

Simulating Reaction Networks together with Digital Protocols

A decorative background consisting of a series of overlapping, semi-transparent, colorful shapes that resemble stylized waves or peaks. The colors include shades of purple, blue, green, yellow, orange, red, and brown. The shapes are arranged in a repeating pattern across the width of the slide.

Luca Cardelli, University of Oxford
DNA 27 Oxford, 2021-09-15

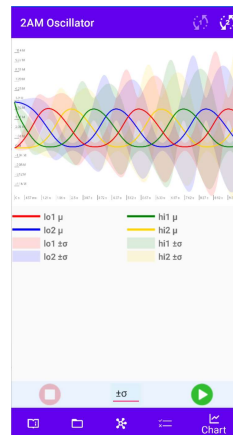
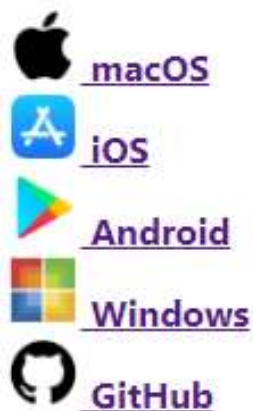
Outline

- The Scientific Method
 - Its eventual automation
- Models (that know nothing about protocols)
 - Chemical Reaction Networks
- Lab Protocols (that know nothing about models)
 - Digital Microfluidics
- Integration
 - Closed-loop modeling and protocol execution
 - The Kaemika App

Kaemika

'kimika/

An integrated language for
chemical models &
experimental protocols

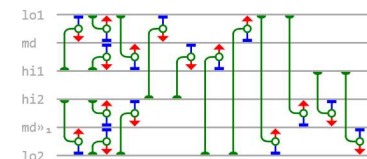


Search "Kaemika" in the app stores
<http://lucacardelli.name/kaemika.html>

Deterministic (ODE) and
stochastic (LNA) simulation

Chemical reaction networks (CRNs)
and liquid-handling protocols

Reaction scores

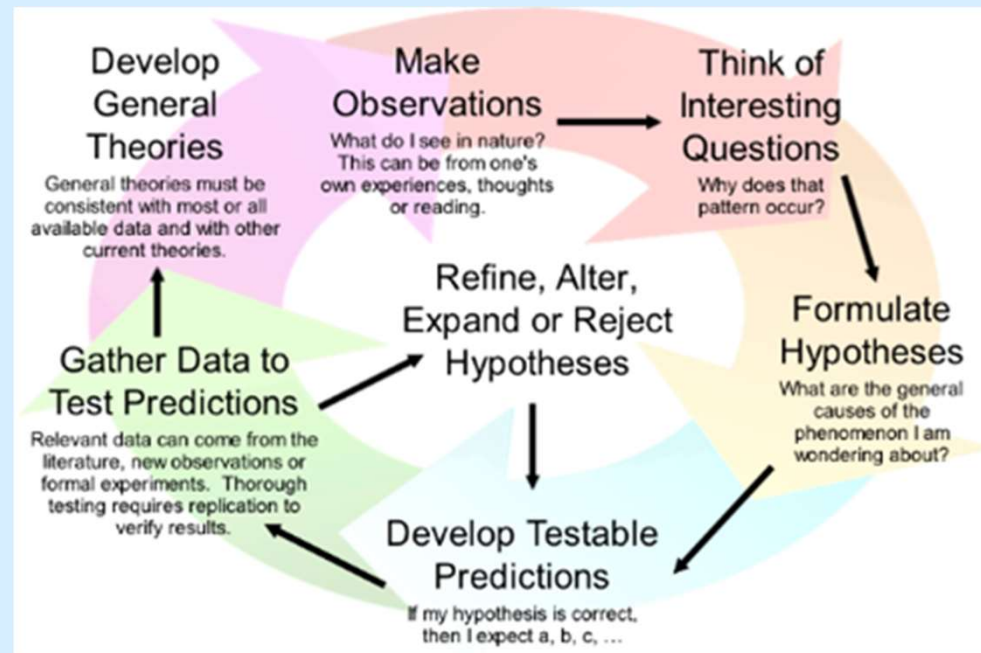


Functional scripting

GUI

The Scientific Method

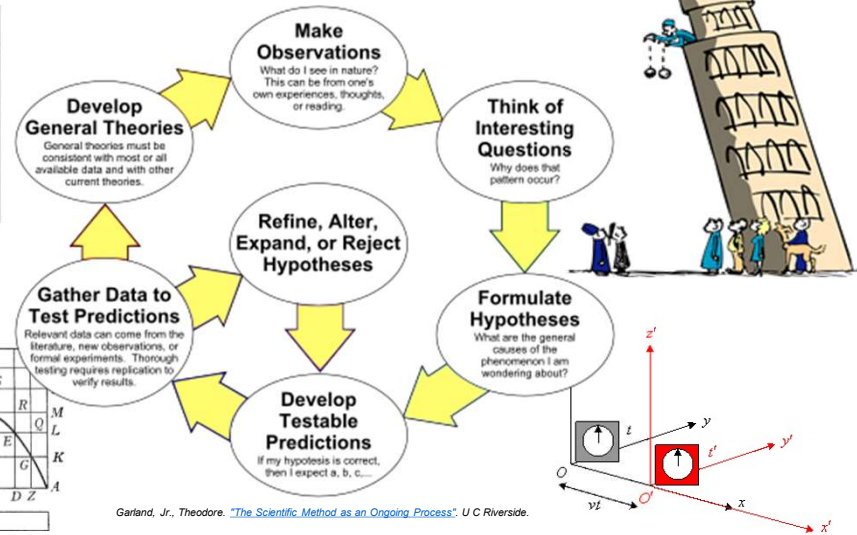
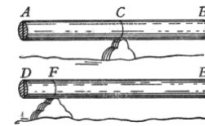
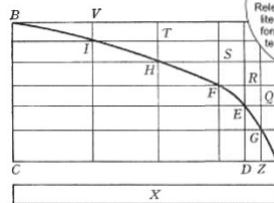
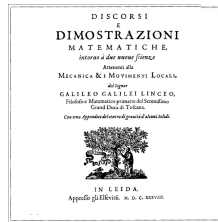
And its automation



Discovery through Observation

The Scientific Method ~ 1638

1 Guy



Garland, Jr., Theodore. "The Scientific Method as an Ongoing Process." U C Riverside.

