

# Chemical Reaction Networks as a Programming Language

Turing-completeness, compiler into finite CRNs, and absolute functional robustness.

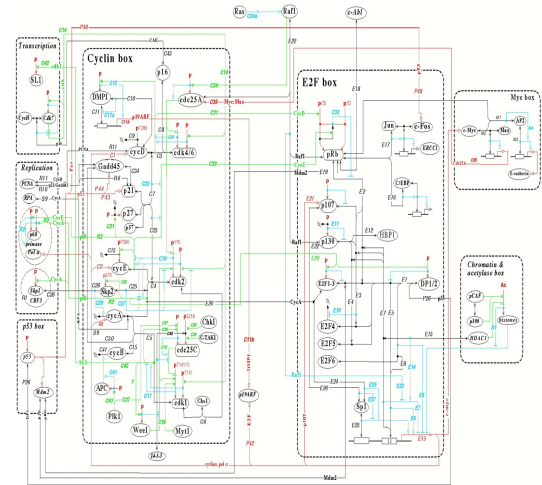


# Outline of my Talk

1. Chemical reaction networks
  - Syntax: reaction rules with well-formed kinetics
  - [Hierarchy of semantics](#): continuous ODE, stochastic CTMC, discrete PN, asynchronous Boolean
2. Static analyses based on hypergraph structure
  - Relating CRN models by subgraph epimorphisms
  - Necessary conditions for multistationarity as positive circuits in the influence graph
3. CRN analog computing
  - [Turing completeness](#), normal form theorem
  - Compiler of real computable functions in elementary CRNs:
  - Quadratization: complexity and open problems
  - [Online computation](#) and [Absolute Functional Robustness \(AFR\)](#)
4. Conclusion on rule-based mathematical modeling

# Motivation for this Work: Cells Compute

- Cells process signals
- Regulate their metabolism
- Take decisions such as
  - Replication
  - Differentiation
  - Move
  - Apoptosis (suicide)
- Control the execution of those processes



But what are the programs ?

Chemical Reaction Networks (CRN)

Analog computation with proteins: gradual concentration levels, continuous time

*Church-Turing thesis*: there is only one notion of mechanistic computability

- What are the links to Turing machines and digital computation?
- Can we understand, beyond describing, natural CRNs ? (Systems Biology)
- Can we synthesize artificial CRN to implement a function ? (Synthetic Biology)

# CRN Syntax

Let  $S = \{x_1, \dots, x_s\}$  be a finite set of molecular species.

**Def.** A reaction is a quadruple  $(R, I, P, f)$ , also noted  $R / I \xrightarrow{f} P$

where  $R$  (resp.  $I, P$ ) is a multiset of reactant species (resp. inhibitors, products)

and  $f: \mathbb{R}_+^S \rightarrow \mathbb{R}_+$  is a rate function (kinetic expression).

- Multisets can be represented by linear expressions (stoichiometric coefficients)
- A reaction catalyst is a molecular species that is both a reactant and a product (can also be an inhibitor).

**Def.** A reaction  $(R, I, P, f)$  is **well-formed** if

- $f: \mathbb{R}_+^S \rightarrow \mathbb{R}_+$  is a partially differentiable function
- $x_i \in R$  if and only if  $\frac{\partial f}{\partial x_i}(x) > 0$  for some value  $x \in \mathbb{R}_+^S$
- $x_i \in I$  if and only if  $\frac{\partial f}{\partial x_i}(x) < 0$  for some value  $x \in \mathbb{R}_+^S$ .

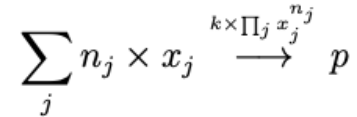
**Def.** A reaction is **strict** if  $R(x_i) > 0$  implies  $f(x_1, \dots, x_s) = 0$  whenever  $x_i = 0$ .

**Def.** A (strict, well-formed) CRN is a finite set of (strict, well-formed) reactions.

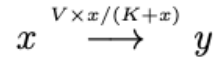
**Prop.** The ODE associated to a well-formed strict CRN defines a positive system

# Standard Well-formed Strict CRN Kinetics

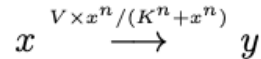
*mass action law kinetics:*



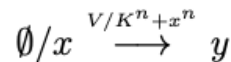
*Michaelis-Menten kinetics:*



*Hill kinetics:*



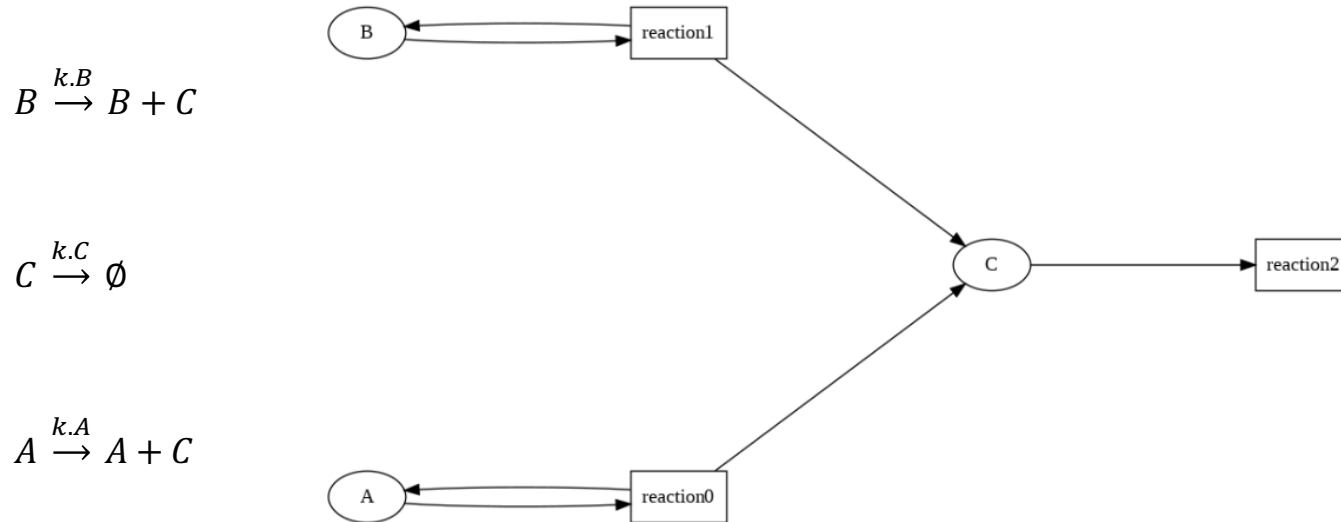
*or negative Hill kinetics:*



*with rate constants  $k, V, K > 0$  and exponent  $n \geq 1$ , are well-formed and strict.*

# CRN Hypergraph Structure

Standard representation of a hypergraph by a [bipartite species/reaction graph](#).



