

# Strong Turing Completeness of Continuous Chemical Reaction Networks

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Joint work with Olivier Bournez, François Fages, Guillaume Le Guludec and Daniel Graça

10 october 2018

# Chemical Reaction Networks

A **reaction system** is a finite set of

- ▶ molecular species  $y_1, \dots, y_n$
- ▶ reactions of the form  $\sum_i a_i y_i \xrightarrow{f} \sum_i b_i y_i$  ( $a_i, b_i \in \mathbb{N}$ ,  $f = \text{rate}$ )

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$$\sum_i a_i y_i \xrightarrow{k} \sum_i b_i y_i \rightsquigarrow f(y) = k \prod_i y_i^{a_i}$$

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Example :

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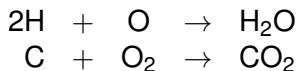
$\rightsquigarrow$  **Polynomial ODE!**

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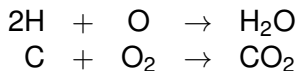


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Not limited to simple chemical reactions :

- ▶ DNA strand displacement
- ▶ RNA
- ▶ protein reactions

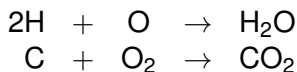


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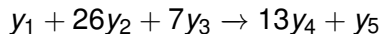
Implementing CRNs is a **recent and active** research field.

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Some reactions are **unrealistic** :

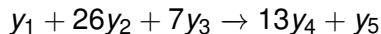


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Only consider **elementary** reactions : at most two reactants

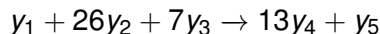
- ▶  $A + B \xrightarrow{k} C$
- ▶  $A \xrightarrow{k} B + C$
- ▶  $A \xrightarrow{k} B$
- ▶  $A \xrightarrow{k} \emptyset$
- ▶  $\emptyset \xrightarrow{k} A$

# Chemical Reaction Networks

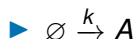
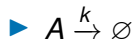
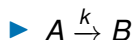
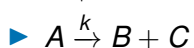
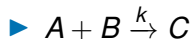
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$$A' = -kAB \quad B' = -kAB \quad C' = kAB$$

~ Quadratic ODE !

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- ▶ **rate** : dependent/independent
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- ▶ **kinetics** : mass action/Michaelis/...
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**Extreme examples :**

rate-independent, differential, any  
kinetics, finite species, value is  
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## Extreme examples :

rate-independent, differential, any kinetics, finite species, value is concentration, stable

↪ **piecewise linear functions**

rate-dependent, stochastic, Markov, finite species, value is molecule count (must be small)

↪ **probabilistic Turing machine**

# Chemical Reaction Networks : main result

A reaction is **elementary** if it has at most two reactants

⇒ can, in principle, be implemented with DNA, RNA or proteins

## Theorem (CMSB 2017)

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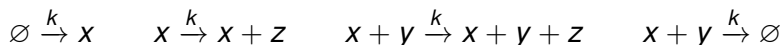
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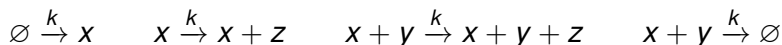
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We can even say something about the **complexity** :

$f \in \text{FPTIME} \Rightarrow \text{CRN computes } f \text{ in } \left\{ \begin{array}{l} \blacktriangleright \text{ polynomial time\&space} \\ \text{or equivalently} \\ \blacktriangleright \text{ polynomial length} \end{array} \right.$

# Chemical Reaction Networks : mathematics

mass-action-law reaction system on finite universes of molecules under the differential semantics



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Polynomial ODE :

$$\begin{cases} y_1' = p_1(y_1, \dots, y_n) \\ \vdots \\ y_n' = p_n(y_1, \dots, y_n) \end{cases}$$

with constraints :

- ▶ nonnegative values (concentration)
- ▶ restricted negative feedback :  $x' = -xyz$

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 clever rewriting  
value encoding :  $y = y^+ - y^-$

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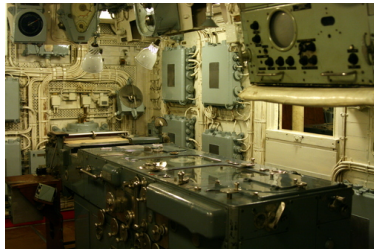
Polynomial ODE :  $y' = p(y)$

What can we compute with polynomial ODEs ?