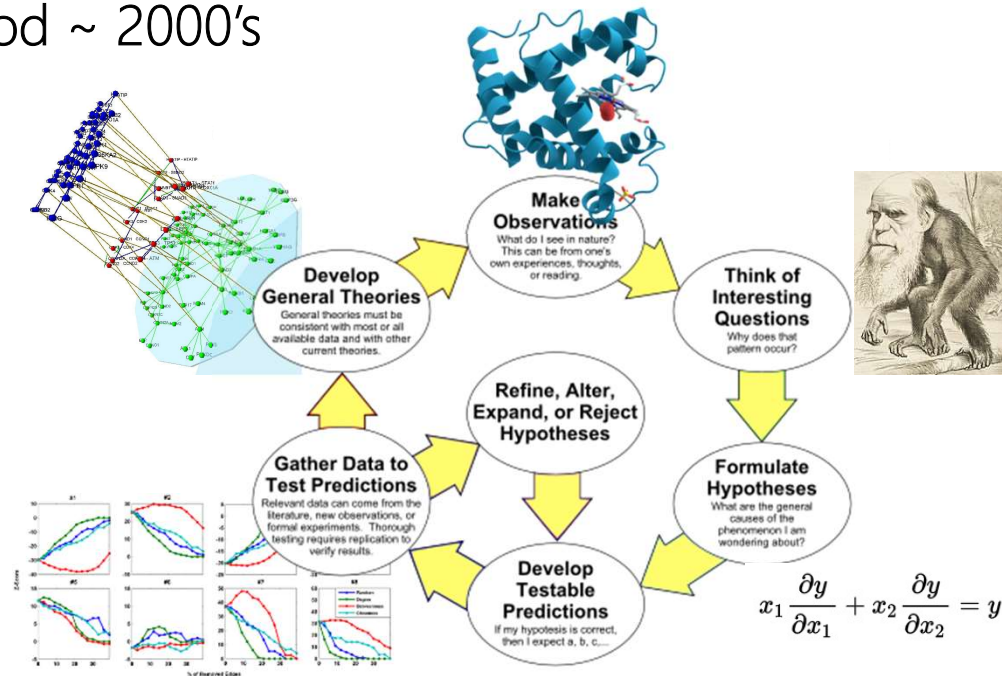
Discovery through Collaboration

The Scientific Method ~ 2000's

1 Lab



1 protein = 30 people / 30 years
 Humans have >250,000 proteins ☹

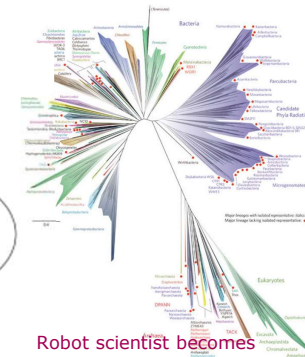
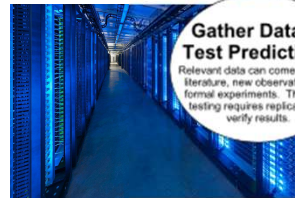
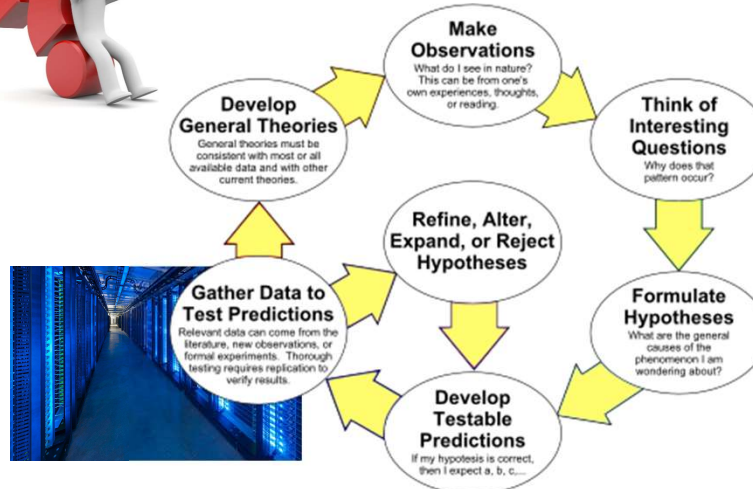
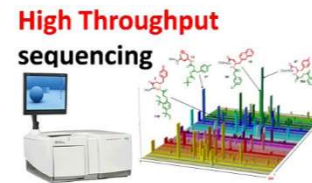


Discovery through Automation

The Scientific Method ~ 2020's

1 Program

```
while (true) {
  predict();
  falsify();
}
```



Robot scientist becomes first machine to discover new scientific knowledge

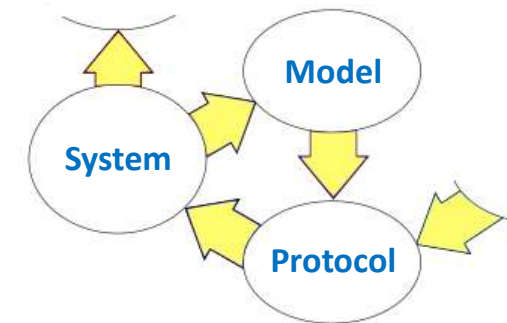


Ross King

Garland, Jr., Theodore. "The Scientific Method as an Ongoing Process." U C Riverside.