# CRN Semantics

One given CRN  $\{(R_r, I_r, P_r, f_r)\}_{r \in C}$  can be interpreted in a **hierarchy of semantics** :

- Continuous interpretation by **ordinary differential equations** (ODE) in explicit form  $x \in \mathbb{R}_+^S$

$$\frac{dx_i}{dt} = \sum_{r \in C} (P_r(x_i) - R_r(x_i)) \cdot f_r(x)$$

- Stochastic interpretation by **continuous-time Markov chain** (CTMC)  $x \in \mathbb{N}_+^S$

$$x \xrightarrow{x \geq R_r, p = \frac{f_r(x)}{\sum f_{r'}(x)}, \tau = \text{Exp}\left(\frac{1}{\sum f_{r'}(x)}\right)} x - R_r + P_r$$

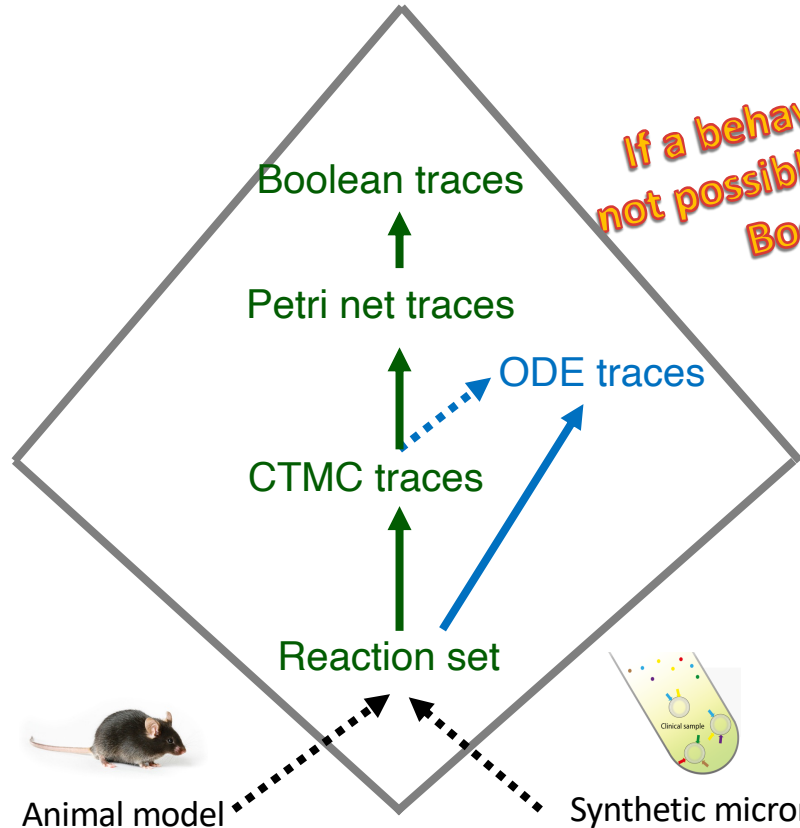
- Rate-independent non-deterministic discrete interpretation by **Petri Net** (PN)  $x \in \mathbb{N}_+^S$

$$x \xrightarrow{x \geq R_r} x - R_r + P_r$$

- Rate-independent non-deterministic **asynchronous Boolean state transition** interpretation  $x \in \mathcal{B}_+^S$

$$x \xrightarrow{x \geq \bar{R}_r, x < \bar{P}_r} x' \quad \text{with} \quad (x \wedge \neg \bar{R}_r) \vee \bar{P}_r \leq x' \leq x \vee \bar{P}_r$$

# Hierarchy of Trace Semantics



**If a behavior is not possible in the Boolean semantics, it is not possible in the stochastic semantics for any reaction rates.**  
**Boolean model-checking can exhibit rare events**

Thm. (abstract interpretation  $\uparrow$ ) Galois connections between the domains of syntactical, stochastic, Petri net and Boolean trace semantics

Fages, Soliman. [Abstract Interpretation and Types for Systems Biology](#). *Theoretical Computer Science*, 403(1):52–70, 2008.

Thm. (approximation  $\uparrow$ ) When the volume tends to infinity the ODE concentration semantics approximates the *mean* stochastic behavior [Gillespie 1971 Kurtz 1978, 1992]

**Limit theorem  
Not the whole story**



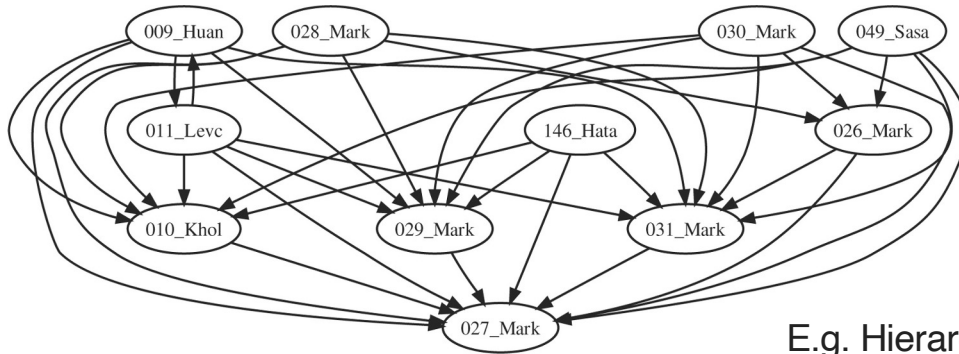
## 2. Static Analyses Based on Hypergraph Structure

# Relating CRN Models by Graph Matching

SBML: markup language, exchange format for CRN models

Biomodels: repository of models of biological and biomedical systems (with typically 10 to  $10^3$  variables)  
*flat list* of thousands of models in SBML with reference to publications and various annotations.

Hierarchy of CRN models related by graph morphisms (purely structural concept)  
= Metamodel of models at different levels of details



E.g. Hierarchy of MAPK signalling CRN models in Biomodels

Steven Gay, Sylvain Soliman, François Fages. [A Graphical Method for Reducing and Relating Models in Systems Biology](#). *Bioinformatics*, 26(18):i575–i581, 2010.

# Subgraph Epimorphisms (SEPI)

**Def.** A graph epimorphism from graph  $G = (V, A)$  to  $G' = (V', A')$  is a **surjective function**  $f : V \rightarrow V'$  s.t.

- for all  $u, v \in V$ , if  $(u, v) \in A$ , then  $(f(u), f(v)) \in A'$  (graph homomorphism),
- and for all  $(u', v') \in A'$ , there exists  $(u, v) \in A$  such that  $f(u) = u'$  and  $f(v) = v'$  (**surjectivity on arcs**).

**Def.** A subgraph epimorphism from  $G$  is graph epimorphism from a subgraph induced by a subset  $U \subseteq V$

**Thm.**  $G \xrightarrow[SEPI]{} G'$  iff  $G'$  is isomorphic to a graph obtained from  $G$  by a **sequence of graph operations** to

- delete species (and incoming/outgoing arcs)
- delete reactions (and incoming/outgoing arcs)
- merge species (and their incoming and outgoing arcs)
- merge reactions (and their incoming and outgoing arcs).

**Thm.** The existence of a SEPI between two graphs is NP-complete.

**Proof.** By reduction of SAT.

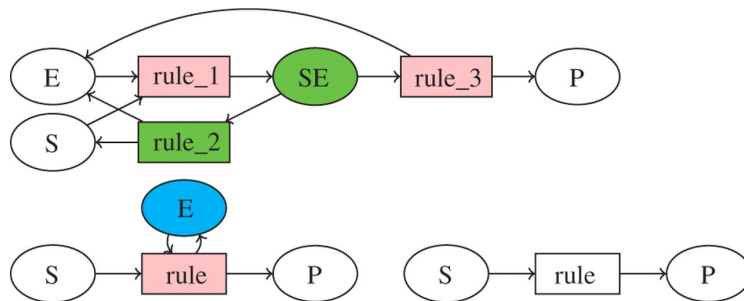
Implemented in Biocham using a Constraint Logic Program (or a SAT solver).

Steven Gay, François Fages, Thierry Martinez, Sylvain Soliman, Christine Solnon. [On the subgraph Epimorphism Problem](#). *Discrete Applied Mathematics*, 162:214–228, 2014.