# Analog Computers

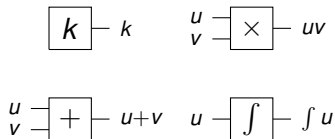


Differential Analyser  
“Mathematica of the 1920s”



Admiralty Fire Control Table  
British Navy ships (WW2)

# Polynomial Differential Equations



General Purpose  
Analog Computer



Differential Analyzer

Newton mechanics

Reaction networks :

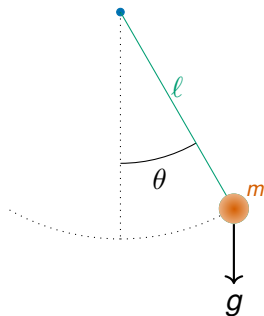
- ▶ chemical
- ▶ enzymatic

polynomial differential  
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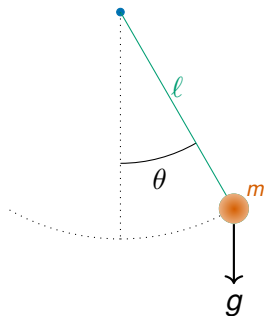
- ▶ Rich class
- ▶ Stable (+,  $\times$ ,  $\circ$ ,  $/$ , ED)
- ▶ No closed-form solution

# Example of dynamical system



$$\ddot{\theta} + \frac{g}{\ell} \sin(\theta) = 0$$

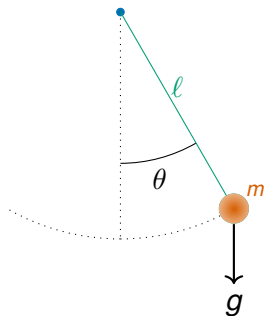
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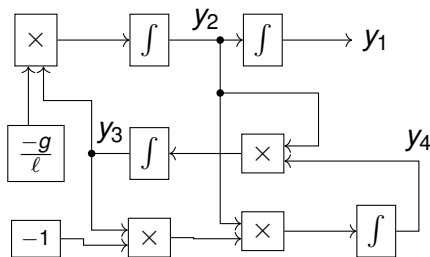
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$$\begin{cases} y_1' = y_2 \\ y_2' = -\frac{g}{\ell} y_3 \\ y_3' = y_2 y_4 \\ y_4' = -y_2 y_3 \end{cases} \Leftrightarrow \begin{cases} y_1 = \theta \\ y_2 = \dot{\theta} \\ y_3 = \sin(\theta) \\ y_4 = \cos(\theta) \end{cases}$$

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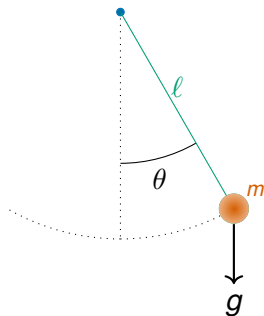


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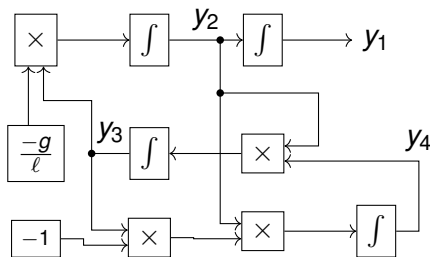


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## Historical remark : the word “analog”

The pendulum and the circuit have the same equation. One can study one using the other by **analogy**.

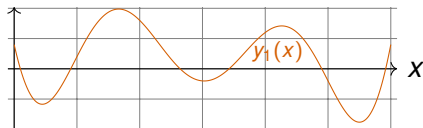


# Computing with differential equations

## Generable functions

$$\begin{cases} y(0) = y_0 \\ y'(x) = p(y(x)) \end{cases} \quad x \in \mathbb{R}$$

$$f(x) = y_1(x)$$



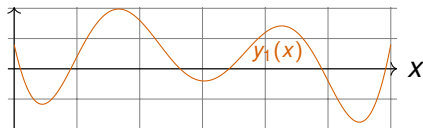
Shannon's notion

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sin, cos, exp, log, ...

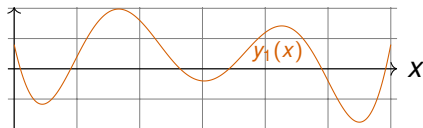
Strictly weaker than Turing  
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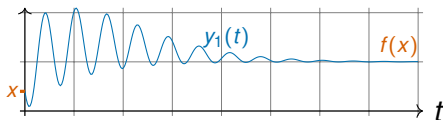
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## Computable

$$\begin{cases} y(0) = q(x) \\ y'(t) = p(y(t)) \end{cases} \quad \begin{matrix} x \in \mathbb{R} \\ t \in \mathbb{R}_+ \end{matrix}$$

$$f(x) = \lim_{t \rightarrow \infty} y_1(t)$$



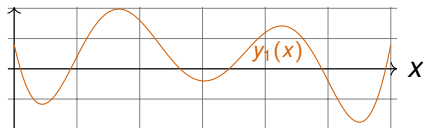
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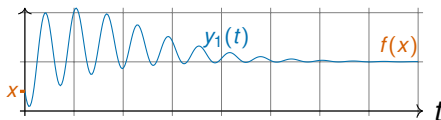
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Modern notion

sin, cos, exp, log,  $\Gamma$ ,  $\zeta$ , ...