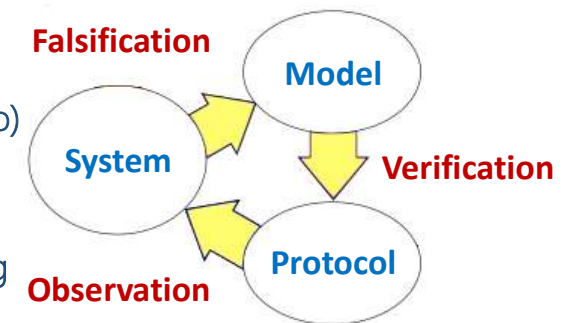
The Inner Loop

- A *model* is refined by testing a (fixed) *protocol* against a *systems*
- A *protocol* is refined by testing a (fixed) *model* against a *systems*
- Today: **publication does not accurately reflect execution**
 - Model: poorly-maintained matlab script
 - Protocol: poorly-described manual steps in the lab
 - System: poorly-characterized and hardly “resettable”
- ⇒ Crisis in biology: experiments are done once and are hard to reproduce
<http://www.nature.com/news/reproducibility-1.17552>



The Inner Loop

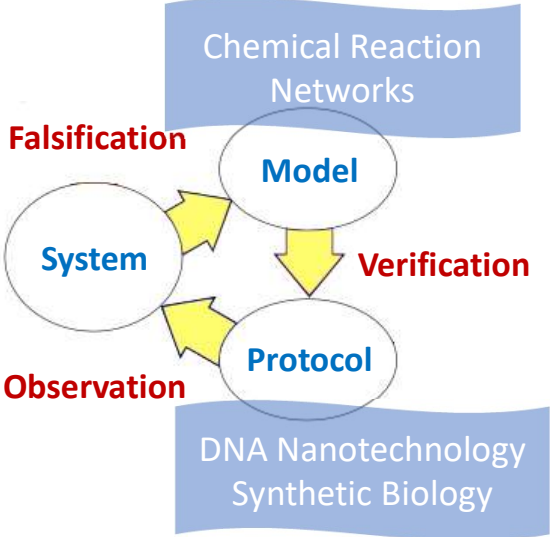
- Tomorrow, **automation**
- | | |
|-----------|--|
| Nodes | <ul style="list-style-type: none">• Model: unambiguous (mathematical) description (CompBio)• Protocol: standardized (engineered) parts and procedures (SynthBio)• System: characterized (biological) organism and foundries (SysBio) |
| Arcs | <ul style="list-style-type: none">• Verification: simulation / analysis / model checking / theorem proving• Observation: lab automation• Falsification: statistical inference / model reduction |
| Lifecycle | <ul style="list-style-type: none">• Performance evaluation/optimization: of model+protocol+system combined• Management: version control, equipment monitoring, data storage |



In this talk

The Inner Loop

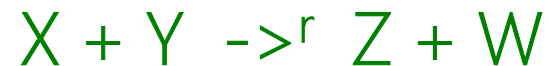
- Tomorrow, **automation** via Molecular Programming
 - Model: unambiguous (mathematical) description (CompBio)
 - Protocol: standardized (engineered) parts and procedures (SynthBio)
 - System: characterized (biological) organism and foundries (SysBio)
- Verification: simulation / analysis / model checking / theorem proving
- Observation: lab automation
- Falsification: statistical inference / model reduction
- Performance evaluation/optimization: of model+protocol+system combined
- Management: version control, equipment monitoring, data storage



Models

(those things that know nothing about protocols)

Chemical Reaction Networks (CRN)



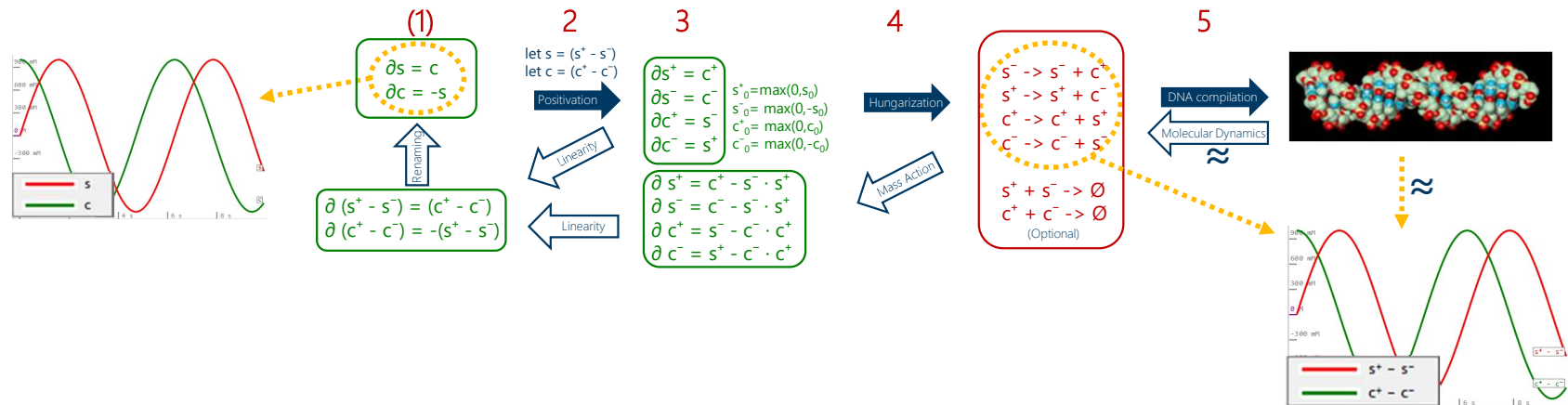
- A *phenomenological model* of kinetics in the natural sciences
By (only) observing naturally occurring reactions
- A *programming language*, finitely encoded in the genome
By which living things manage the *unbounded* processing of matter and information
- A *mathematical structure*, rediscovered in 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 *any* ^{"elementary"} dynamical system as a CRN