# Example of SEPI for Michaelis-Menten Reduction

Unlike subgraph isomorphisms (SISO), SEPI does capture Michaelis-Menten CRN reductions from 3 reactions with mass action kinetics  $k_1.E.S$ ,  $k_2.SE$ ,  $k_3.SE$  with the enzyme  $E$  and complex  $SE$  SEPI corresponding for instance to

- Merge rule\_1 rule\_3
- Delete rule\_2
- Delete SE
- (Delete E)



to 1 reaction  $S \rightarrow P$  with(out) the enzyme

Justified with Michaelian kinetics  $\frac{V_M S}{K_M + S}$  where  $V_M = k_3 E_{(0)}$  and  $K_M = \frac{k_2 + k_3}{k_1}$

by quasi-steady state approximation on SE if  $E \ll S$  (or by quasi-equilibrium if  $k_3 \ll k_2$  with  $K_M = \frac{k_2}{k_1}$ )

Assuming  $\frac{dES}{dt} = 0$  we get  $\frac{dP}{dt} = -\frac{dS}{dt} = \frac{V_M S}{K_M + S}$  and can show preservation of time scales [Segel 84]

More precise proof than Tikhonov theorem of perturbation theory ? Error control ? Transseries ?

Conjecture: compositionality of QSSA/QE reductions with reduction to automaton of reduced dynamics.

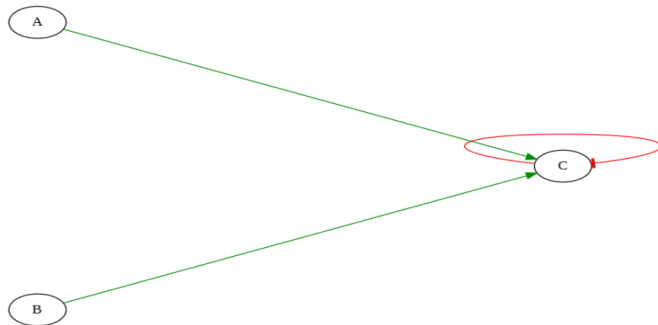
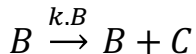
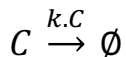
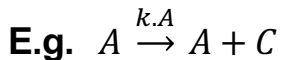
# Influence Graph of a CRN

**Def.** The **differential influence graph (DIG)** of a CRN is the graph of signs of the Jacobian matrix:

$$\begin{aligned} & \{A \rightarrow^+ B \mid \partial x'_B / \partial x_A > 0 \text{ for some value } x \in R_+^s\} \\ & \cup \{A \rightarrow^- B \mid \partial x'_B / \partial x_A < 0 \text{ for some value } x \in R_+^s\} \end{aligned}$$

**Def.** The **syntactical influence graph (SIG)** of a CRN is the graph with signed arcs

$$\begin{aligned} & \{A \rightarrow^+ B \mid \exists(R_i, I_i, P_i, f_i) (R_i(A) > 0 \text{ and } P_i(B) - R_i(B) > 0) \text{ or } (I_i(A) > 0 \text{ and } P_i(B) - R_i(B) < 0)\} \\ & \cup \{A \rightarrow^- B \mid \exists(R_i, I_i, P_i, f_i) (R_i(A) > 0 \text{ and } P_i(B) - R_i(B) < 0) \text{ or } (I_i(A) > 0 \text{ and } P_i(B) - R_i(B) > 0)\} \end{aligned}$$



**Prop.** The DIG of a well-formed CRN is included in its SIG, and equal if the SIG contains no conflict pair.

# Positive Circuits as Necessary Condition for Multistationarity

**Thm.** [Soliman 2013] A [necessary condition for non-zero multistability](#) in the ODE semantics of a CRN is the existence of a positive circuit in the reaction-labelled DIG of the CRN

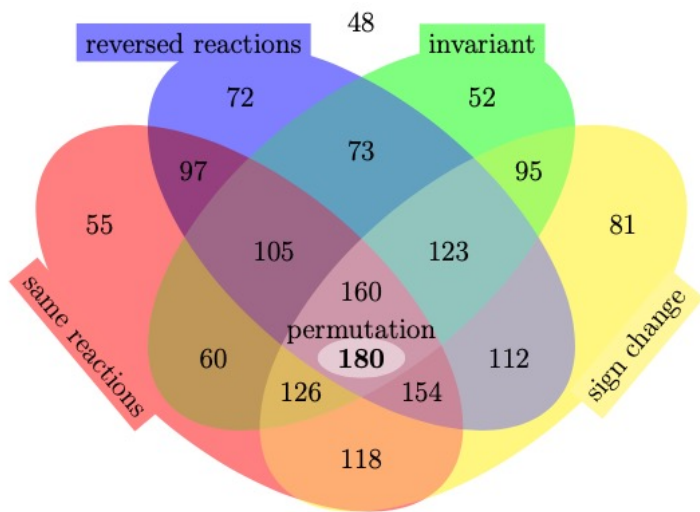
- using at most once each reaction
- not both forward and backward reactions of a reversible reaction
- not all species involved in a conservation law
- and in every DIG obtained by reversing the arcs targeting some technical subsets of species
- and in every DIG obtained by rewiring the arcs targeting any permutation of some subsets of species.

Sylvain Soliman. [A stronger necessary condition for the multistationarity of chemical reaction networks](#). *Bulletin of Mathematical Biology*, 75(11):2289-2303, 2013.

Implemented in Biocham by graph rewriting algorithms.

Adrien Baudier, François Fages, Sylvain Soliman. [Graphical Requirements for Multistationarity in Reaction Networks and their Verification in BioModels](#). *Journal of Theoretical Biology*, 459:79-89, 2018.

# Evaluation of Multistability Necessary Condition on Biomodels



$r$	Jacobian Method [13]	Graphical Method (Alg 2 & 4)		
	model (1)	(1)	(2)	(3)
1	4	0.3	1.6	0.6
2	75	0.5	2	1
3	44	1	2	1
4	81	1	3	1
5	191	1	4	1
6	256	1	4	1
7	444	1	5	1
8	795	2	5	1
9	1169	2	6	2
10	2195	2	6	2
11	3998	2	6	2
12	7696	2	7	2
13	15180	2	7	2
14	32180	3	7	2
15	67740	3	7	2
16	171700	3	8	2
17	1199000	4	8	2
50	×	12	17	4
100	×	26	40	6
500	×	343	549	34
1000	×	1200	1874	98

Figure 6: Number of models among the 506 curated reaction models of BioModels for which multistationarity can be ruled out by using respectively original Thomas's positive circuit condition, Cor. 2.3 (no same reactions), 2.4 (no reversed reactions), 2.5 (no invariant) and 2.6 (sign change), plus 2.7 (permutation).

Table 2: Execution times given in milliseconds for the analysis of the  $r$ -site phosphorylation system of [34], first as reported in [13] for the Jacobian method using symbolic computation, then obtained with our graphical algorithm on the same model and on two variants concerning the writing of the dephosphorylation and phosphorylation reactions.

## 3. CRN Synthesis



# Computable Real Numbers and Functions

Classical definitions of **computable analysis** based on Turing machines

**Definition.** A **real number**  $r$  is **computable** if there exists a Turing machine with

Input: precision  $p \in \mathbb{N}$

Output: rational number  $q \in \mathbb{Q}$  with  $|r - q| < 2^{-p}$

**Examples.** Rational numbers, limits of computable Cauchy sequences  $\pi$ ,  $e$ , ...

**Definition.** A **real function**  $f: \mathbb{R} \rightarrow \mathbb{R}$  is **computable** if there exists a Turing machine that computes  $f(x)$  with an oracle for  $x$ .