Turing powerful  
[Bournez et al., 2007]

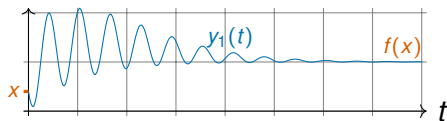
# Equivalence with computable analysis

Definition (Bournez et al, 2007)

$f$  **computable by GPAC** if  $\exists p$  polynomial such that  $\forall x \in [a, b]$

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satisfies  $|f(x) - y_1(t)| \leq y_2(t)$  et  $y_2(t) \xrightarrow[t \rightarrow \infty]{} 0$ .



$$y_1(t) \xrightarrow[t \rightarrow \infty]{} f(x)$$

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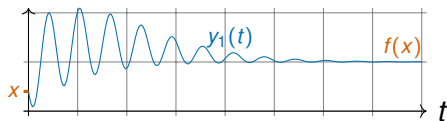
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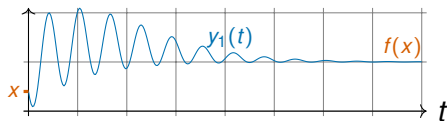
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1. In Computable Analysis, a standard model over reals built from Turing machines.

# Complexity of analog systems

- ▶ Turing machines :  $T(x)$  = number of steps to compute on  $x$

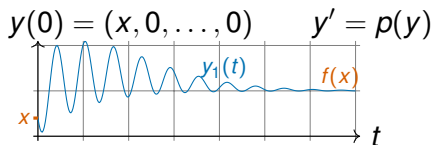


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- ▶ GPAC :

## Tentative definition

$$T(x) = ??$$

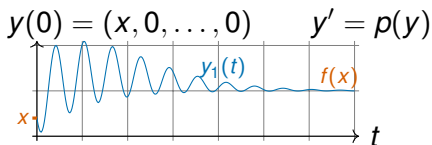


# Complexity of analog systems

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$$T(x, \mu) =$$

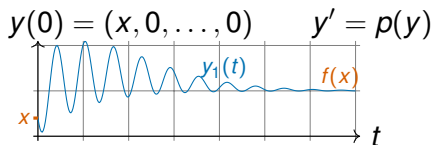


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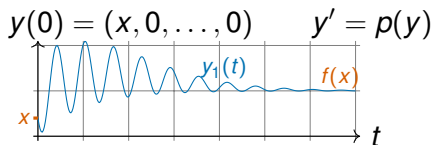


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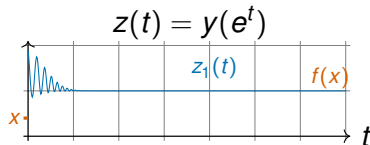
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$\leadsto$



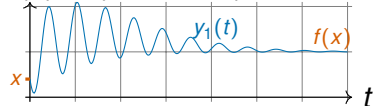
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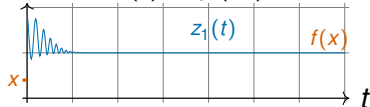
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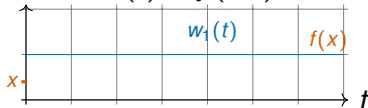


$\leadsto$

$$z(t) = y(e^t)$$



$$w(t) = y(e^{e^t})$$



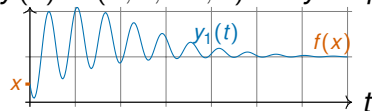
# Complexity of analog systems

- ▶ Turing machines :  $T(x)$  = number of steps to compute on  $x$
- ▶ GPAC : time contraction problem → **open problem**

## Tentative definition

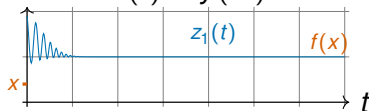
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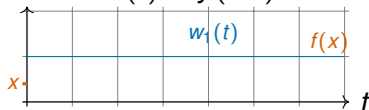
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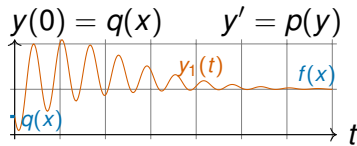
Something is wrong...

All functions have constant time complexity.