For example, take *the* canonical oscillator: sine/cosine

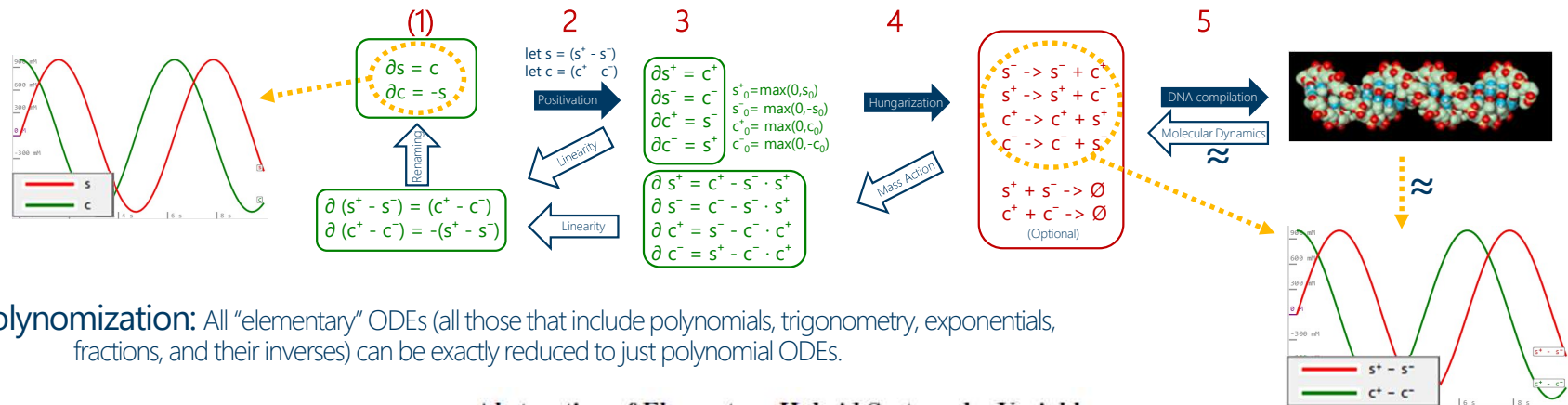
$$\partial^2 \theta = -g/l \sin \theta$$

Equation of motion of a simple pendulum



Programming ^{"elementary"} *any* dynamical system as a CRN

For example, take *the* canonical oscillator: sine/cosine



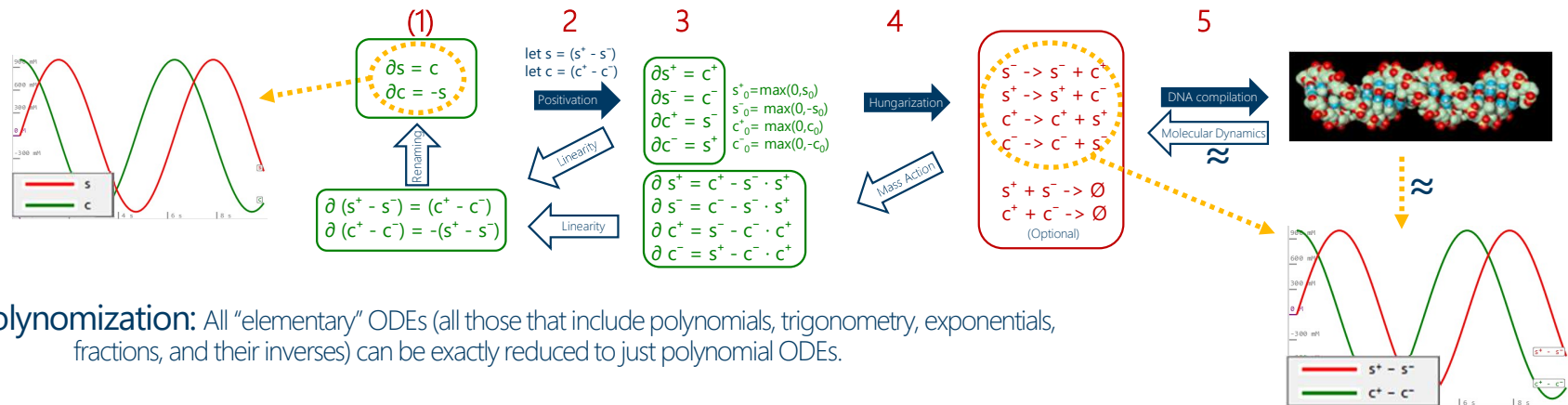
1. Polynomization: All "elementary" ODEs (all those that include polynomials, trigonometry, exponentials, fractions, and their inverses) can be exactly reduced to just polynomial ODEs.