**Examples.** Polynomials, trigonometric functions, analytic functions...

**Counter-examples.**  $x=0$ ,  $\lceil x \rceil$  are not computable (undecidable on  $x=0.000\dots$ ) discontinuous functions are not computable

**Decision problem**  $w \in \mathcal{L}$ : analog encoding by a real function  $f: \mathbb{R} \rightarrow \mathbb{R}$  ?

Input encoding  $e: \mathcal{L} \rightarrow \mathbb{R}$  problem encoding by  $f$ : accept  $w$  if  $f(e(w)) > 1$  reject if  $< -1$

# Analog Computer? Differential Analyzer [Bush 1931]

Underlying principles: Lord Kelvin, 1876

First ever built: Vannevar Bush, MIT, 1931



Applications: from gunfire control up to aircraft design

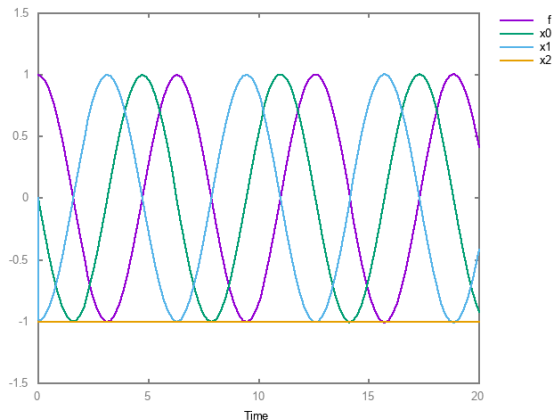
- Intensively used by the U.S. and Japanese armies during world war II
- Electronic versions from late 40s, used until 70s

# General Purpose Analog Computer [Shannon 1941]

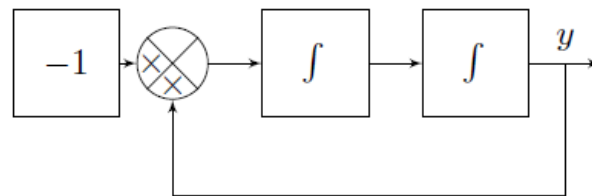
Shannon's formalization of the [Differential Analyser](#) by GPAC circuits

A time function is GPAC-generated if it is the output of some unit of a GPAC circuit built from:

1. Constant unit
2. Sum unit
3. Product unit
4. Integral  $\int y dx$  unit ( $dt$  by default)



What does this GPAC circuit compute ?



$$y_1 = \frac{dy}{dt}$$

$$\frac{dy_1}{dt} = -y = y''$$

$$\text{if } y(0) = 1, y_1(0) = 0$$

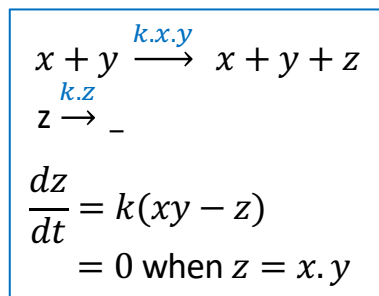
$$y(t) = \cos(t) \quad y_1(t) = \sin(t)$$

# CRN Implementation of GPAC Units

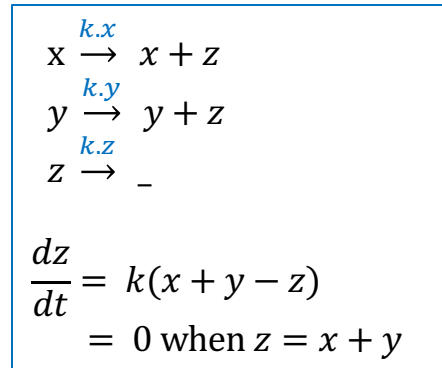
Mass action law kinetics reaction network with output concentration stabilizing on the result of the operation applied to the input concentrations

Positive constant units: molecular concentrations

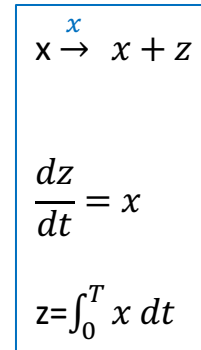
Product unit  $z = x \cdot y$



Sum unit  $z = x + y$



Time integral  $z = \int x dt$  unit

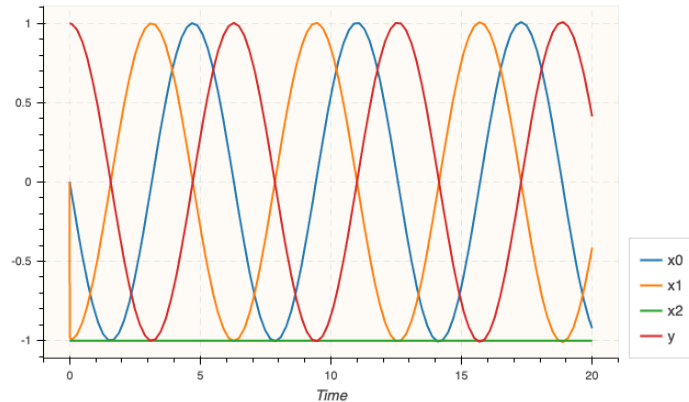


# Polynomial ODE Initial Value Problems (PIVP)

Graça and Costa 2003's formalization of GPAC generated functions

**Definition.** A real time function  $f: \mathbb{R}_+ \rightarrow \mathbb{R}$  is **PIVP-generable** iff there exist a **vector of polynomials**  $p \in \mathbb{R}^n[\mathbb{R}^n]$  and of initial values  $y(0) \in \mathbb{R}^n$  and a solution function  $y: \mathbb{R}_+ \rightarrow \mathbb{R}^n$  such that  $y'(t) = p(y(t))$  and  $f(t) = y_1(t)$

**Example.**  $y = \cos(t)$



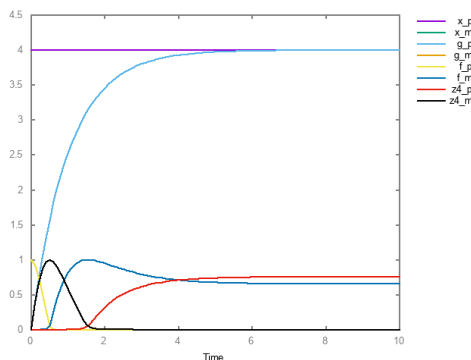
Closure properties:

$f+g$ ,  $f-g$ ,  $f.g$ ,  $1/f$ ,  $f \circ g$ ,  $y$  s.t.  $y' = f(y)$  are GPAC-generable if  $f$ ,  $g$  are.

# PIVP-Computable Function $f(x)$

**Definition.** [Graça Costa 03 J. Complexity] A real function  $f: \mathbb{R} \rightarrow \mathbb{R}$  is **PIVP-computable** if there exists vectors of polynomials  $p \in \mathbb{R}^n[\mathbb{R}^n]$  and  $q \in \mathbb{R}^n[\mathbb{R}]$  and a function  $y: \mathbb{R}^n \rightarrow \mathbb{R}^n$  such that  $y'(t) = p(y(t))$ ,  $y(0) = q(x)$  and  $|y_1(t) - f(x)| < y_2(t)$  with  $y_2(t) \geq 0$  decreasing for  $t > 1$  and  $\lim_{t \rightarrow \infty} y_2(t) = 0$

**Example.**  $y = \cos(4)$



**Reconciles  
Digital and Analog  
Turing and Shannon  
Computation !**

**Theorem (analog characterization of Turing computability).** [Bournez Campagnolo Graça Hairny 07 J. Complex]

A real function is **computable (by Turing machine)** iff it is **PIVP-computable**.

