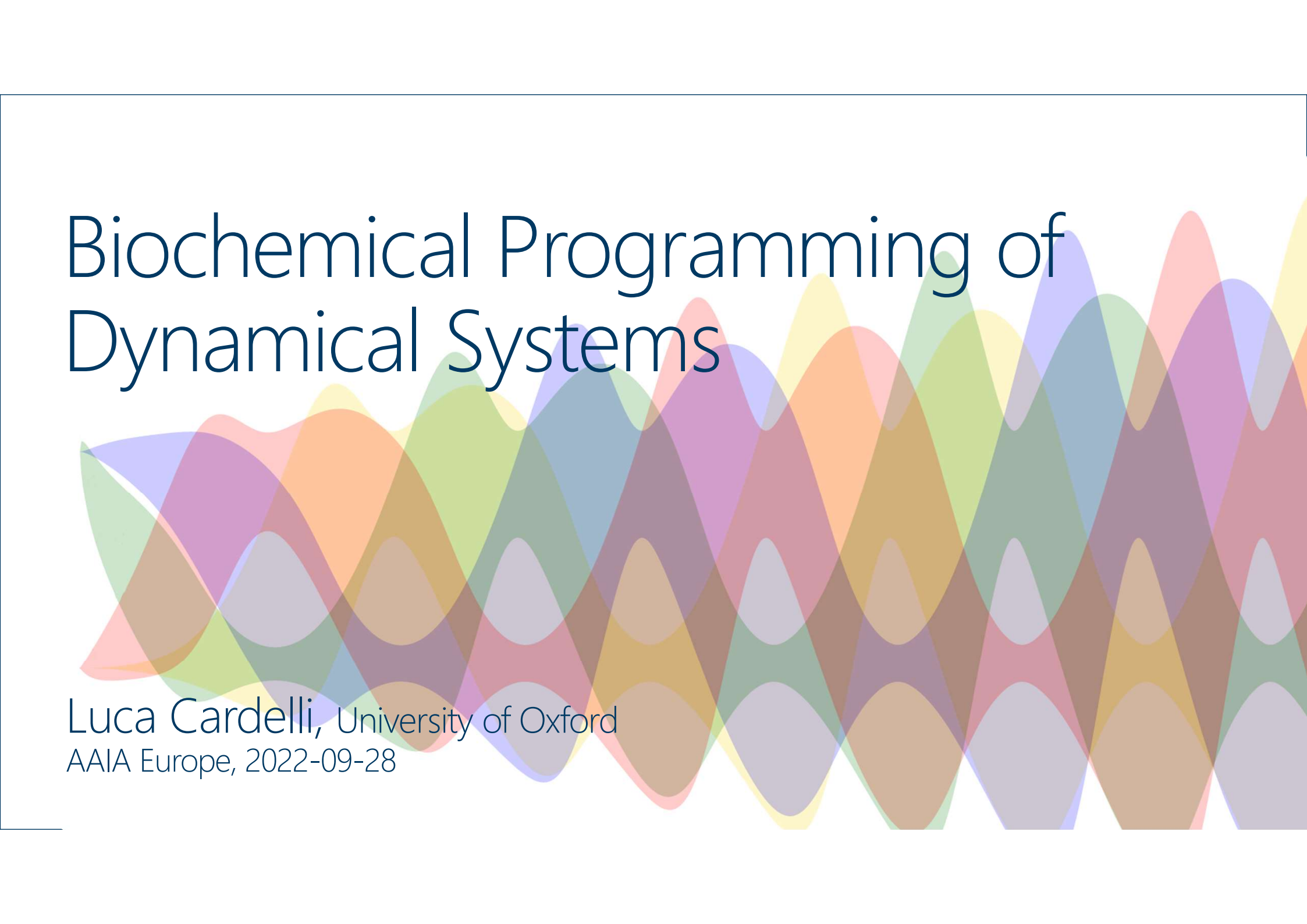
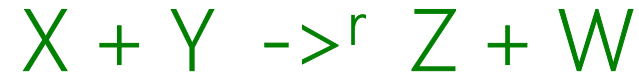


# Biochemical Programming of Dynamical Systems

The background of the slide is a complex, colorful pattern of overlapping, semi-transparent waveforms. These waves are arranged in a regular, repeating pattern across the width of the slide. The colors used include shades of purple, blue, green, yellow, orange, red, and brown. The waves vary in amplitude and phase, creating a rich, textured visual effect.

