R1}{2 C1 L1} \end{bmatrix} = \begin{bmatrix} p1 & 0 \\ 0 & p2 \end{bmatrix}$$

$$A = M J M^{-1} \quad \text{similarity transform}$$

$$A^k = (M J M^{-1})^k = M J^k M^{-1}$$

$$M = \begin{bmatrix} 1 & 1 \\ \frac{1}{p1 C1} & \frac{1}{p2 C1} \end{bmatrix}$$

$$e^{At} = \sum_{k=0}^{\infty} \frac{t^k A^k}{k!} = M e^{Jt} M^{-1} =$$

transition matrix

$$\begin{bmatrix} \frac{t \sqrt{C1^2 R1^2 - 4 C1 L1} + C1 R1}{2 C1 L1} & 0 \\ 0 & \frac{t \sqrt{C1^2 R1^2 - 4 C1 L1} - C1 R1}{2 C1 L1} \end{bmatrix}$$

that could be rewritten

based on the poles **p1** & **p2**

(that are **conjugates**)

$$\begin{bmatrix} \frac{p2 \cdot e^{p2 t} - p1 \cdot e^{p1 t}}{p2 - p1} & -\frac{e^{p2 t} - e^{p1 t}}{(p2 - p1) L1} \\ \frac{e^{p2 t} - e^{p1 t}}{(p2 - p1) C1} & \frac{p1 \cdot e^{p2 t} - p2 \cdot e^{p1 t}}{p2 - p1} \end{bmatrix}$$

even with **complex poles**, **A** **real** $\rightarrow$ **transition matrix** **real**

Inverse Laplace Transform or **Jordan**

$p1$ & $p2$ conjugates

Either way ...

$$e^{At} = \begin{bmatrix} \frac{p_2 e^{p_2 t} - p_1 e^{p_1 t}}{p_2 - p_1} & \frac{e^{p_2 t} - e^{p_1 t}}{(p_2 - p_1)L_1} \\ \frac{e^{p_2 t} - e^{p_1 t}}{(p_2 - p_1)C_1} & \frac{p_1 e^{p_2 t} - p_2 e^{p_1 t}}{p_2 - p_1} \end{bmatrix}$$

poles
"distance"

let

$$p = r \pm v = r \pm z w$$

r real
 z 1 or i
 w real > 0

r must be negative for stability

$$r = \frac{p_1 + p_2}{2} =$$

$$-\frac{R_1}{2L_1}$$

$$v = z w = z \left| f \left(z, \frac{p_1 - p_2}{2} \right) \right| =$$

$$\frac{z \sqrt{z^2 (C_1^2 R_1^2 - 4 C_1 L_1)}}{2 C_1 L_1}$$

$$e^{At} = \begin{bmatrix} \frac{(v e^{2vt} + r e^{2vt} + v - r) e^{rt - vt}}{2v} & \frac{(e^{vt} - 1)(e^{vt} + 1) e^{rt - vt}}{2L_1 v} \\ \frac{(e^{vt} - 1)(e^{vt} + 1) e^{rt - vt}}{2C_1 v} & \frac{(v e^{2vt} - r e^{2vt} + v + r) e^{rt - vt}}{2v} \end{bmatrix}$$

$$V = \frac{1 \sqrt{1^2 (C1^2 R1^2 - 4 C1 L1)}}{2 C1 L1} \quad e^{At} = \begin{bmatrix} \frac{(w e^{2wt} + r e^{2wt} + w - r) e^{rt-wt}}{2w} & \frac{(e^{wt}-1)(e^{wt}+1) e^{rt-wt}}{2L1w} \\ \frac{(e^{wt}-1)(e^{wt}+1) e^{rt-wt}}{2C1w} & \frac{(w e^{2wt} - r e^{2wt} + w + r) e^{rt-wt}}{2w} \end{bmatrix}$$

exponential

$$V = \frac{z \sqrt{z^2 (C1^2 R1^2 - 4 C1 L1)}}{2 C1 L1} \quad e^{At} = \infty$$

natural frequency

$$\frac{\sqrt{4 C1 L1 - C1^2 R1^2}}{4 \pi C1 L1}$$

Euler $e^{ix} = \cos x + i \sin x$ $\cos x = \frac{e^{ix} + e^{-ix}}{2}$ $\sin x = \frac{e^{ix} - e^{-ix}}{2i}$

$$V = \frac{i \sqrt{i^2 (C1^2 R1^2 - 4 C1 L1)}}{2 C1 L1} \quad e^{At} = \begin{bmatrix} \frac{e^{rt} (r \sin(wt) + w \cos(wt))}{w} & \frac{e^{rt} \sin(wt)}{L1w} \\ \frac{e^{rt} \sin(wt)}{C1w} & \frac{e^{rt} (r \sin(wt) - w \cos(wt))}{w} \end{bmatrix}$$

periodic

decay *pulsation*

Those expressions are clearly highlighting the **intrinsic influence** of the **Transition Matrix** on the overall **state** behavior **regardless** of the system **input activation**

Nineteen Dubious Ways to Compute the Exponential of a Matrix, *Twenty-Five Years Later*

Cleve Moler , Charles Van Loan

SIAM review 1978 and then 2003



1 *Taylor series*

2 *Padé approximation*

3 *Scaling and squaring*

4 *Chebyshev rational approximation*

5 *General purpose ODE solver*

6 *Single step ODE method*

7 *Multistep ODE solver*

8 *Cayley - Hamilton*

9 *Lagrange interpolation*

10 *Newton interpolation*

11 *Vandermonde matrix*

academic

12 *Inverse Laplace transform*

13 *Companion matrix*

14 *Eigenvectors*

15 *Triangular systems of eigenvectors*

academic

16 *Jordan canonical form*

not e^x specialized

17 *Schur decomposition*

18 *Block diagonal matrix*

19 *Splitting method*

20 *Krylov space methods* (added in 2003)





Brook Taylor (1685 – 1731)

“**Taylor series**” order 3

$$e^x \approx 1 + x + \frac{x^2}{2} + \frac{x^3}{6} \cdot \text{X}$$

series development
approximation



Henri Padé (1863 – 1953)

“**Padé approximants**” order 3

$$e^z \approx \frac{1 + \frac{1}{2}z + \frac{1}{10}z^2 + \frac{1}{120}z^3}{1 - \frac{1}{2}z + \frac{1}{10}z^2 - \frac{1}{120}z^3}$$

“**rational fraction**”

$$e^z = \frac{1}{1 - \frac{z}{1 + \frac{\frac{1}{2}z}{1 - \frac{\frac{1}{6}z}{1 + \frac{\frac{1}{6}z}{1 - \frac{\frac{1}{10}z}{1 + \frac{\frac{1}{10}z}{1 - \dots}}}}}}}$$

“**continued fraction**”
form

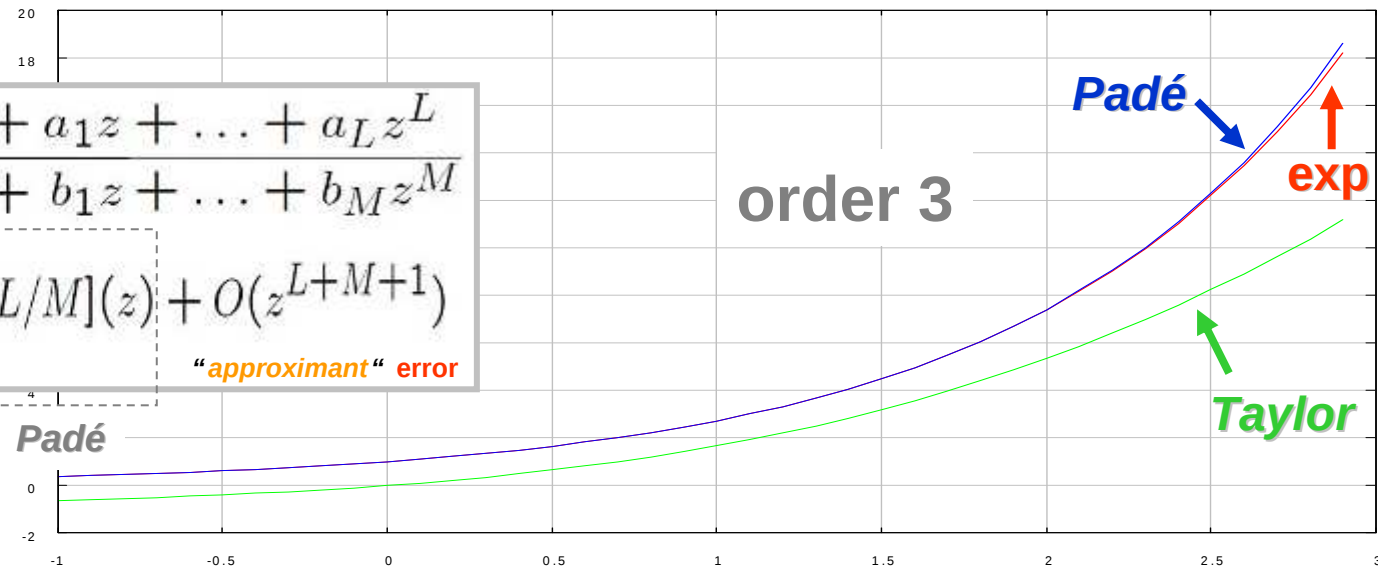
$$[L/M](z) = \frac{a_0 + a_1z + \dots + a_Lz^L}{b_0 + b_1z + \dots + b_Mz^M}$$

$$f(z) = \sum_{i=0}^{\infty} c_i z^i = [L/M](z) + O(z^{L+M+1})$$

“**approximant**” error

Taylor

Padé



STATE MODEL



$\text{expression}(t) \Rightarrow x(t)$



$x(t) \Rightarrow x(t+h)$



$\text{expression}(t) \Rightarrow e^{At}$



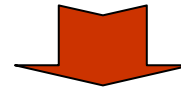
unknown input $u(\tau) \Rightarrow \text{approximation}$



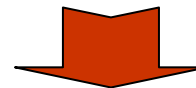
Semi Analytical Recursive Convolution algorithm (**SARC**)



$$\int_{t-h}^t e^{a(t-\tau)} u(\tau) d\tau$$



$$e^{at} \int_{t-h}^t e^{-a\tau} u(\tau) d\tau$$



$$e^{at} \int_{t-h}^t e^{-a\tau} \textit{interpolation}(\tau) d\tau$$

Lagrange polynomial

$$u(\tau) = \sum_{i=0}^n L_i(\tau) u_i$$

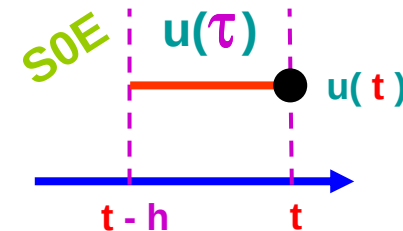
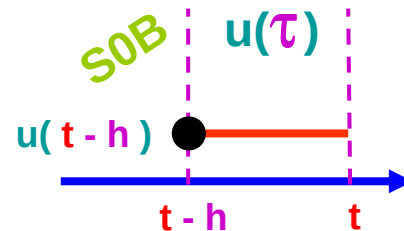
with

$$L_i(\tau) = \prod_{\substack{j=0 \\ j \neq i}}^n \frac{(\tau - \tau_j)}{(\tau_i - \tau_j)}$$

$$\begin{aligned} L_i(\tau_i) &= 1 \\ L_i(\tau_j) &= 0 \quad (j \neq i) \end{aligned}$$

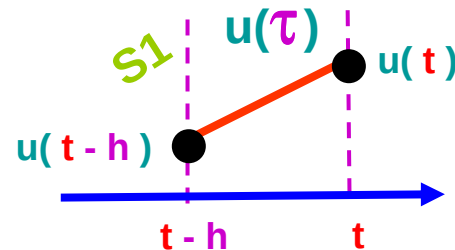
order $n = 0$

$u(\tau)$ is approximated on the interval $[t-h, t]$ by a **constant**



order $n = 1$

$u(\tau)$ is approximated on the interval $[t-h, t]$ by a **slope**



etc ...

interpolation **error**

$$O(h^{n+1}) \quad \frac{h}{2} \rightarrow \frac{\text{error}}{2^{n+1}}$$

$$e(\tau) = \frac{u^{(n+1)}(\xi)}{(n+1)!} \lambda(\tau)$$

with $\lambda(\tau) = (\tau - \tau_0)(\tau - \tau_1) \dots (\tau - \tau_n)$

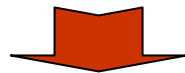
$$\xi \in [\tau_0, \tau_n]$$

interpolation points

$$\int_{t-h}^t e^{a(t-\tau)} u(\tau) d\tau \Rightarrow e^{at} \int_{t-h}^t e^{-a\tau} \text{interpolation}(\tau) d\tau$$

order 0

$$e^{at} \int_{t-h}^t e^{-a\tau} \left[u_t \right] d\tau$$



$$\Phi = e^{ah} \quad \frac{\Phi - 1}{a} u_t$$

constant

order 1

$$e^{at} \int_{t-h}^t e^{-a\tau} \left[\frac{t-\tau}{h} u_{t-h} + \frac{\tau-(t-h)}{h} u_t \right] d\tau$$



$$\frac{1}{a} \left[\Phi - \frac{\Phi - 1}{ah} \right] u_{t-h} + \frac{1}{a} \left[\frac{\Phi - 1}{ah} - 1 \right] u_t$$



does not depend on

$$e^{at}$$

!!!

(Markov process)

The "shaking" due to $[t-h, t]$ input is independent of the current oldness of the system

putting everything together

$$x(t) = e^{at} x(0) + \int_0^t e^{a(t-\tau)} b u(\tau) d\tau$$

$$x(t) = e^{a(t-t_0)} x(t_0) + \int_{t_0}^t e^{a(t-\tau)} b u(\tau) d\tau$$

$$x(t) = e^{ah} x(t-h) + \int_{t-h}^t e^{a(t-\tau)} b u(\tau) d\tau$$

$$x(t) = e^{ah} x(t-h) + b e^{at} \int_{t-h}^t e^{-a\tau} \text{interpolation}(\tau) d\tau$$

order 0

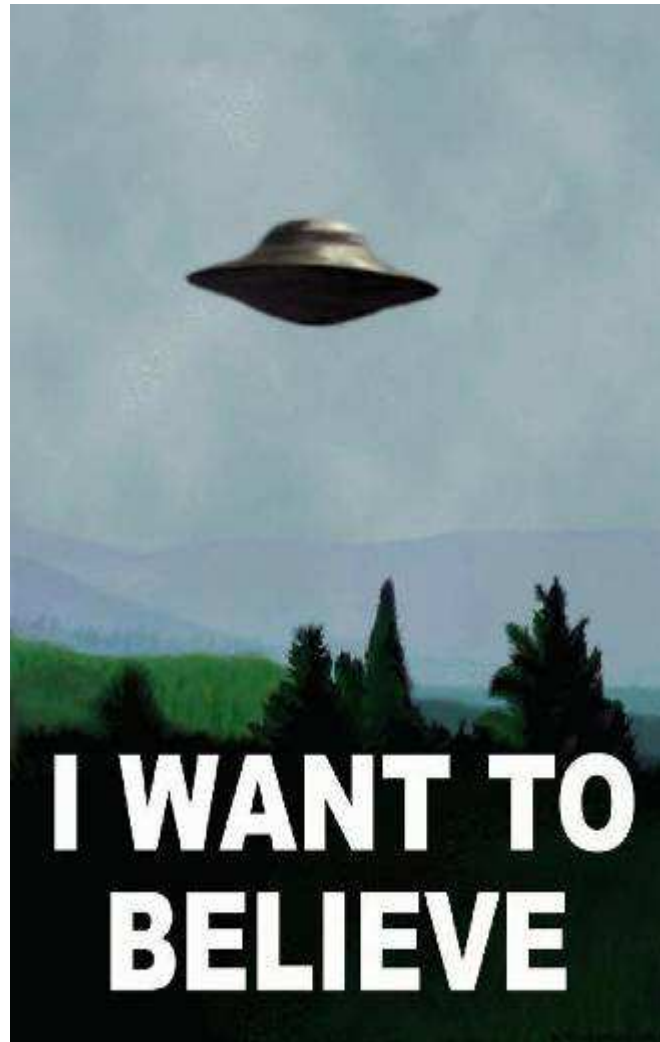
$$x_t = \Phi x_{t-h} + \frac{\Phi - 1}{a} b u_t$$

$$\Phi = e^{ah}$$

amazingly trivial
compared to *Spice*

constant

constant



x	→	X
a	→	A
b	→	B
$\frac{1}{a}$	→	<i>invert(A)</i>
*	→	● (<i>matricial product</i>)
a ^t	→	A ^t
e	→	e (<i>19+1 methods ...</i>)

$$A e^{A^t} = e^{A^t} A$$

SARC equations are also applicable to **STATE MODEL** matrices



SARC ERROR



Can I help ???

input interpolation

order

step

$$\Phi = e^{Ah}$$

choose finer calculation

method

(may be jeopardized by poles configuration)

Euler

$$e^{ix} = \cos x + i \sin x$$

complex number

floating number

bits
digits
exponent

IEEE STANDARD 754

float

double

long double

32

7

≈ ±38

64

16

≈ ±308

80 96

19

≈ ±4932

128

34

≈ ±4932

x87 standard



contribution of *interpolation* to

SARC *step* *ERROR*

$$x(t) = e^{a(t-t_0)} x(t_0) + \int_{t_0}^t e^{a(t-\tau)} b u(\tau) d\tau$$

$$x(t) = e^{ah} x(t-h) + b e^{at} \int_{t-h}^t e^{-a\tau} \text{interpolation}(\tau) d\tau$$

Lagrange interpolation *step* *error*

$$\mathcal{E} = b e^{at} \frac{u^{(n+1)}(\xi)}{(n+1)!} \int_{t-h}^t e^{-a\tau} \lambda(\tau) d\tau$$

step
interpolation

with

$$\lambda(\tau) = (\tau - t_0) \dots * (\tau - t_n)$$

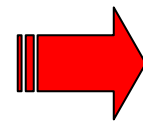
$\xi \in$ interpolation involved segments

order $n = 0$

S0B

step constant = u_{t-h}

$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{b e^{a t}}{1!} \sup_{[t-h, t]} \left\{ \frac{d u(t)}{d t} \right\} \int_{t-h}^t e^{-a \tau} \left[\tau - (t-h) \right] d\tau$$



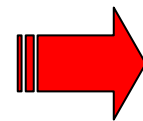
$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{1}{a^2} \left[\Phi - ah - 1 \right] \sup_{[t-h, t]} \left\{ b \frac{d u(t)}{d t} \right\}$$

order $n = 0$

S0E

step constant = u_t

$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{b e^{a t}}{1!} \sup_{[t-h, t]} \left\{ \frac{d u(t)}{d t} \right\} \int_{t-h}^t e^{-a \tau} \left[\tau - (t) \right] d\tau$$



$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{1}{a^2} \left[(1 - ah) \Phi - 1 \right] \sup_{[t-h, t]} \left\{ b \frac{d u(t)}{d t} \right\}$$

order $n = 1$

slope between

u_{t-h}

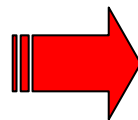
&

u_t

interpolation points

S1

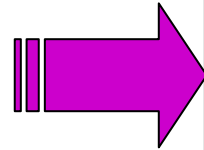
$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{b e^{a t}}{2!} \sup_{[t-h, t]} \left\{ \frac{d^2 u(t)}{d t^2} \right\} \int_{t-h}^t e^{-a \tau} \left[\tau - (t-h) \right] \left[\tau - (t) \right] d\tau$$



$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{1}{2 a^3} \left[(2 - a h) \Phi - a h - 2 \right] \sup_{[t-h, t]} \left\{ b \frac{d^2 u(t)}{d t^2} \right\}$$

SARC cumulated interpolation ERROR

RECURSIVITY



$$\underset{\text{cumulated interpolation}}{\varepsilon(k)} = \sum_{i=1}^k \left[\phi^{k-i} \underset{\text{step interpolation}}{\varepsilon(i)} \right]$$

constant "majorant"

MAJORANT

$$\underset{\text{cumulated max interpolation}}{\varepsilon(k)} < \left(\frac{1 - \phi^k}{1 - \phi} \right) \left| \underset{\text{step interpolation}}{\varepsilon_{\max}} \right|$$

for large k



$$t = kh$$

$$\underset{\text{cumulated max interpolation}}{\varepsilon(\text{only large } t)} < \frac{\left| \underset{\text{step interpolation}}{\varepsilon_{\max}} \right|}{1 - \phi}$$

$$\frac{\left| \underset{\text{step interpolation}}{\varepsilon_{\max}} \right|}{1 - \phi}$$

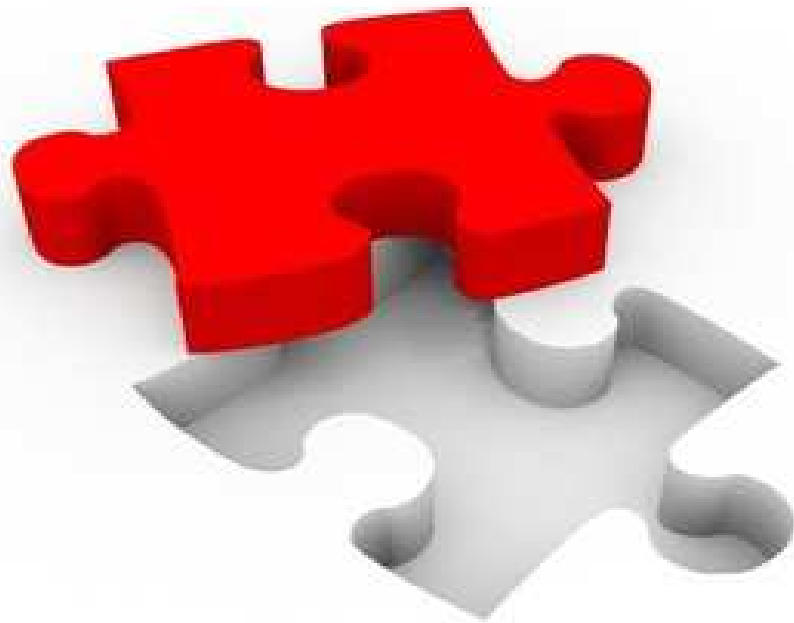
asymptotic

error majorant

(proof of absolute stability)



*estimated **SARC error**, in practice*



cumulated SARC error can be **estimated** at **run time**
 due to interpolation

example, for **order $n = 0$** with **step constant** =

$$\frac{u}{t-h}$$

$$\frac{\Delta u}{\Delta t}$$



$$\mathcal{E}(k)_{\text{step interpolation}} = \frac{1}{a^2} \left[\Phi - ah - 1 \right] \left[b \frac{u(t) - u(t-h)}{h} \right]$$

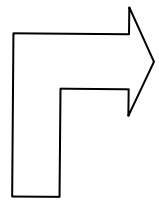
$$\mathcal{E}(k)_{\text{cumulated interpolation}} = \phi \mathcal{E}(k-1)_{\text{cumulated interpolation}} + \mathcal{E}(k)_{\text{step interpolation}}$$

$$\mathcal{E}(k)_{\text{cumulated max interpolation}} \approx \max \left(\left| \mathcal{E}(k-1)_{\text{cumulated interpolation}} \right|, \left| \mathcal{E}(k)_{\text{cumulated interpolation}} \right| \right)$$

SARC SUMMARY

$$\Phi = e^{A h}$$

recursive convolution



$$\Phi X_{t-h} + \sum_{n=0}^{\text{order}} \left(c_n U_{t-n*h} \right)$$

$$X_t \Rightarrow Y$$

likely to engender low noise

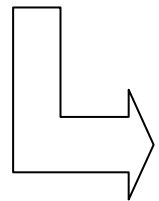


multi **inputs**
vector

estimated U_t (order + 1)

multi **outputs**
vectors

recursive convolution



$$\Phi \varepsilon_{t-h} + \hat{C}_{\text{order}} \left\{ \frac{\Delta U}{h} \right\}^{\text{order} + 1}$$

$$\varepsilon_t \Rightarrow Y_{\text{err}}$$

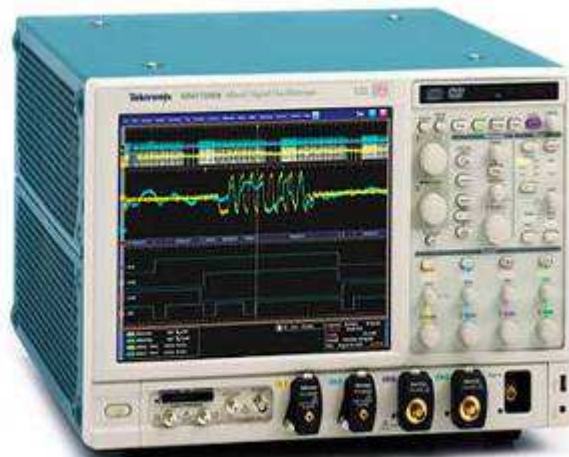
interpolation

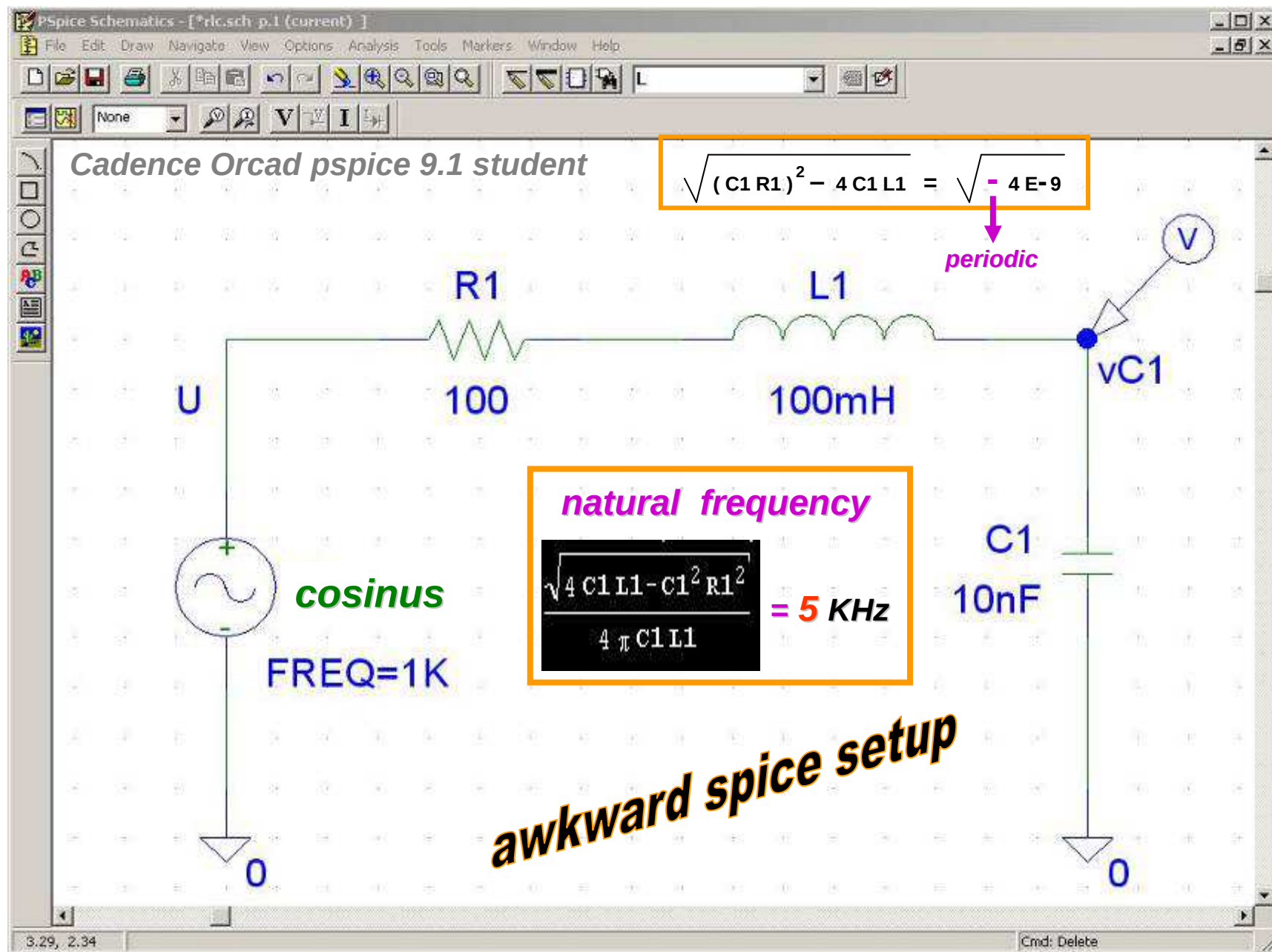
*runtime
estimated*

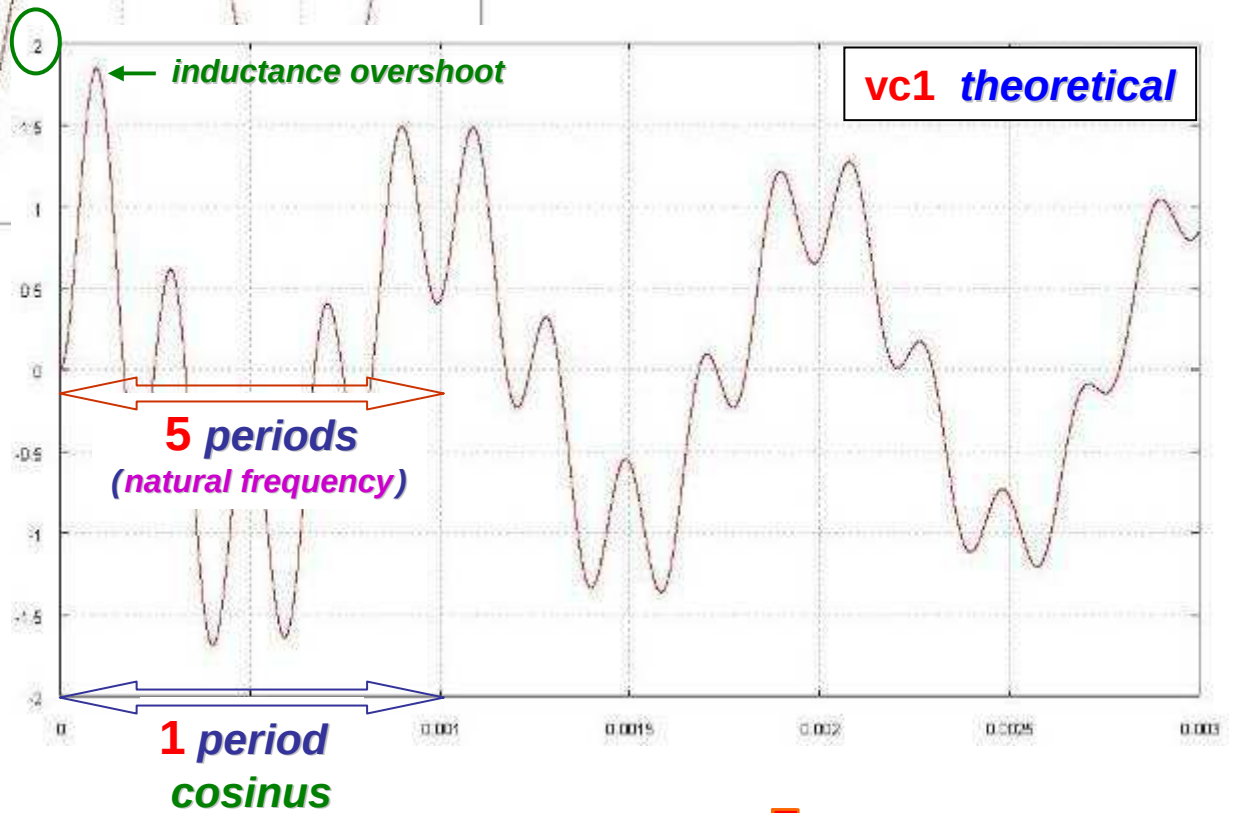
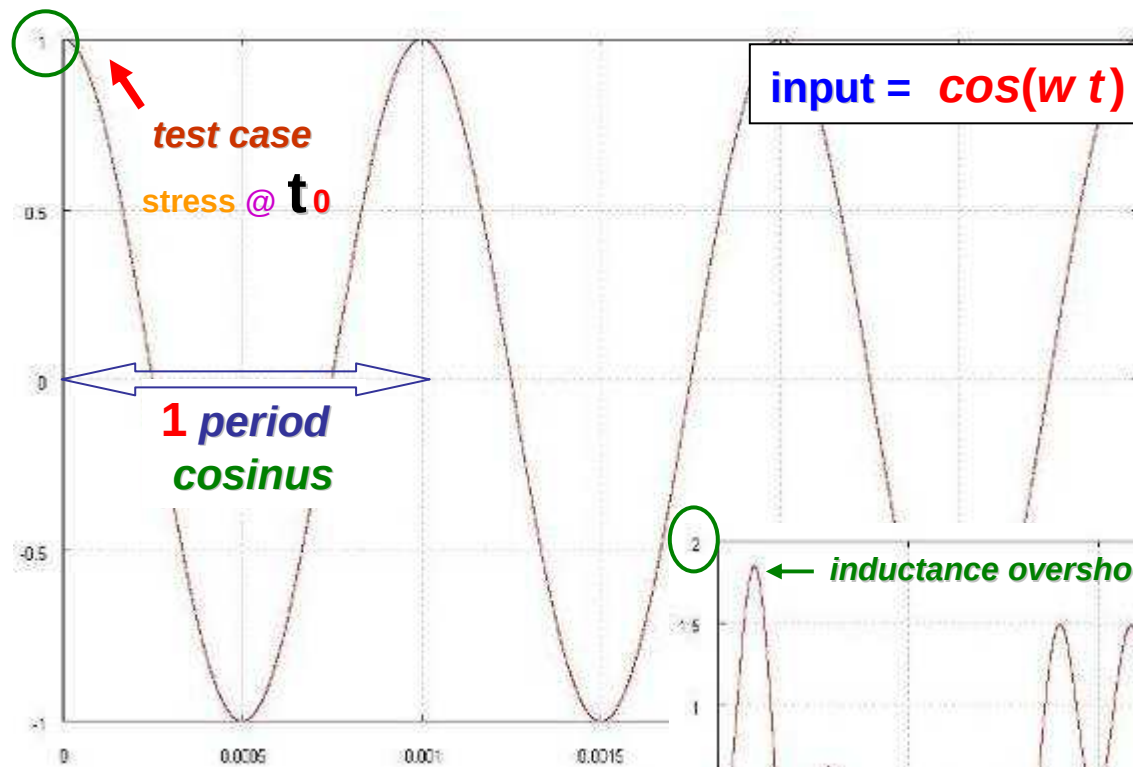
interpolation error