# Analog characterization of Ptime

Reconciles Digital and Analog  
Polynomial Time Complexity!

**Theorem** [Pouly PhD thesis 2015, Bournez Graca Pouly 16 ICALP]

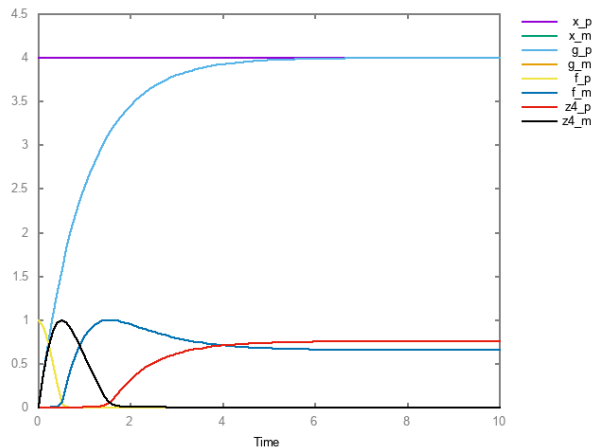
A real function is in **Ptime** iff it is PIVP-computable with a trajectory of polynomial length (i.e. polynomial time and polynomial amplitude)

Time in ODE is a bad measure of complexity

- Exponential speedup by changing time variable  $t' = e^t$
- But price to pay in the amplitude of  $t'$

The computational complexity measure here combines time and space-amplitude:

- length in the  $n$  dimensions of the trajectory to compute the result



# Turing Completeness of Continuous CRN

- Consider mass action law kinetics
  - polynomial ODEs
  - PIVP computation of input/output function
- Molecular concentration are positive real values
  - Restriction to positive dynamical systems
- Elementary reactions with at most two reactants
  - Restriction to PIVP of degree at most 2



# Turing Completeness of Continuous CRNs

**Lemma (positive systems)** Any PIVP-computable function can be encoded by a PIVP of double dimension on  $\mathbb{R}^+$ , preserving polynomial length complexity.

**Proof.** Encode  $y_i \in \mathbb{R}$  by  $y_i^+, y_i^- \in \mathbb{R}^+$  such that  $y_i = y_i^+ - y_i^-$  (dual-rail encoding of [Hars Toth 79] used in [Oishi Klavins 2011] for encoding linear I/O systems)

For a PIVP  $p[y_1, \dots, y_n]$  let  $\underline{p}(y_1^+, y_1^-, \dots, y_n^+, y_n^-) = p[y = y_i^+ - y_i^-]$

Let  $\underline{p}_i = \underline{p}_i^+ - \underline{p}_i^-$  where  $\underline{p}_i^+, \underline{p}_i^-$  are positive coefficient polynomials.

Let us consider the positive PIVP defined by

$$\begin{aligned} y_i^{+'} &= \underline{p}_i^+ - f_i y_i^+ y_i^- & y_i^+(0) &= \max(0, y_i(0)) \\ y_i^{-'} &= \underline{p}_i^- - f_i y_i^+ y_i^- & y_i^-(0) &= \max(0, -y_i(0)) \end{aligned}$$

where  $f_i$  is chosen large enough such that  $f_i y_i^+ y_i^- \geq \max(\underline{p}_i^+, \underline{p}_i^-)$

At any time time we have  $y_i(t) = y_i^+(t) - y_i^-(t)$ .

- **Fast annihilation** reactions:  $y_i^+ + y_i^- \xrightarrow{f_i} \_$
- **n-ary catalytic synthesis** reactions for each monomial  $\underline{m}_{i,j}^+$  in  $\underline{p}_i^+$ ,  $\underline{m}_{i,j}^-$  in  $\underline{p}_i^-$ :



# Turing Completeness of Continuous CRNs

**Lemma (quadratic systems)** [Carothers Parker Sochacki Warne 2005]

Any PIVP can be encoded by a PIVP of degree  $\leq 2$ .

**Proof.** Introduce **variable**  $v_{i_1, \dots, i_n}$  for each possible **monomial**  $y_1^{i_1} \dots y_n^{i_n}$

We have  $y_1 = v_{1,0,\dots,0}$ ,  $y_2 = v_{0,1,0,\dots,0}$ , ...  $y'_i$  is of degree one in  $v_{i_1, \dots, i_n}$

$v'_{i_1, \dots, i_n} = \sum_{k=1}^n i_k v_{i_1, \dots, i_{k-1}, \dots, i_n} y'_k$  is of degree at most 2. Trade high dimension for low degrees.

That algorithm may introduce an exponential number of variables.

**Thm.** Deciding the existence of such monomial quadratization of dimension  $k$  is NP-complete in the non-succinct (matrix) representation (in NExp with succinct symbolic representation).

**Proof.** By reduction of the feedback vertex set problem.

Mathieu Hemery, François Fages, Sylvain Soliman. [On the Complexity of Quadratzation for Polynomial Differential Equations](#). In *CMSB'20: Proceedings of the eighteenth international conference on Computational Methods in Systems Biology, Lecture Notes in Computer Science*. Springer-Verlag, 2020.

# Optimal Quadratizations

Suboptimal quadratization implemented in Biocham using a SAT solver and solution preserving heuristics.

Mathieu Hemery, François Fages, Sylvain Soliman. [On the Complexity of Quadratization for Polynomial Differential Equations](#). In *CMSB'20: Proceedings of the eighteenth international conference on Computational Methods in Systems Biology, Lecture Notes in Computer Science*. Springer-Verlag, 2020.

Optimal branch-and-bound algorithm for monomial quadratization

Andrey Bychkov and Gleb Pogudin. Optimal monomial quadratization for ode systems. In Proceedings of the IWCCA 2021 - 32nd International Workshop on Combinatorial Algorithms, July 2021.

Existence of non-monomial quadratizations of smaller dimension

Alauddin, F.: Quadratization of ODEs: monomial vs. non-monomial. *SIAM Undergraduate Res. Online* **14** (2021).

Could the existence of an unrestricted quadratization of dimension  $k$  be undecidable?

Similarly to Matiyasevich 1971 Hilbert 10<sup>th</sup> problem, solving of Diophantine Equations, is undecidable,

# Turing Completeness of Continuous CRNs

Fages, François, Le Guludec, Guillaume and Bournez, Olivier, Pouly, Amaury. [Strong Turing Completeness of Continuous Chemical Reaction Networks and Compilation of Mixed Analog-Digital Programs](#). In Proc. *CMSB'17*: pages 108–127, volume 10545 of *LNCS*. Springer-Verlag, 2017.

**Theorem** Any computable function over the reals can be computed by a continuous CRN over a finite set of molecular species (no polymerization, no compartments)

In this view, the (protein) concentrations are the information carriers.

The programs of a cell are implicitly defined by the set of all possible reactions

- with the proteins encoded in its genome
- and the chemicals of the environment.

Program change is determined by gene expression which can be seen as a (digital) metaprogram

- No artificial construct (no polymers)
- Compatible with natural cells: making of programming a “natural science” !