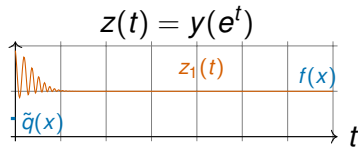
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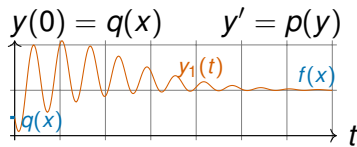
# Time-space correlation of the GPAC



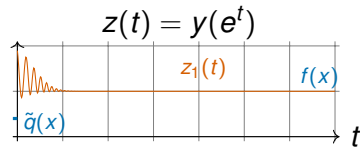
$\leadsto$



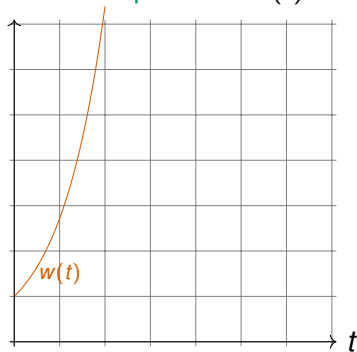
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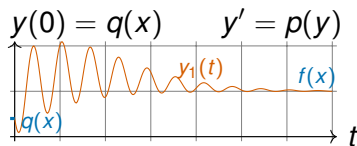


extra component :  $w(t) = e^t$

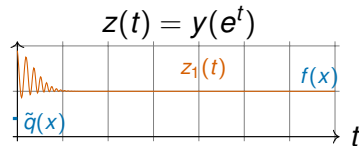




# Time-space correlation of the GPAC



$\rightsquigarrow$



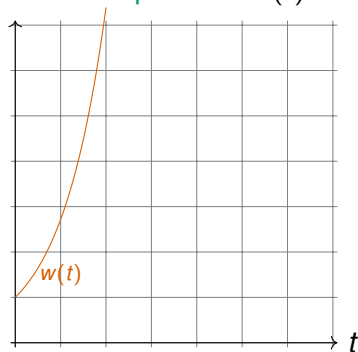
## Observation

Time scaling costs “space”.

$\rightsquigarrow$

Time complexity for the GPAC must involve time and **space**!

extra component :  $w(t) = e^t$



# Complexity of solving polynomial ODEs

$$y(0) = x \quad y'(t) = p(y(t))$$



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## Theorem

If  $y(t)$  exists, one can compute  $p, q$  such that  $\left| \frac{p}{q} - y(t) \right| \leq 2^{-n}$  in time  
 $\text{poly}(\text{size of } x \text{ and } p, n, \ell(t))$

where  $\ell(t) \approx$  length of the curve (between  $x$  and  $y(t)$ )

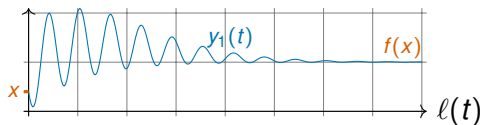


length of the curve = complexity = resource

# Characterization of real polynomial time

**Definition :**  $f : [a, b] \rightarrow \mathbb{R}$  in  $\text{ANALOG-P}_{\mathbb{R}} \Leftrightarrow \exists p$  polynomial,  $\forall x \in [a, b]$

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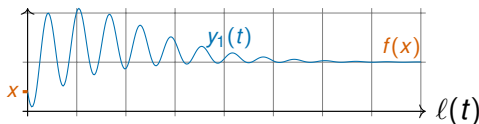
satisfies :

1.  $|y_1(t) - f(x)| \leq 2^{-\ell(t)}$

«greater length  $\Rightarrow$  greater precision»

2.  $\ell(t) \geq t$

«length increases with time»



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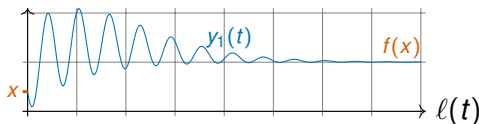
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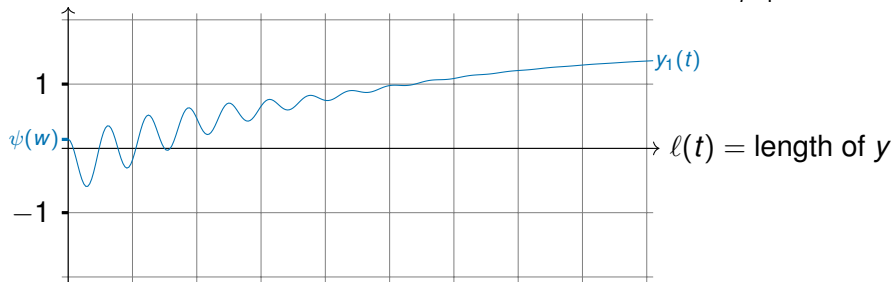
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**Definition :**  $\mathcal{L} \in \text{ANALOG-PTIME} \Leftrightarrow \exists p \text{ polynomial, } \forall \text{ word } w$