X_{error}_{t-h}

TEST

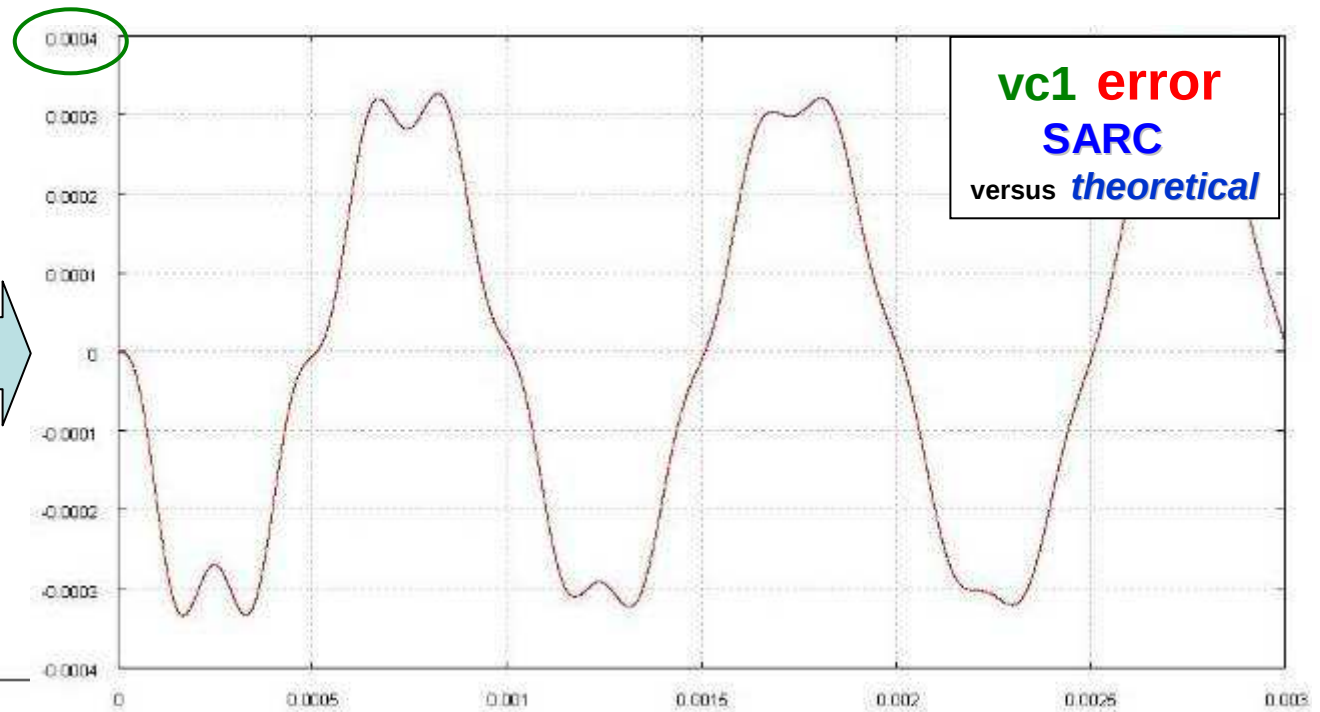
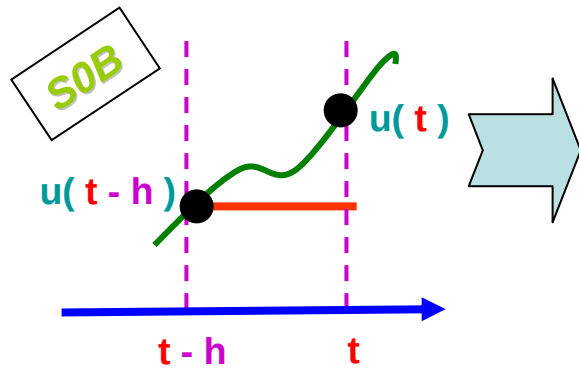




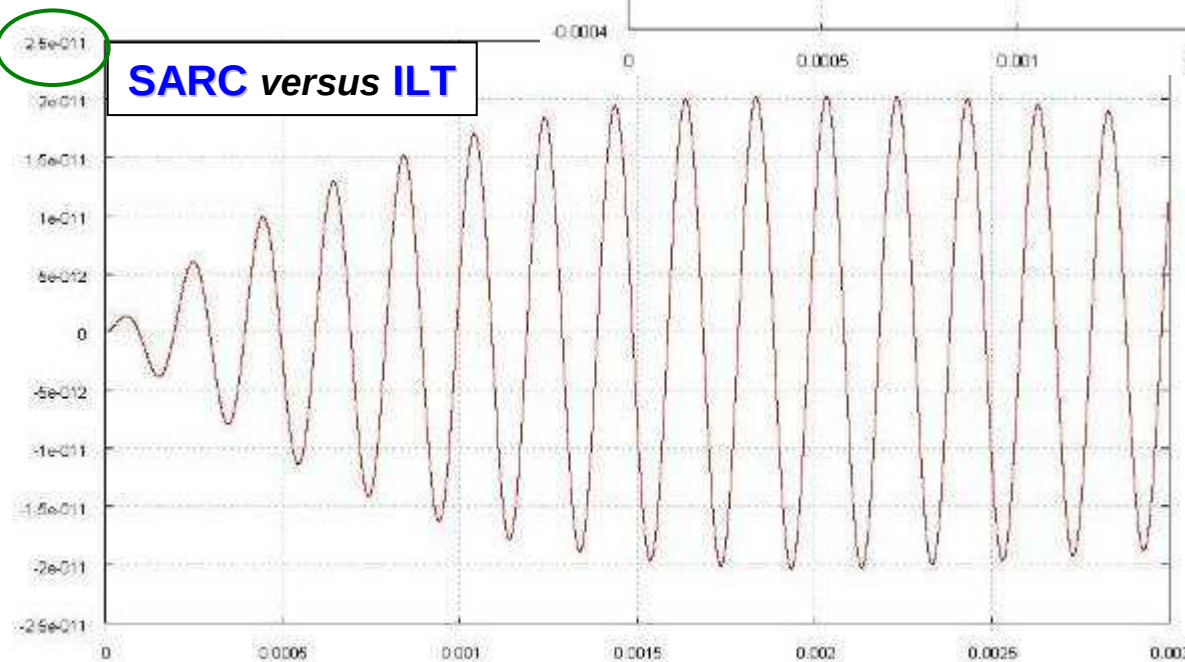


Lagrange interpolation

order 0

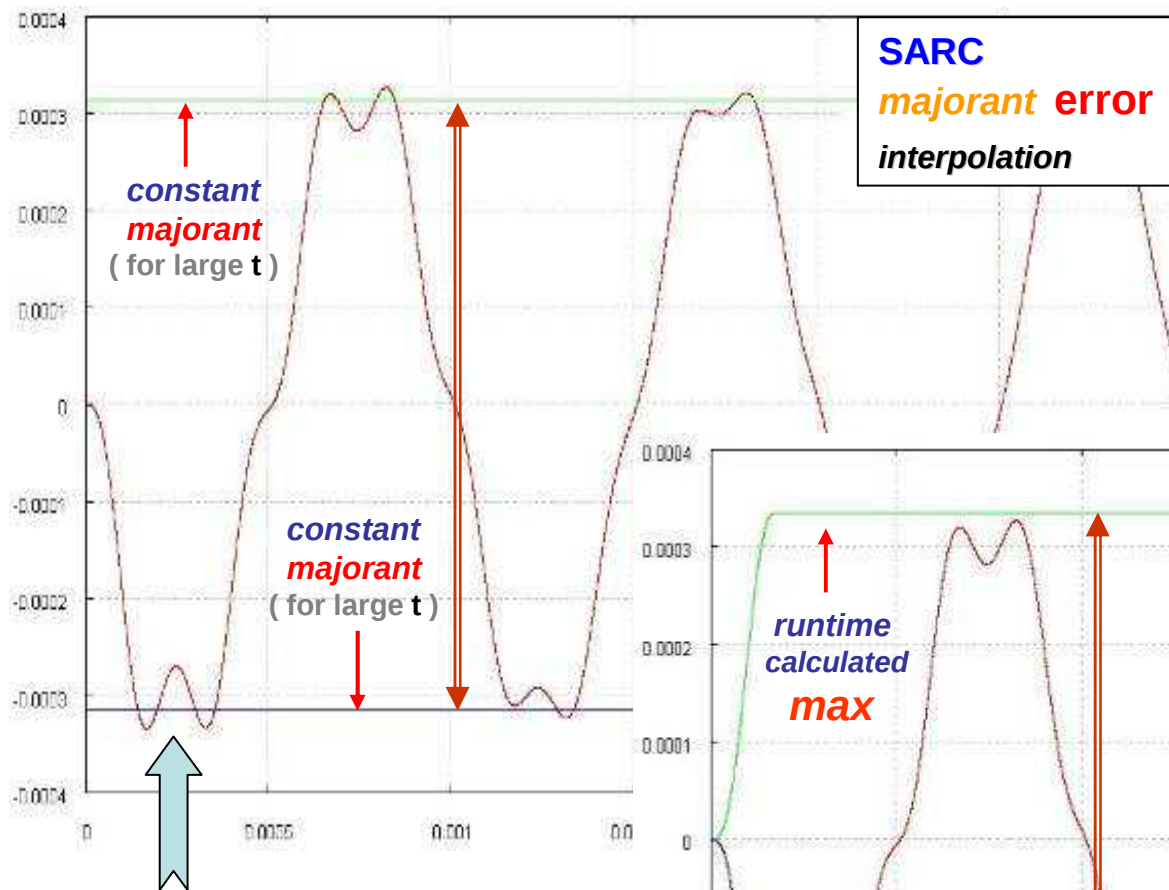


SARC versus ILT

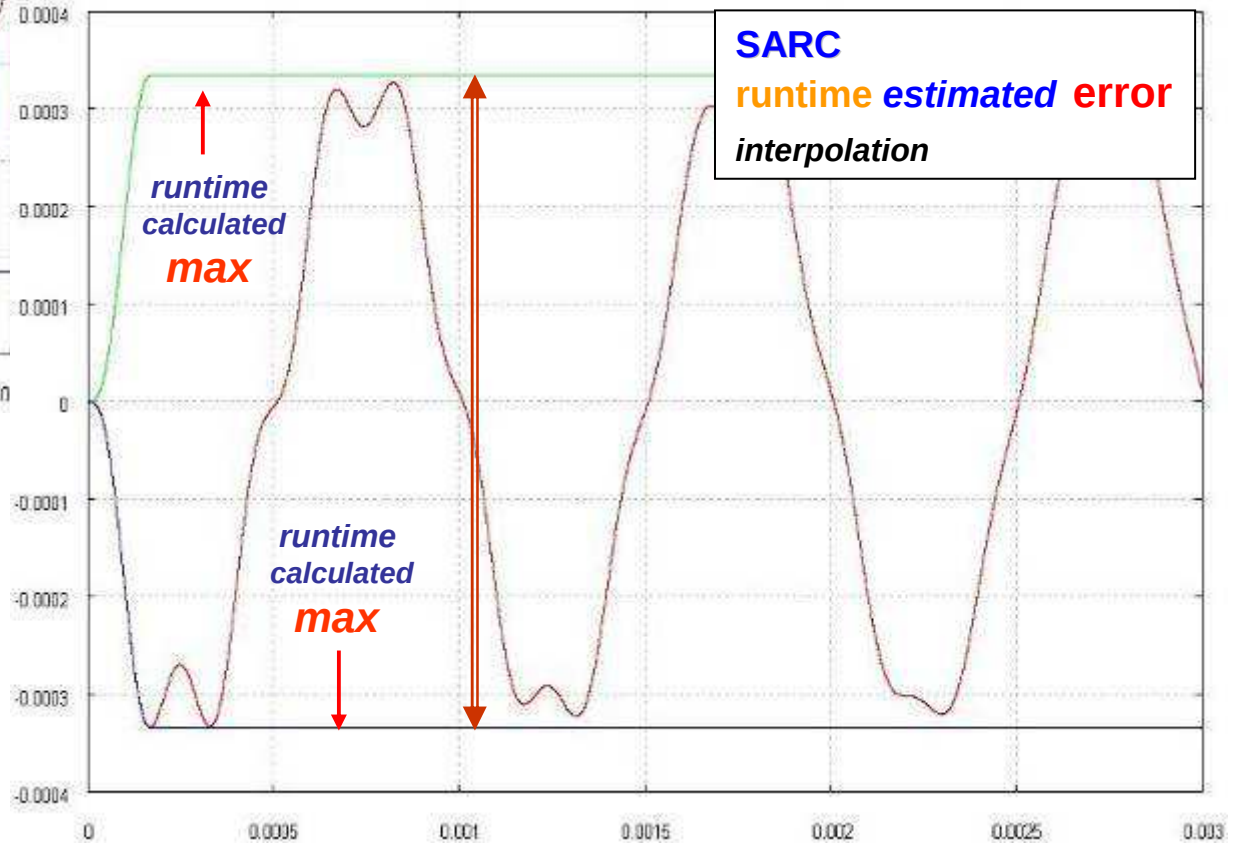


methods comparison

$$\approx 10^{-11} !!!$$

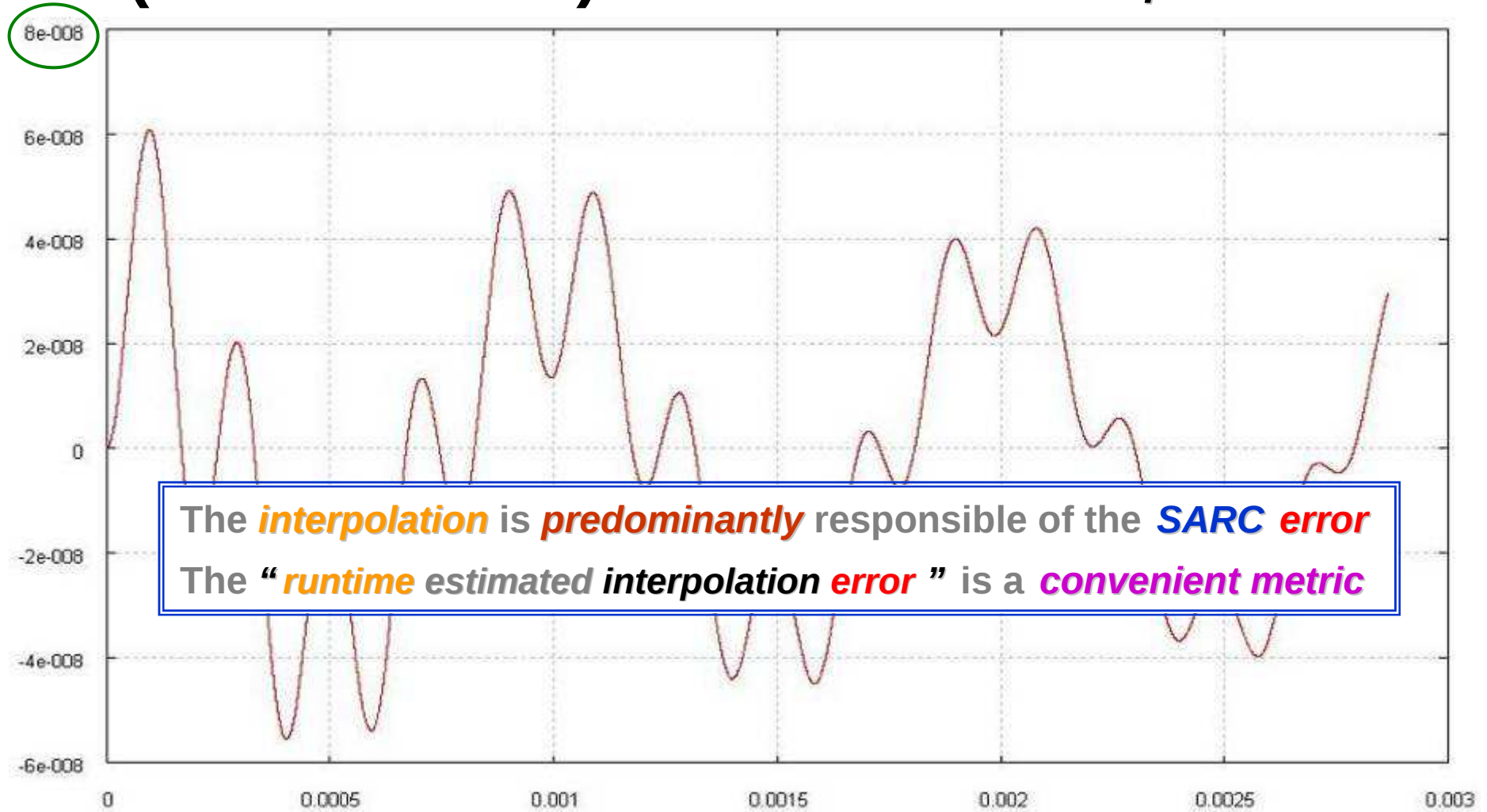


inaccurate
for small t



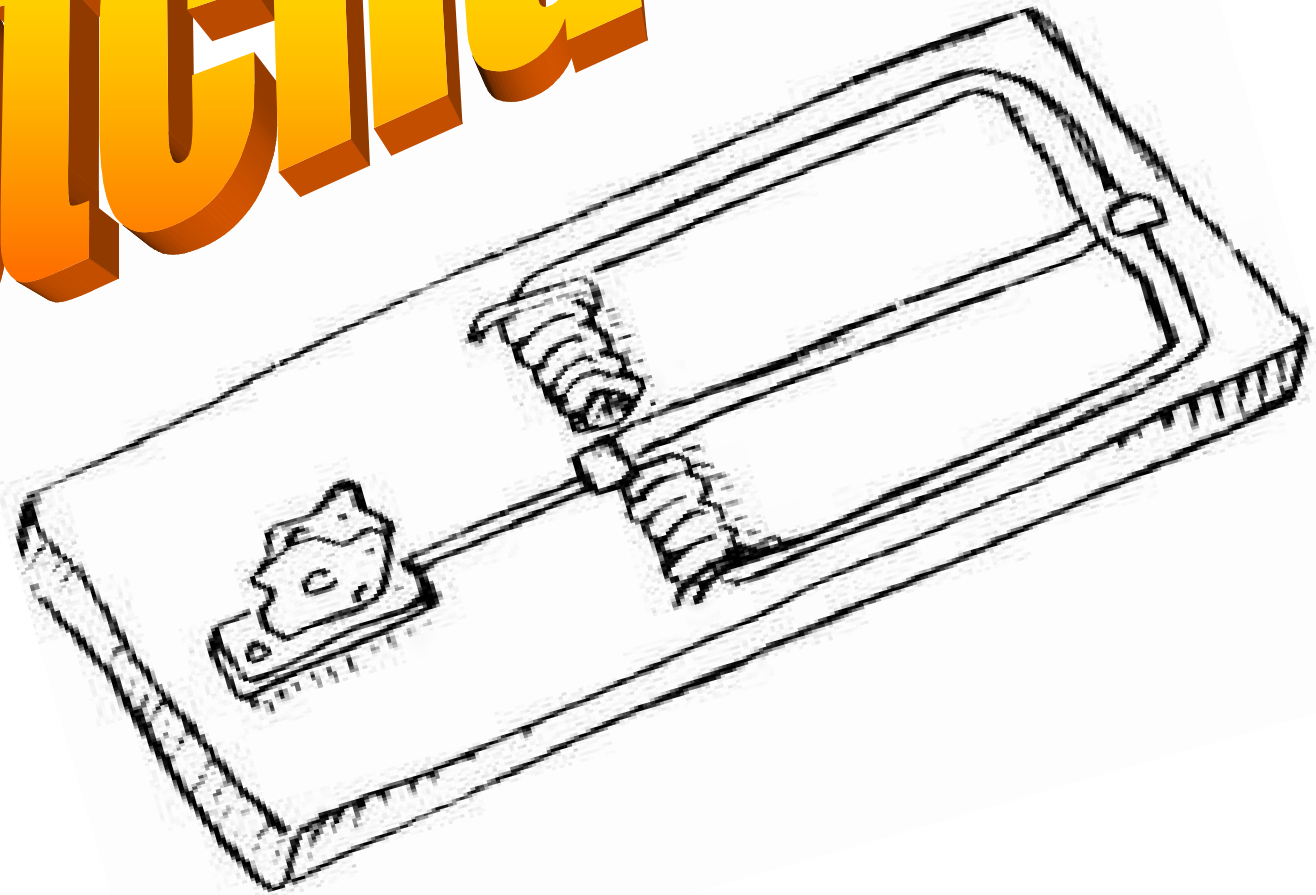
full SARC error

$(\text{theoretical} - \text{SARC}) - \text{runtime estimated interpolation error}$





Gotcha



INITIALIZATION

a single

missing

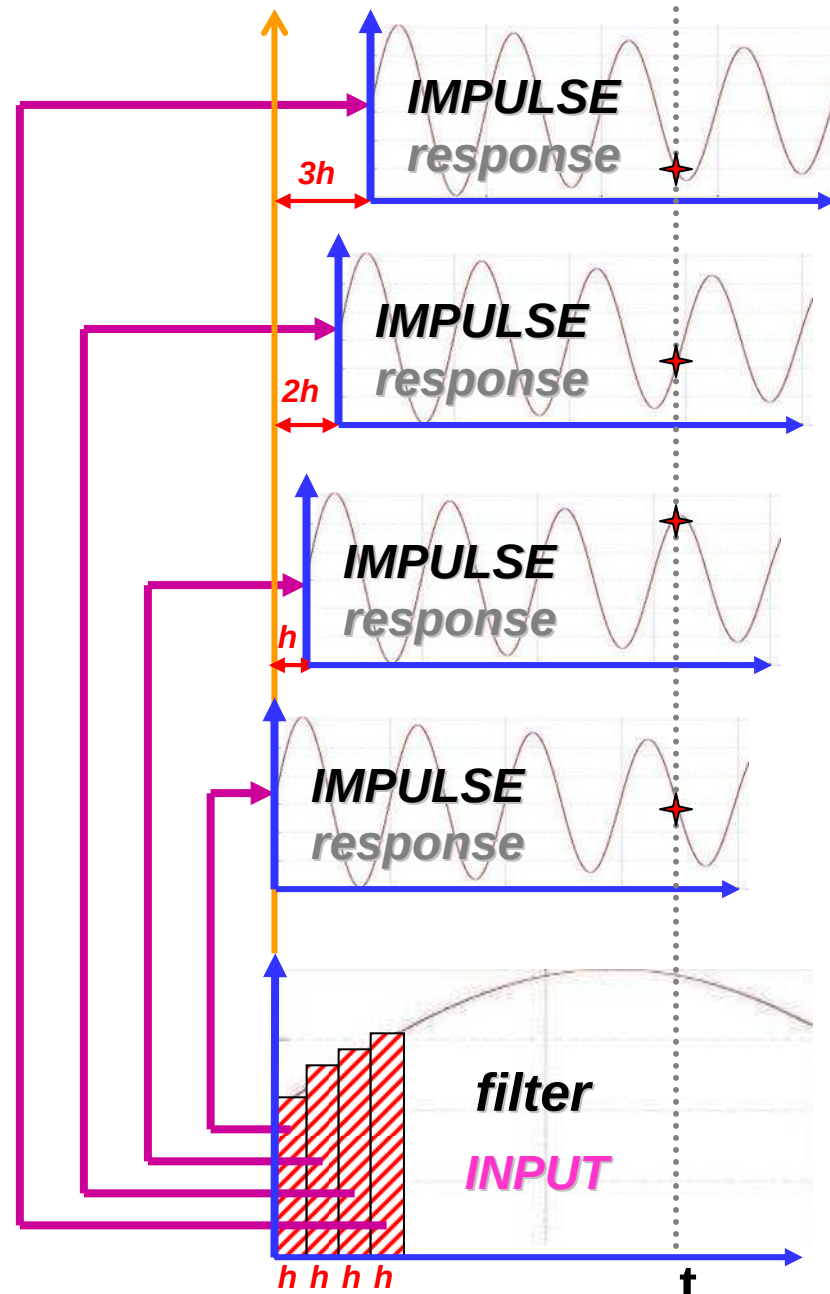
misplaced

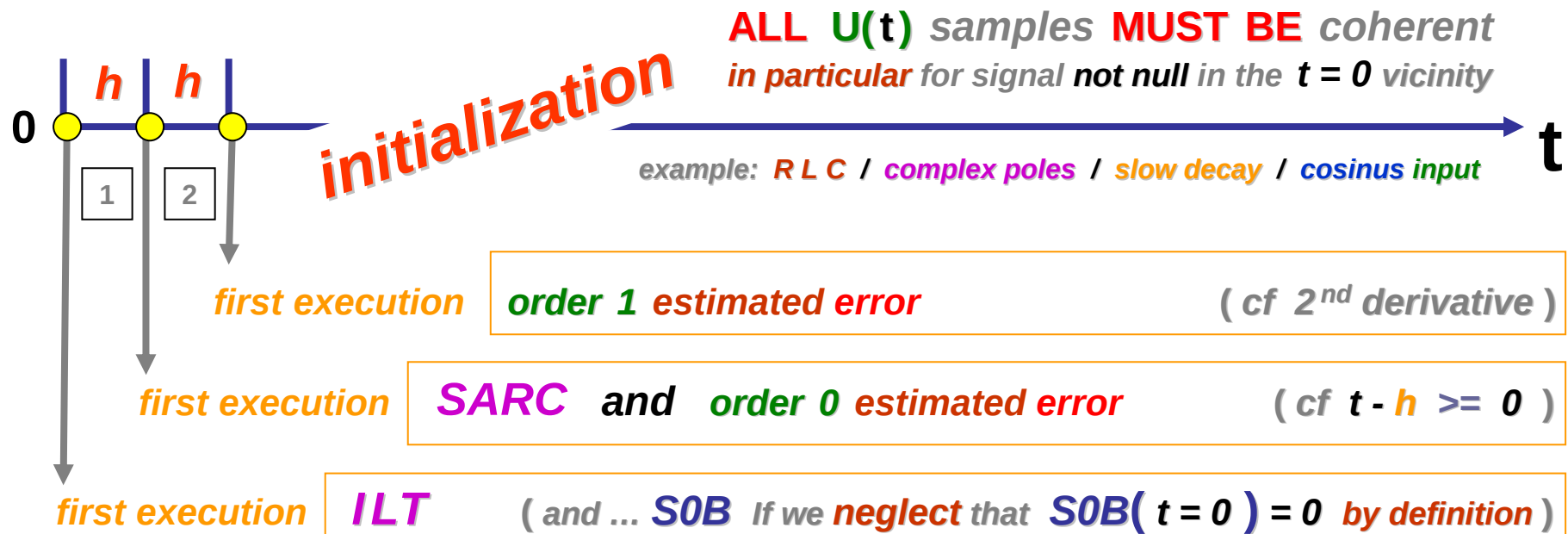
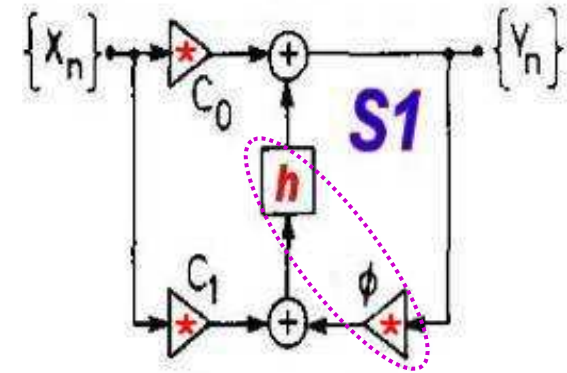
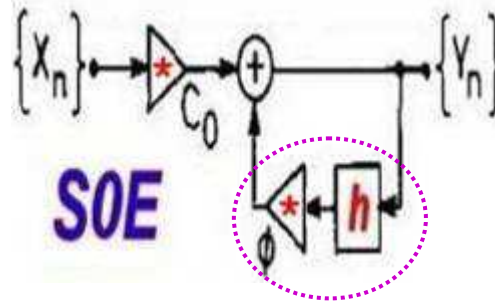
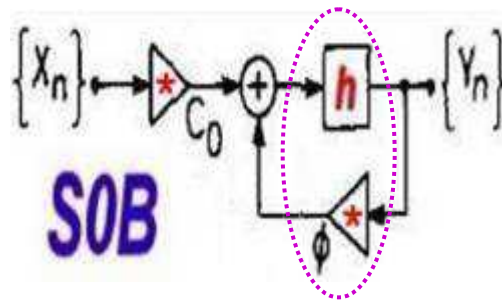
or