# Normal Form Theorem

## Theorem (abstract CRN normal form)

A real function is computable if and only if it is computable by a system of elementary reactions of the form



plus annihilation reactions  $x+y \Rightarrow \_$  all with mass action law kinetics

## Realistic CRN:

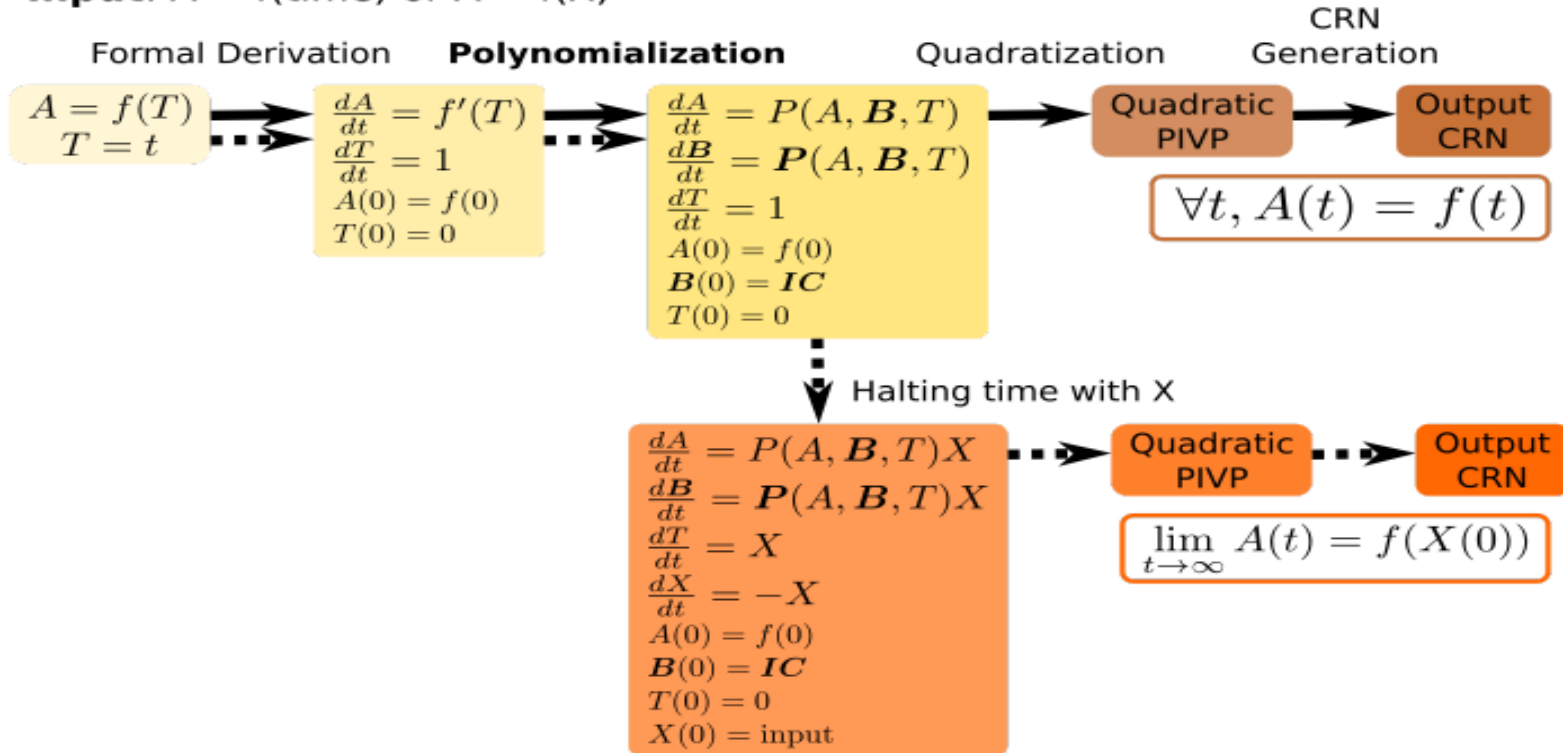
- formal annihilations by complexations (e.g. in a stable inactive complex)
- formal syntheses by modifications (e.g. phosphorylation with kinases)

## Concrete CRN: search mapping with real enzymes (e.g. Brenda database)

- Easier for CRN with rate independence property (ensured by graphical conditions [Degrand Fages Soliman CMSB 2020])
- Robustness w.r.t. parameter perturbations (extrinsic variability)
- Robustness w.r.t. stochastic simulations (intrinsic variability)

# Compiler of Real Functions in Elementary CRNs

**Input:**  $A = f(\text{time})$  or  $A = f(X)$





# Compiling Cosine(time)

`biocham`: `compile_from_expression(cos,time,f)`.

`initial_state(f_p=1)`.

MA(fast) for `f_m+f_p=>_`.

MA(fast) for `A_m+A_p=>_`.

MA(1.0) for `A_p=>A_p+f_p`.

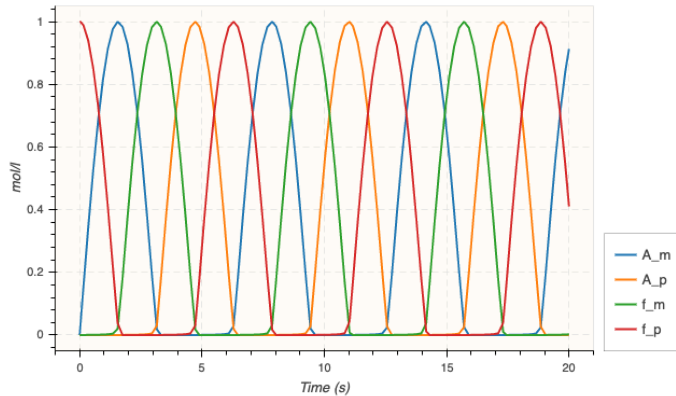
MA(1.0) for `A_m=>A_m+f_m`.

MA(1.0) for `f_m=>A_p+f_m`.

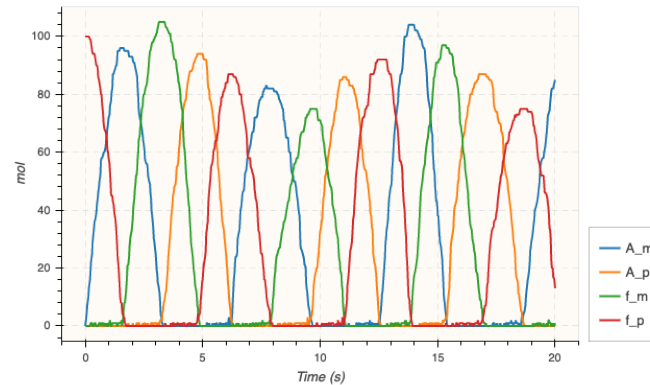
MA(1.0) for `f_p=>A_m+f_p`.