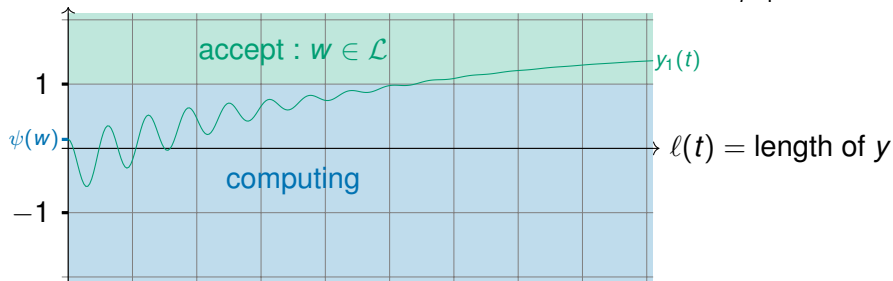
$$y(0) = (\psi(w), |w|, 0, \dots, 0) \quad y' = p(y) \quad \psi(w) = \sum_{i=1}^{|w|} w_i 2^{-i}$$



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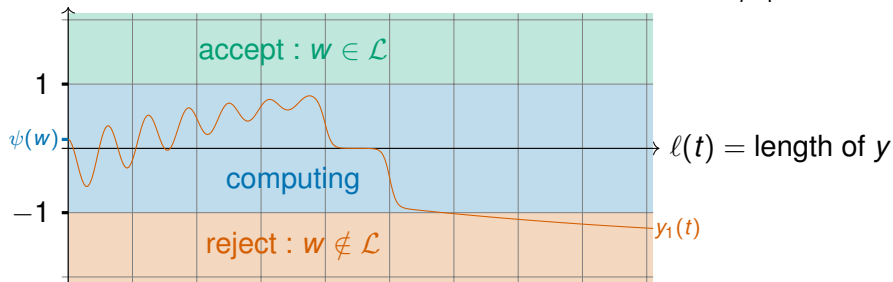
1. if  $y_1(t) \geq 1$  then  $w \in \mathcal{L}$



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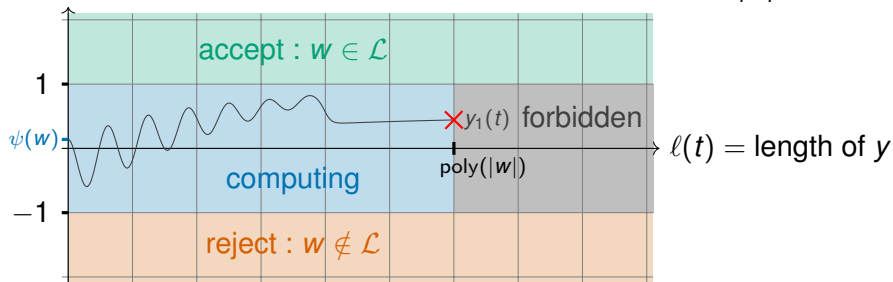
satisfies

2. if  $y_1(t) \leq -1$  then  $w \notin \mathcal{L}$

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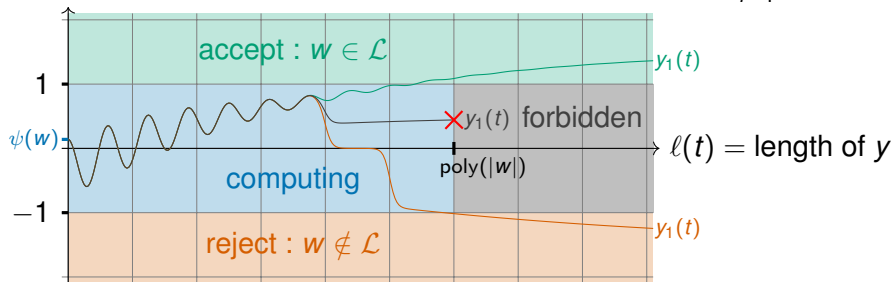
satisfies

3. if  $\ell(t) \geq \text{poly}(|w|)$  then  $|y_1(t)| \geq 1$

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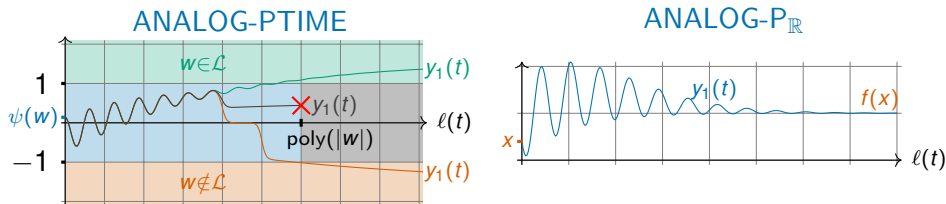
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**Theorem**

$$\text{PTIME} = \text{ANALOG-PTIME}$$

# Summary

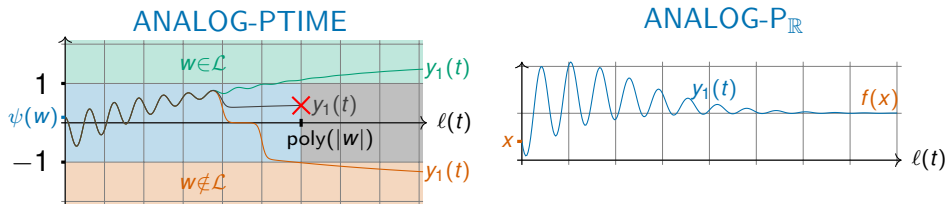


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- ▶ Only **rational coefficients** needed