aberrant

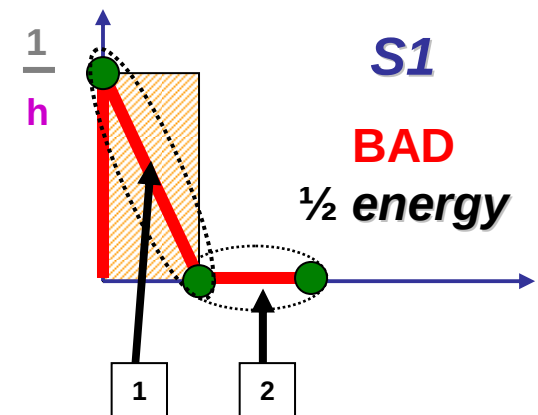
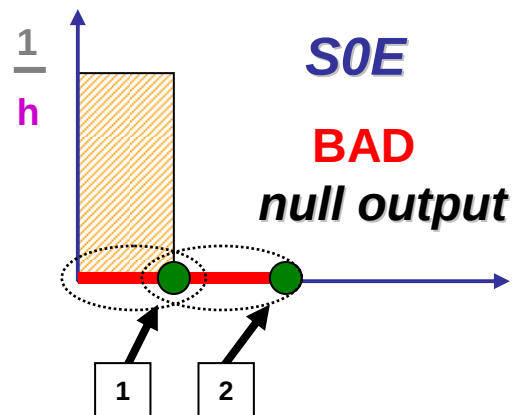
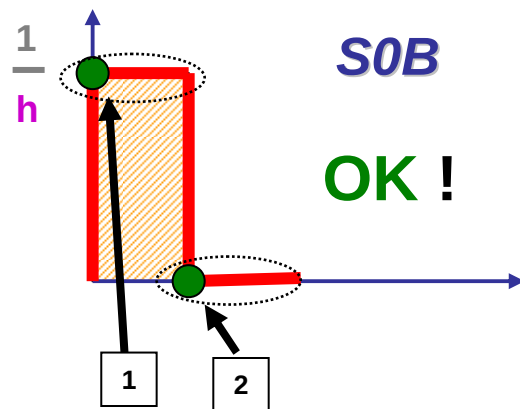
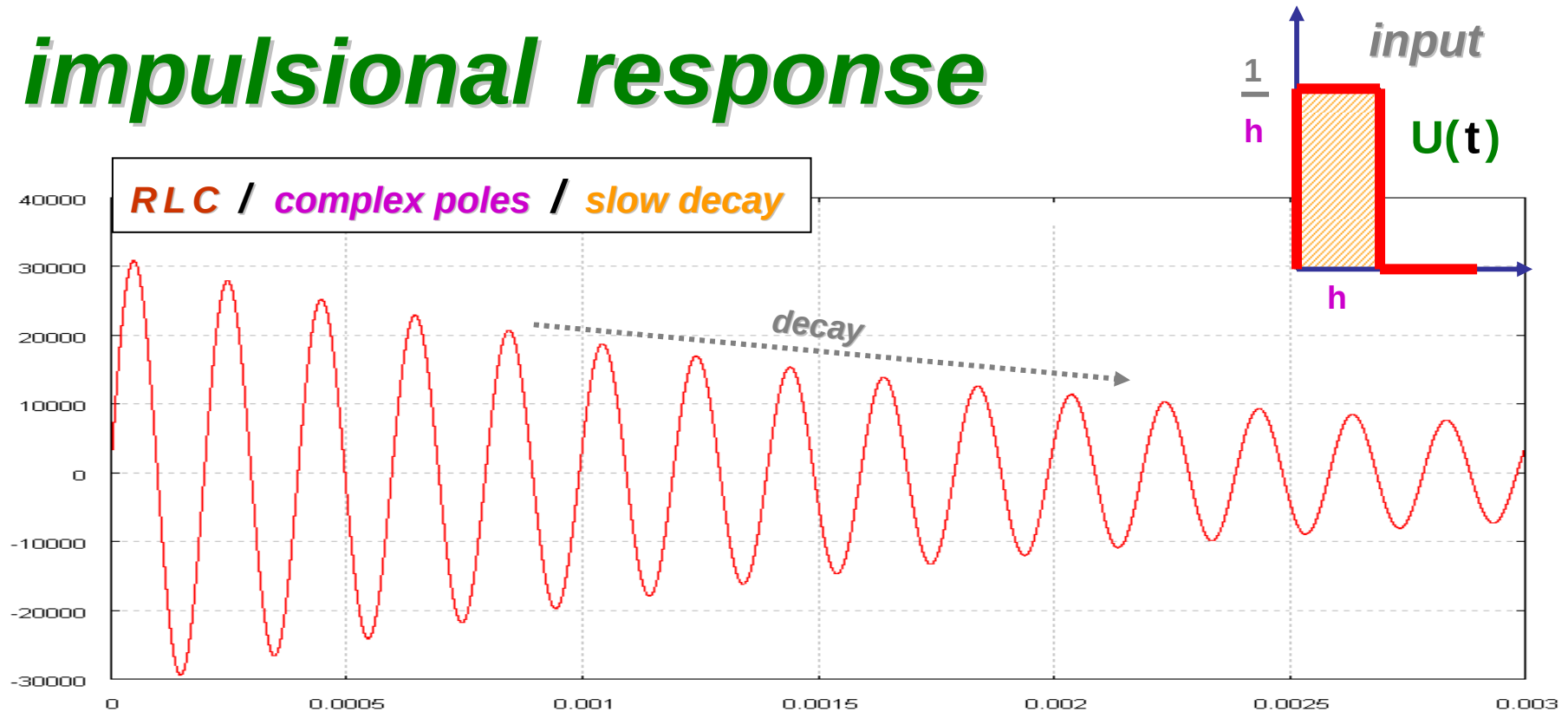
input sample response

may result in a significant **ERROR**
over a long interval





impulsional response

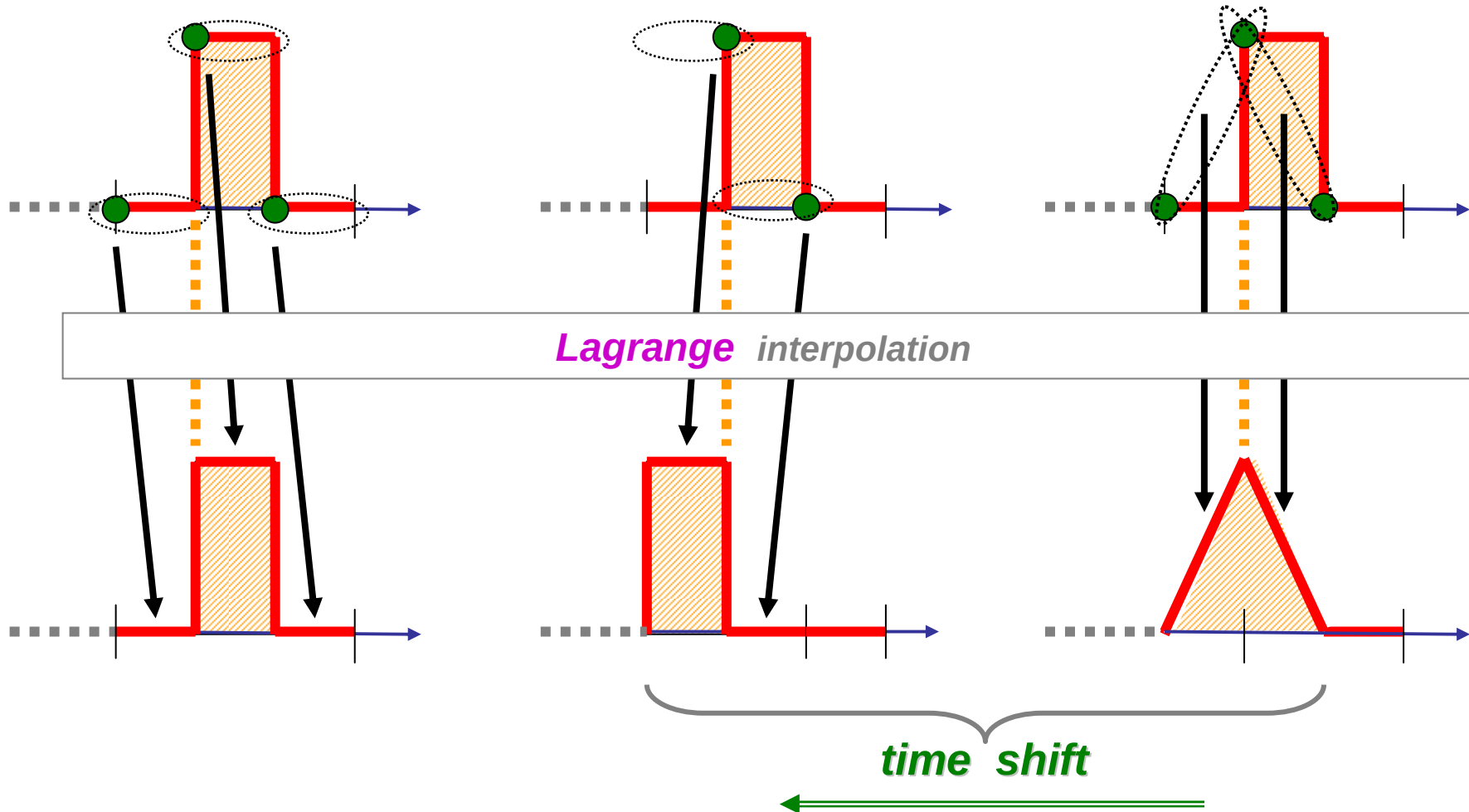


S0B

S0E

S1

input signal $U(t)$

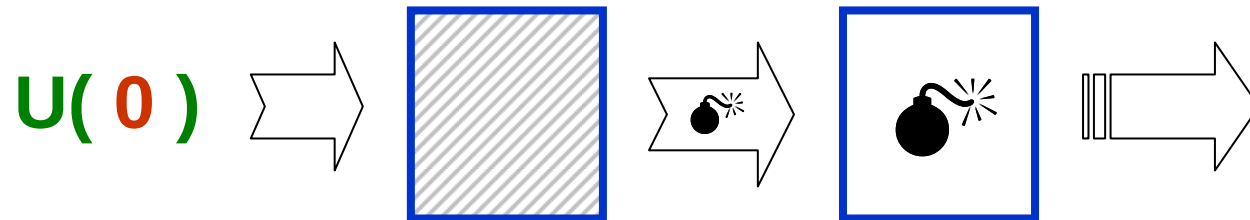


order 1 is better than **order 0**

BUT

general purpose simulators cascading sub-circuits

cannot tolerate initialization missing output samples



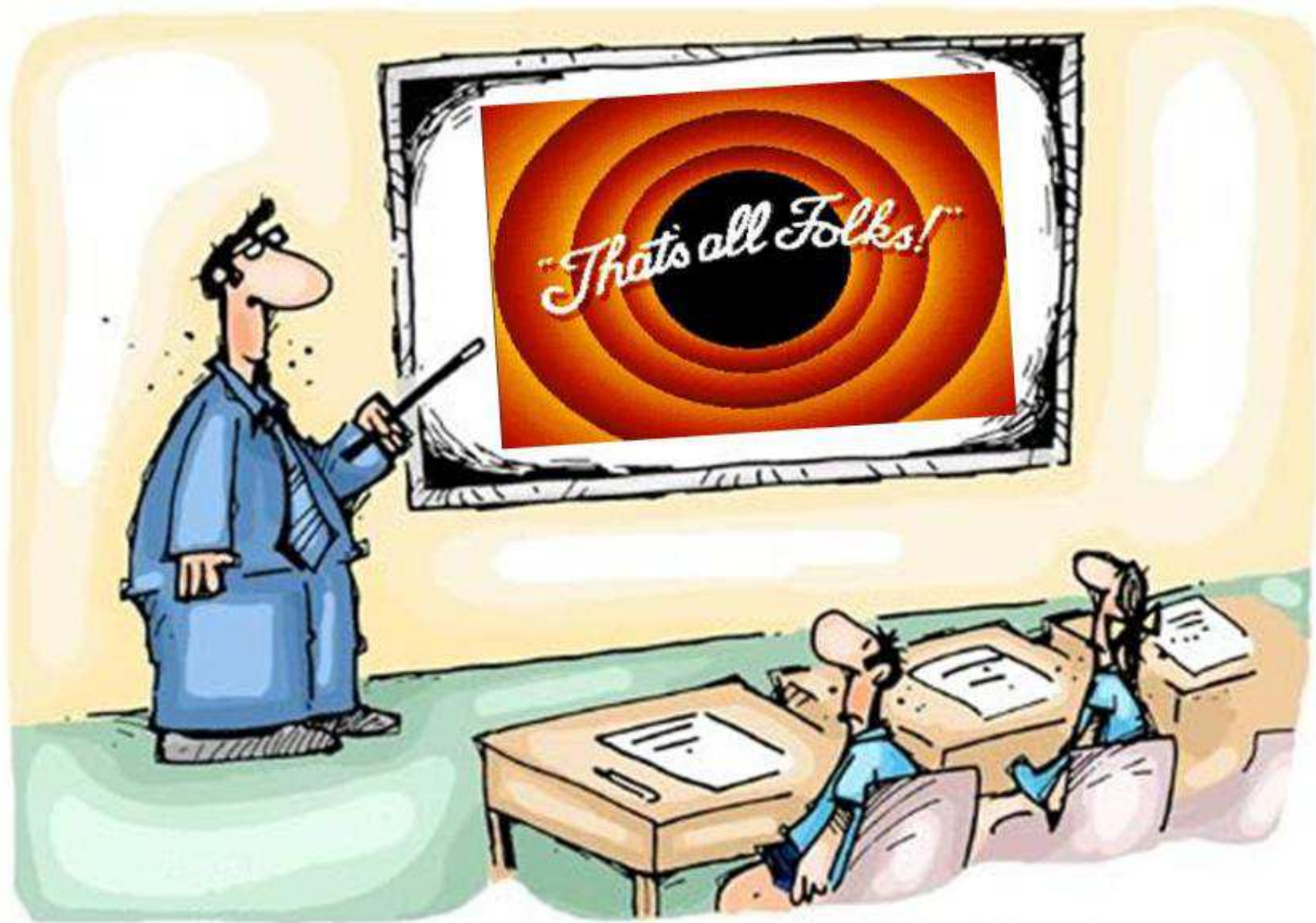
So, the



is more likely to be

S0B

similar to **ILT**





*Wait a minute ,
we have not talked
about
PERCUSSIONS*

CHARGE conservation

$$Q = C V$$

$$Q = C' V'$$

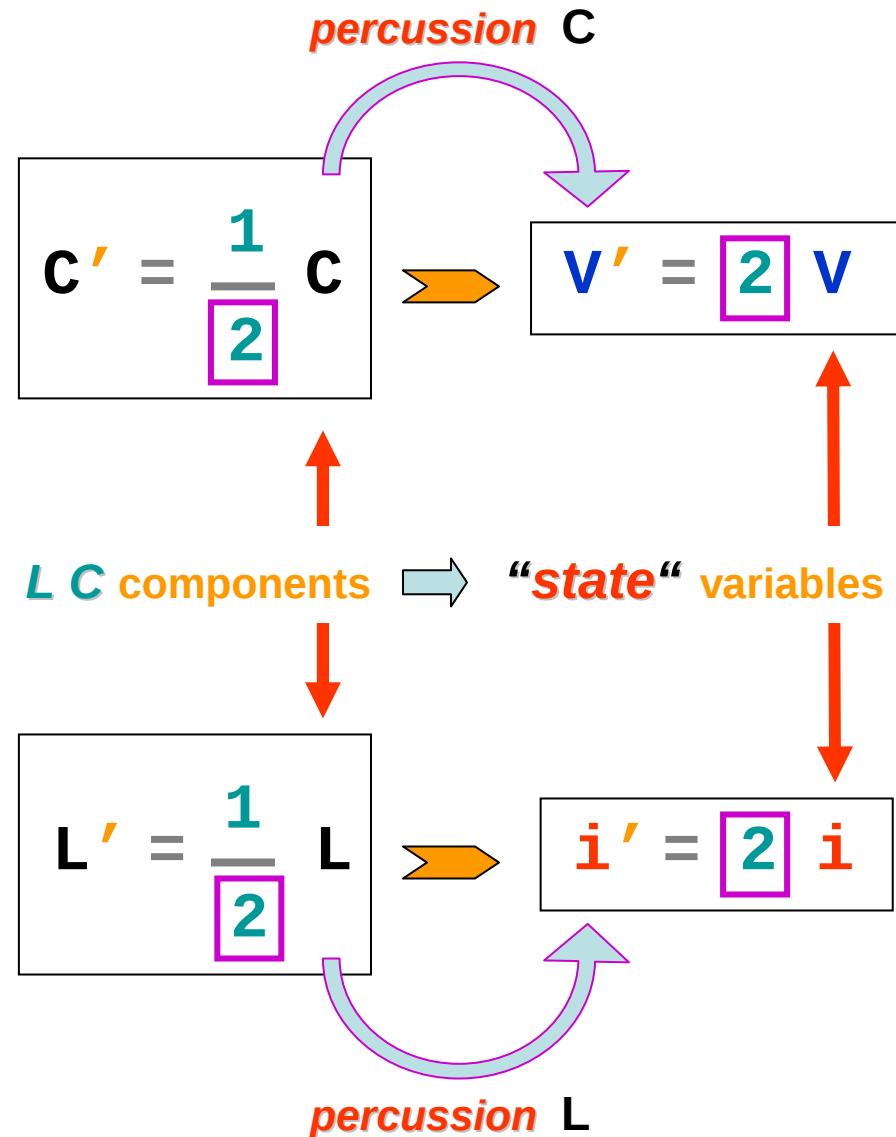
$$V' = \frac{C}{C'} V$$

FLUX conservation

$$\phi = L i$$

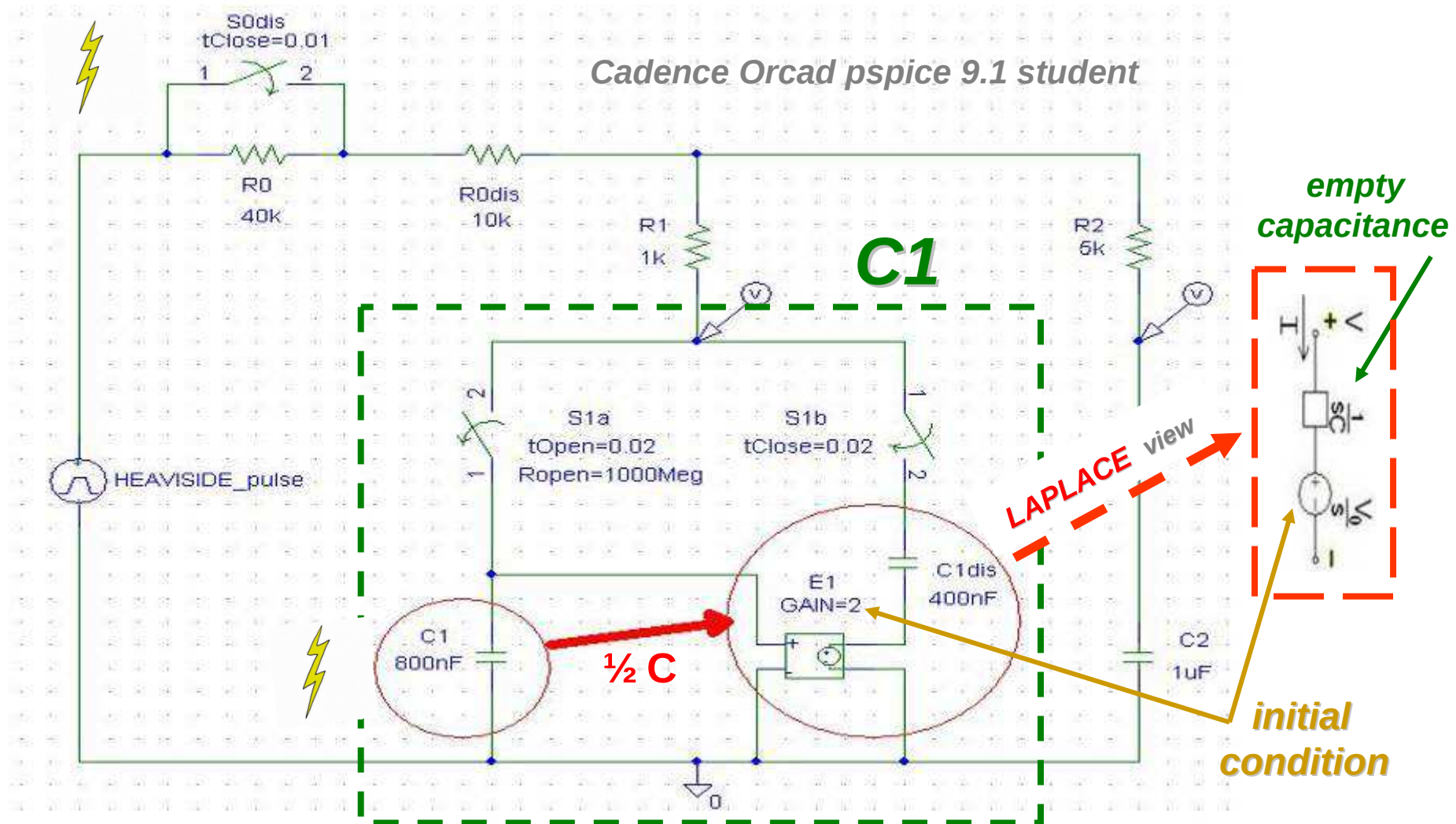
$$\phi = L' i'$$

$$i' = \frac{L}{L'} i$$



PERCUSSION

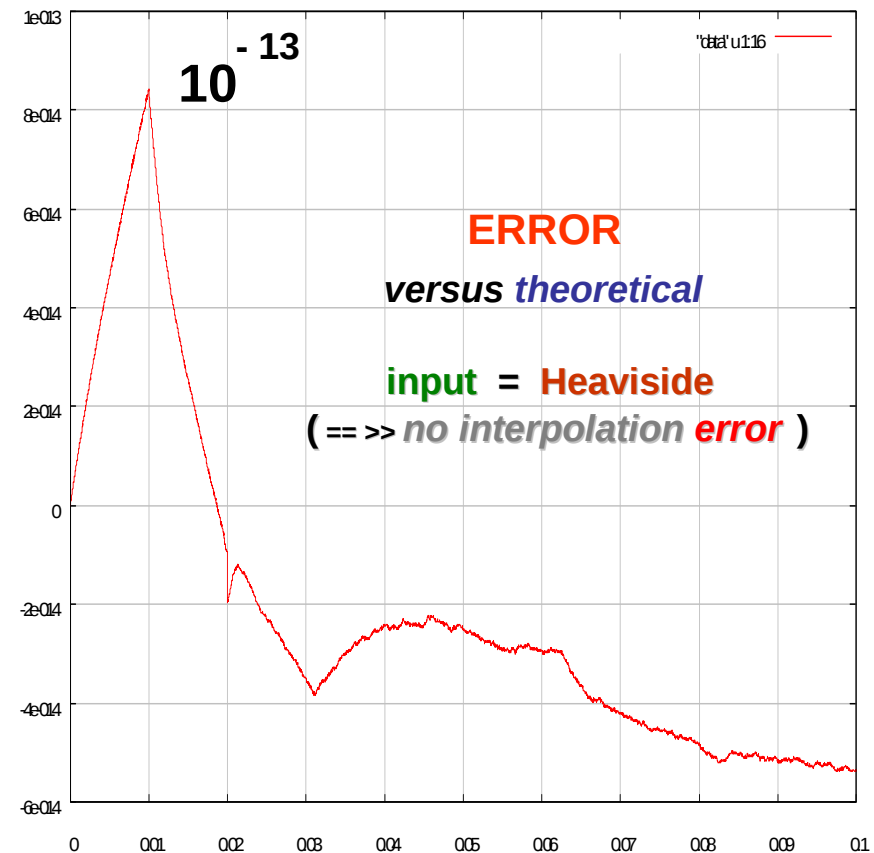
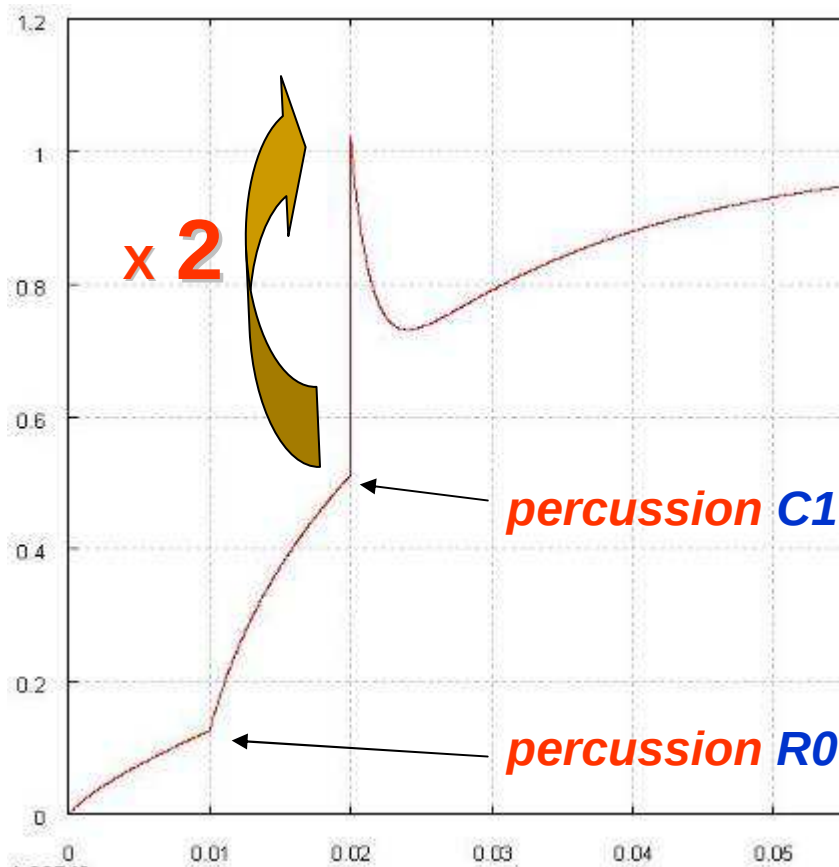
at $t = 0.01$ R0 50 Kohm ==>> 10 Kohm
 at $t = 0.02$ C1 800 nF ==>> 400 nF