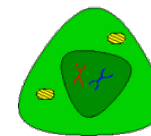
$$\begin{aligned}\frac{dA_m}{dt} &= f_p - fast * A_m * A_p \\ \frac{dA_p}{dt} &= f_m - fast * A_m * A_p \\ \frac{df_m}{dt} &= A_m - fast * f_m * f_p \\ \frac{df_p}{dt} &= A_p - fast * f_m * f_p\end{aligned}$$

ODE simulation (design)



Stochastic simulation (test)





# Compiling Cosine(input)

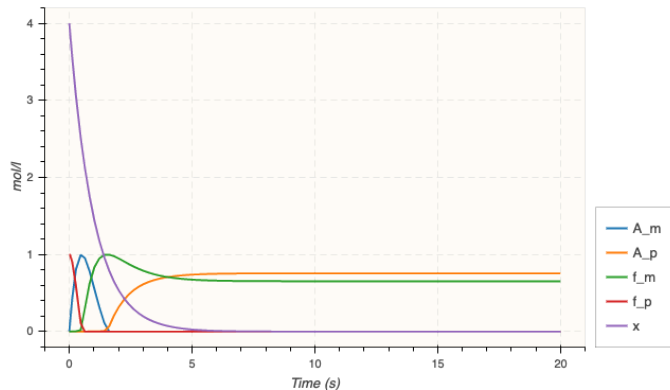
```
biocham: parameter(input=4).  
biocham: compile_from_expression(cos,x,f).  
initial_state(f_p=1, x=input).  
MA(fast) for f_m+f_p=>_.  
MA(fast) for A_m+A_p=>_.  
MA(1.0) for A_p+x=>A_p+f_p+x.  
MA(1.0) for A_m+x=>A_m+f_m+x.  
MA(1.0) for f_m+x=>A_p+f_m+x.  
MA(1.0) for f_p+x=>A_m+f_p+x.  
MA(1.0) for x=>_.
```

PIVP that generates  $f(g(t))$   
with  $\lim_{t \rightarrow \infty} g(t) = x$

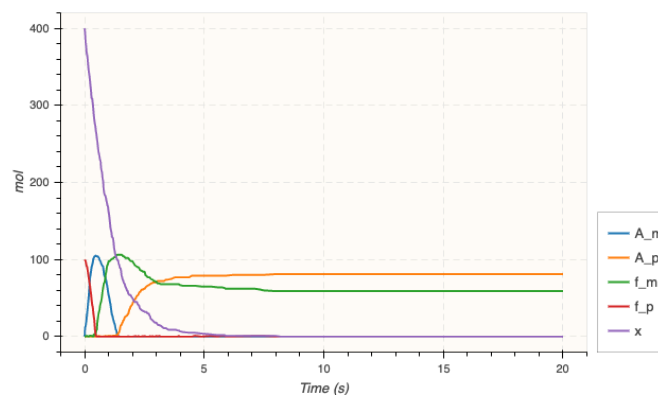
$$g'(t) = x - g(t)$$

$$g(t) = x + (x_0 - x)e^{-t}$$

ODE simulation (design)



Stochastic simulation (test)





# Sequentiality and Iteration

```

Division(A, B)
begin
01  while  $A \geq B$ 
02     $A := A - B$ 
03     $Q := Q + 1$ 
04     $R := A$ 
end
    
```

Main Reactions	Preconditions
01  while $[A] \geq [B]$	
02 $(A + B \rightarrow D)$	$\neg G_\theta$
03 $C \rightarrow Q + E$	$A_\theta \wedge \neg B_\theta$
04 $D \rightarrow F$	$\neg C_\theta$
05 $E \rightarrow G$	$\neg D_\theta$
06 $F \rightarrow B$	$\neg E_\theta$
07 $G \rightarrow C$	$\neg F_\theta$
08 $D \rightarrow R$	$\neg A_\theta$

## 1. Asynchronous (precondition) CRN programming

[Huang Jiang Huang Cheng 2012 ICCAD]

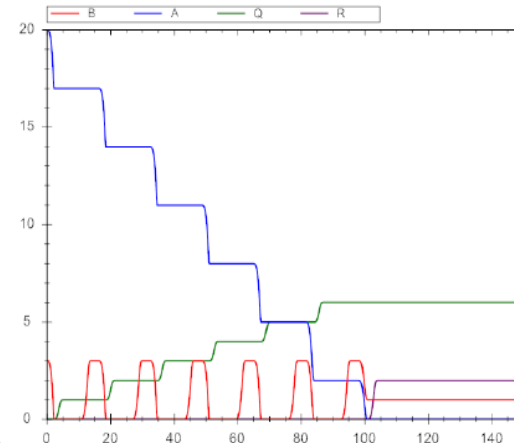
[Huang Huang Chiang Jiang Fages 2013 IWBD]

**many species and reactions**

## 2. Synchronous (clock) CRN programming

[ Vasic, David Soloveichik, Sarfraz Khurshid 2018 CRN++]

**many reactions with the clock**



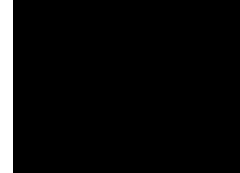
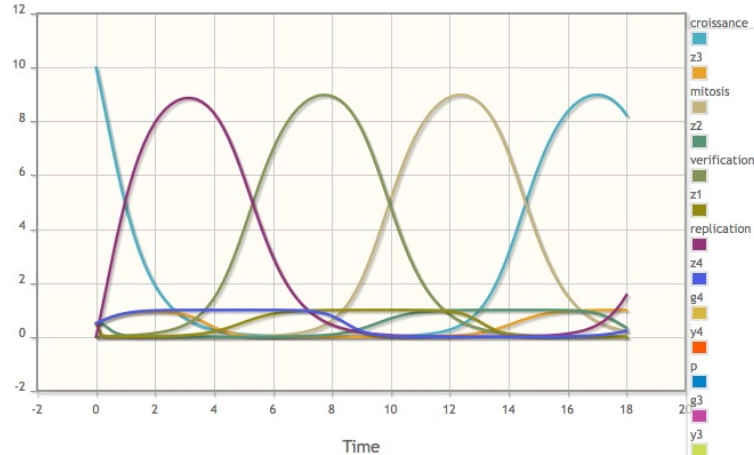
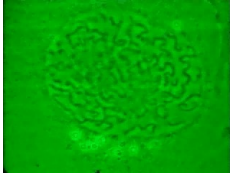
# Cell Division Cycle Program

```
while true {growing; replication; verification; mitosis}
```

→ compilation of **sequentiality** and **loops** with **program control variables**

→ 50 reactions

→ 13 variables



Cyclins D, E, A, B appear as necessary markers for implementing sequentiality

# TD Chemical Arithmetic

<http://lifeware.inria.fr/biocham4/online/notebooks/C2-19-Biochemical-Programming/22arith.ipynb>

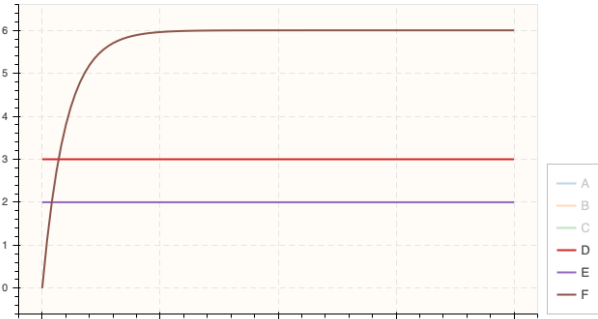
**Product  $F = E * F$**

```
In [11]: D+E => D+E+F.
```

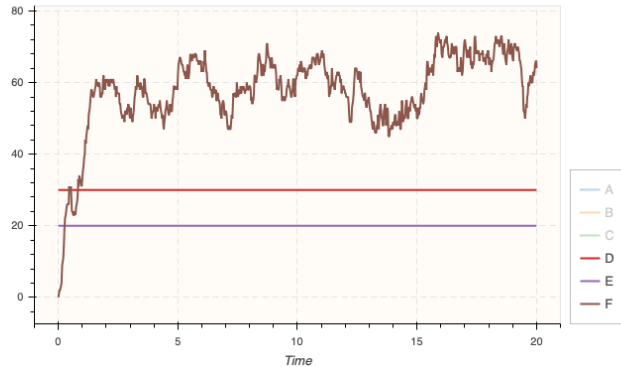
```
In [12]: F => _.
```

```
In [13]: present(D,d). parameter(d=3).  
present(E,e). parameter(e=2).
```

```
In [17]: option(show={D,E,F}).  
numerical_simulation.plot.
```