Luca Cardelli, University of Oxford  
AAIA Europe, 2022-09-28

# Chemical Reaction Networks (CRN)



- A *phenomenological model* of kinetics in the natural sciences  
By (only) observing naturally occurring reactions
- A *programming language*, *finitely* encoded in the genome  
By which living things manage the *unbounded* processing of matter and information
- A *mathematical structure*, rediscovered in many forms  
Vector Addition Systems, Petri Nets, Bounded Context-Free Languages, Population Protocols, ...
- A description of *mechanism* (“instructions” / “interactions”) rather than *behavior* (“equations” / “approximations”)  
Although the two are related in precise ways  
Enabling, e.g., the study of the evolution of *mechanism* through unchanging *behavior*

# Programming Examples

*spec*

$$Y := 2X$$

$$Y := \lfloor X/2 \rfloor$$

$$Y := X1 + X2$$

$$Y := \min(X1, X2)$$

*program*

$$X \rightarrow Y + Y$$

$$X + X \rightarrow Y$$

$$X1 \rightarrow Y$$

$$X2 \rightarrow Y$$

$$X1 + X2 \rightarrow Y$$

# Advanced Programming Examples

*spec*

$Y := \max(X1, X2)$

*program*

$X1 \rightarrow L1 + Y$

$X2 \rightarrow L2 + Y$

$L1 + L2 \rightarrow K$

$Y + K \rightarrow 0$

$\max(X1, X2) =$   
 $(X1 + X2) - \min(X1, X2)$

(but is not computed  
"sequentially")

Approximate Majority

$(X, Y) :=$

if  $X \geq Y$  then  $(X + Y, 0)$

if  $Y \geq X$  then  $(0, X + Y)$

$X + Y \rightarrow Y + B$

$Y + X \rightarrow X + B$

$B + X \rightarrow X + X$

$B + Y \rightarrow Y + Y$

# Finally, Some *Bad Bad* Programs



Violates thermodynamics.

(No biggie, assume there is a tiny reverse reaction.)



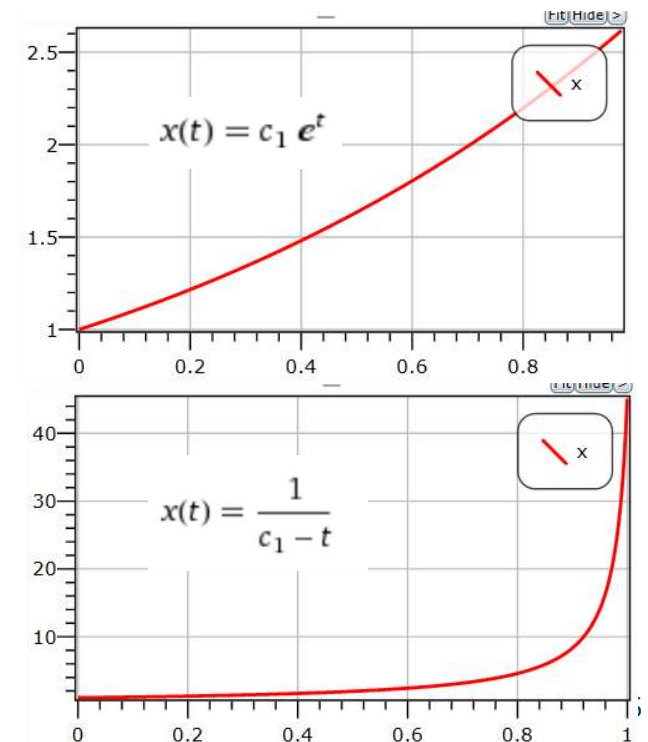
Violates conservation of mass.