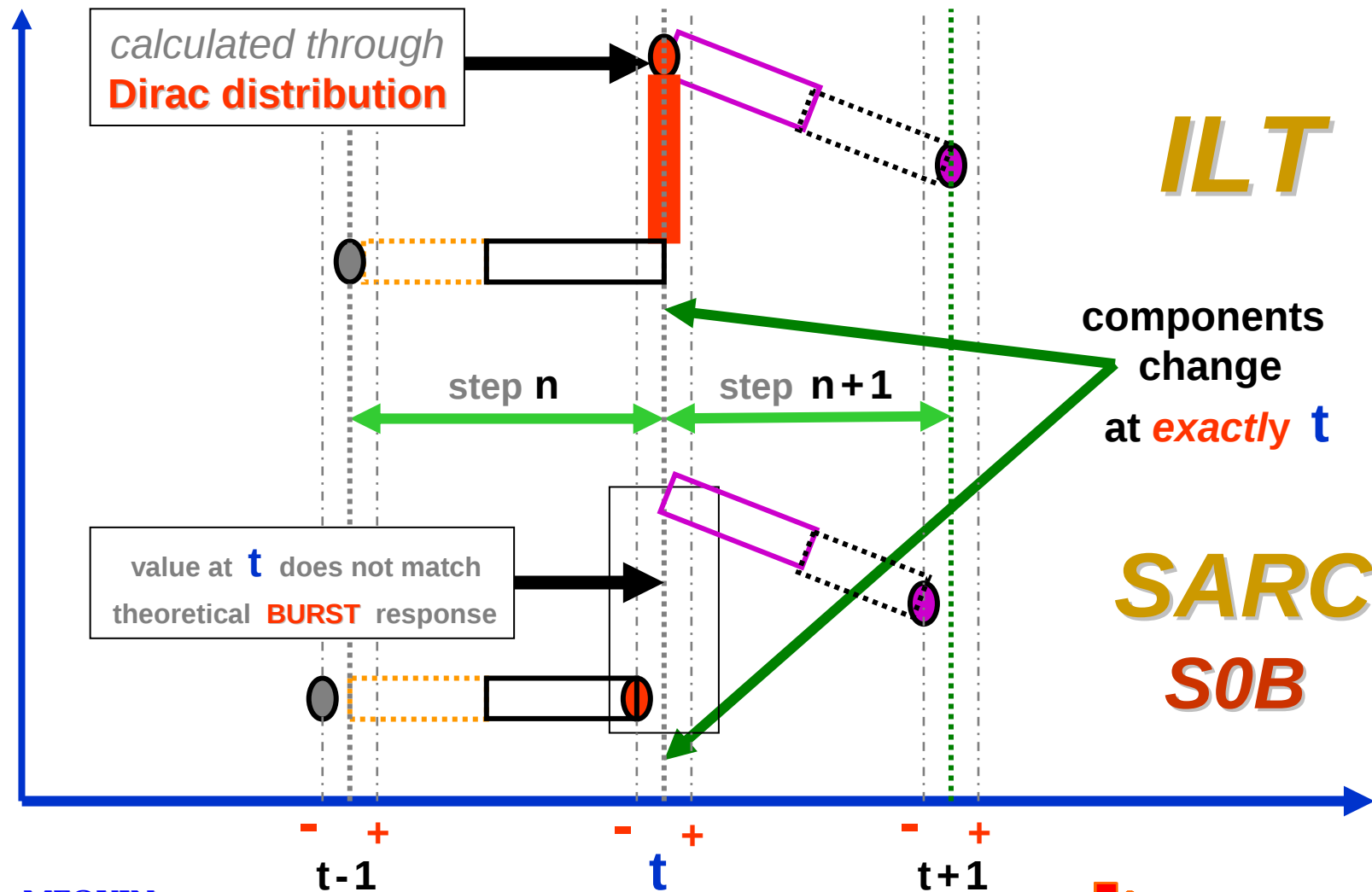
In fact, at least for *network* not “degenerate”, **it is trivial !!!**

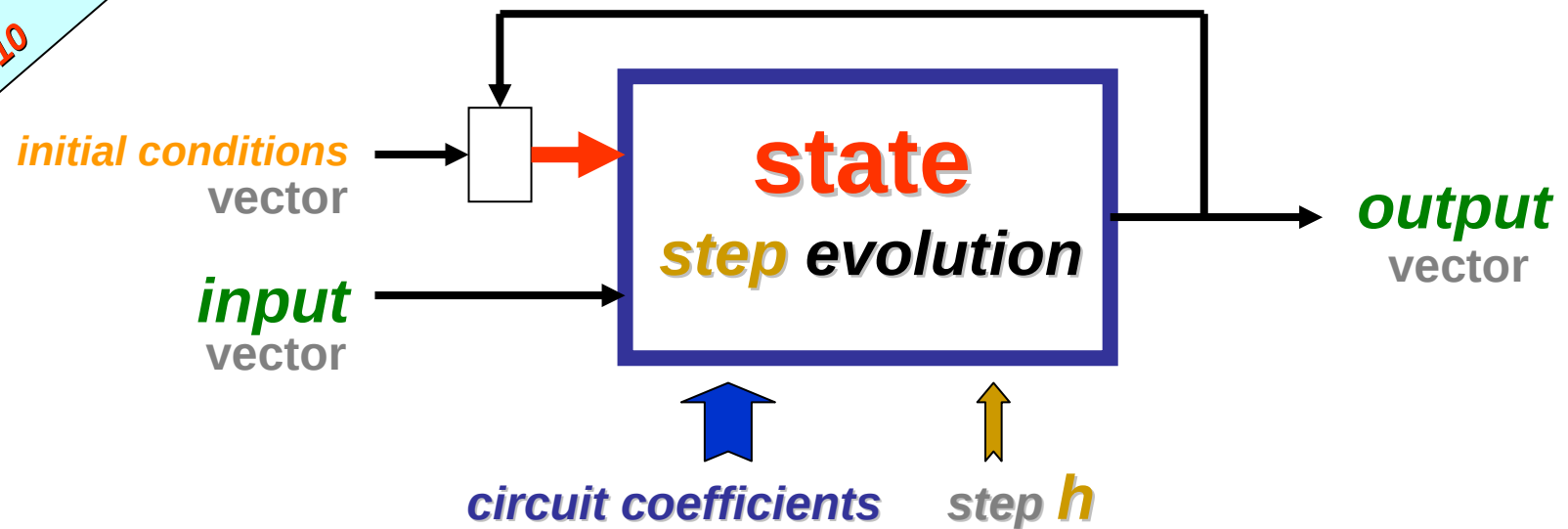
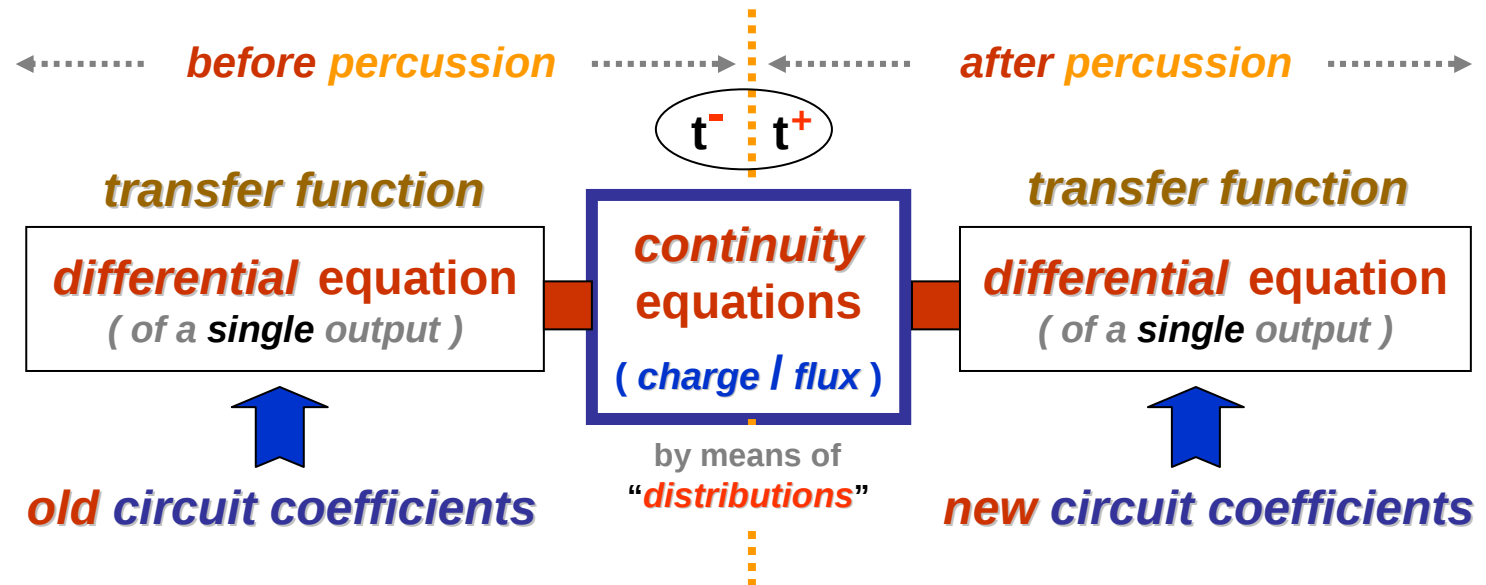
At the correct instant, update the **percuted** component values (*R0* or *C1*)

and for the capacitance *C1* “double” its **STATE VARIABLE** voltage



SARC versus **ILT** percussion





QUESTIONS ?



Component PERCUSSION

of

Linear Time Invariant

analog circuit





What's going on between

t^- and t^+ ???

Digital and Analog simulations are the 2 faces of a same coin

In the **digital world** architects require (through **ESL**) **high level functional** simulations to perform numerous “**what if**” and **analysis** in order to challenge the concept and to **increase confidence** about its validity

Unfortunately in the **analog world** architects do have much less tools and methodologies available to perform equivalent **high level functional** simulations

In many cases, **SPICE** turns to be **the only practical approach**

This is unfortunate because **SPICE** is so **low level** that its **runtime** quickly becomes a bridle to the architect ambition

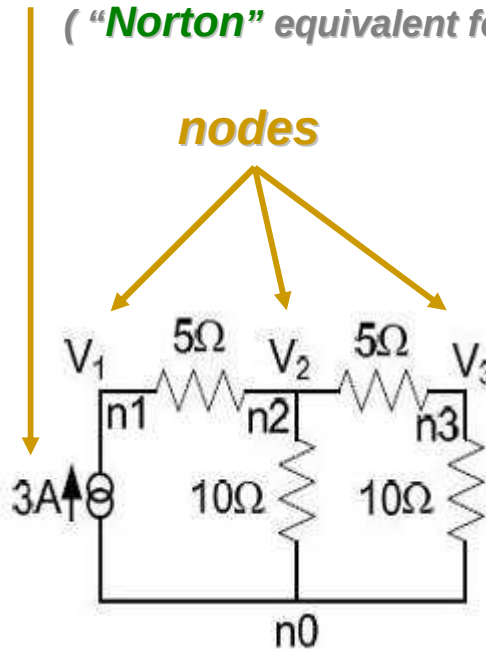
Less experiments, Less tested alternatives, Less confidence ...

SPICE

Simulation **P**rogram with **I**ntegrated **C**ircuit **E**mphasis

current source

(“**Norton**” equivalent for **Voltage**)



“**NODAL**” equations based on **KCL**

(Kirchhoff **C**urrent **L**aw $\sum I = 0$)

$$[G] * [V] = [I]$$

conductance

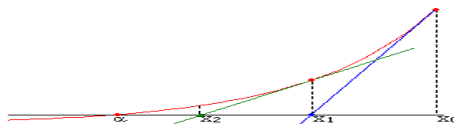
$$\begin{bmatrix} 0.2 & -0.2 & 0 \\ -0.2 & 0.5 & -0.2 \\ 0 & -0.2 & 0.3 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix}$$

“**Gaussian**” elimination

to make it “**triangular**”
then substitutions ...

$$\begin{bmatrix} 0.2 & -0.2 & 0 \\ 0 & 0.3 & -0.2 \\ 0 & 0 & 0.25 \end{bmatrix} \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix}$$

NON-LINEAR COMPONENT behavior is approximated by doing iterations (**convergence**)
based on **Newton-Raphson** method (improvement of an equation root “**guess**”)



$$f(x) \cong f(x_0) + f'(x_0)(x - x_0)$$

$\implies$ **recurrence**

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

NUMERICAL INTEGRATION (for **linear** inductances and capacitances) is approximated
according to **3 user-selectable** methods : **Trapezoidal** , **backward-Euler** , **Gear**

NUMERICAL INTEGRATION (for *linear* inductances and capacitances)

Why several different methods ?

Depending on the “*expected*” waveform, one method may have an advantage over another

What makes them strong or weak ?

Accuracy How much *error* (Local Truncation Error) each method introduces ?

Stability Does the *error* accumulate (that's bad) or decrease to zero (that's good) over time ?

Each method has its own strengths and weaknesses

Trapezoidal

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

*Better accuracy than stability
may oscillate in some cases*

Backward Euler

$$x_{n+1} = x_n + h \cdot \dot{x}_{n+1}$$

Average accuracy and stability

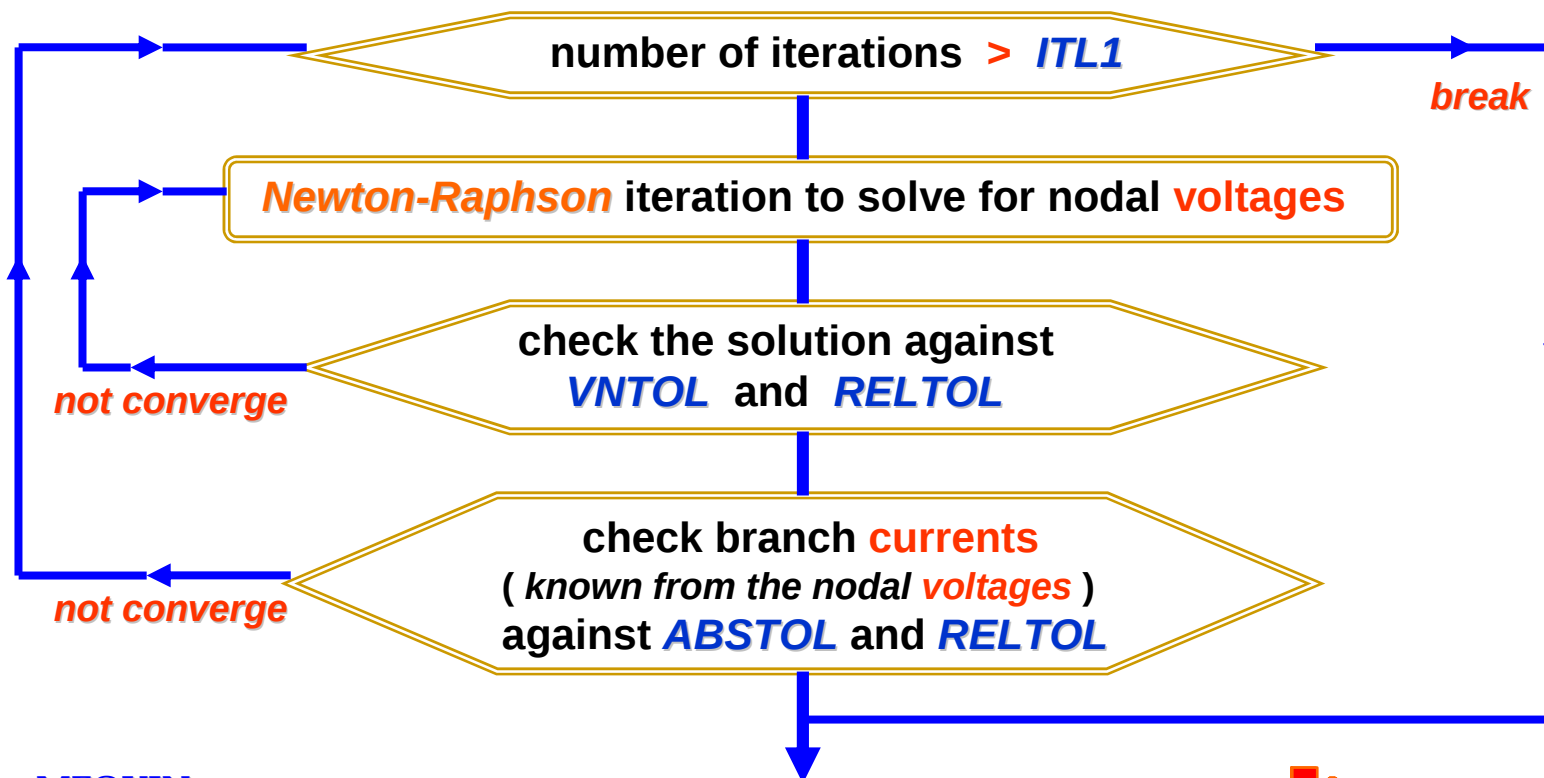
Gear-2

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

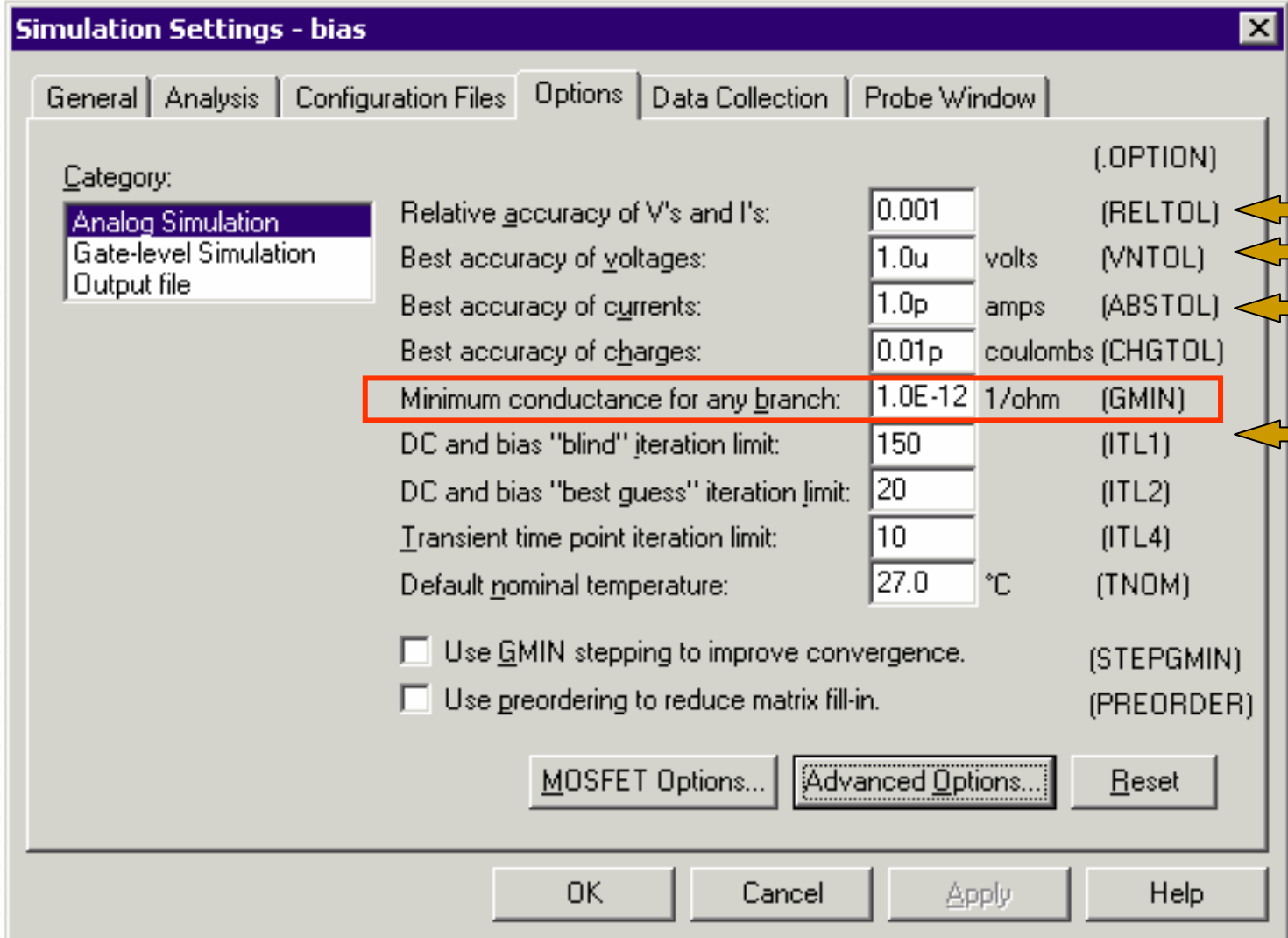
*Better stability than accuracy
Most stable method*

SPICE “convergence” control variables

ITL1	number of iterations allowed for a bias point calculation
VNTOL	node voltage tolerance
RELTOL	fractional tolerance of voltages and currents
ABSTOL	current branch tolerance



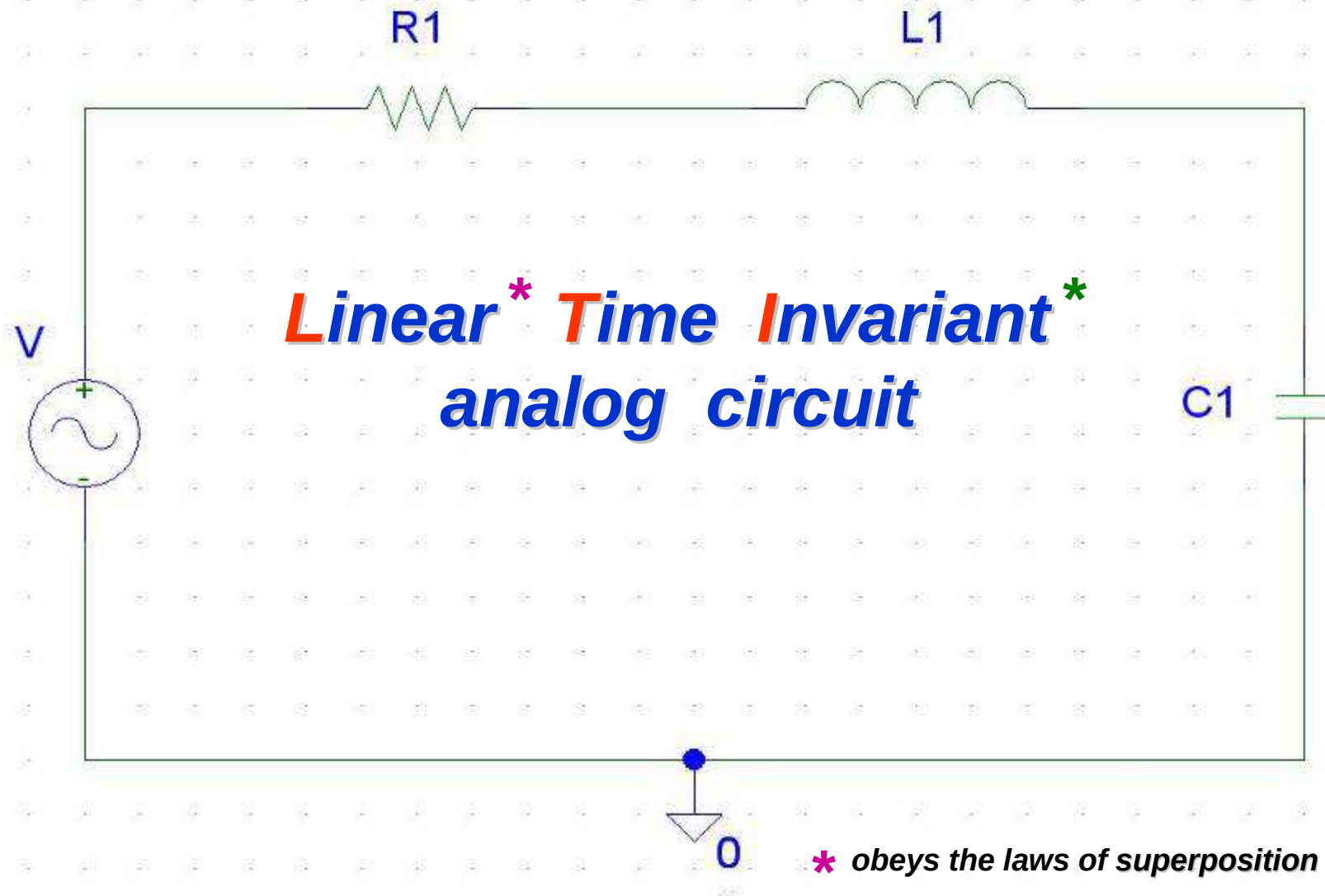
SPICE



The image shows a screenshot of the 'Simulation Settings - bias' dialog box in a SPICE simulation environment. The 'Options' tab is selected, showing various simulation parameters. The 'Category' list on the left includes 'Analog Simulation', 'Gate-level Simulation', and 'Output file'. The 'Analog Simulation' category is selected. The main area contains several parameters with input fields and labels. The 'Minimum conductance for any branch' parameter is highlighted with a red box, and its value '1.0E-12' is shown in the input field. Four yellow arrows point to the right-hand labels of the parameters: (RELTOL), (VNTOL), (ABSTOL), and (GMIN). At the bottom, there are buttons for 'MOSFET Options...', 'Advanced Options...', 'Reset', 'OK', 'Cancel', 'Apply', and 'Help'.

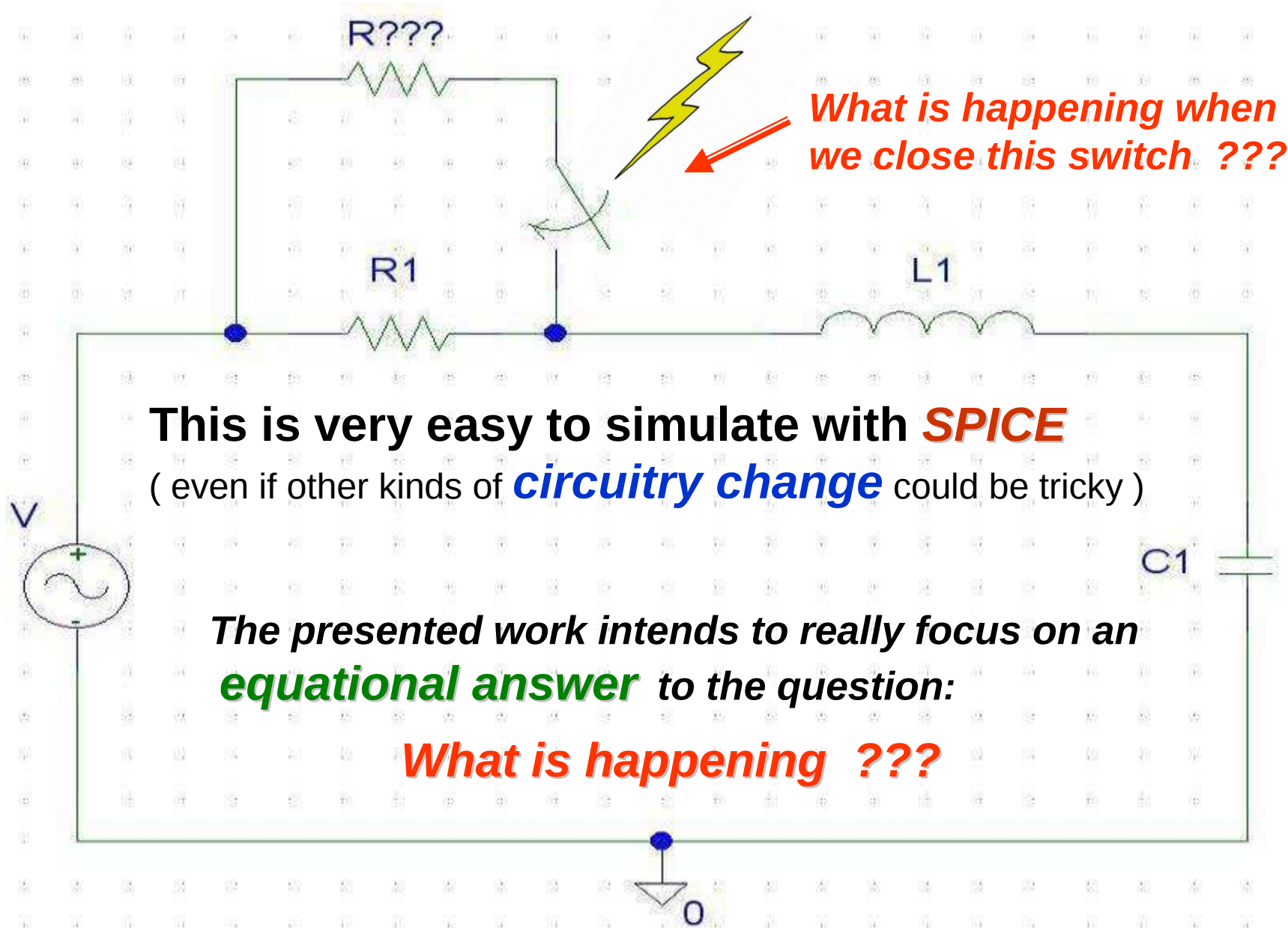
Category:	Parameter	Value	Unit	Label
Analog Simulation	Relative accuracy of V's and I's:	0.001		(RELTOL)
	Best accuracy of voltages:	1.0u	volts	(VNTOL)
	Best accuracy of currents:	1.0p	amps	(ABSTOL)
	Best accuracy of charges:	0.01p	coulombs	(CHGTOL)
	Minimum conductance for any branch:	1.0E-12	1/ohm	(GMIN)
	DC and bias "blind" iteration limit:	150		(ITL1)
	DC and bias "best guess" iteration limit:	20		(ITL2)
	Transient time point iteration limit:	10		(ITL4)
	Default nominal temperature:	27.0	°C	(TNOM)
	☐ Use GMIN stepping to improve convergence.			(STEPGMIN)
☐ Use reordering to reduce matrix fill-in.			(PREORDER)	

Buttons: MOSFET Options..., Advanced Options..., Reset, OK, Cancel, Apply, Help



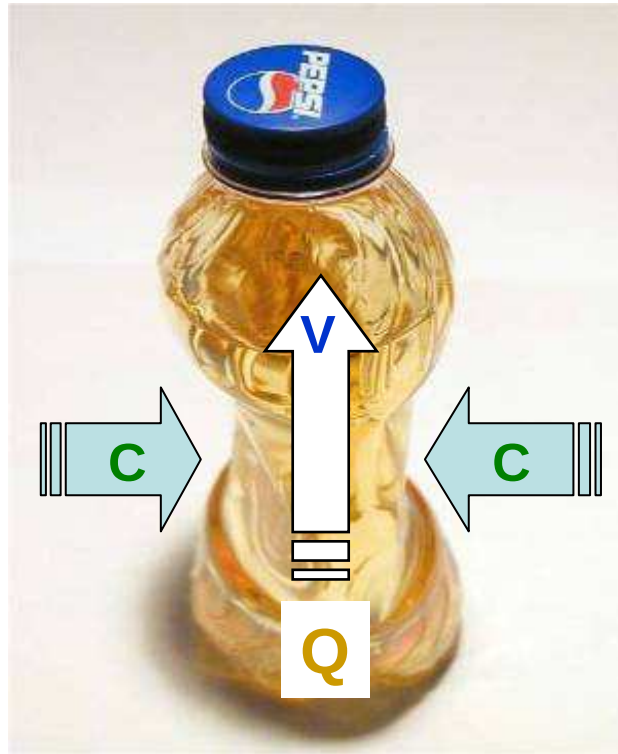
* obeys the laws of superposition

* does not change its behavior over time
(i.e. constant intrinsic parameters)



