On the Computational Complexity of Rate-Independent CRN/ODE Computations

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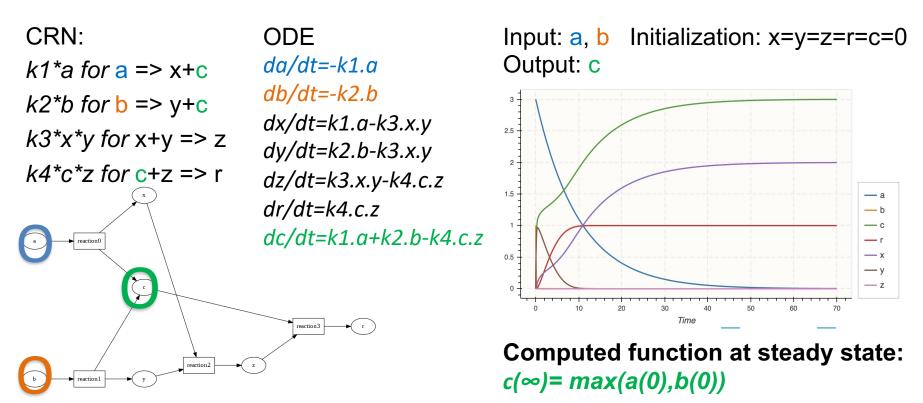
Joint work with Elisabeth Degrand and Sylvain Soliman CMSB 2020

INRIA Saclay

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Analog Computations with Chemical Reaction Networks (CRN)

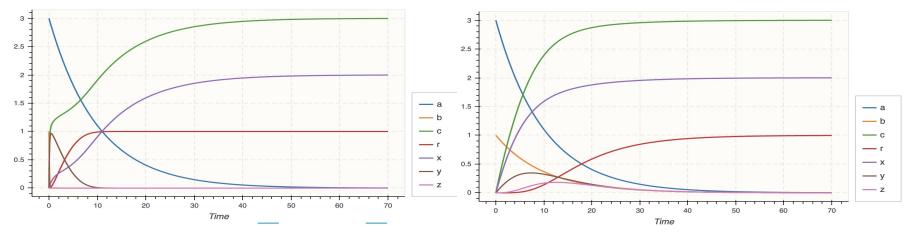


Theorem [F, Le Guludec, Bournez, Pouly CMSB 2017]

A real function is Turing-computable (in Ptime) if and only if it can be computed by a CRN over a finite set of molecular species (with polynomial length trajectories)

Rate-Independent CRN/ODE Computation

Input: a(0)=3 b(0)=1 Result c*=3 independently of the reaction rates k1=0.1, k2=10.0, k3=1, k4=100.0: k1=0.1, k2=0.1, k3=0.1, k4=0.1:



The I/O functions computed by that CRN are independent of the kinetics terms

a => x+c

b => y+c

x+y => Z

C+Z => **r**

 $c^{*} = \max(a(0), b(0)) = a(0) + b(0) - \min(a(0), b(0))$ $x^{*} = \max(0, a(0) - b(0))$ $r^{*} = \min(a(0), b(0))$ $z^{*} = 0, a^{*} = 0, b^{*} = 0$ Absolute robustness Ideal circuit designs Ideal circuit biology

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Mathematical Characterization of the **Functions Computed by Rate-Independent CRNs**

Theorem [Chen Doty Soloveichik 2014 ITCS]

A real function is computable by a rate-independent CRN if and only if it is a positive continuous piecewise linear function with rational coefficients.

Theorem [Chalk Kornerup Reeves Soloveichik 2018 CMSB]

A real function is computable by a composable (i.e. not consuming its Does not help to show that a given CRN is rate-independent inputs) rate-independent CRN iff it is a superadditive (i.e. $f(x + y) \ge 1$ f(x) + f(y) positive continuous piecewise rational linear.



Graphical conditions on the CRN

ensuring rate-independence ?

Monomolecular Rate-Independent CRN Structures

A => B

 $A \Rightarrow C$

output B: computes the *identity function* $B(\infty)=A(0)+B(0)$ rate-independent ! output A: computes the zero function $A(\infty)=0$ rate-independent !

Harmless join $B \Rightarrow C$ output C: computes the sum $C(\infty)=A(0)+B(0)+C(0)$ rate-independent output A: computes the zero function $A(\infty)=0$ rate-independent !