(No biggie, assume there is inflow/outflow.)



Violates finite density.

(This is *really* bad.)



# Programming <sup>"approximately"</sup> *any* algorithm as a FSCRN

A FSCRN is a *finite* set of reactions over a *finite* set of species  
with a *stochastic* reaction activation rule base on the reaction rates

FSCRNs are not Turing complete

Like Petri nets: reachability is decidable

But unlike Petri nets, FSCRNs are *approximately* Turing complete

Because reactions have also *rates*

This make it possible to approximate Turing completeness by approximating test-for-zero in a register machine.  
The probability of error (in test-for-zero) can be made arbitrarily small over the entire (undecidably long) computation.

Computation with Finite Stochastic Chemical Reaction  
Networks  
David Soloveichik\* Matthew Cook† Erik Winfree‡ Jehoshua Bruck§

Adding polymerization to the model makes it fully Turing complete

Formal Molecular Biology  
Vincent Danos\* Cosimo Laneve†

# “Elementary” (NOT!) dynamical systems

A *dynamical systems* is anything characterized by a system of differential equations (ODEs).

*Elementary dynamical systems* are those that include on the r.h.s. only polynomials, trigonometry, exponentials, fractions, and their inverses.

E.g., *physics*: the equation of the simple pendulum has trigonometry on the r.h.s.:

$$\partial^2\theta = -g/l \sin(\theta)$$

E.g., *biology*: the enzyme kinetics equation has fractions on the r.h.s.:

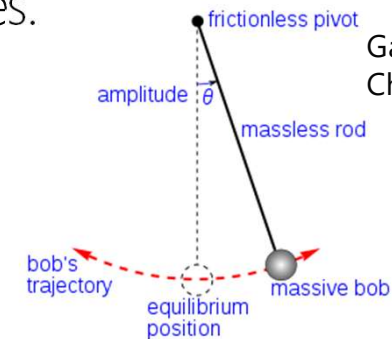
$$\partial[P] = V_{\max} [S] / (K_M + [S])$$

E.g., *metereology*: the chaotic Lorenz attractor has just 3 polynomial equations:

$$\partial x = ay - ax \quad \partial y = cx - xz - y \quad \partial z = xy - bz$$

E.g., *chemistry*: the law of mass action for CRNs implies that their ODEs are

(a restricted “Hungarian” class) of polynomials



Galileo Galilei 1602  
Christiaan Huygens 1673

<https://en.wikipedia.org/wiki/Pendulum>

**STEP 1, Polynomization:** All elementary ODEs can be exactly reduced to polynomial ODEs.

MATHEMATICAL THEORY OF THE DIFFERENTIAL  
ANALYZER  
BY CLAUDE E. SHANNON

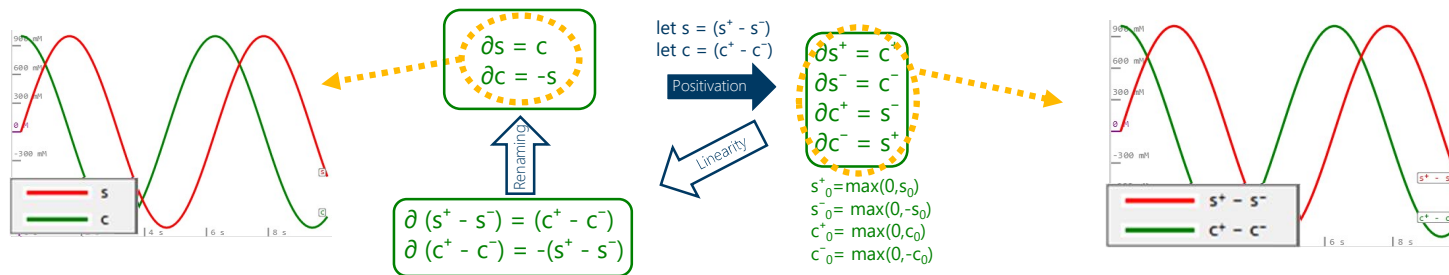
Abstraction of Elementary Hybrid Systems by Variable  
Transformation