# Back to Chemical Reaction Networks

## Theorem (CMSB 2017)

*Elementary mass-action-law reaction system on finite universes of molecules are Turing-complete under the differential semantics.*

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2. Disclaimer : not in the paper, I haven't checked the details.



Reaction networks :

- ▶ chemical
- ▶ enzymatic

$$y' = p(y)$$

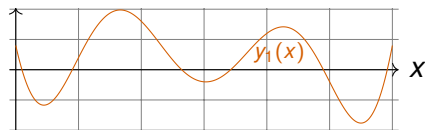
?

$$y' = p(y) + e(t)$$

- ▶ Finer time complexity (linear)
- ▶ Nondeterminism
- ▶ Robustness
- ▶ « Space » complexity
- ▶ Other models
- ▶ Stochastic

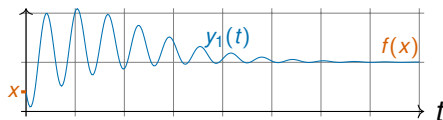
# Universal differential equations

## Generable functions



subclass of analytic functions

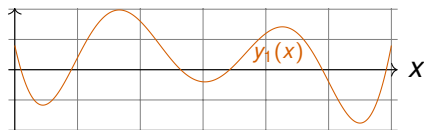
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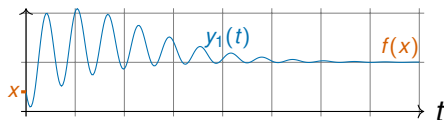
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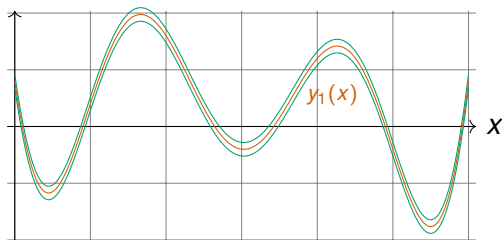


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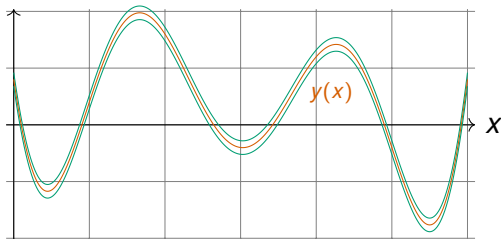


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# Universal differential algebraic equation (DAE)



## Theorem (Rubel, 1981)

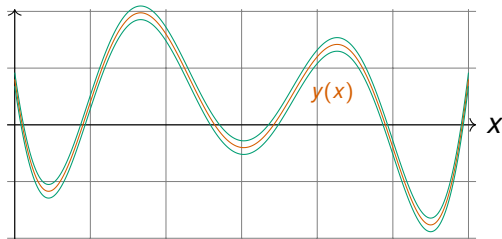
For any continuous functions  $f$  and  $\varepsilon$ , there exists  $y : \mathbb{R} \rightarrow \mathbb{R}$  solution to

$$\begin{aligned} &3y'^4 y'' y''''^2 - 4y'^4 y'''^2 y'''' + 6y'^3 y''^2 y''' y'''' + 24y'^2 y''^4 y'''' \\ &- 12y'^3 y'' y'''^3 - 29y'^2 y''^3 y'''^2 + 12y''^7 = 0 \end{aligned}$$

such that  $\forall t \in \mathbb{R}$ ,

$$|y(t) - f(t)| \leq \varepsilon(t).$$

# Universal differential algebraic equation (DAE)



## Theorem (Rubel, 1981)

There exists a **fixed** polynomial  $p$  and  $k \in \mathbb{N}$  such that for any continuous functions  $f$  and  $\varepsilon$ , there exists a solution  $y : \mathbb{R} \rightarrow \mathbb{R}$  to

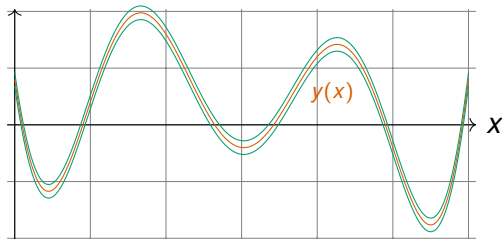
$$p(y, y', \dots, y^{(k)}) = 0$$

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such that  $\forall t \in \mathbb{R}$ ,

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**Problem** : this is «weak» result.