C => A

C => B

Harmfull fork

output A: computes $A(\infty) = \frac{\alpha}{\beta}C(0) + A(0)$ not rate-independent ! Harmless fork with a circuit

- C => A
- C => B
- $B \Rightarrow C$

output A: computes the sum $A(\infty)=C(0)+B(0)+A(0)$ rate-independent !



Bimolecular Rate-Independent CRN Structures

A+B => C

output C: computes $C(\infty)=minimum(A(0),B(0))+C(0)$ rate-independent ! output B: computes $B(\infty)=max(0,B(0)-A(0))$ rate-independent !

C => A+B

output A: makes copies $A(\infty)=C(0)+A(0)$ rate-independent !

A => X+C

B => Y+C Rate-independent on all species, why?

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X+Y => Z
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C+Z => R

Definition A funnel CRN is a CRN that is:

- fork-free on species nodes
- circuit-free
- synthesis-free

Theorem [Degrand F Soliman CMSB 2020] A funnel CRN (ODE) is rate-independent for any output species.

Sufficient condition, not a necessary condition (e.g. harmless fork with circuit)

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Global Rate-Independence Condition

Lemma The structure of a funnel CRN C is a DAG with no reaction source node **Lemma** All steady fluxes of a funnel CRN C are equal to 0.

Proof: by induction on the topological order of the graph.

Definition We shall denote x_i^* the total amount of species x_i available in an execution of the corresponding ODE system.

$$x_i^+ = x_i^0 + \int_0^{+\infty} \frac{dx_i^+}{dt} = x_i^0 + \int_0^{+\infty} \sum_{P_j(x_i) > R_j(x_i)} (P_j(x_i) - R_j(x_i)) f_j$$

Theorem [Degrand F Soliman CMSB 2020] The ODE system associated to a funnel CRN has a single steady state x^* that does not depend on the kinetic functions f_i of C.

Corollary A funnel CRN is globally rate-independent for all species.

Theorem [Degrand F Soliman CMSB 2020] Any function computable by a rate-independent CRN/ODE is computable by a funnel CRN/ODE.

Proof: by Chen-Doty-Soloveichik's characterization and Ovchinnikov's max-min representation of continuous piecewise linear functions with rational coefficients.

Rate-Independence for « Persistent » Outputs

The harmless-fork-with-circuit CRN is rate-independent on outputs A, B, C

 $C \Rightarrow A$ $C \Rightarrow B$ $B \Rightarrow C$

Def. A species *x*

- is a product of a CRN if it can only increase: $\forall i R_i(x) \le P_i(x)$
- is structurally persistent if it is covered by a P-invariant S, $\forall i S$. $R_i = S$. P_i , and does not belong to a critical (gets empty) siphon (when empty remains empty)

Theorem [Degrand F Soliman CMSB 2020] Any CRN (ODE) is rate-independent on its structurally persistent products.

Proof: P-invariant covering ensures boundedness and convergence for products. The species reaching 0 are localized in siphons and exclude persistent outputs.

Implemented in BIOCHAM using Constraint Logic Programming for computing P-invariants and siphons [Nabli, Martinez, F, Soliman 2016 Constraints]

Evaluation against BioModels repository

590 CRNs from SBML models (many not well-formed CRNs) [F Gay Soliman 2011 *TCS*]

94 with rate-independent products29 with non trivial rate-ind. products2 globally rate-ind. CRNs

Size of those 29 models:

- 4-136 species
- 2-316 reactions

Constraint solving time:

- between 0.07 and 151 seconds
- except 2 timeouts >240s

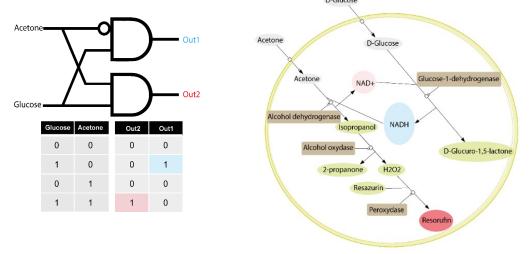
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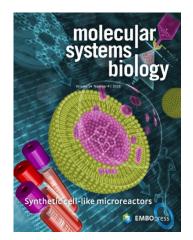




Relevance to Synthetic Biology

- Absolute robustness through rate-independence by design
 - Graphical constraints for CRN design
 - Constraint-based synthesis method
- Concrete chemical implementation with « morally » rate-independent CRNs
 - Rate-independent CRN kernel
 - Plus reverse reactions breaking formal rate-independence (limited robustness)
 - Boolean function CRNs for coma diagnosis [Courbet Amar F Renard Molina 2018 MSB]





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Time Complexity of the Functions Computed by Rate-independent CRN/ODE

Proposition. The input-output function computed by a rate-independent CRN/ODE with rational kinetics and rational inputs is computable in time $O(n \log n)$ and log space on a multitape Turing machine where *n* is the number of bits to encode the rational inputs with exact precision.