Abstraction of Elementary Hybrid Systems by Variable Transformation

Jiang Liu¹, Naijun Zhan², Hengjun Zhao¹, and Liang Zou²

Programming ^{"elementary"} *any* dynamical system as a CRN

For example, take *the* canonical oscillator: sine/cosine



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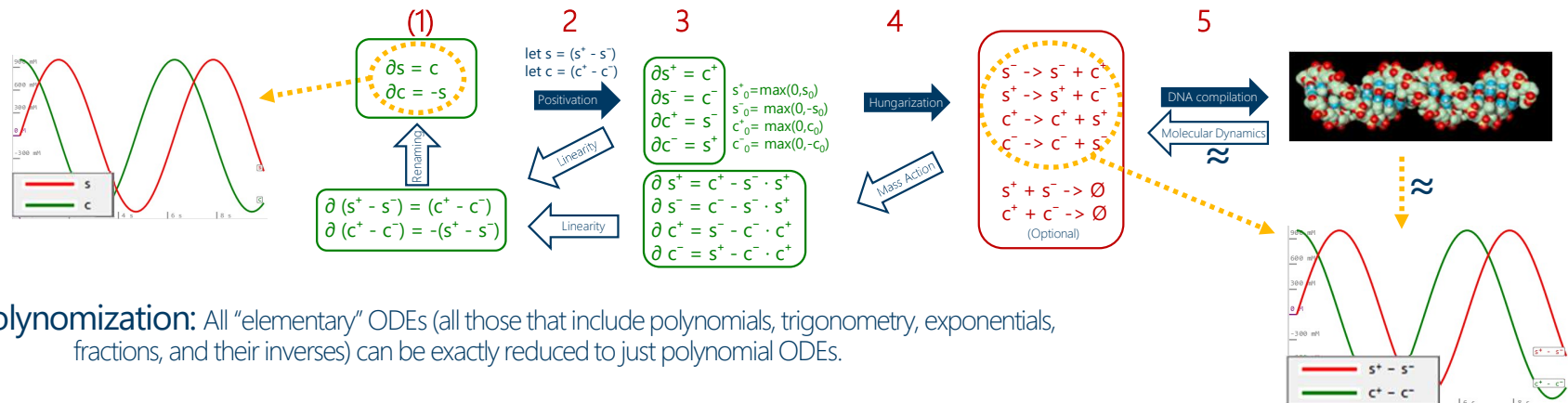
2. **Positivation:** All polynomial ODEs can be exactly reduced to polynomial ODEs in the positive quadrant (as differences).

Biomolecular implementation of linear I/O systems

K. Oishi E. Klavins

Programming ^{"elementary"} any dynamical system as a CRN

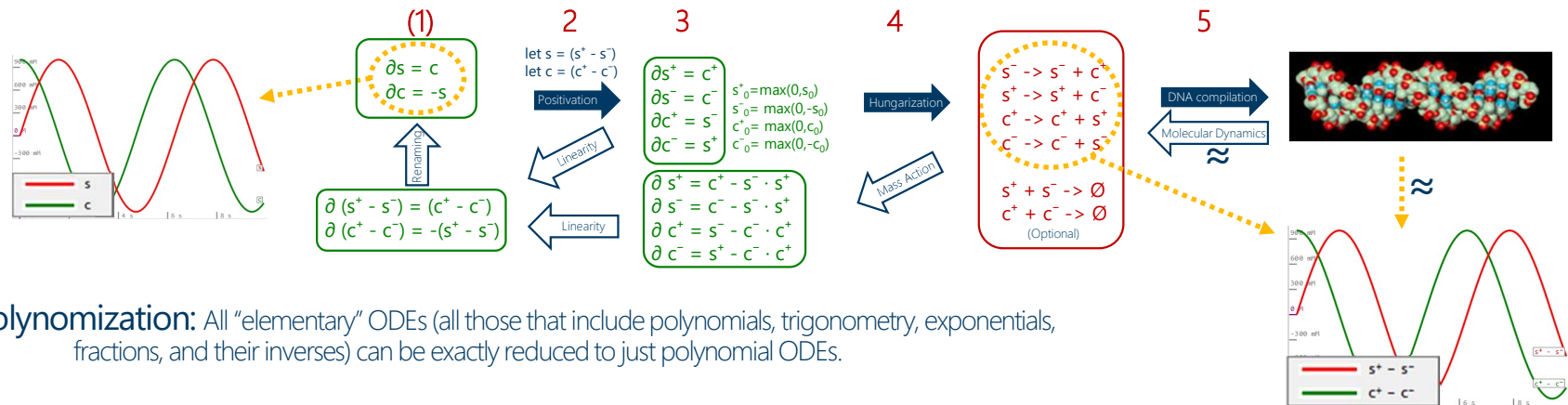
For example, take *the* canonical oscillator: sine/cosine



- 1. Polynomization:** All "elementary" ODEs (all those that include polynomials, trigonometry, exponentials, fractions, and their inverses) can be exactly reduced to just polynomial ODEs.
- 2. Positivation:** All polynomial ODEs can be exactly reduced to polynomial ODEs in the positive quadrant (as differences).
- 3. All positivized ODEs are Hungarian:** I.e., all negative monomials have their l.h.s. differential variable as a factor.

Programming *any* ^{"elementary"} dynamical system as a CRN

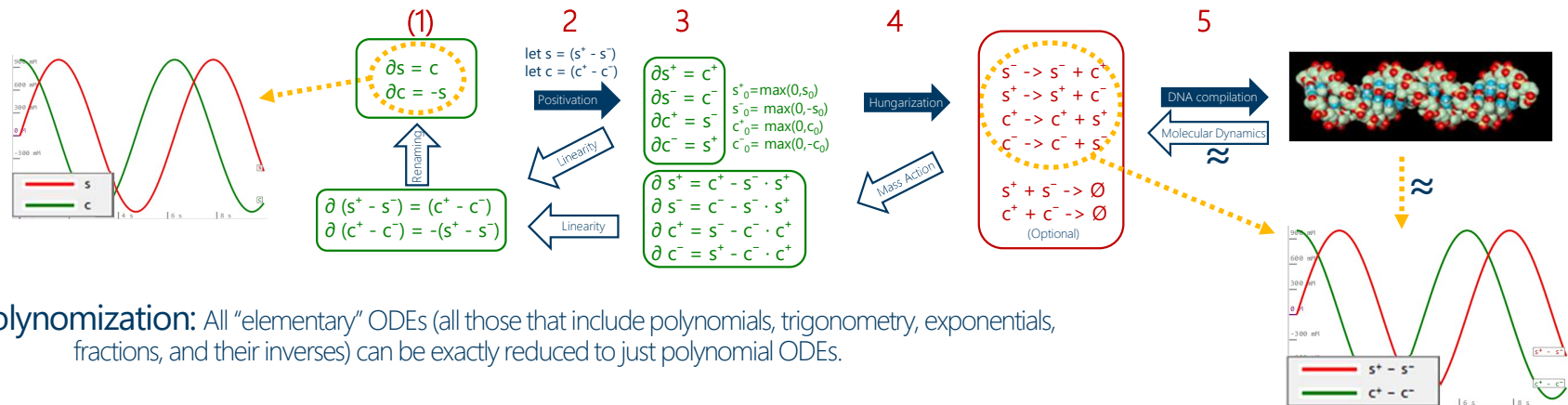
For example, take *the* canonical oscillator: sine/cosine