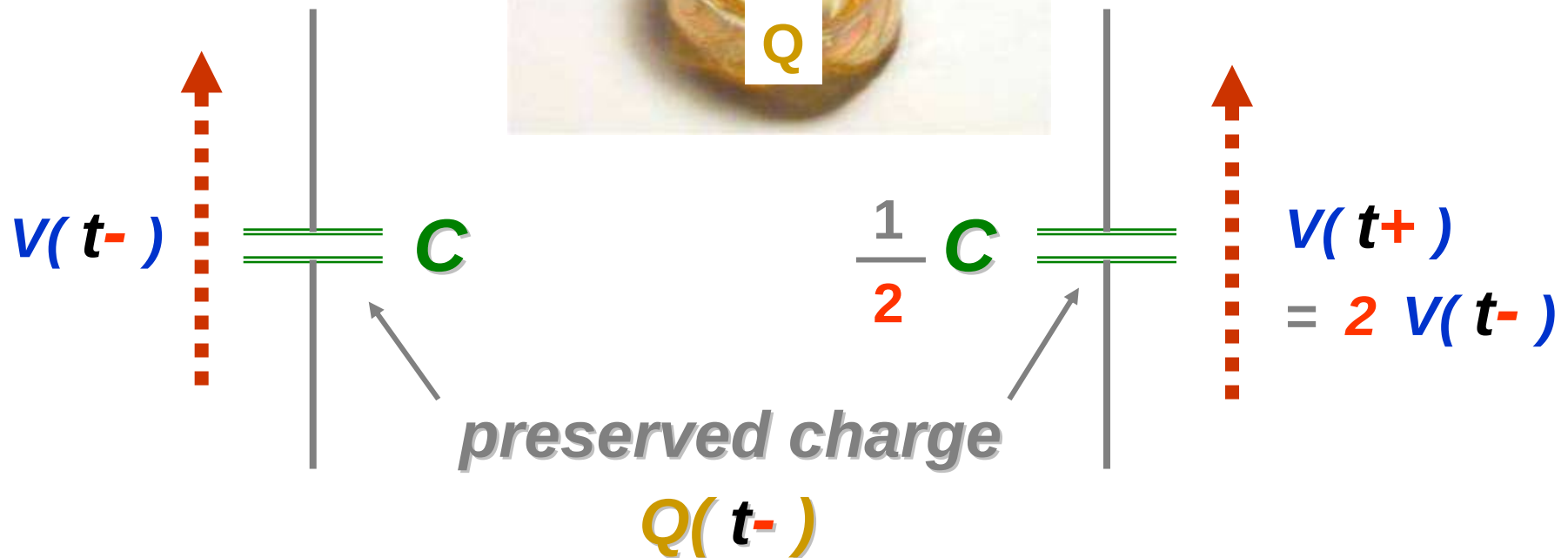
other example:

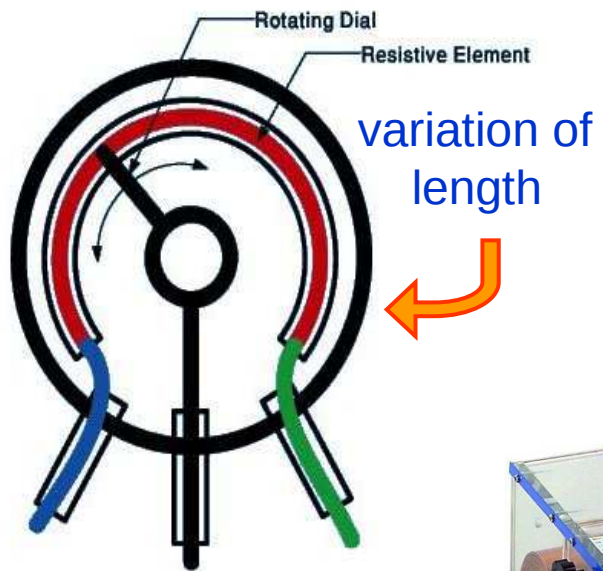
capacitance
change ?



$$Q = C V$$

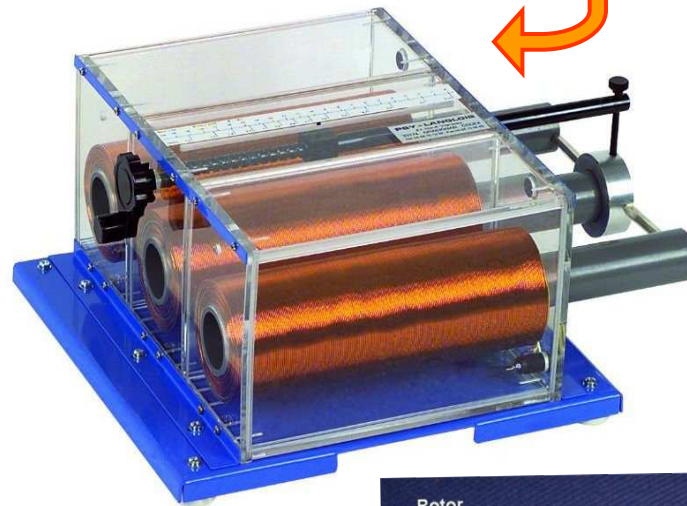
$$V = \frac{Q}{C}$$



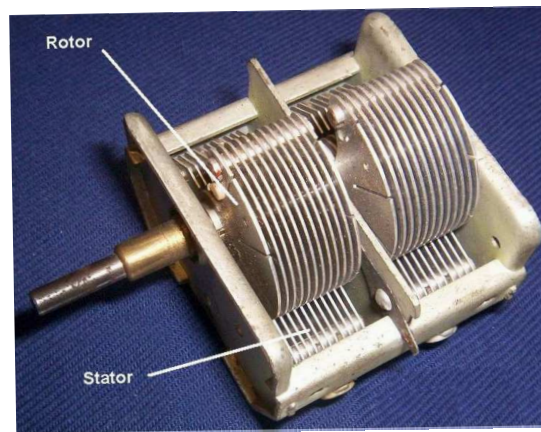


R

variation of permeability (μ)



L



variation of surface

C

percussion

Definition:



1) **musical instruments that are hit:** the group of musical instruments that produce sound by being struck, including drums and cymbals, or the section of the orchestra playing such instruments

2) **act of detonating percussion cap:** the striking or detonating of a percussion cap in a firearm

III → 3) **impact:** the impact of one object striking another, or the noise or ***shock created when two objects hit each other***

4) **medicine tapping of body:** examination of part of a patient's body by tapping with the fingers to assess the presence of fluid, the enlargement of organs, or the solidification of normally hollow parts

Wait a minute, is there anything **wrong** with the title of this presentation ???

Component *PERCUSSION*
of
***Linear* * *Time Invariant* ***
analog circuit

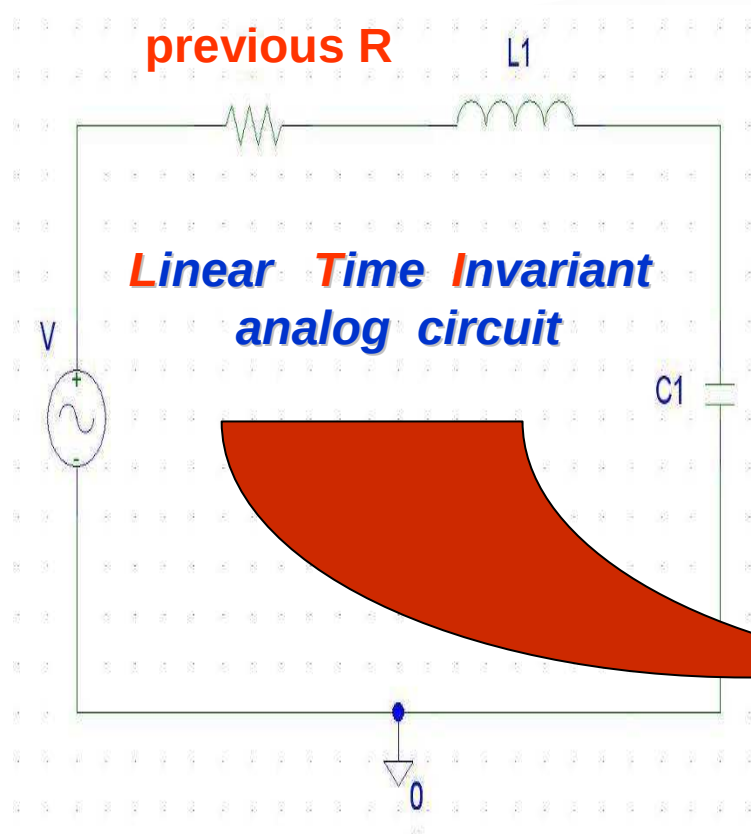
* obeys the laws of superposition

* does not change its behavior over time
(i.e. **constant** intrinsic parameters)

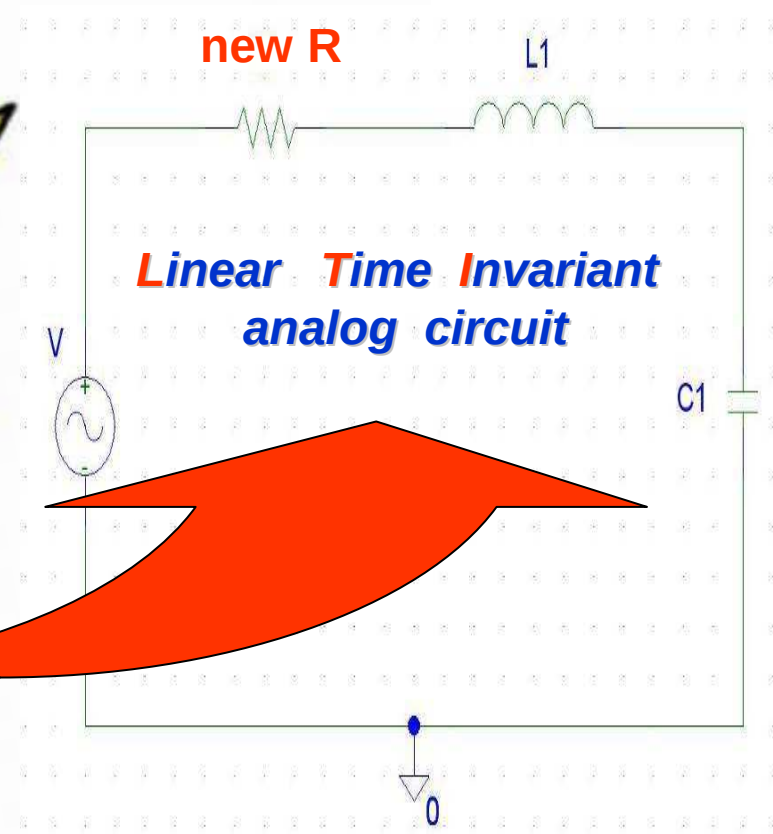
t^-

t

t^+



???

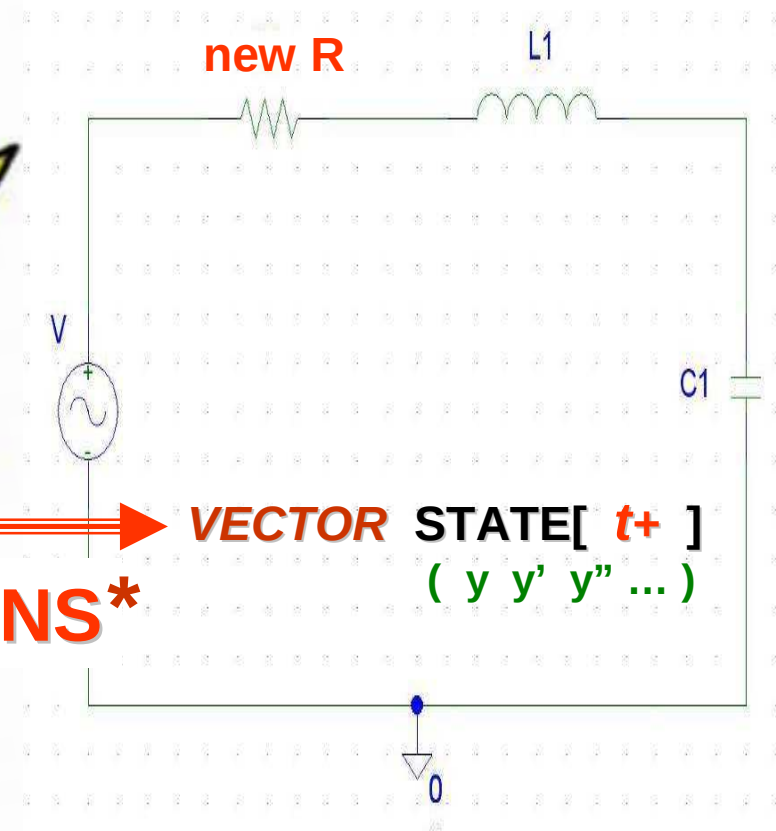
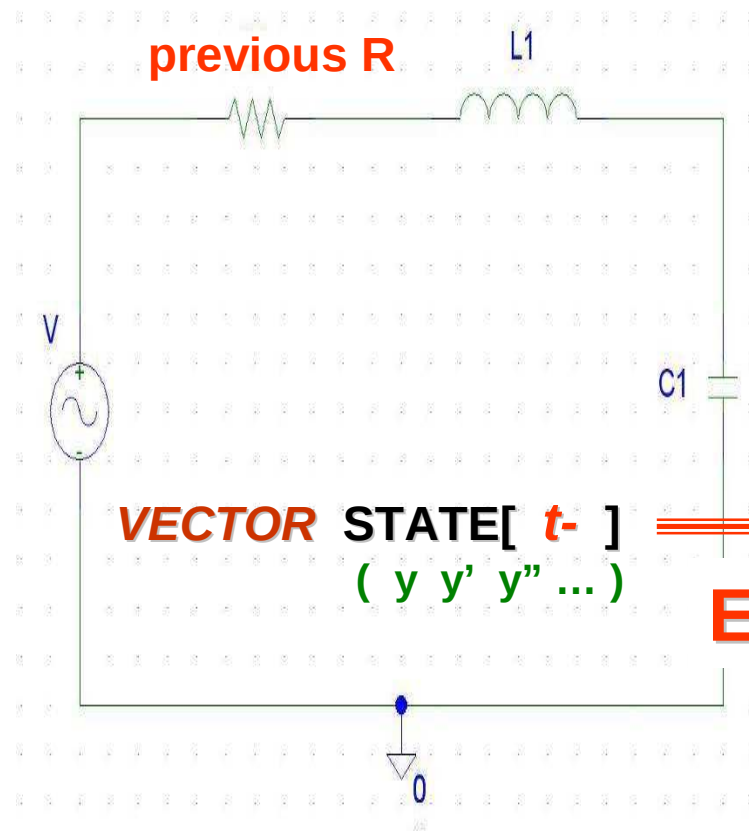


percussion

t^-

t

t^+



EQUATIONS*

percussion

* "zero time" algorithm
based on the circuit
intrinsic parameters
not a convergence !

Moreover,

the circuit **“initialization”**
is just a *particular case* of such
instantaneous state change

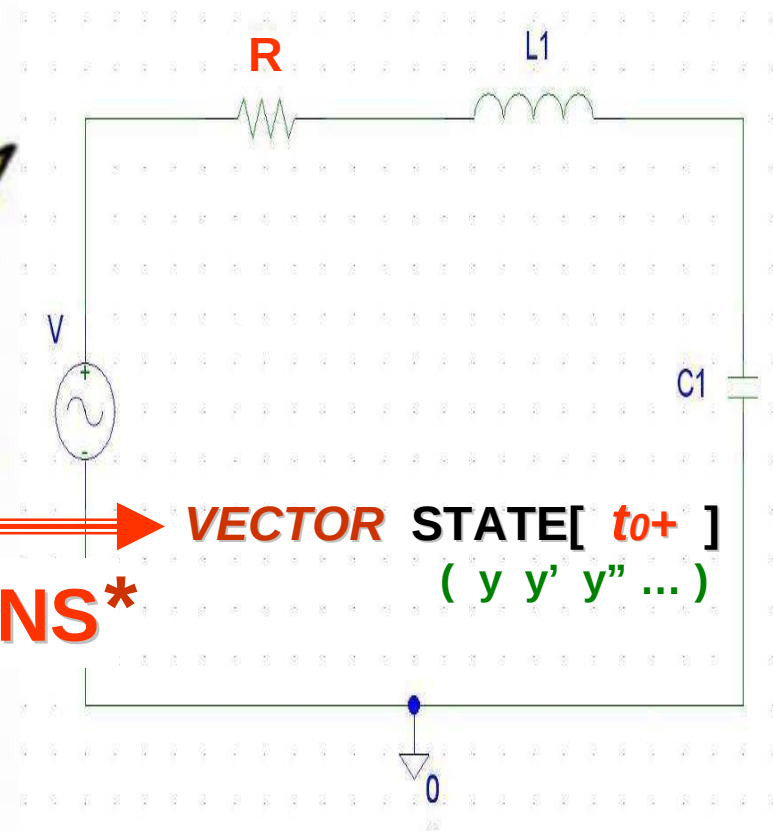
circuit
“at rest”

EQUATIONS*

percussion

t_0

t_0^+



* “zero time” algorithm
based on the circuit
intrinsic parameters
not a convergence !



In summary, a **PERCUSSION** can be

⇒ a “**sharp**” change of the **INPUT** (including **INITIALIZATION**)

⇒ an “**INSTANTANEOUS**” change of **R** , **L** or **C**

However,

the processing of **INPUT** and **COMPONENT PERCUSSIONS** being “similar”
but one modeling an **external shock**
and the other one an **internal structural change**

in this **Inverse Laplace Transform software package**

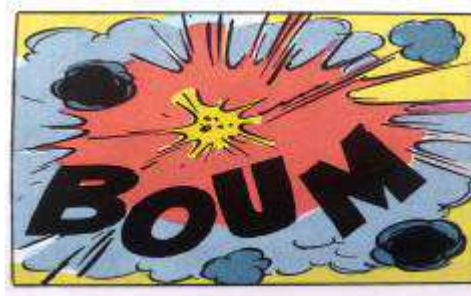
a “**sharp**” change of an **INPUT** is referenced as a **DISCONTINUITY**
but a modification of **R**, **L** or **C** is effectively called a **PERCUSSION**

In fact, we are playing with words



a **discontinuity** on the **INPUT** results in a **percussion** of the circuitry

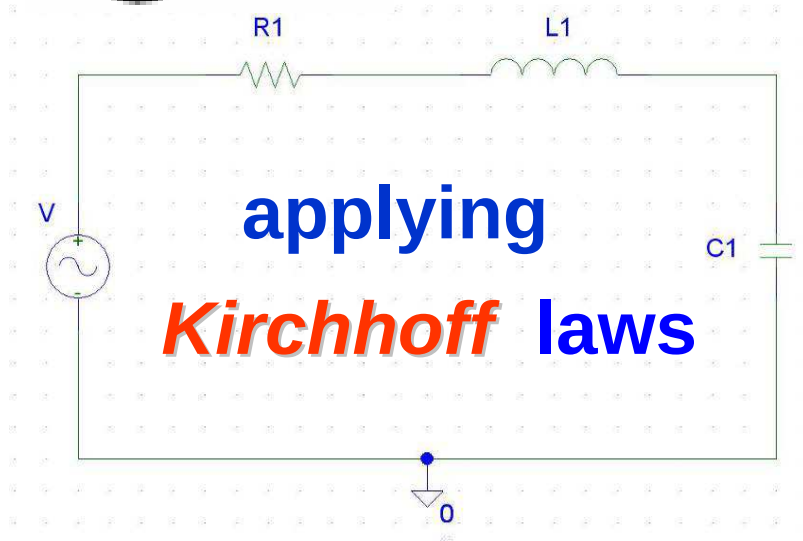
a **percussion** of a **COMPONENT** results in a **discontinuity** on the **OUTPUT**





MAXIMA

Symbolic
(**Computer Algebra System**)



```
EQ : [
    vr1 = R1 * ir1 ,
    vl1 = ZL1 * il1 ,
    vc1 = ZC1 * ic1 ,
    ir1 = il1 ,
    ir1 = ic1 ,
    V = vr1 + vl1 + vc1
] ;
```

```
solve( eliminate( EQ , [ vl1, vc1, ir1, il1, ic1 ] ) , vr1 ) ;
solve( eliminate( EQ , [ vr1, vl1, vc1, ir1, ic1 ] ) , il1 ) ;
solve( eliminate( EQ , [ vr1, vl1, ir1, il1, ic1 ] ) , vc1 ) ;
```

$$[\text{vr1} = \frac{R1 V}{ZL1 + ZC1 + R1}]$$

$$[\text{il1} = \frac{V}{ZL1 + ZC1 + R1}]$$

$$[\text{vc1} = \frac{V ZC1}{ZL1 + ZC1 + R1}]$$



UNFORTUNATLY,



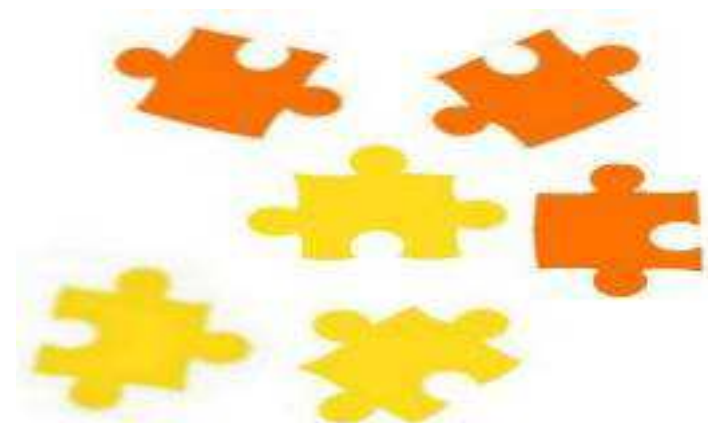
Multiple approaches exist **to model** such kind of filters :

★ *Laplace Transform*

★ *State Model*

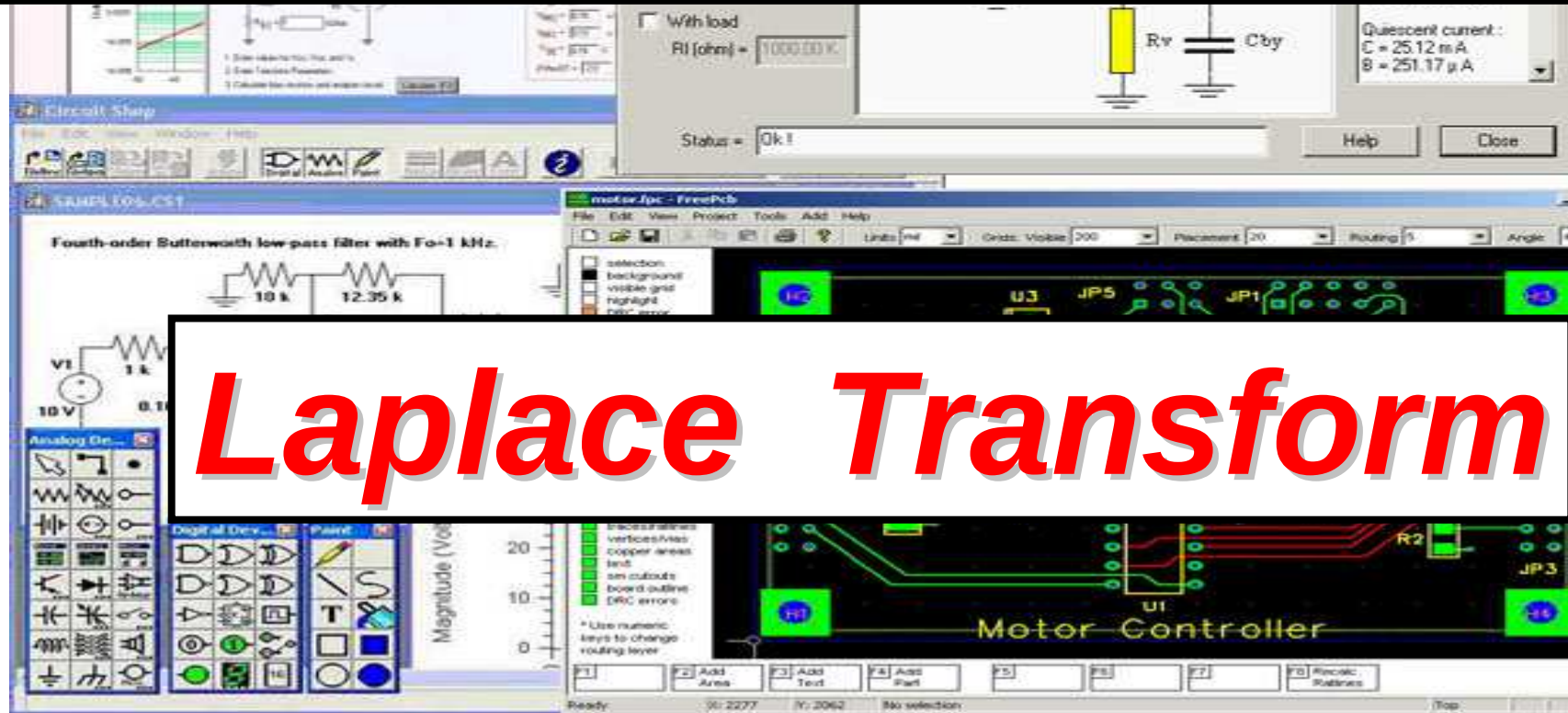
★ *Differential equations*

⊕ *Differential equations* by means of **Distributions**





first modeling approach



Laplace Transform

LAPLACE transform

$$L\{f(t)\} = F(s) = \int_0^{\infty} f(t) e^{-st} dt$$

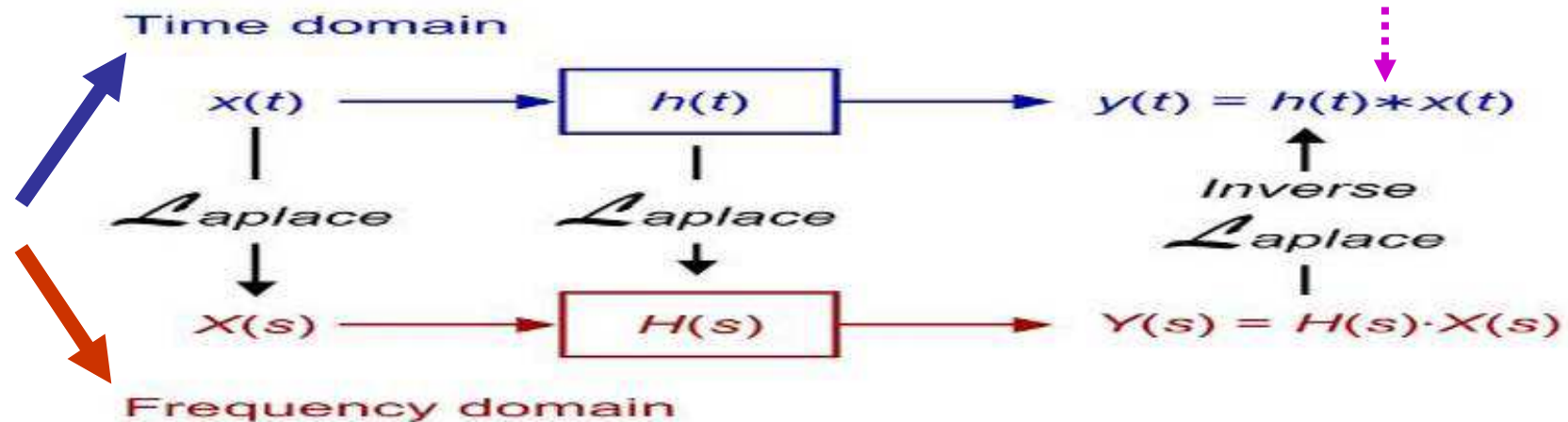
$$s = \alpha + j\omega$$



1749 - 1827

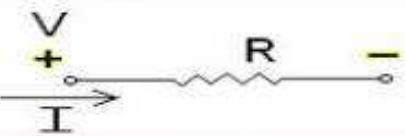
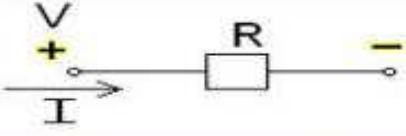
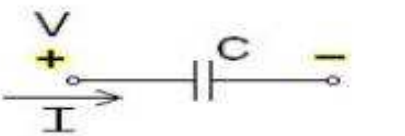
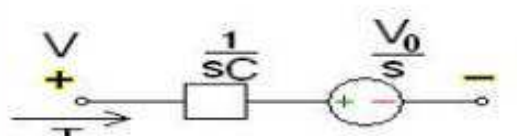
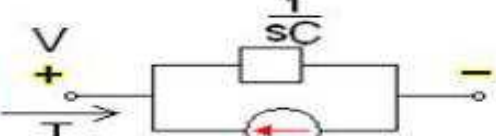
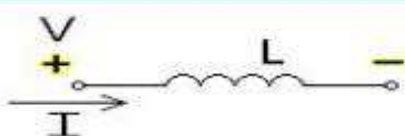
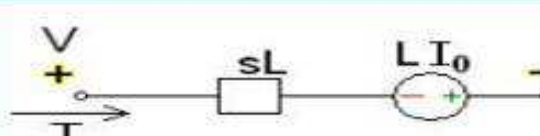
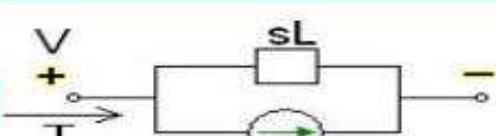
for **causal** **energy bounded** **signals**

$$\begin{array}{lcl} \text{Fourier} & = & \text{Laplace} \\ \mathcal{F}x(f) & = & \mathcal{L}x(j2\pi f) \end{array}$$