Proof. By Chen Doty Soloveichik 2014 ICTS the function computed by a rate independent CRN is a continuous positive piecewise linear function

$$y = \sum_{i=1}^{k} a_i x_i = \sum_{i=1}^{k} a_i x_i$$
 where

- The fixed k rational coefficients a_i 's can be encoded with by pairs of integers
- the x_i 's are the input rational numbers encodable by pairs of integers of n bits
- and *y* is the output rational number similarly encoded with a pair of integers

By Harvey and Van der Hoeven 2021 multiplication is in $O(n \log n)$ on a multitape Turing machine (required for the output with common denominator), addition and substraction are in O(n) as piece-wise discrimination. Hence the function computed by a rate-independent CRN/ODE is computable in $O(n \log n)$ time and log space.

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Analog Characterization of Low Complexity Class?

Rational Polynomials can also be evaluated in $O(n \log n)$ time and log space on a multitape Turing machine with binary encoding of the inputs. There is no notion of $O(n \log n)$ nor O(n) complete problems.

Characterizing L the Logspace complexity class for decision problems ?

- Turing machine with 3 tapes: R input n, RW work $\log n$, W output P(n)
- $L = DSPACE(O(\log n)) \subseteq DTIME(2^{O(\log n)}) = DTIME(n^{O(1)}) = P$
- $L \subseteq NL \subseteq P$ but it is unknown wheter L=NL, NL=P, even L=NP
- 2-SAT and ST-connectivity are NL-complete under log-space reduction
- ST-connectivity is also NL-complete under First-Order reduction
- L = languages expressible in FO + transitive closure predicate (data complexity)



Logspace reduction \leq_L

Defined on languages $L_1 \leq_L L_2$ such that $L_2 \in L$ implies $L_1 \in L$ $L_1 \leq_L L_2$ if $x \in L_1 \leftrightarrow f(x) \in L_2$ where *f* is computable in LOGSPACE.

Space complexity for computing a function:

- Turing machine with 3 tapes: R input n, RW work $\log n$, W output P(n)
- Problem f(x) may be large, so consider TM computing f(x) bitwise on the fly f ∈ SPACE(S(n)) if both languages L' and L" are in SPACE(S(n))

L'= { $(x, i) : f_i(x) = 1$ } and L" = { $(x, i) : i \le |f(x)|$ }

This gives

- Transitivity: $L_1 \leq_L L_2$ and $L_2 \leq_L L_3$ implies $L_1 \leq_L L_3$
- The composition of two L computable functions is L computable
- Yet every decision problem in L is L-complete under \leq_L (useful w.r.t. P not L)
- Stronger notion of reduction for meaningful L-completeness, e.g. \leq_{FO}

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Conclusion

- Funnel CRN/ODE structure: fork-free, circuit-free, synthesis-free
 - Theorem. Any funnel CRN/ODE is rate-independent on all outputs
 - Theorem [Soloveichik et al. 2014]. Any function computed by a rate independent CRN/ODE is a continuous piecewise linear rational function.
 - Theorem. Any function computed by a rate independent CRN/ODE can be computed by a funnel CRN/ODE
- Relevance to low analog computational complexity classes
 - Functions computed by rate-independent CRN/ODE are in $O(n \log n)$ time and logspace on a multitape Turing machine with binary representation of the inputs
 - Investigate Logspace completeness with an appropriate notion of reduction? \leq_{QFFO} ?
- Meaning for rate-independent difference equations?



Lifeware team at INRIA Saclay

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 - ANR δifference project: Computing with Discrete Differential Equations
- IBM on rule learning from temporal data, CIFRE Johnson&Johnson, CIFRE Servier
- F. Molina CNRS Sys2diag, Montpellier, J.H. Jiang, NTU, Taiwan
 - ANR-MOST BIOPSY project: Biochemical Programming Systems
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