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- 3. All positivized ODEs are Hungarian:** I.e., all negative monomials have their l.h.s. differential variable as a factor.
- 4. Hungarization:** All Hungarian ODEs can be exactly reduced to mass action CRNs.

Programming ^{"elementary"} any dynamical system as a CRN

For example, take *the* canonical oscillator: sine/cosine



1. **Polynomization:** All "elementary" ODEs (all those that include polynomials, trigonometry, exponentials, fractions, and their inverses) can be exactly reduced to just polynomial ODEs.

2. **Positivation:** All polynomial ODEs can be exactly reduced to polynomial ODEs in the positive quadrant (as differences).

3. **All positivized ODEs are Hungarian:** I.e., all negative monomials have their l.h.s. differential variable as a factor.

4. **Hungarization:** All Hungarian ODEs can be exactly reduced to mass action CRNs.

5. **Molecular Programming:** All mass action CRNs, up to time rescaling, can be arbitrarily approximated by engineered DNA molecules.

DNA as a universal substrate for chemical kinetics

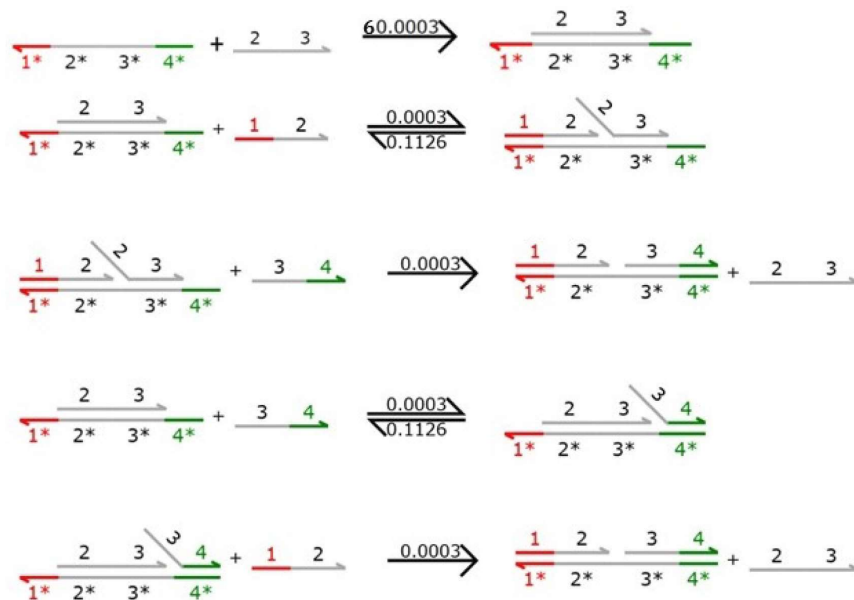
David Soloveichik, Georg Seelig, and Erik Winfree
PNAS March 23, 2010 107 (12): 5393-5398; <https://doi.org/10.1073/pnas.0909380107>

Chemistry is (also) a formal language that we can use to implement *any* dynamical system with *real* (DNA) molecules

- 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.

A Model

A Chemical Reaction Network, provided explicitly or (in this case) generated from a higher-level description of the initial strands, according to the DNA strand displacement rules



Model Semantics (deterministic)

- ODE semantics of CRNs

State produced by a CRN $\mathcal{C} = (\mathcal{A}, \mathcal{R})$ (species \mathcal{A} , reactions \mathcal{R})
with flux F (r.h.s. of its mass action ODEs) at time t ,
from initial state (x_0, V, T) (initial concentrations x_0 , volume V , temperature T):

$$\llbracket ((\mathcal{A}, \mathcal{R}, x_0), V, T) \rrbracket (H)(t) = (G(t), V, T)$$

$$\text{let } G : [0 \dots H] \rightarrow \mathbb{R}^{|\mathcal{A}|} \text{ be the solution of } G(t') = x_0 + \int_0^{t'} F(V, T)(G(s)) ds$$