Laplace

“**s**” (or “**p**”) transform

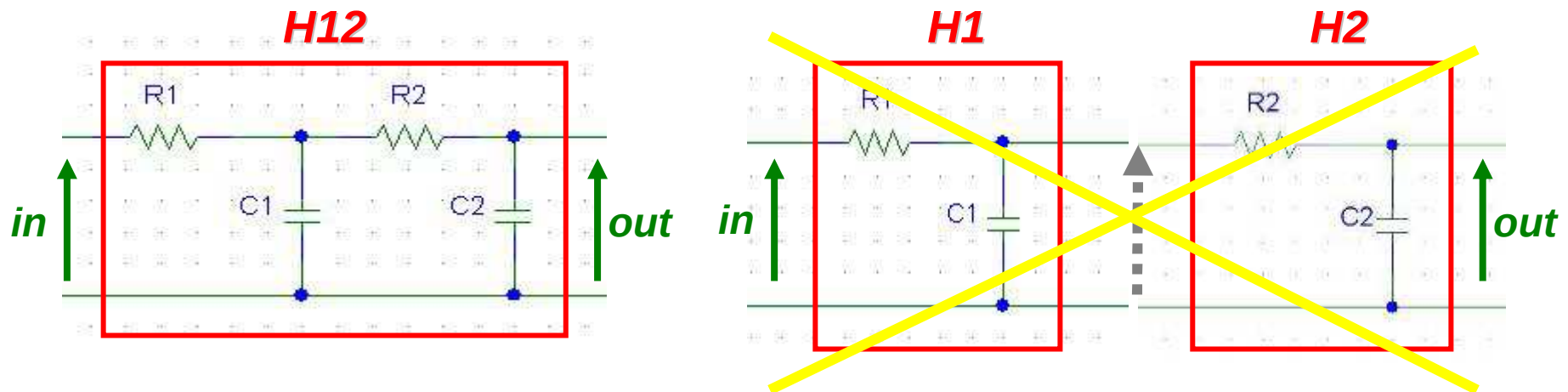
Time Domain	s-Domain	
		(same as normal)
		or 
		or 

terminology

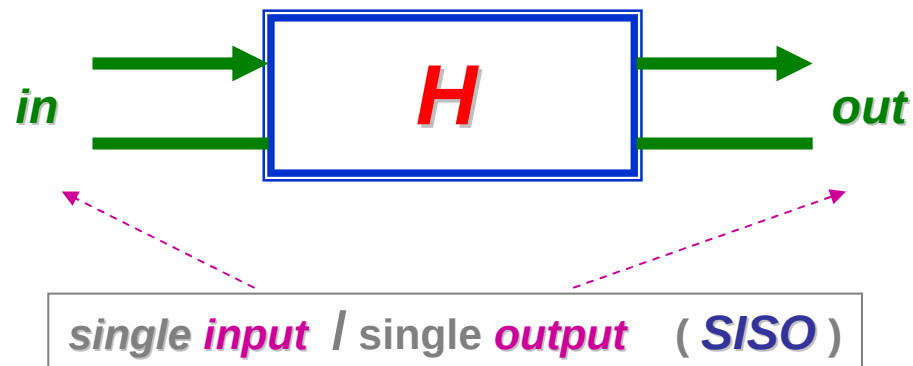
resistance R
conductance G

impedance $Z(s)$
admittance $Y(s)$ } immittance

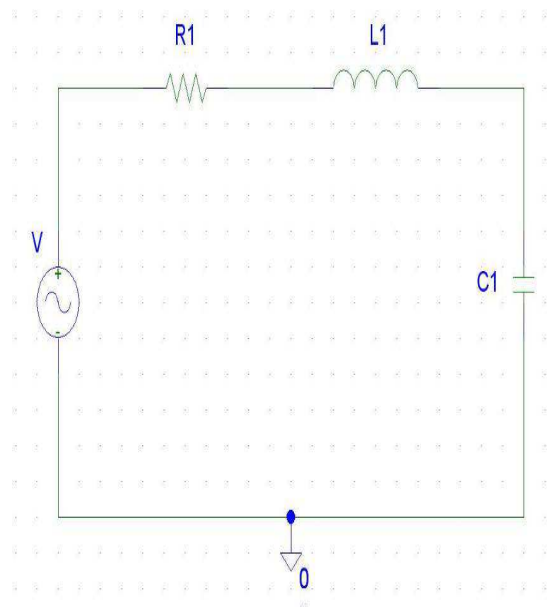
important NOTE



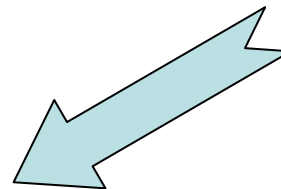
The concept of **Transfer Function** assumes that the **input** and **output** are **ideal** (= no **interfering impedance**)



UNIT gain **OPAMP** may be used to safely cascade **FILTERS**



$$\begin{aligned} [v_{r1} &= \frac{R1 V}{Z_{L1} + Z_{C1} + R1}] \\ [i_{l1} &= \frac{V}{Z_{L1} + Z_{C1} + R1}] \\ [v_{c1} &= \frac{V Z_{C1}}{Z_{L1} + Z_{C1} + R1}] \end{aligned}$$



$$\begin{aligned} Z_{L1} &= L1 * S \\ Z_{C1} &= \frac{1}{C1 * S} \end{aligned}$$



$$\begin{aligned} \frac{v_{r1}}{V} &= \frac{C1 R1 s}{C1 L1 s^2 + C1 R1 s + 1} \\ \frac{i_{l1}}{V} &= \frac{C1 s}{C1 L1 s^2 + C1 R1 s + 1} \\ \frac{v_{c1}}{V} &= \frac{1}{C1 L1 s^2 + C1 R1 s + 1} \end{aligned}$$

Those **3** “**transfer functions**” have the general form
 (**transmittance**)

$$\frac{Y(s)}{X(s)} = \frac{s^2 N_2 + s N_1 + N_0}{s^2 D_2 + s D_1 + D_0}$$

An expression of the **Y(t)** “**time response**” can be determined

when

$$\frac{s^2 N_2 + s N_1 + N_0}{s^2 D_2 + s D_1 + D_0} X(s)$$

(or its **decomposition**) can be
 retrieved from an

Inverse Laplace Transform Table

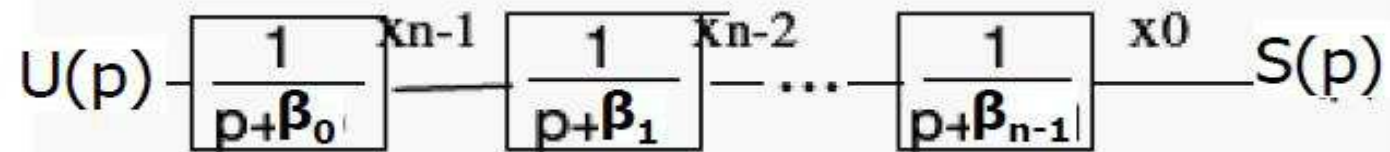


Obviously, this supposes that a “**S**” **transform** of the **INPUT X(t)** is known !!!

DECOMPOSITION

CASCADE

$$H(p) = \frac{1}{(p + \beta_0)(p + \beta_1) \dots (p + \beta_{n-1})}$$



MODAL

$$H(p) = \frac{\alpha_0}{p + \beta_0} + \frac{\alpha_1}{p + \beta_1} + \dots + \frac{\alpha_{n-1}}{p + \beta_{n-1}}$$

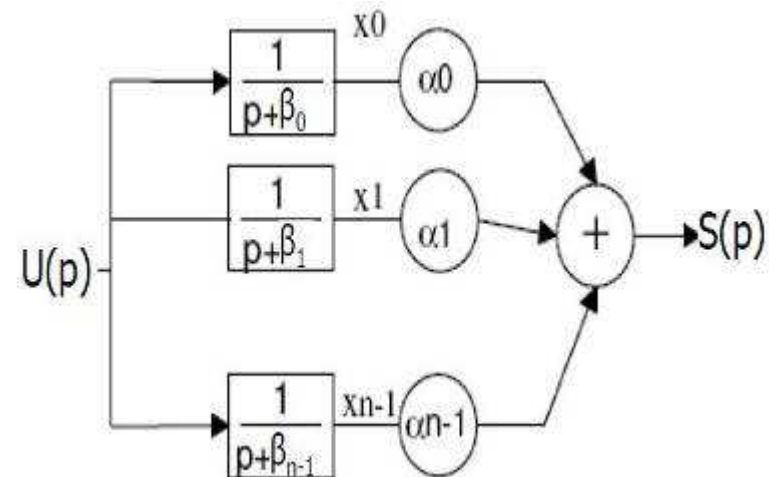
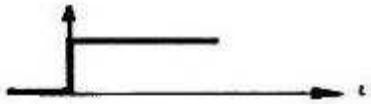
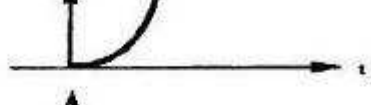
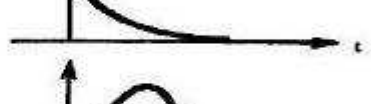
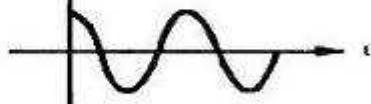
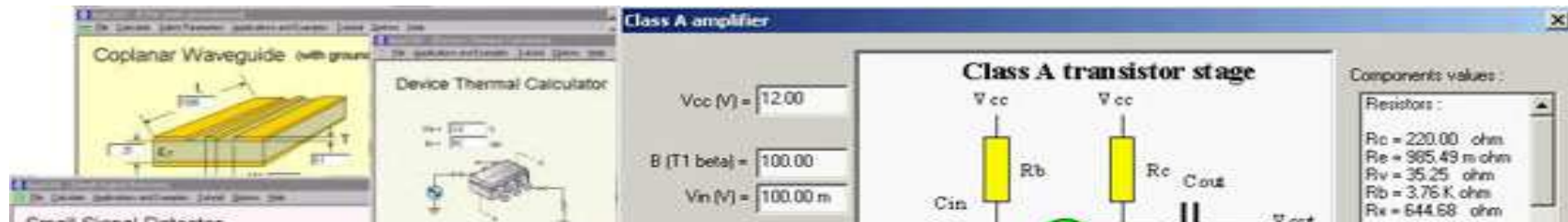


TABLE EXTRACT

Unit $\frac{1}{p}$	HEAVISIDE	1	
Speed $\frac{1}{p^2}$		t	
Acceleration $\frac{1}{p^n}$		$\frac{t^{n-1}}{(n-1)!}$	
Dirac 1	DIRAC	at	
1 order $\frac{1}{p+a}$		e^{-at}	
2 order $\frac{1}{(p+a)^2}$		te^{-at}	
(n+1) order $\frac{1}{(p+a)^{n+1}}$		$\frac{t^n e^{-at}}{n!}$	
Cosinus ... $\frac{p}{p^2 + \omega^2}$		$\cos \omega t$	
$\frac{p}{p^2 - \omega^2}$		$\text{ch } \omega t$	
Sinus ... $\frac{\omega}{p^2 + \omega^2}$		$\sin \omega t$	
$\frac{\omega}{p^2 - \omega^2}$		$\text{sh } \omega t$	
$\frac{p+a}{(p+a)^2 + \omega^2}$		$e^{-at} \cos \omega t$	
$\frac{ap+\beta}{(p+a)^2 + \omega^2}$		$\Lambda e^{-at} \cos(\omega t + \varphi)$ $A = \frac{1}{\omega} \sqrt{a^2 \omega^2 + (\beta - a\alpha)^2}$ $\varphi = -\arctan(\beta - a\alpha)$	



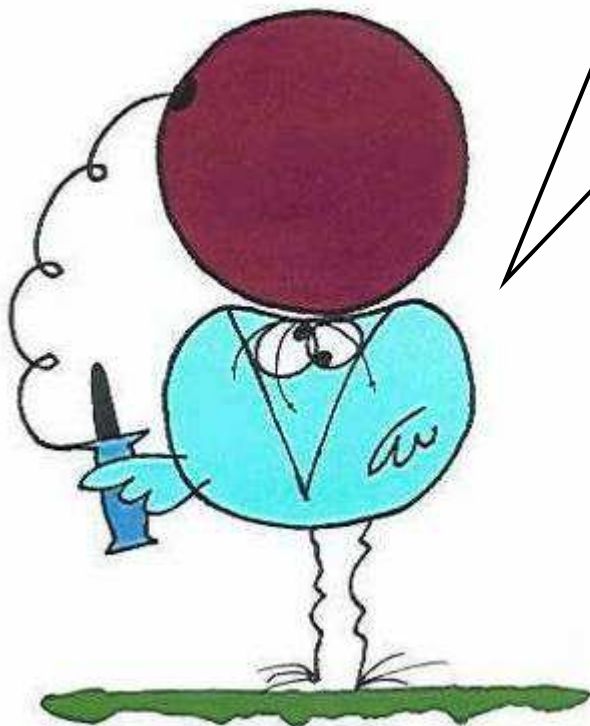
second modeling approach



STATE MODEL

STATE MODEL

*look at its dedicated
power point presentation*



Component PERCUSSION

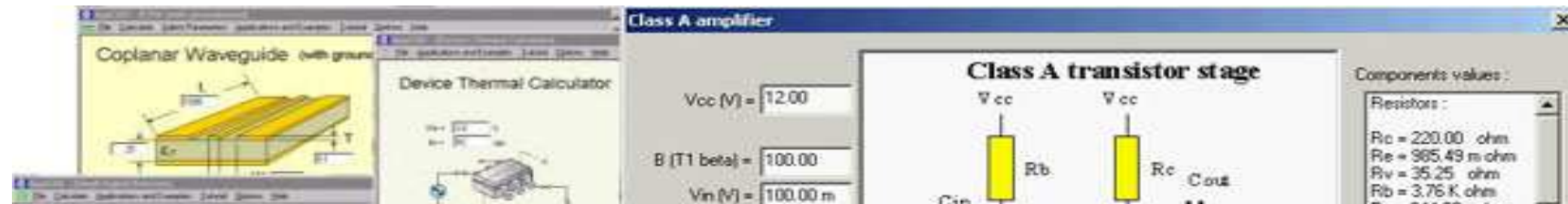
episode II

Jacques MEQUIN
NICE EWTBU / DTE

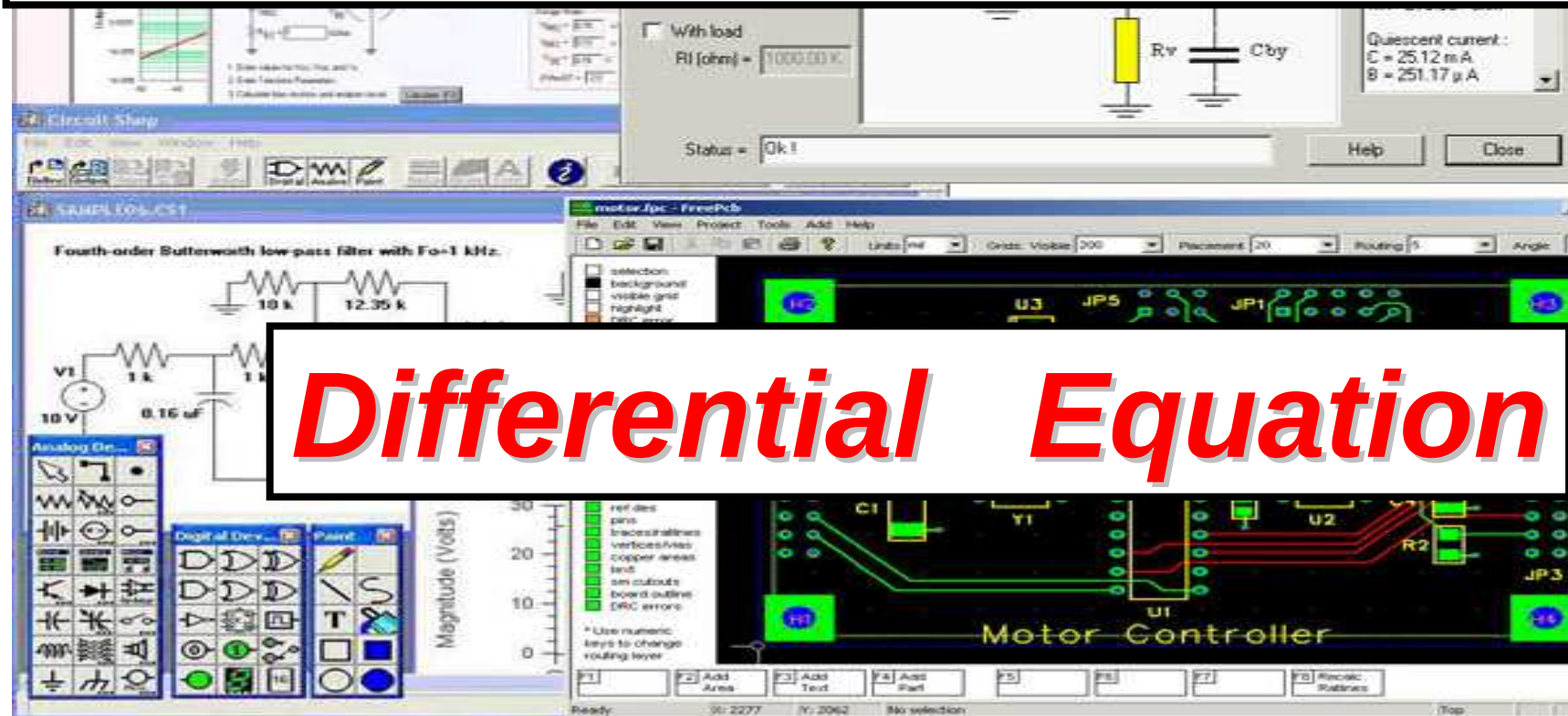


TEXAS INSTRUMENTS
PROPRIETARY INFORMATION
INTERNAL DATA





third modeling approach



Differential Equation

Differential Equation

A **generic method** to **simulate** the “**time domain**” response of the **transfer function**

$$\frac{Y(s)}{X(s)} = \frac{s^2 N2 + s N1 + N0}{s^2 D2 + s D1 + D0}$$

is to use its equivalent **time domain** “**differential equation**”[★]

$$\left(\frac{d^2}{dt^2} Y(t) \right) D2 + \left(\frac{d}{dt} Y(t) \right) D1 + Y(t) D0 = \left(\frac{d^2}{dt^2} X(t) \right) N2 + \left(\frac{d}{dt} X(t) \right) N1 + X(t) N0$$

This only requires that **samples** of the **INPUT X(t)** can be evaluated

★ **LCCODE** : linear constant coefficient ordinary differential equation

They are numerous methods available to **integrate Ordinary Differential Equation (ODE)**

Among them, the **RK4** method (“**fourth order**” **Runge-Kutta**) is both **accurate** (error **$O(h^5)$**) and **easy** to implement

Expecting for **$y(t + h)$** an expression such as

$$y_{i+1} = y_i + (a_1 k_1 + a_2 k_2 + a_3 k_3 + a_4 k_4) (x_{i+1} - x_i) \quad \text{assuming} \quad \frac{dy}{dx} = f(x, y)$$

it is based on the determination of the value of the coefficients that “**match**” the **Taylor series** expansion at **order 4**

$$y_{i+1} = y_i + \frac{dy}{dx} \Big|_{x_i, y_i} (x_{i+1} - x_i) + \frac{1}{2!} \frac{d^2 y}{dx^2} \Big|_{x_i, y_i} (x_{i+1} - x_i)^2 + \frac{1}{3!} \frac{d^3 y}{dx^3} \Big|_{x_i, y_i} (x_{i+1} - x_i)^3 + \frac{1}{4!} \frac{d^4 y}{dx^4} \Big|_{x_i, y_i} (x_{i+1} - x_i)^4$$

due to supernumerary (*free*) variables,
several possible **set of coefficients $k_1, \dots, k_4$**
are acceptable solutions

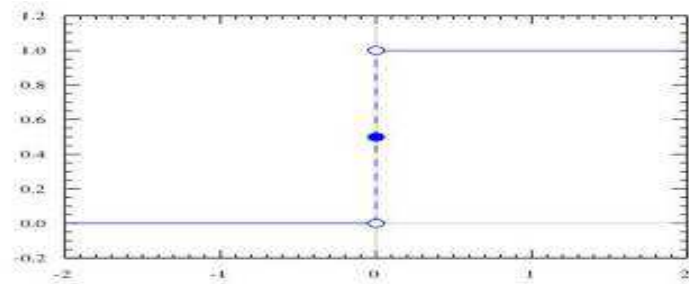
The following choice is generally very convenient

$$\begin{aligned} k_1 &= \Delta t f(t_i, y_i) \\ k_2 &= \Delta t f\left(t_i + \frac{1}{2} \Delta t, y_i + \frac{1}{2} k_1\right) \\ k_3 &= \Delta t f\left(t_i + \frac{1}{2} \Delta t, y_i + \frac{1}{2} k_2\right) \\ k_4 &= \Delta t f(t_i + \Delta t, y_i + k_3) \\ y_{i+1} &= y_i + \frac{1}{6} (k_1 + k_2 + k_3 + k_4) \end{aligned}$$

DISTRIBUTIONS

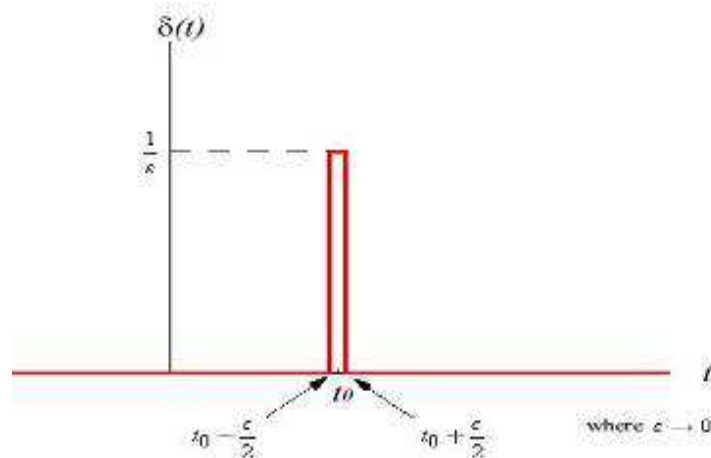
< T , ϕ >

HEAVISIDE “generalized function”



$$H(x) = \begin{cases} 0 & x < 0 \\ 1 & x > 0 \end{cases}$$

DIRAC ✦ “generalized function”



$$\delta(x) \equiv \frac{dH}{dx} = \begin{cases} 0 & x < 0 \\ \infty & x = 0 \\ 0 & x > 0 \end{cases}$$

$$x\delta'(x) = -\delta(x)$$

$$\int_{-\infty}^{\infty} \delta(x) f(x) dx = f(0)$$

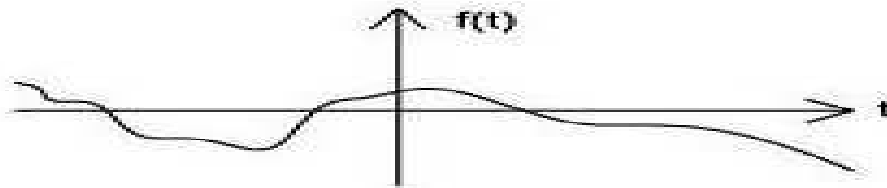
$$\int_{-\infty}^{\infty} a\delta(x) dx = a$$

✦ **Laurent Schwartz** established it as a “**measure**” (a particular subspecies of his “**distributions**” theory)

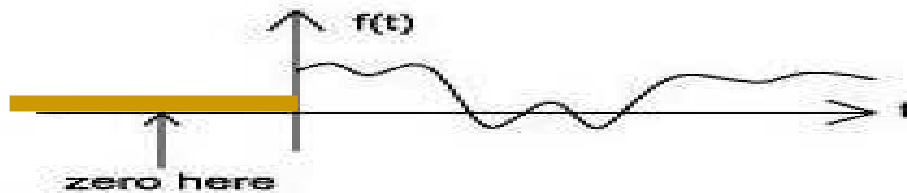
(**Fields medal** in 1950)

Jacques MEQUIN
NICE EWTBU / DTE
2012 j-mequin@ti.com

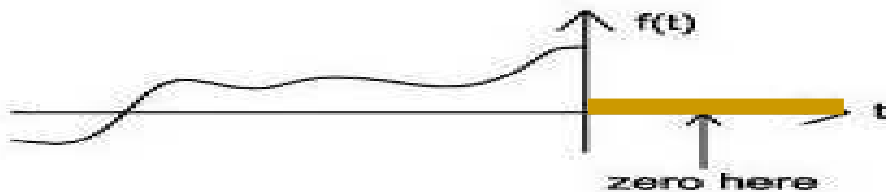
DEFINITIONS



noncausal signal



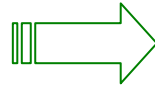
causal signal



anticausal signal

important relations

$$\begin{aligned} H'(t) &= \delta(t) \\ f(t) \delta(t) &= f(0) \delta(t) \end{aligned}$$



$$\begin{aligned} H'(t-a) &= \delta(t-a) \\ f(t) \delta(t-a) &= f(a) \delta(t-a) \end{aligned}$$

Let us consider a signal $y(t)$, its “CAUSAL representation” is $Y(t)$

$$Y(t) = y(t) H(t)$$

$$\text{Note : } (uv)' = u'v + uv'$$

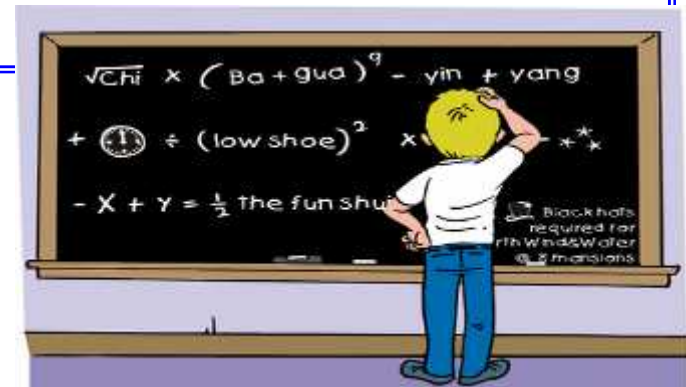
$$Y'(t) = y'(t) H(t) + y(t) H'(t) = y'(t) H(t) + y(0) \delta(t)$$

the expression of next derivative will simplify as

$$Y''(t) = y''(t) H(t) + y'(0) \delta(t) + y(0) \delta'(t)$$

etc . . .

The next thing to do would be to show what is happening when we inject those definitions into the differential equations of the circuit maybe another day . . .



STABILITY



★ Poles and stability of a system (of any order)

➤ *Look at the real part of each pole:*

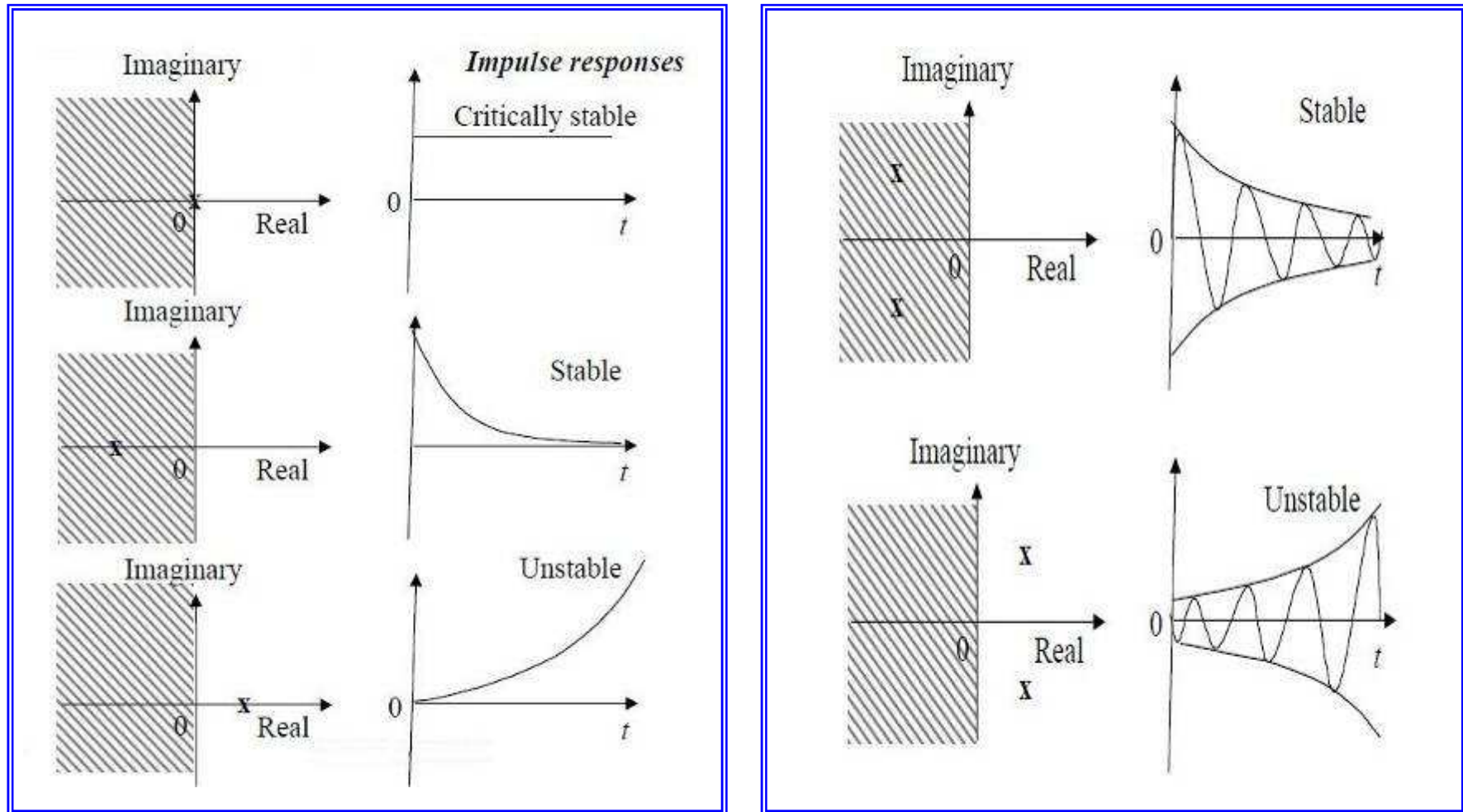
- If the **real part of ALL poles** is **negative** then all components of the system are **stable** and the **overall system is stable**
- If **ONE pole (or more)** has a **zero real part** then that component is critically/marginally stable and the **overall system is critically stable**
- If **ONE pole (or more)** has a **positive real part** then that component lead the **overall system to instability**

➤ *Look at the imaginary part for the presence of oscillations:*

- If the imaginary part of a pole is zero then that component does not have any oscillatory contribution
- If the imaginary part is not zero then its value is the frequency of oscillation of the corresponding component of the system

★ The zeros of a system do not affect the stability

- The zeros affect the transient response of the system



For **higher order** systems that have many poles,
the poles that are **further to the right** of the **s -plan** will **dominate** the response

Numerically, there is another issue

pole order



$$\begin{aligned}
 f = 100 \text{ Mhz} & \Rightarrow \omega = 2\pi f = 6.28 \cdot 10^8 \\
 & \omega^2 = 3.94 \cdot 10^{17} \\
 & \omega^3 = 2.48 \cdot 10^{26} \\
 & \dots
 \end{aligned}$$

Such very large $\pm$ *exponent* are also found in the $Y(t)$ *high order derivatives*

($f \uparrow \rightarrow \text{step } h \downarrow$ *very small* and *powered* as $h, h^2, h^3, \dots$)

*quantum slice of the period
must be SHANNON compliant to avoid aliasing*

accordingly to TAYLOR expansion

NOTE: To minimize this issue the *ILT package* is coded "*internally*" with floating
"*long double*" (*x87 standard* : 80 bits / 19 digits / $\approx \pm 4932$ *exponent*)

None of **3 methods** presented **structurally** handle **percussions** since, *by hypothesis*, the circuit is assumed **Linear Time Invariant**

Nevertheless,

simulating the original circuit “up to” t^-

then, “starting at” t^+ , simulating a *modified circuit* having as “**initial conditions**” the **final values** reached just **before** the **percussion**

does play the trick !!!