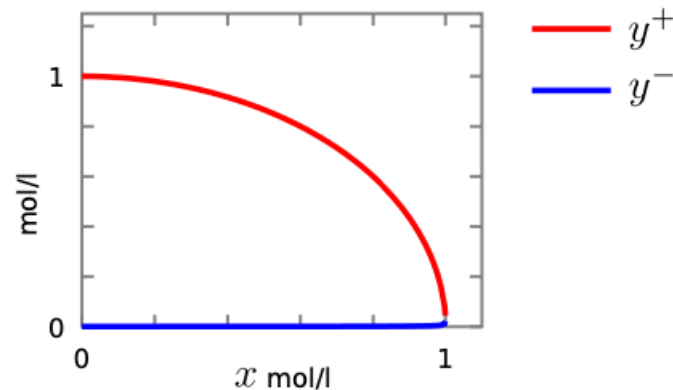
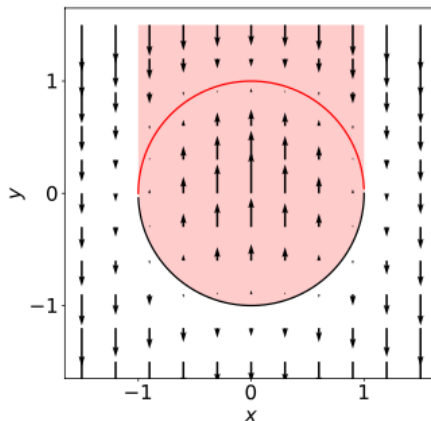
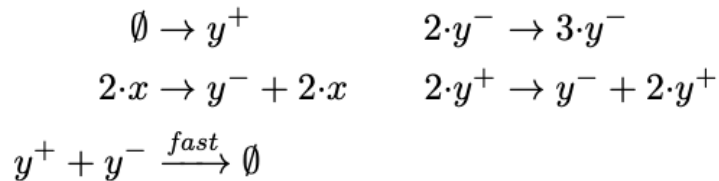
```
In [19]: option(method:ssa, stochastic_conversion: 10).  
numerical_simulation.plot.
```



# Robust Online Computation by Stabilization

`stabilize_expression(x^2+y^2-1, y, [x=0, y=1]).`

*Our pipeline gives us the following CRN.*

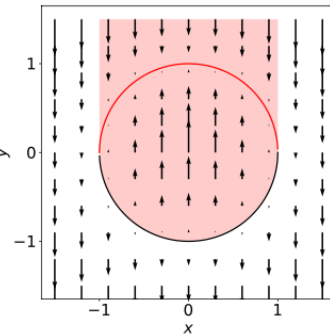


# Stabilization

**Definition 4.** We say that an open CRN over a set of  $m+1+n$  species  $\{X, y, Z\}$  with environment inputs  $X$  of cardinality  $m$  and distinguished outputs  $y$ , stabilizes the function  $f : I \rightarrow \mathbb{R}_+$ , with  $I \subset \mathbb{R}_+^m$ , over the domain  $\mathcal{D} \subset \mathbb{R}_+^{m+1+n}$  if:

1.  $\forall X^0 \in I$  the restriction of the domain  $\mathcal{D}$  to the slice  $X = X^0$  is of plain dimension  $n + 1$ , and
2.  $\forall (X^0, y^0, Z^0) \in \mathcal{D}$  the Polynomial Initial Value Problem (PIVP) given by the differential semantic with constant input species  $\forall t, X(t) = X^0$  and the initial conditions  $y^0, Z^0$  is such that:  $\lim_{t \rightarrow \infty} y(t) = f(X^0)$ .

**Definition 5.** The basin of attraction of a CRN stabilizing a function  $f : I \rightarrow \mathbb{R}_+$ , with  $I \subset \mathbb{R}_+^m$ , is the maximum domain (i.e. union of the domains) over which the CRN stabilizes  $f$ .



# Absolute Functional Robustness

**Definition 7.**  $(X^*, y^*, Z^*)$  is a partial equilibrium point of a system with distinguished species  $y$  if, fixing the environment inputs at  $\forall t, X(t) = X^*$  the trajectory initialized at  $y = y^*, Z = Z^*$  is such that  $\forall t \in \mathbb{R}_+, y(t) = y^*$ .

**Definition 8.** A CRN with distinguished inputs  $X$ , output  $y$  and intermediate species  $Z$  displays Absolute Functional Robustness (AFR) of the function  $f$  on the domain  $\mathcal{D}$  if for all choices of  $X$  in  $\mathcal{D}$ , there exist a partial stable equilibrium point  $(X, y)$  included in an open subset of  $\mathcal{D}$  and verifying:  $y = f(X)$ .

**Proposition 9.** A CRN displays AFR of  $f$  on  $\mathcal{D}$  if and only if it stabilizes  $f$  on  $\mathcal{D}$ .

# Characterization by Real Algebraic Functions

**Definition 10.** A function  $f : I \subset \mathbb{R}^m \rightarrow \mathbb{R}$  is algebraic if there exists a polynomial  $P_f$  of  $m + 1$  variables such that:

$$\forall X \in I, P_f(X, f(X)) = 0, \quad (6)$$

and  $\forall X \in I, (X, f(X))$  is a point of multiplicity 1.

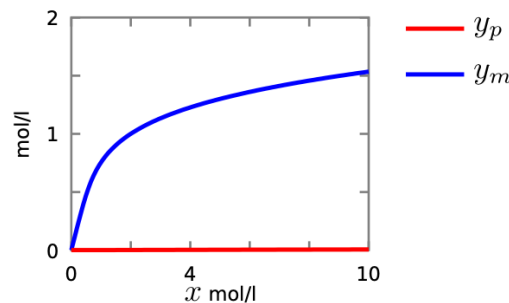
**Theorem 11.** The set of functions stabilized by a CRN with mass action law kinetics is the set of algebraic real functions:  $\mathcal{F}_S = \mathcal{F}_A$ .

**Example.** Bring radical  $y^5 + y + x = 0$

No analytic solution

CRN generated with 6 species

18 reactions + 2 annihilation reactions



# Wrapup

- A **rule-based** CRN model has an **explicit graphical structure**
- It is **higher-level** than a flat ODE model (similarly to structured program versus assembly code)
- **Hierarchy of CRN models** at different levels of details computed by **graph epimorphisms**
- **Hierarchy of semantics**: interpretations by ODE, CTMC, PN, Bool according to the question
- **Graphical analyses** provide efficient necessary or sufficient conditions for several dynamical properties
  
- **Turing-completeness**: any computable real function can be computed by a finite CRN
- **Compiler of real functions in elementary CRNs** through symbolic transformation steps
- Importance of the **quadratization problem** to minimize the dimension of the generated CRN
- **Absolute functional robustness of stabilizing CRN** characterized by **real algebraic functions**
  
- Open problems about
  - Optimal unrestricted quadratization (i.e. introducing variables for not necessarily monomial expressions)
  - **Restricting to a catalogue of concrete reactions** along the compilation pipeline.
- Practical applications for the **design of biosensors**

Alexis Courbet, Patrick Amar, François Fages, Eric Renard, Franck Molina. [Computer-aided biochemical programming of synthetic microreactors as diagnostic devices](#). *Molecular Systems Biology*, 14(4), 2018.