Summarizing

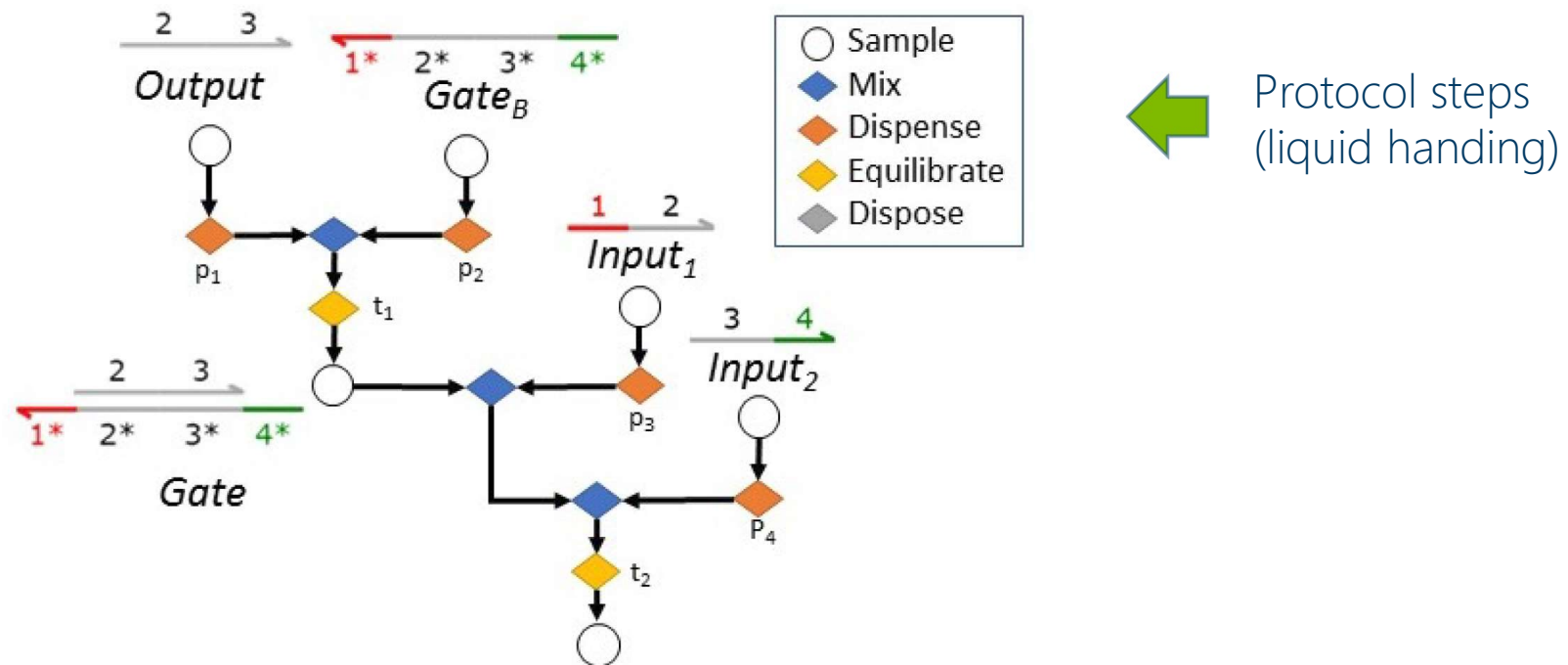
- Our models are (chemical) programs
- We can compute their behavior (their final state)
- We can (virtually) run them by integration of the ODEs
- We can (physically) run them by DNA nanotech

Protocols

(those things that know nothing about models)

A Protocol

For DNA gate assembly and activation in vitro



Digital Microfluidics

OpenDrop

<https://www.youtube.com/watch?v=ncfZWqPm7-4>



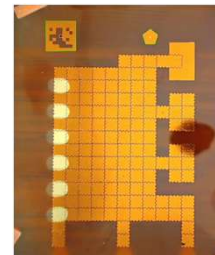
OpenDrop speed test

https://www.youtube.com/watch?v=pSls9L_h3Q0



Purple Drop (UW)

<https://misl.cs.washington.edu/projects/fluidics.html>



Digital Microfluidics

- A general, *programmable*, platform to execute the main liquid-handling operations
- To close the cycle, it can support many automated observation techniques on-board or off-board via peripheral pumps (sequencing, mass spec, ...) although these are all very hardware-dependent.

A Protocol Language

Samples: containers with volume, temperature, concentrations

$P =$

x	<i>(a sample variable)</i>
(x_0, V, T)	<i>(initial condition)</i>
$let\ x = P_1\ in\ P_2$	<i>(define local variable)</i>
$Mix(P_1, P_2)$	<i>(mix samples)</i>
$let\ x, y = Split(P_1, p)\ in\ P_2$	<i>(split samples)</i>
$Equilibrate(P, t)$	<i>(equilibrate sample for t seconds)</i>
$Dispose(P)$	<i>(discard sample)</i>

Experimental Biological Protocols with Formal Semantics

Alessandro Abate², Luca Cardelli^{1,2}, Marta Kwiatkowska², Luca Laurenti²,
and Boyan Yordanov¹

¹ Microsoft Research Cambridge

² Department of Computer Science, University of Oxford

Protocol Semantics (deterministic)

Each program denotes a *final* state <concentrations, volume, temperature>

$\llbracket P \rrbracket^\rho$ is the final state produced by a protocol P where ρ binds its free variables:

$$\llbracket x \rrbracket^\rho = \rho(x)$$

$$\llbracket x_0, V, T \rrbracket^\rho = (x_0, V, T)$$

$$\llbracket Mix(P_1, P_2) \rrbracket^\rho =$$

$$let (x_0^1, V_1, T_1) = \llbracket P_1 \rrbracket^\rho$$

$$let (x_0^2, V_2, T_2) = \llbracket P_2 \rrbracket^\rho$$

$$\left(\frac{x_0^1 V_1 + x_0^2 V_2}{V_1 + V_2}, V_1 + V_2, \frac{T_1 V_1 + T_2 V_2}{V_1 + V_2} \right)$$

$$\llbracket let x = P_1 in P_2 \rrbracket^\rho =$$

$$let (x_0, V, T) = \llbracket P_1 \rrbracket^\rho$$

$$let \rho_1 = \rho \{ x \leftarrow (x_0, V, T) \}$$

$$\llbracket P_2 \rrbracket^{\rho_1}$$

$$\llbracket let x, y = Split(P_1, p) in P_2 \rrbracket^\rho =$$

$$let (x_0, V, T) = \llbracket P_1 \rrbracket^\rho$$

$$let \rho_1 = \rho \{ x \leftarrow (x_0, V \cdot p, T), y \leftarrow (x_0, V \cdot (1 - p), T) \}$$

$$\llbracket P_2 \rrbracket^{\rho_1}$$

(*Equilibrate semantics*)

$$\llbracket Dispose(P) \rrbracket^\rho = (0^{|A|}, 0, 0),$$

(*CRN semantics*)