This is OK !!!, but the making of such **simulator** changing the **user defined circuit during the runtime** is likely to be **clumsy**

The aim of this presentation is precisely to claim that

an **equational solution exist** to stick, **all along the simulation**, with a more **seamless approach**

NAPA



15 years ago,

Yves Leduc who is the **owner**, **developer** and **enhancer** of an **"in-house"** **analog functional simulator** (named **NAPA**) has asked me to develop a **plug-in** performing the **Inverse Laplace Transform**

Along the years, collecting new algorithms in the **literature** and on the **web** **NAPA** has succeed to become a unique collection of **state of the art** **mathematical library** feature (sophisticated **FFT** windows, **noise** analysis, etc ...)

NAPA “ILT” plug-in hypothesis

For the *ILT package*, the output $Y[t]$ of a *transfer function* is the same

if $INPUT[t]$ is changing while $INPUT[t - h]$ did also change value
or if $INPUT[t - h]$ was maintaining a steady value

such *heuristic* (that neglects step “input variation”) enables the *ILT package* to behave independently of the *INPUT derivatives*
(in line with the *STATE MODEL* activated by a steady input during a step interval)

Therefore, such *ILT processing* can be viewed as “Markovian”★

This is particularly well adapted to *non-vectorial* simulator such as *NAPA* that can consequently consider any filter as an ordinary function to which it has trivially to provide just a simple “scalar” *INPUT* “sample” at each call

★ the state of such process at time t is independent of the history of the process before its previous state

$$\frac{Y(s)}{X(s)} = \frac{s^2 N2 + s N1 + N0}{s^2 D2 + s D1 + D0}$$

$N2$
 $N1$
 $N0$

$$\left(\frac{d^2}{dt^2} Y(t) \right) D2 + \left(\frac{d}{dt} Y(t) \right) D1 + Y(t) D0 = \left(\frac{d^2}{dt^2} X(t) \right) N2 + \left(\frac{d}{dt} X(t) \right) N1 + X(t) N0$$

$= 0$

$= 0$

$$X(t) \text{ causal} \Rightarrow X(t) = x(t) H(t)$$

developing this differential equation even with input $x''(t) = 0$ and $x'(t) = 0$

still leave the opportunity for the **numerator coefficients** $N2$ and $N1$

to contribute to the $Y(t)$ **output** response equation

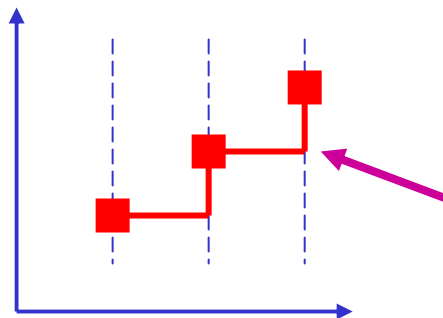
The “**Markovian**” approach of the **ILT package** is motivated by **another tactical choice**

➡ **To assume** that the **INPUT** potentially may contain some **sharp discontinuities** (**Initialization** being one of them **➡** **ILT curves** at “ **t_{0+}** ” **match the theory**)

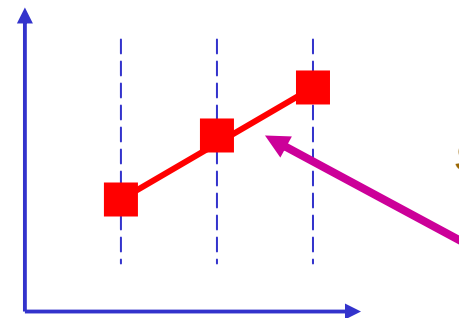
➡ and consequently, **not to assume**, estimate (nor “**interpolate**”), any **smoothness** of the **INPUT**

This would **cost** to store (to an *arbitrary chosen depth*) a vector of the previous **INPUTs** and to **interpolate** them to compute **associated derivatives**

This would also add some extra pain in case of **STEP** changes (**re-sampling**)



*belief of a
sharp INPUT
based on
discontinuity
assumptions*

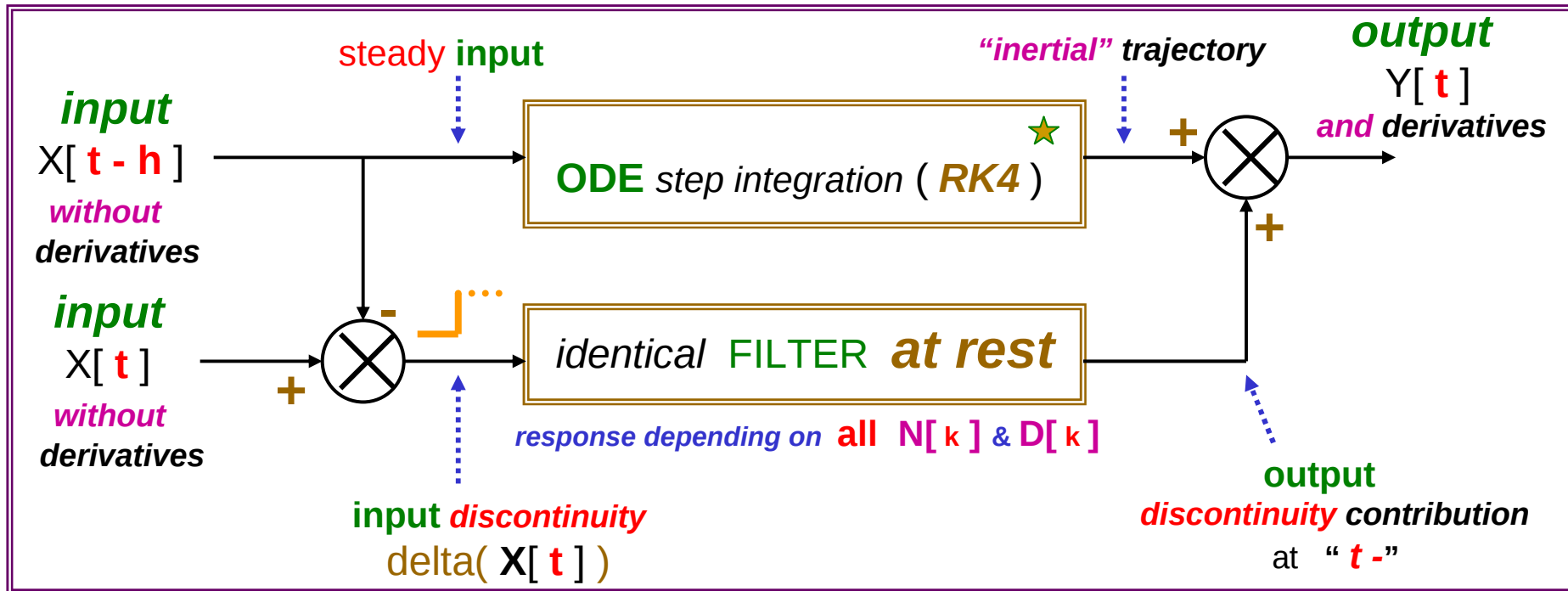


*belief of a
smooth INPUT
based on
derivatives*

In other words, the **expectation** to accurately calculate the response to some **sharp** signal is somewhat tight to the **necessity** of a **brut force** processing of the **INPUT**

Quality of the **LINEAR** response ???

So, the **output** of the **ILT package** is calculated as



In summary, the hypothesis of **sharp INPUT** based on **discontinuity** assumptions enables accurate responses of “**switched**” type of circuits (in particular, the proper calculation of **filter “impulse response”**)

However, for “**smooth**” **input**, the **error** is closely tight to the **user STEP** granularity

★ the **ILT package** uses **RK4** for historical reasons, a simpler integration should be OK

ILT plug-in package functions :

ILT_init_LAPLACE() to define the *transfer function* **NUMERATOR**
DENOMINATOR
and the “*time step*”

ILT_change_h() enable ‘*on the fly*’ change of the “*time step*”

ILT_step_LAPLACE() **input(k + 1) == >> output(k + 1)**

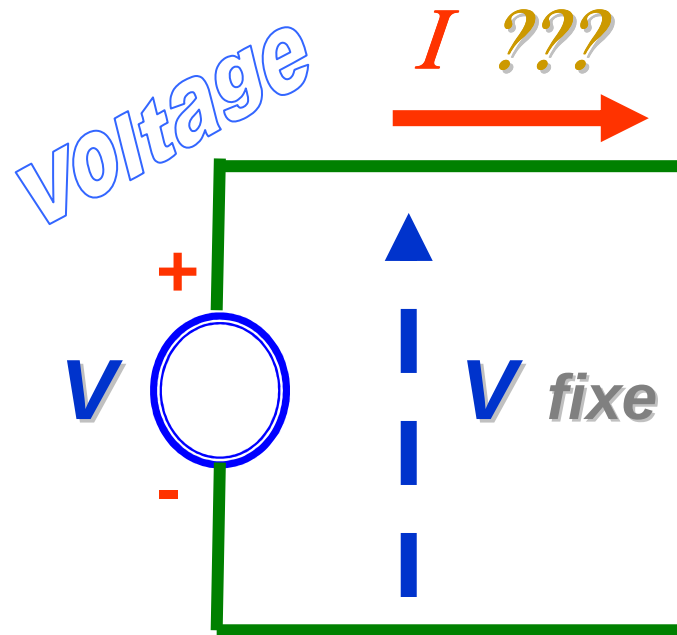
After all those fascinating (?) slides of introduction,

Now, we can go back to our frightening question :

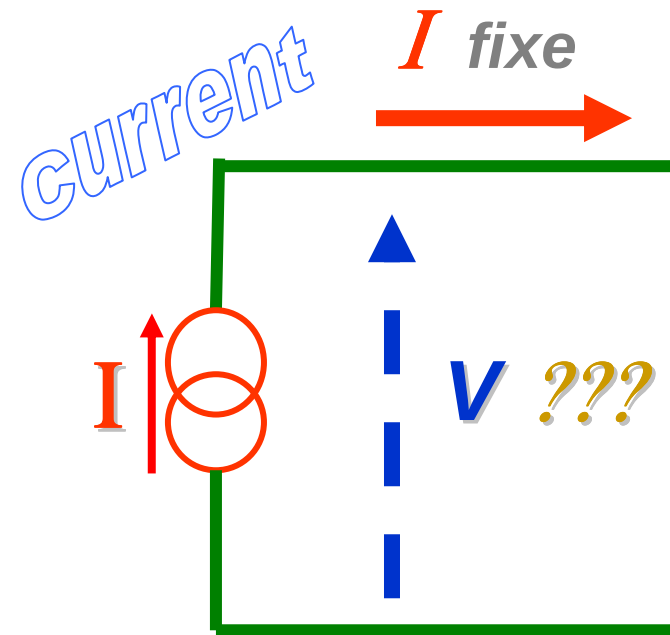
***Component PERCUSSION
of
Linear Time Invariant
analog circuit***

What is happening ???

“IDEAL” SOURCES

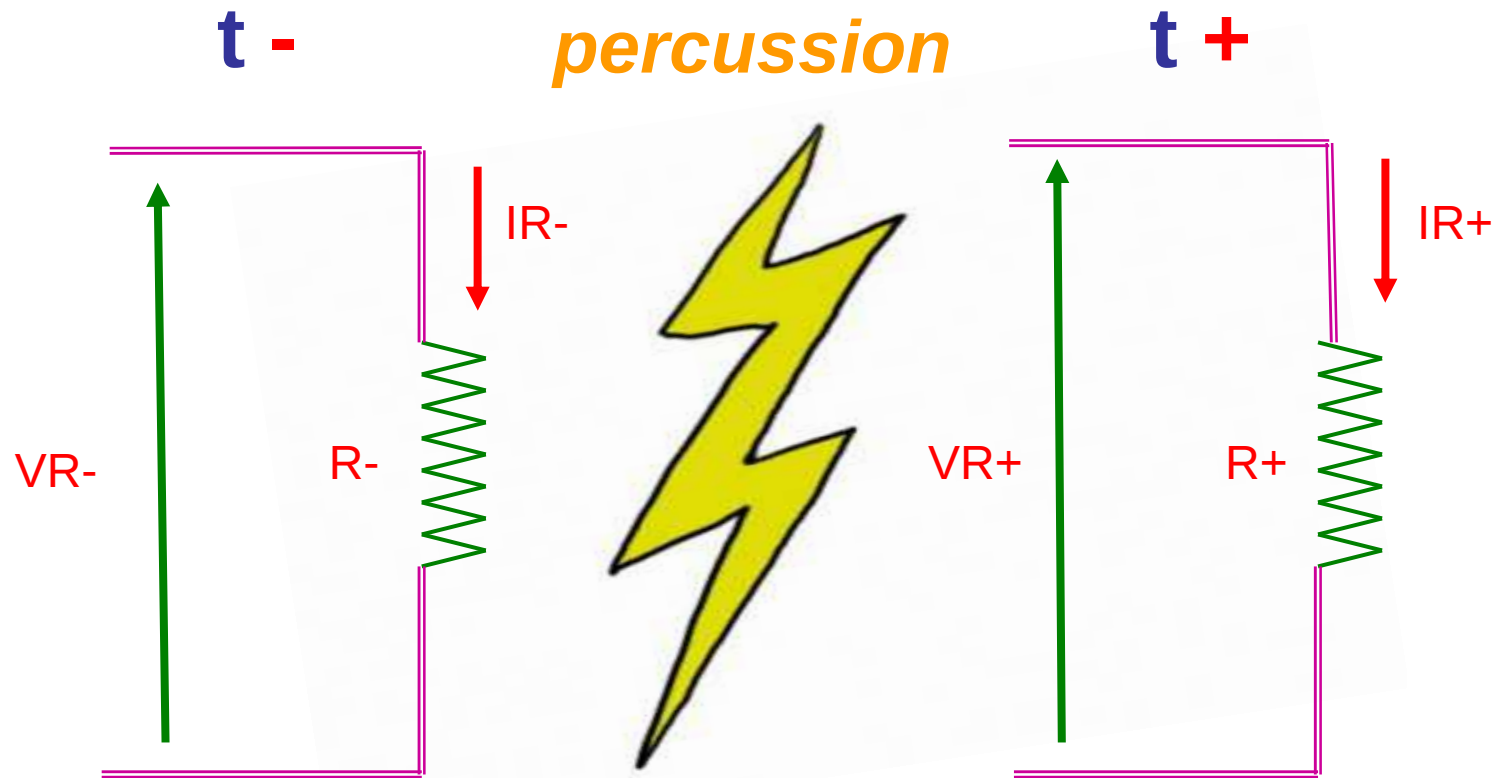


the **current**
will take whatever value
as a consequence of the
connected circuit



the **voltage**
will take whatever value
as a consequence of the
connected circuit

In both cases, the **VOLTAGE** and **CURRENT** are not “interdependent”



At the precise *instant* of the *percussion*,

$VR\delta$ and $IR\delta$ are *paradoxically* no longer “*interdependent*”

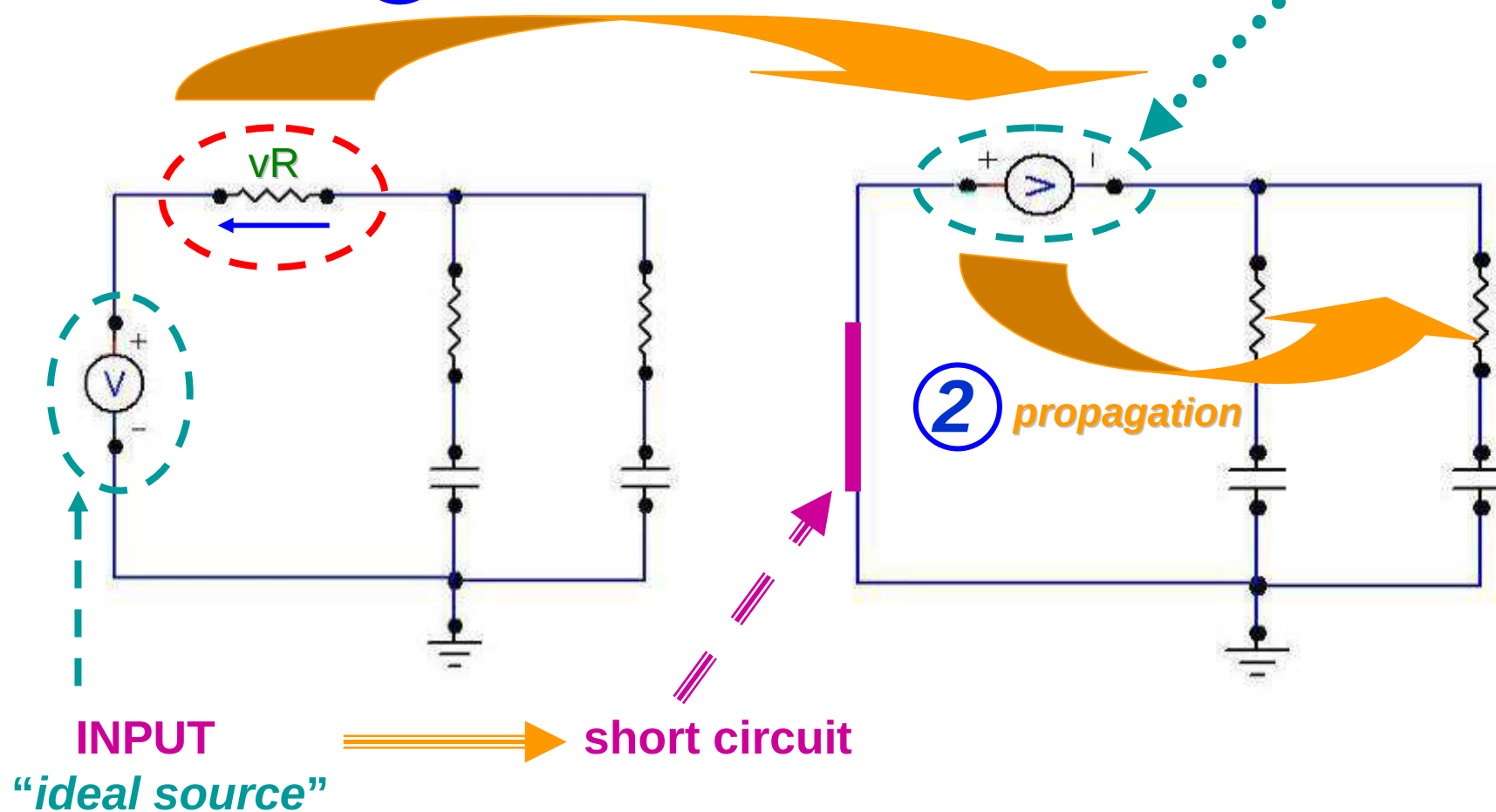
since the “*percuted*” component can be viewed (during such *null time*) as an “*ideal source*”

2 stages mystery . . .

① PERCUSSION

"ideal source"
but for a null time !!!

$f(vR, \text{filter coeffs})$



ILT PERCUSSION processing

FIRST stage : "COMPONENT" PERCUSSION

ILT plug-in package function: `ILT_percussion( )`

Please remember that the **ILT** package **initialization** has only defined the **coefficients Numerator / Denominator** of some **transfer functions**

for example, **the numerical values**
of the polynomial coefficients

$N = [n1, 0]$ and $D = [d2, d1, 1]$

$$\frac{vr1}{V} = \frac{C1 R1 s}{C1 L1 s^2 + C1 R1 s + 1}$$

Can we just make a call to a function with an updated set of polynomial coefficients (computed, for example, using a new value of "R1") **without passing**★ the information that a **R**, a **L**, or a **C** has changed ???

★ This is a very ambitious **assumption** regarding the assumed impact of the coefficients on the behavior of the **targeted differential equation variables**, demonstration is beyond this presentation

Yes !!!, equations can be found using “**DISTRIBUTION**” formalism
(as expected, they predominantly depend on the **old** & **new denominator coefficients**)
but . . .

it is OK to “**percute the Transfer Function**” representative of the **voltage**,
or the **current**, across a **RESISTOR**



for a **CAPACITANCE** value change, it is mandatory to “**percute**” the **voltage**
since the **current** and the **charge Q** will remain unchanged by the **percussion**
It is the **voltage** that will **evolve**, according to the **new C**, to maintain **$Q = CV$**

for an **INDUCTANCE** value change, it is mandatory to “**percute**” the **current**
since the **voltage** and the **flux ϕ** will remain unchanged by the **percussion**
It is the **current** that will **evolve**, according to the **new L**, to maintain **$\phi = LI$**



If several components have **simultaneously** to change their values
they must do it through **several consecutive (2 stages) percussions**

VALID SIMULATIONS



never percuted

sometimes percuted

R	$V_R(s) = H(s) \text{ input}(s)$	$I_R(s) = H(s) \text{ input}(s)$	$V_R(s) = H(s) \text{ input}(s)$	$I_R(s) = H(s) \text{ input}(s)$
	$V_L(s) = H(s) \text{ input}(s)$	$I_L(s) = H(s) \text{ input}(s)$	$V_L(s) = H(s) \text{ input}(s)$	$I_L(s) = H(s) \text{ input}(s)$
	$V_C(s) = H(s) \text{ input}(s)$	$I_C(s) = H(s) \text{ input}(s)$	$V_C(s) = H(s) \text{ input}(s)$	$I_C(s) = H(s) \text{ input}(s)$

ILT PERCUSSION processing

SECOND stage : **PERCUSSION “PROPAGATION”**

ILT plug-in package function: `ILT_export_percussion( )`

Now that an **updated value** of the **voltage** (or **current**) across the “**percuted**” **component** has been computed, **we must update** the **voltage** (or **current**) across **any other output** for which the user has defined, **at initialization**, an **observable transfer function**

That is, for our **RLC** example, **iL1** & **vC1**

$$\frac{i_{L1}}{V} = \frac{C1 s}{C1 L1 s^2 + C1 R1 s + 1}$$
$$\frac{v_{C1}}{V} = \frac{1}{C1 L1 s^2 + C1 R1 s + 1}$$

HOW TO DO IT ???



The “**percuted**” component **will act** (but for a **null time**)
as an **ideal INPUT source** within a **virtual circuit** having
a **short circuit** in place of the **regular INPUT !!!**



So, we have to predefine, for those **OUTPUTs** requiring a **side effect**
of the **main percussion**, a set of **null execution time (!!!)**
transfer functions modeling those **OUTPUTs** in the **virtual circuit**



equations can be found using “**DISTRIBUTION**” formalism
to **superpose** to their current **OUTPUT** value at “**t**” **the effect**
of the **main discontinuity** propagated through the **virtual circuit**

simulation model:

ALL percuted component ILTs

+

user observed signal ILTs

INIT {

```
ILT_init_LAPLACE( pILT_vr0_x , h , N_vr0_x , D_vr0_x , 2 , 2 , 0 ) ;
ILT_init_LAPLACE( pILT_vc1_x , h , N_vc1_x , D_vc1_x , 1 , 2 , 0 ) ;
ILT_init_LAPLACE( pILT_vc2_x , h , N_vc2_x , D_vc2_x , 1 , 2 , 0 ) ;
```

```
vr0 = ILT_step_LAPLACE( pILT_vr0_x , x ) ;
vc1 = ILT_step_LAPLACE( pILT_vc1_x , x ) ;
vc2 = ILT_step_LAPLACE( pILT_vc2_x , x ) ;
```

step from
previous t to $t-$

```
if ( percussion_R0 ) {
    UPDATE_ALL_COEFFS( new_R0 ) ;
```

virtual circuit
 $vc1 = H(vr0)$

stage #1 R0 percussion
stage #2 propagation

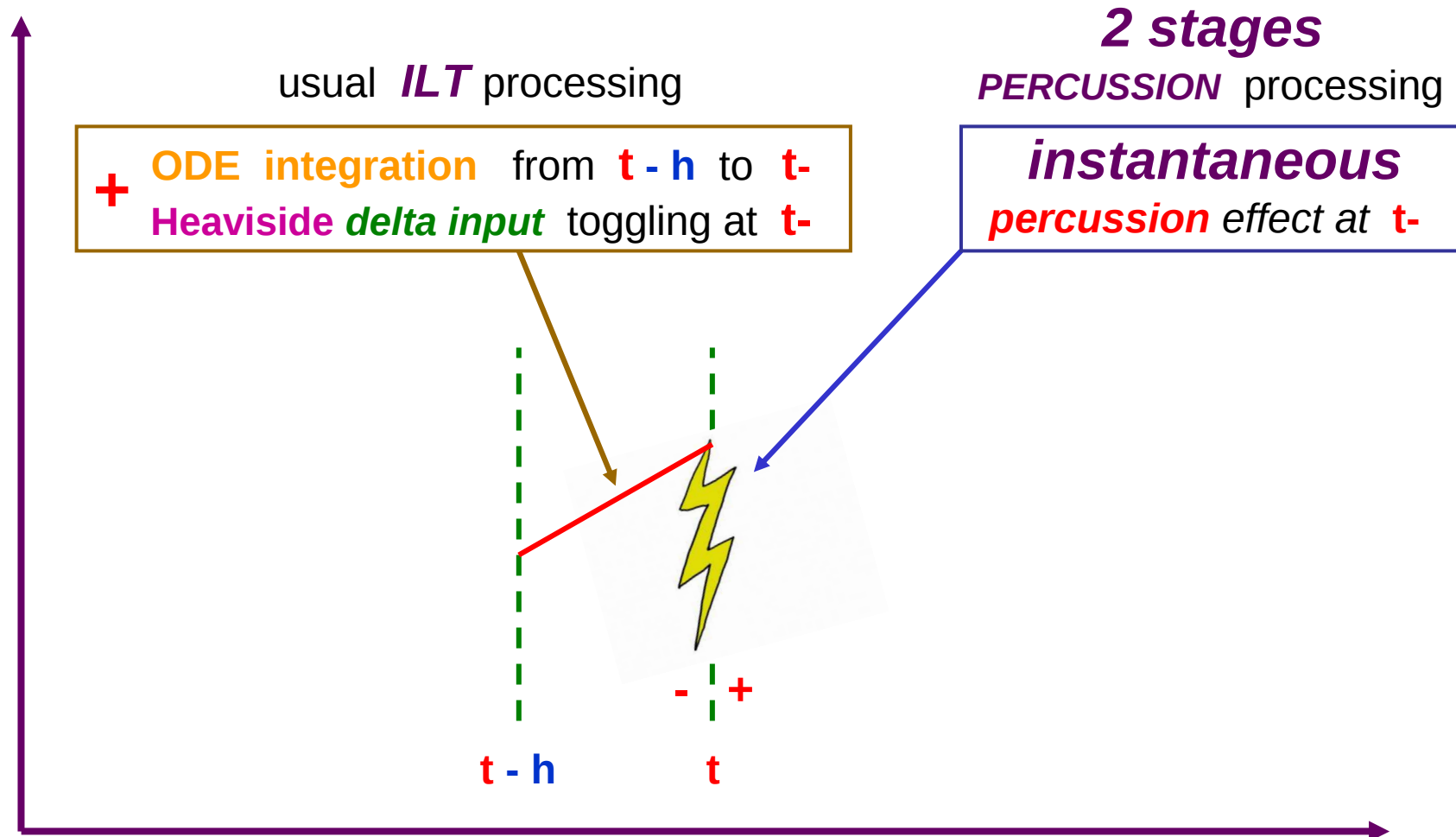
```
vr0 = ILT_percussion( pILT_vr0_x ) ;
vc1 = ILT_export_percussion( pILT_vr0_x , pILT_vc1_x , pILT_vc1_$vr0 ) ;
vc2 = ILT_export_percussion( pILT_vr0_x , pILT_vc2_x , pILT_vc2_$vr0 ) ;
}
```

```
if ( percussion_C1 ) {
    UPDATE_ALL_COEFFS( new_C1 ) ;
```