Jiang Liu<sup>1</sup>, Naijun Zhan<sup>2</sup>, Hengjun Zhao<sup>1</sup>, and Liang Zou<sup>2</sup>

"elementary"

# Programming *any* dynamical system as a CRN

Consider *the* canonical polynomial oscillator: sine/cosine



A very simple *elementary* ODE system.

But variables go negative: we can't have that in a CRN (no negative concentrations).

**STEP 2, Positivation:** Split potentially negative variables of polynomial ODEs into the difference of two positive variables. Obtain the same trajectories as differences.

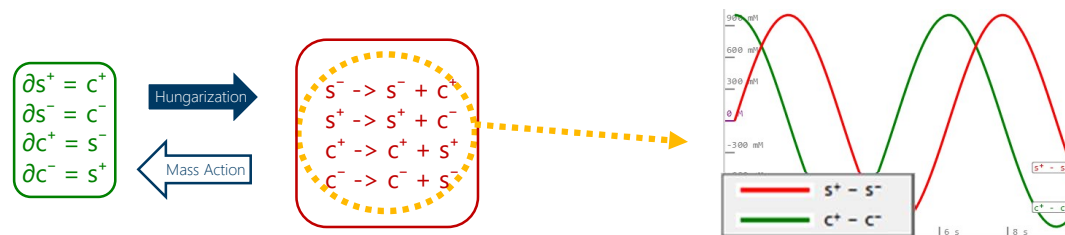
**Biomolecular implementation of linear I/O systems**

K. Oishi E. Klavins



# Programming <sup>"elementary"</sup> *any* dynamical system as a CRN

Translate positive ODEs to chemical reactions



The Law of Mass Action tells us how to produce polynomial ODEs from CRNs. The inverse process is called Hungarization, it works for *Hungarian* ODEs (polynomial ODEs where each negative monomial has the l.h.s. differentiated variable as a factor).

**STEP 3, Hungarization:** Translate polynomial ODEs to chemical reaction networks: each monomial on the r.h.s. produces one reaction.

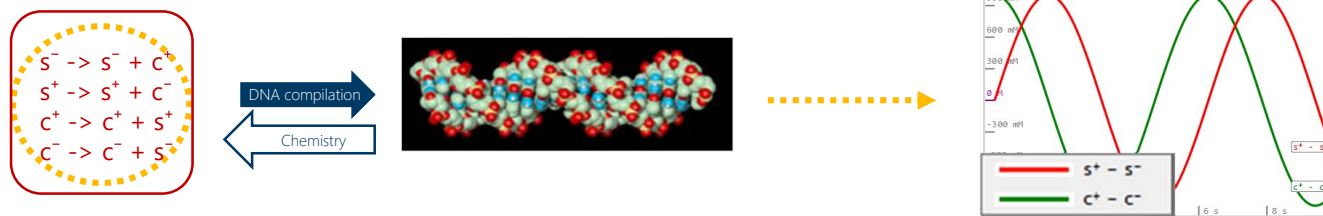
ON THE INVERSE PROBLEM OF REACTION KINETICS  
V. HÁRS - J. TÓTH

Subject to the ODEs being *Hungarian*, but that is always satisfied after positivation!

E.g. the Lorenz attractor is already polynomial but not Hungarian, it cannot be translated to mass action reactions without first doing positivation.