# The problem with Rubel's DAE

The solution  $y$  is not unique, **even with added initial conditions** :

$$p(y, y', \dots, y^{(k)}) = 0, \quad y(0) = \alpha_0, y'(0) = \alpha_1, \dots, y^{(k)}(0) = \alpha_k$$

In fact, this is fundamental for Rubel's proof to work !

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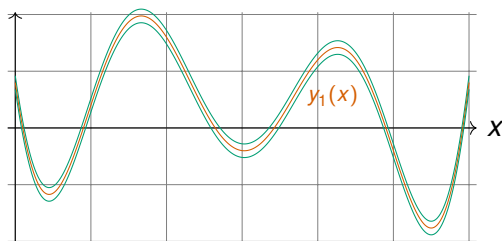
- ▶ Rubel's statement : this DAE is universal
- ▶ More realistic interpretation : this DAE allows almost anything

## Open Problem (Rubel, 1981)

Is there a universal ODE  $y' = p(y)$  ?

**Note** : explicit polynomial ODE  $\Rightarrow$  unique solution

# Universal initial value problem (IVP)



## Theorem

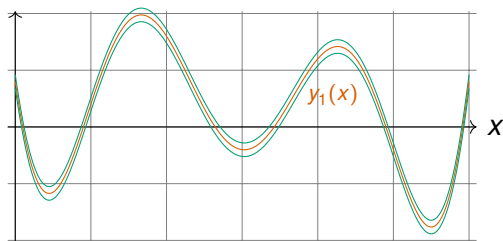
There exists a **fixed** (vector of) polynomial  $p$  such that for any continuous functions  $f$  and  $\varepsilon$ , there exists  $\alpha \in \mathbb{R}^d$  such that

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has a **unique solution**  $y : \mathbb{R} \rightarrow \mathbb{R}^d$  and  $\forall t \in \mathbb{R}$ ,

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# Universal initial value problem (IVP)



Notes :

- ▶ **system** of ODEs,
- ▶  $y$  is analytic,
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## Theorem

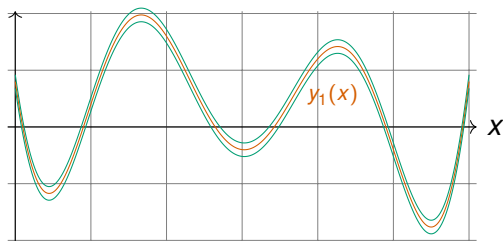
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**Remark :**  $\alpha$  is usually transcendental, but computable from  $f$  and  $\varepsilon$

# What is a computer ?

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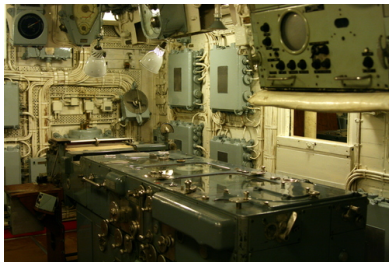




# What is a computer ?

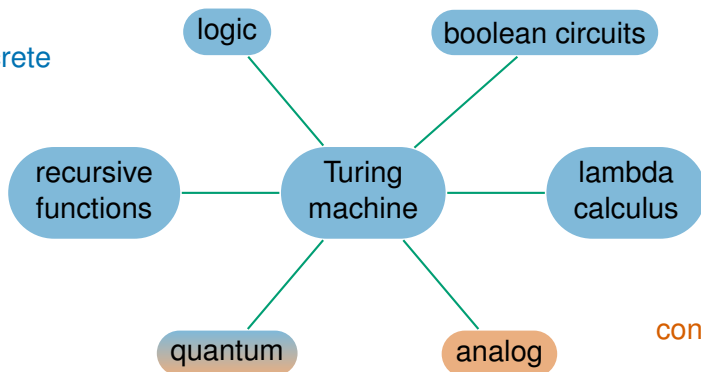


VS



## Computability

discrete



continuous

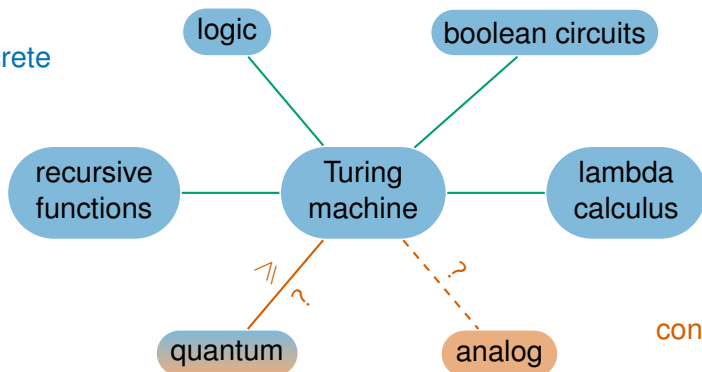
## Church Thesis

All **reasonable** models of computation are equivalent.