stage #1 C1 percussion
stage #2 propagation

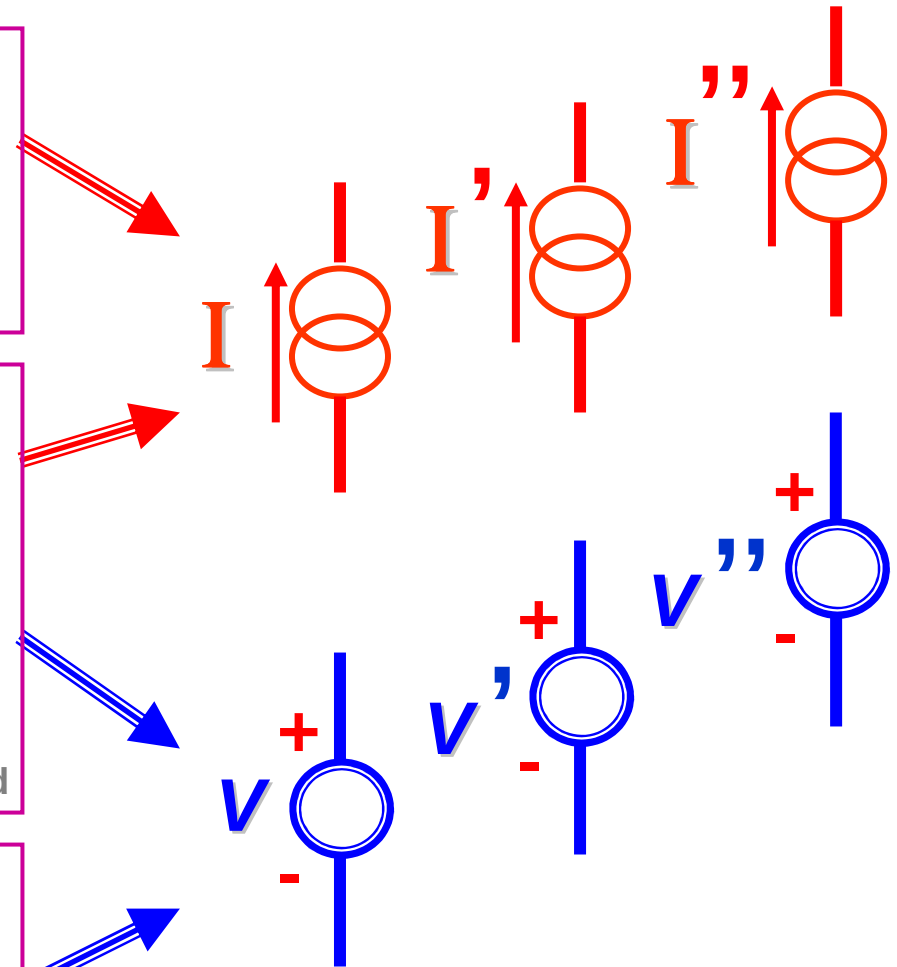
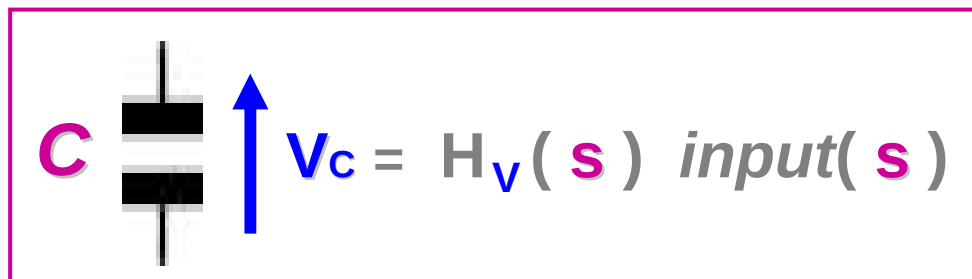
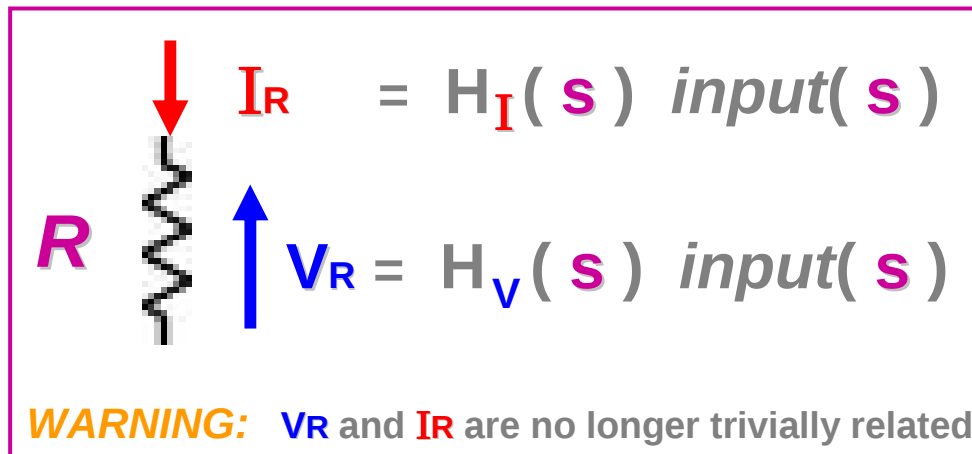
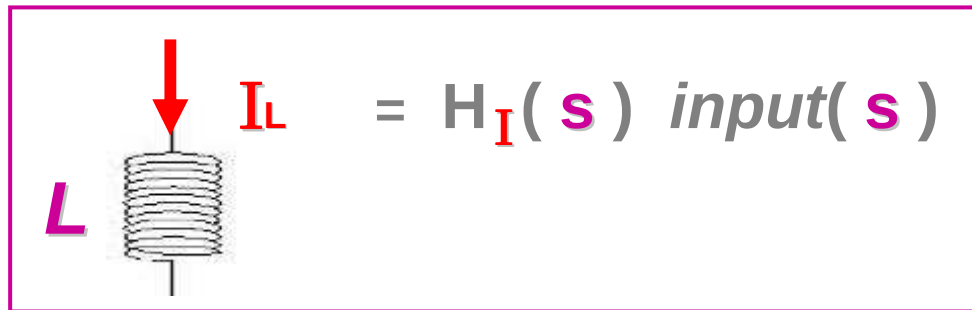
```
vc1 = ILT_percussion( pILT_vc1_x ) ;
vr0 = ILT_export_percussion( pILT_vc1_x , pILT_vr0_x , pILT_vr0_$vc1 ) ;
vc2 = ILT_export_percussion( pILT_vc1_x , pILT_vc2_x , pILT_vc2_$vc1 ) ;
}
```

ILT PERCUSSION processing summary



COMPONENT

PERCUSSION

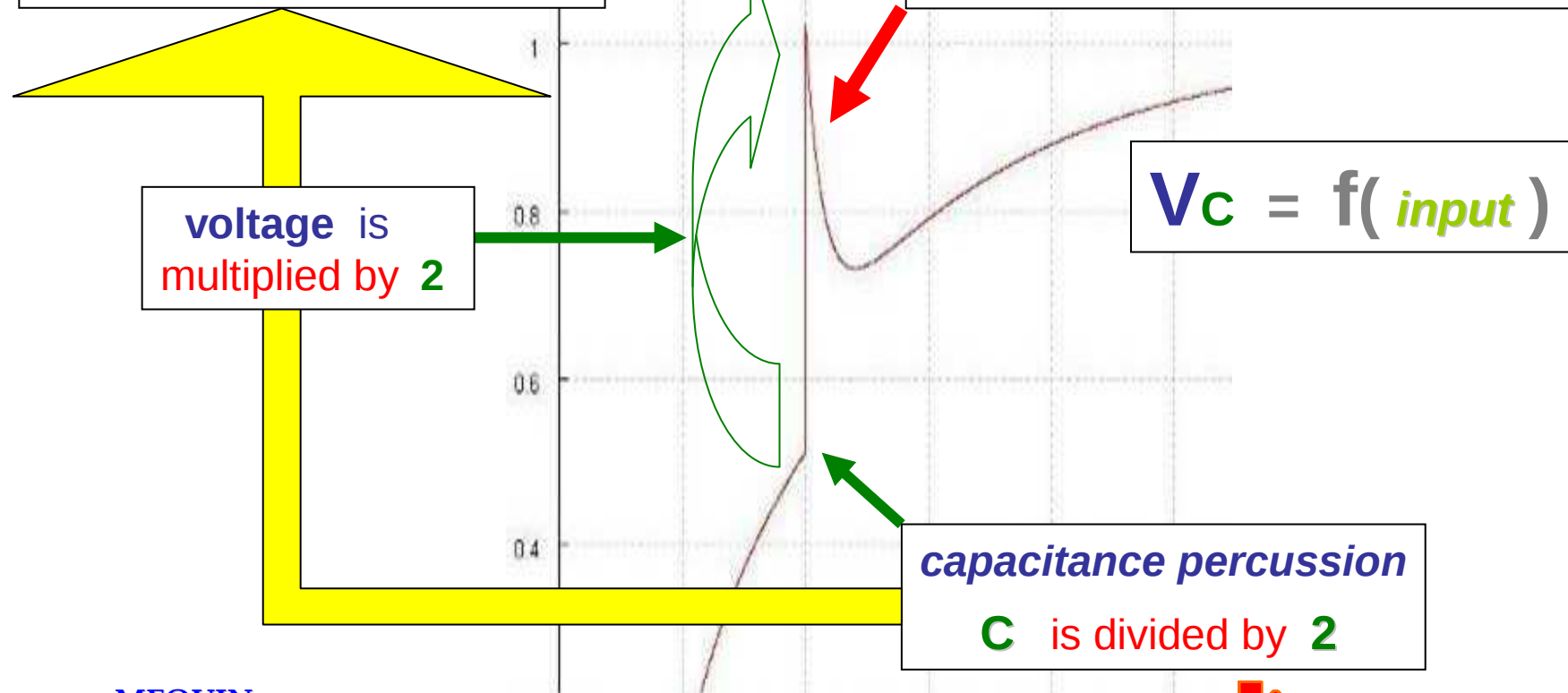


*derivatives of those IDEAL SOURCES
do also have their own equations*

from now on the curve becomes **the resultant** of **a new free-mode + forced-mode**

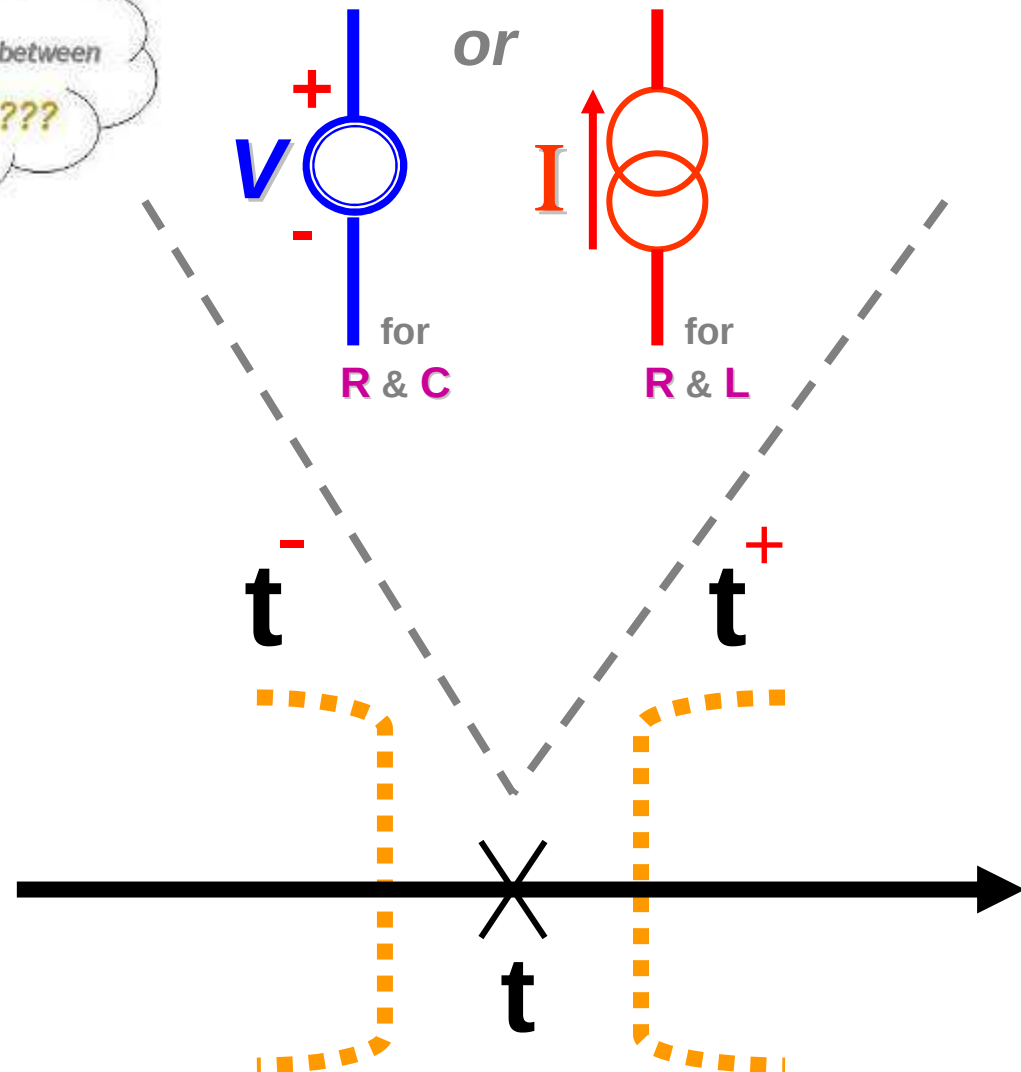
*at the precise instant
of the percussion
there is nothing very
impressive in the fact that
the equations predict
the doubling of the voltage*

but the equations are priceless
in predicting how the circuit
will behave immediately after
the percussion due to
the symbolic expressions
of the **derivatives**

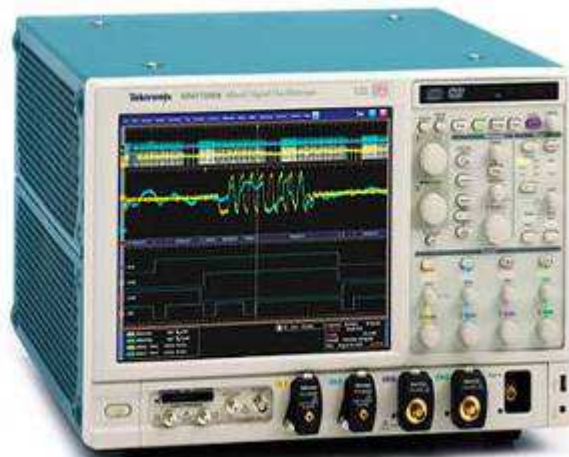




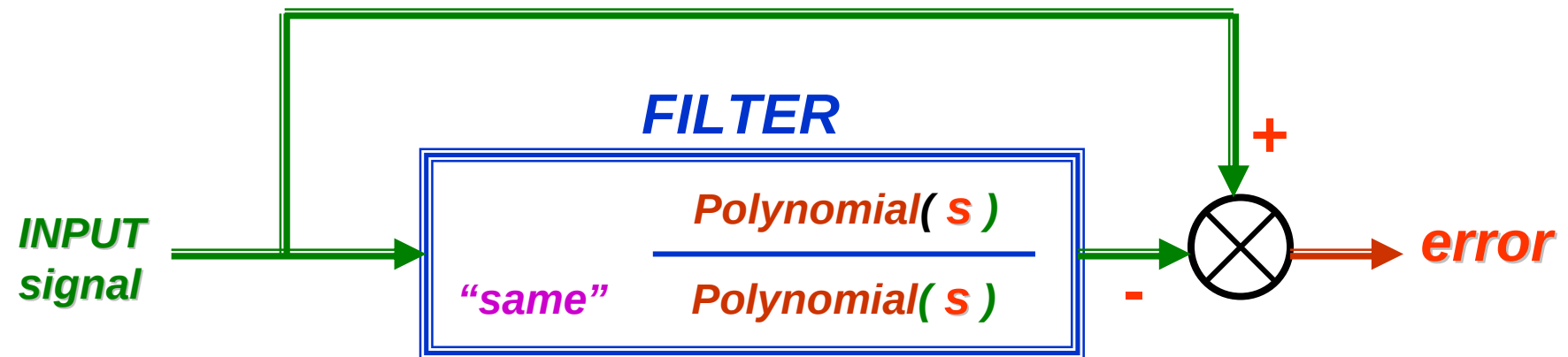
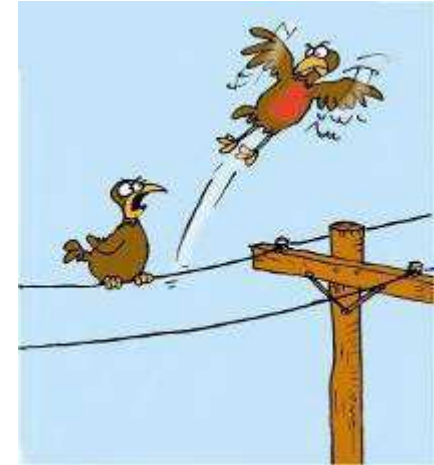
percuted component



TEST



“ **WIRE** ” test



Assuming that the **FILTER** is **stable**, this is expected to behave as an **ideal “wire”**

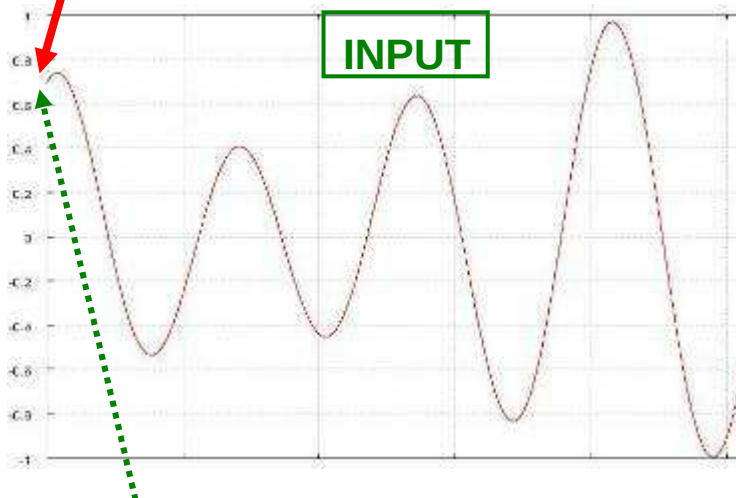
“ WIRE ” test

this filter is “**stable**” by construction

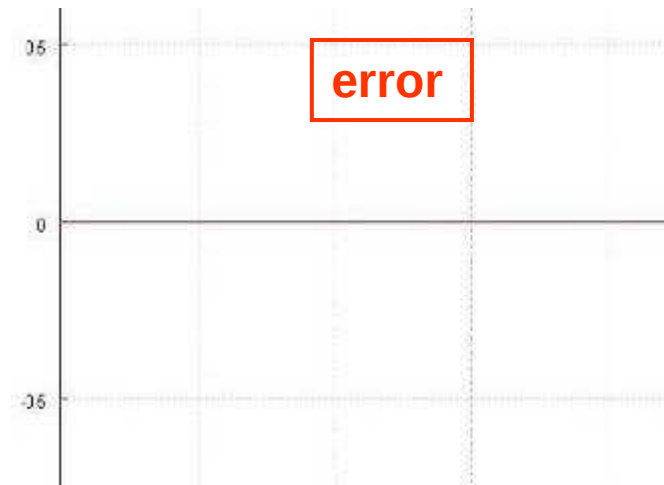
$$ka * \cos(wa * t) + kb * \sin(wb * t)$$

$$\frac{(s + 1)(s + 2)(s + 3)(s + 4)(s + 5)}{(s + 1)(s + 2)(s + 3)(s + 4)(s + 5)}$$

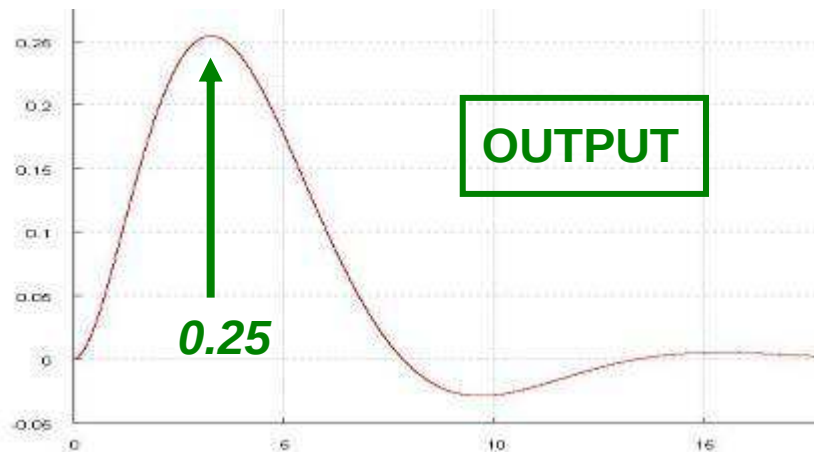
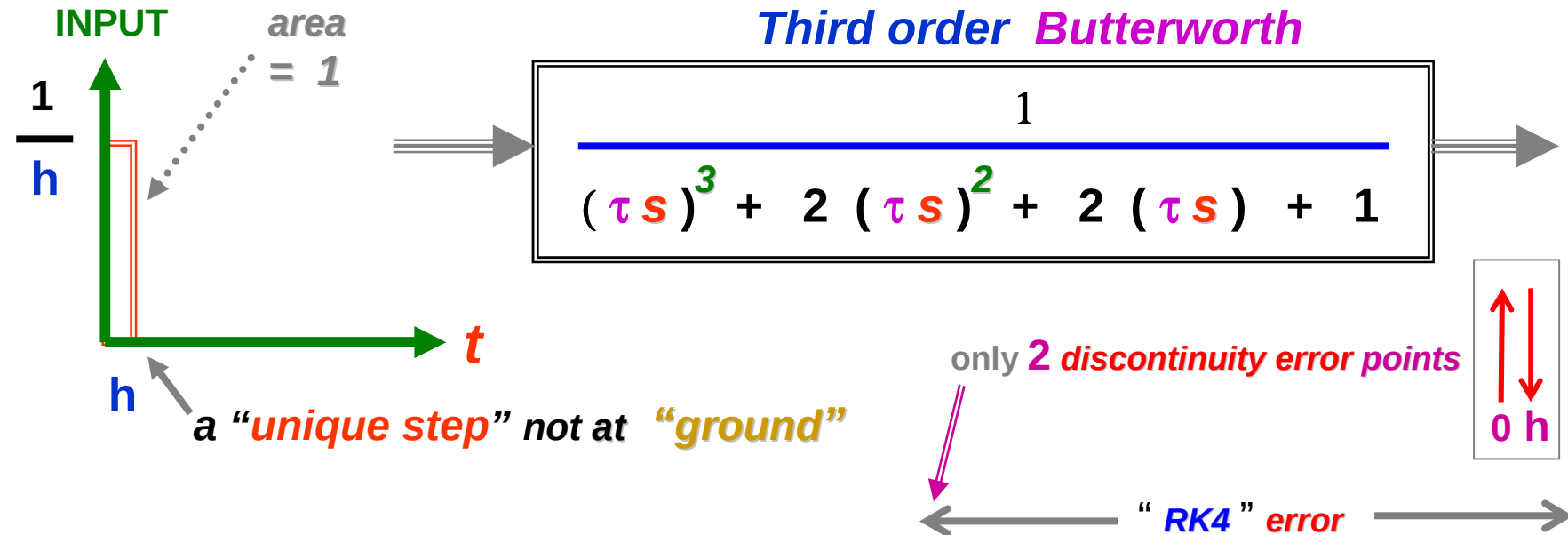
output(0) will only depends on
this first input sample
and the **filter coefficients**



to increase **stress** this **activation** does not start at 0

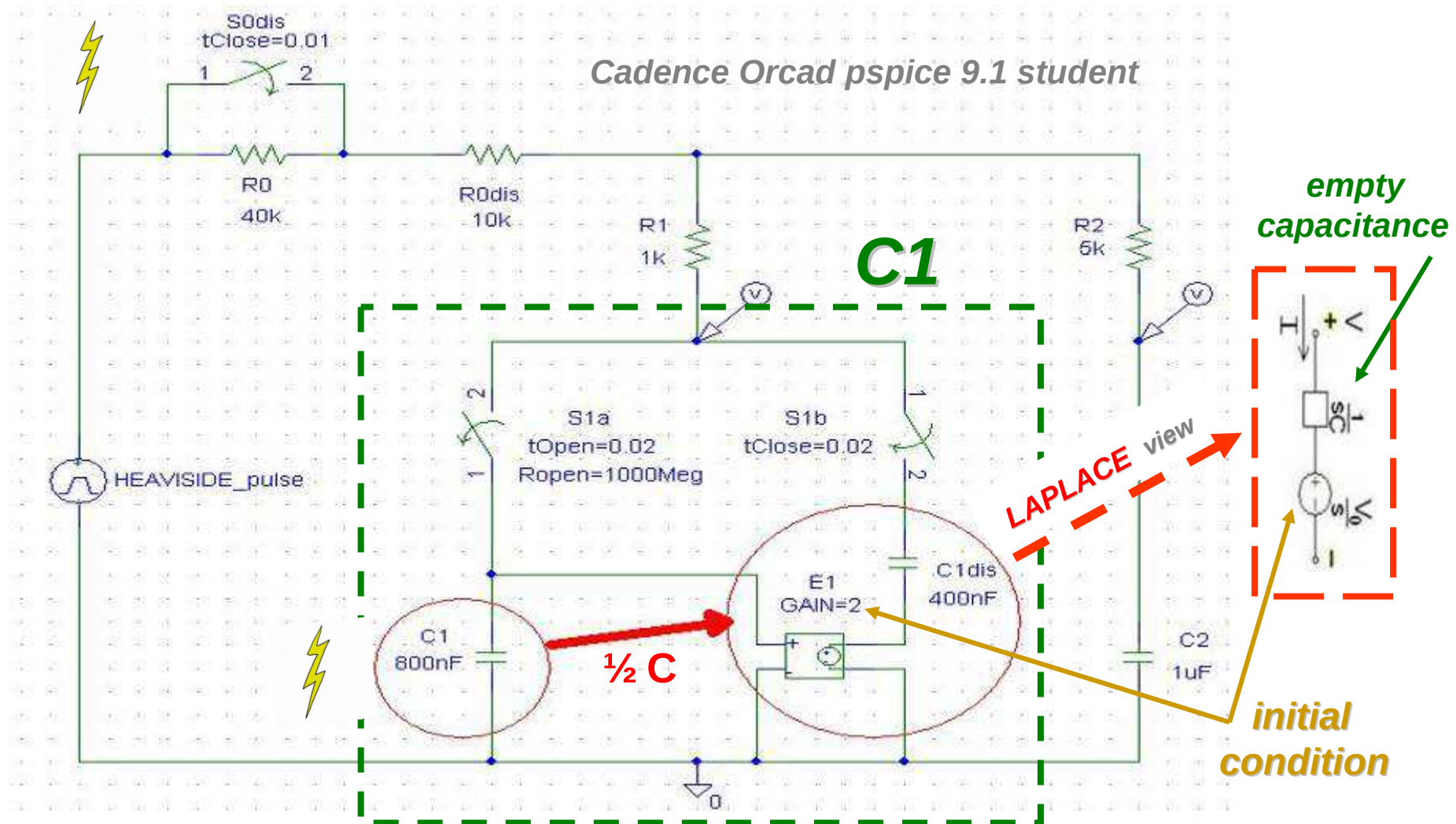


“IMPULSIONAL” response

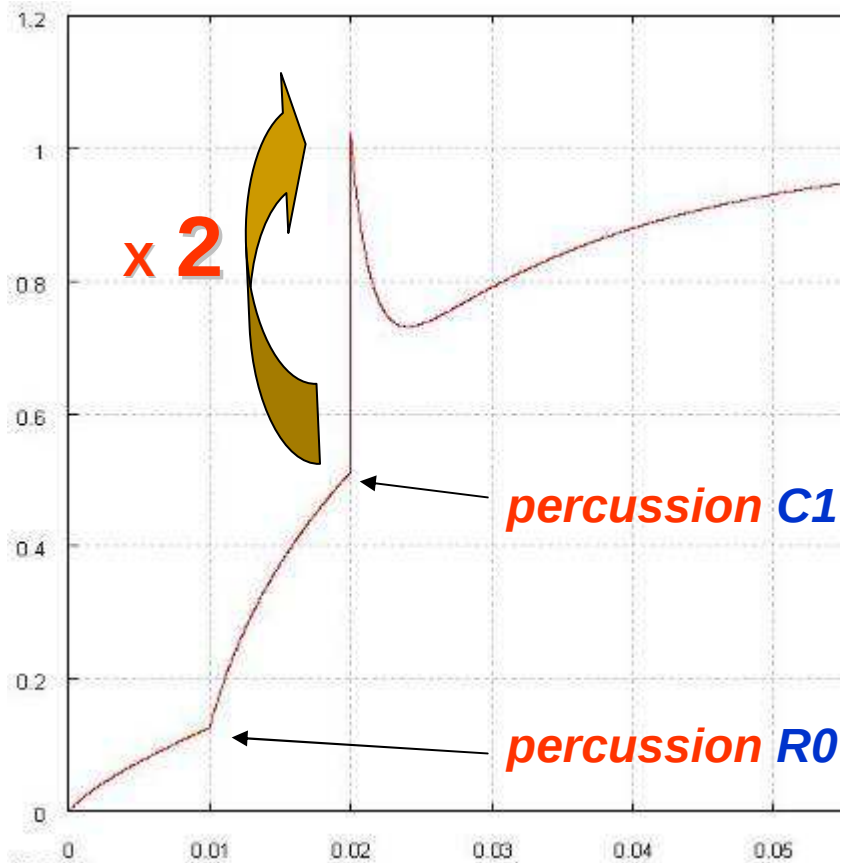


PERCUSSION

at $t = 0.01$ $R0$ 50 *Kohm* ==>> 10 *Kohm*
 at $t = 0.02$ $C1$ 800 *nF* ==>> 400 *nF*



VC1 output



error on VC1 output



The **INPUT** of this test case is a trivial **HEAVISIDE**

Therefore, there is **no contribution** of the "ILT package" **INPUT** **intrinsic sampling error method** to this filter **output response**

This is a favorable situation to measure on its own the percussion processing quality



SPICE is a **matrix based solver**

computing the behavior of some (**linear** and/or **non-linear**) **global circuit** by the **full evaluation** of its **possibly very (very) large** number of individual **NODES**

INTEGRATION (and **NON-LINEARITY**) are handled through “**iterative**” **heuristics**

for **Linear Time Invariant (LTI)** circuit,

TRANSFER FUNCTION models the behavior of a **single observed signal**

The **DENOMINATOR** reflects the complexity of the **circuit STATE** “**as a whole**”
Therefore, to limit the **polynomial order** this method is reserved to **small circuit**

Nevertheless, the advantage of a **TRANSFER FUNCTION** is that the associated **DIFFERENTIAL EQUATION** allows the manipulation of the overall **circuit STATE** through “**algorithmic equations**”

This is why this facilitates the modeling of **PERCUSSIONS**

PERCUSSIONS also occur in a “pinball”

By analogy with a **Differential Equation**, the metal ball has a **trajectory**

An **instantaneous shock** on an **obstacle** is enough to change it **forever** . . .



HALL of Fame



Brook **Taylor**

1685 - 1731



Marquis Pierre-Simon **Laplace**

1749 - 1827



Joseph **Fourier**

1768 - 1830



Gustav Robert **Kirchhoff**

1824 - 1887



Oliver **Heaviside**

1850 - 1925



Carl **Runge** & Martin Wilhelm **Kutta** method

1901



Paul Adrien Maurice **Dirac**

1902 - 1984



Laurent **Schwartz**

1915 - 2002

MIT **Macsyma** (now **Maxima**)

late 1960s

SPICE 1 public domain

1972

QUESTIONS ?