# Programming *any*<sup>"elementary"</sup> dynamical system as a CRN

Translate those CNRs to (real, DNA) molecules



Chemistry tells us (sometimes) what reactions molecules obey.

The inverse process is possible for DNA molecules, because we can "program" them.

**STEP 4, Molecular programming:** Translate any mass action chemical reaction network into a set of DNA molecules that obey those reactions.

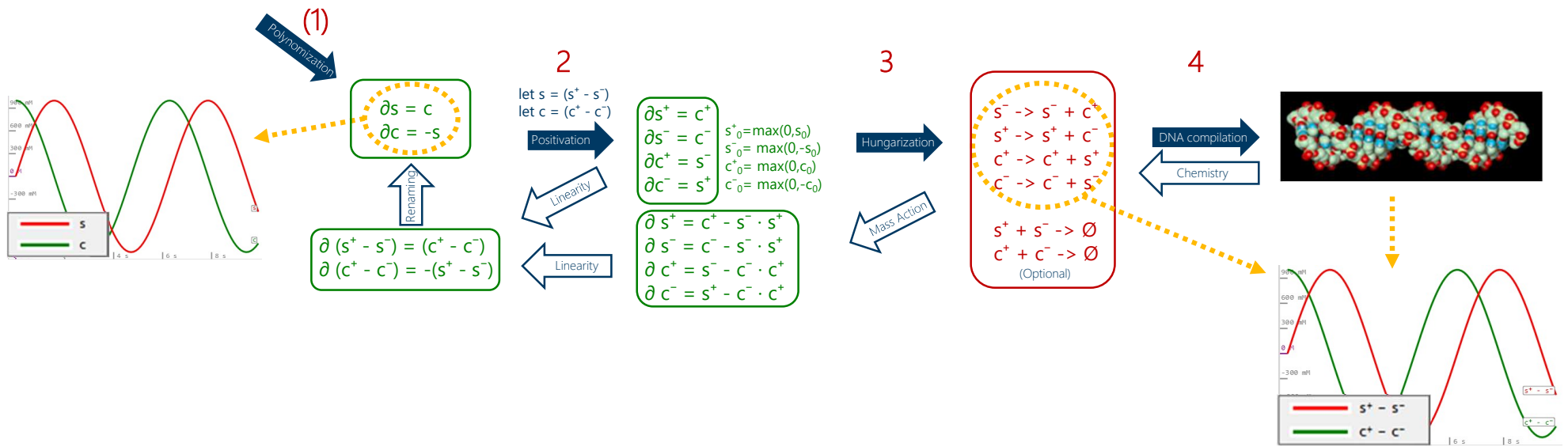
Works up to an arbitrarily good approximation of Mass Action kinetics, and up to time rescaling.

DNA as a universal substrate for chemical kinetics

David Soloveichik, Georg Seelig, and Erik Winfree  
PNAS March 23, 2010 107 (12) 5393-5398; <https://doi.org/10.1073/pnas.0909380107>