# Church Thesis

## Complexity

discrete



continuous

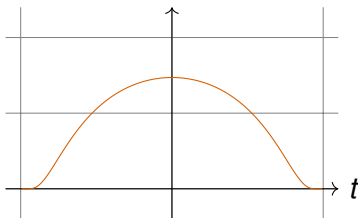
## Effective Church Thesis

All **reasonable** models of computation are equivalent for complexity.

# Rubel's proof in one slide

- Take  $f(t) = e^{\frac{-1}{1-t^2}}$  for  $-1 < t < 1$  and  $f(t) = 0$  otherwise.

It satisfies  $(1 - t^2)^2 f''(t) + 2t f'(t) = 0$ .



# Rubel's proof in one slide

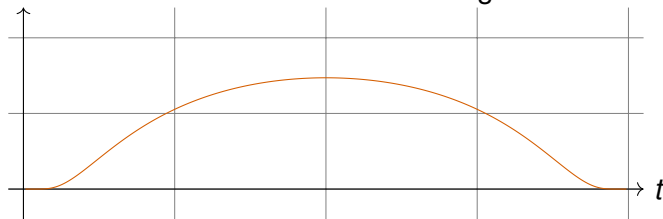
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- For any  $a, b, c \in \mathbb{R}$ ,  $y(t) = cf(at + b)$  satisfies

$$\begin{aligned} 3y'^4 y'' y''''^2 & - 4y'^4 y''^2 y'''' + 6y'^3 y''^2 y''' y'''' + 24y'^2 y''^4 y'''' \\ & - 12y'^3 y'' y''''^3 - 29y'^2 y''^3 y''''^2 + 12y''^7 = 0 \end{aligned}$$

Translation and rescaling :



# Rubel's proof in one slide

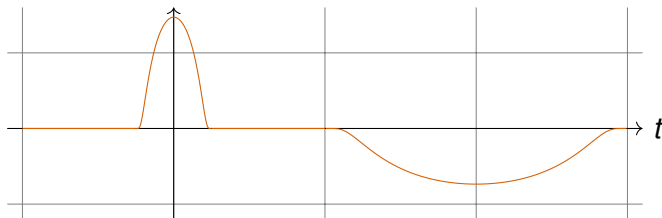
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- Can glue together arbitrary many such pieces



# Rubel's proof in one slide

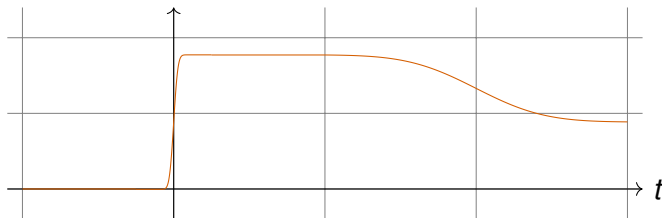
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- Can arrange so that  $\int f$  is solution : **piecewise pseudo-linear**



# Rubel's proof in one slide

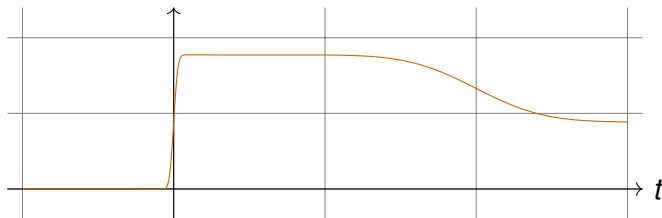
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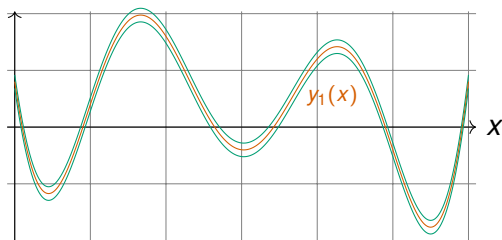
- Can glue together arbitrary many such pieces
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**Conclusion** : Rubel's equation allows any piecewise pseudo-linear functions, and those are **dense in  $C^0$**



# Universal DAE revisited



## Theorem

There exists a **fixed** polynomial  $p$  and  $k \in \mathbb{N}$  such that for any continuous functions  $f$  and  $\varepsilon$ , there exists  $\alpha_0, \dots, \alpha_k \in \mathbb{R}$  such that

$$p(y, y', \dots, y^{(k)}) = 0, \quad y(0) = \alpha_0, y'(0) = \alpha_1, \dots, y^{(k)}(0) = \alpha_k$$

has a **unique analytic solution** and this solution satisfies such that

$$|y(t) - f(t)| \leq \varepsilon(t).$$