# Programming *any* <sup>"elementary"</sup> dynamical system as a CRN

Thus, CNRs are "Shannon complete", and can be physically realized

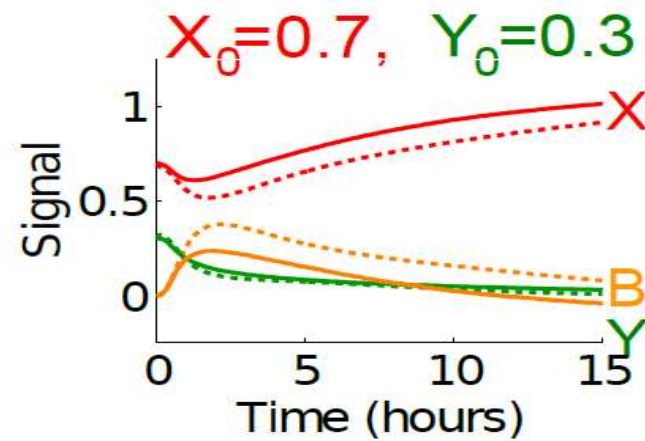
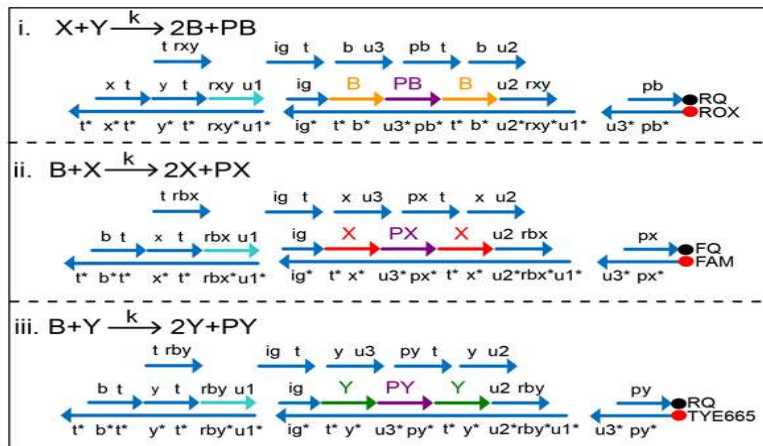


# DNA Implementation of the Approximate Majority algorithm

nature  
nanotechnology

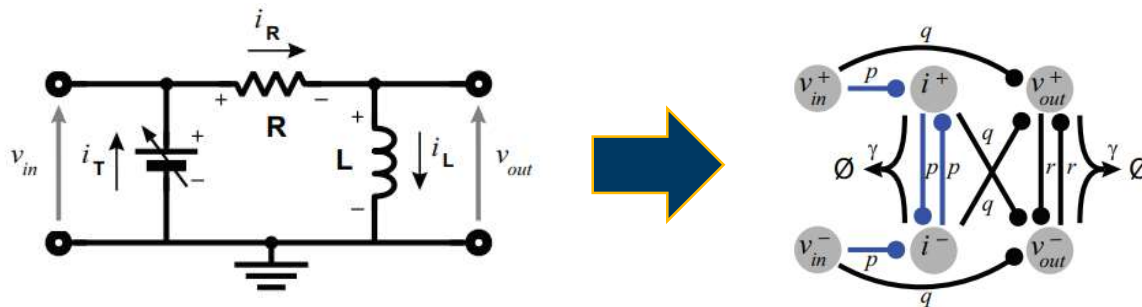
Programmable chemical controllers made from DNA

Yuan-Jyue Chen, Neil Dalchau, Niranjan Srinivas, Andrew Phillips, Luca Cardelli, David Soloveichik & Georg Seelig



# Extended to Linear DAE (differential-algebraic equations)

- An electric (RLC) circuit requires a mixture of
  - differential equations (Faraday's law of induction)
  - algebraic equations (Ohm's and Kirchhoff's laws)
- In general, of the DAE form  $E\partial_t x = Ax + Bu$  where  $E$  produces a linear combination of derivatives on the l.h.s. (not always reduceable to semi-explicit form)
- A reduction (involving an approximation) exists:



From Electric Circuits to Chemical Networks

Luca Cardelli<sup>1</sup>, Mirco Tribastone<sup>2</sup>, and Max Tschaikowski<sup>3</sup>

# Conclusions

- Chemistry is (also) a formal language that we can use to implement *~any* algorithm and *~any* dynamical system with *real* (DNA) molecules
- Turing complete and “Shannon complete”
- ANY collection of abstract chemical reactions can be implemented with specially designed DNA molecules, with accurate kinetics (up to time scaling).
- Approaching a situation where we can “systematically compile” (synthesize) a model to DNA molecules, run an (automated) protocol, and observe (sequence) the results in a closed loop.

Join  $x+y \rightarrow z$