Kaemika Microfluidics Compiler

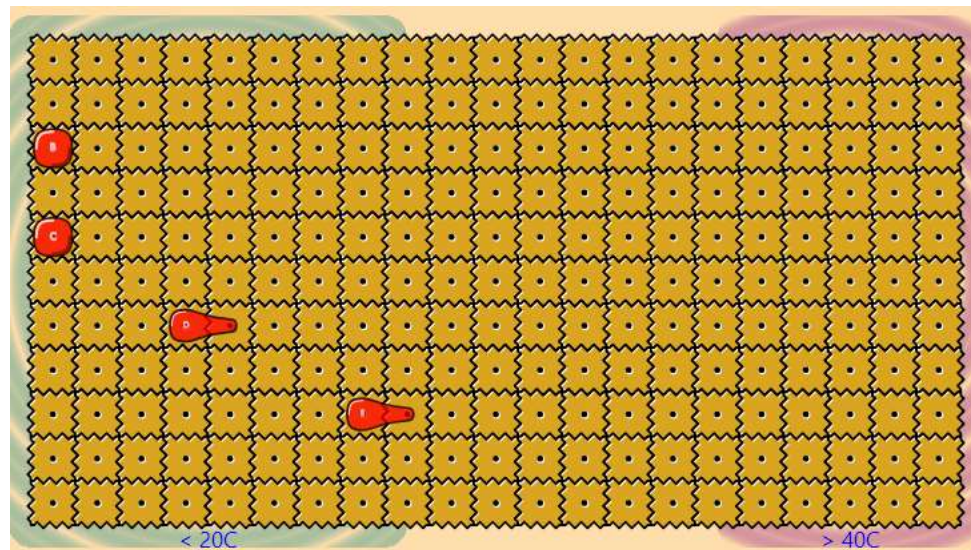
- Mix, split, equilibrate, dispose
- Automatic routing – no geometrical information
- Hot/cold zones

sample A {3 μ L, 20C}

split B,C,D,E = A

mix F = E,C,B,D

dispose F



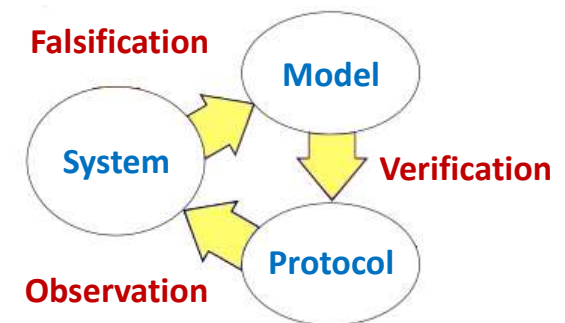
Summarizing

- Our protocols are (liquid handling) programs
- We can compute their behavior (their final state)
- We can (virtually) run them (by simulation)
- We can (physically) run them (by digital microfluidics)

Models *together with* Protocols

Automating “the whole thing”

- Protocols: sets of steps to direct lab machinery (or people)
 - Published in specialized journals. With varying accuracy.
- Models: sets of equations to predict the results of lab experiments
 - Published in Auxiliary Online Materials. With lots of typos.
- Protocols know nothing about models
 - What hypothesis is the protocol trying to test? It is not written in the protocol.
- Models know nothing about protocols
 - What lab conditions are being used to test the model? It is not written in the model.
- While presumably talking about the same system
 - Through the experiment.
- Reproducibility crisis
 - Experiments are hard to reproduce. (materials, conditions, shortcuts)
 - Even models are hard to reproduce! (typos in equations, sketchy diagrams, unexplained graphs, mysterious scripts)
- Similar to classical lifecycle problems in C.S.
 - Documentation (model) gets out of step from code (protocol) if their integration is not automated.



An Integrated Description

Samples: containers with volume, temperature, concentrations

$P =$

 x (a sample variable)

 (x_0, V, T) (initial condition)

 $let\ x = P_1\ in\ P_2$ (define local variable)

 $Mix(P_1, P_2)$ (mix samples)

 $let\ x, y = Split(P_1, p)$ (split samples)

 $Equilibrate(P, t)$ (equilibrate sample for t seconds)

 $Dispose(P)$ (discard sample)

each sample evolves (via *Equilibrate*) according to a given overall CRN:

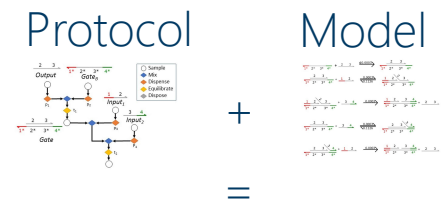
$\mathcal{C} = (\mathcal{A}, \mathcal{R})$ (species, reactions)

Experimental Biological Protocols with Formal Semantics

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Joint script

```

Input1 =< 1* 2 >  Output =< 2 3 >
Input2 =< 3 4* >  Gate = {1*}[2 3]{4*}

P1 = let In1 = ((Input1, 100.0nM), 0.1mL, 25.0°C) in
let In2 = ((Input2, 100.0nM), 0.1mL, 25.0°C) in
let GA = ((Output, 100.0nM), 0.1mL, 25.0°C) in
let GB = ((GateB, 100.0nM), 0.1mL, 25.0°C) in
let sGA, = Dispense(GA, p1) in
let sGB, = Dispense(GB, p2) in
let sIn1, = Dispense(In1, p3) in
let sIn2, = Dispense(In1, p4) in
Observe(Equilibrate(Mix(Mix(Equilibrate(
Mix(sGA, sGB), t1), sIn1), sIn2), t2), idn).
  
```

Program Semantics (deterministic)

Each program denotes a *final* state <concentrations, volume, temperature>

$\llbracket P \rrbracket^\rho$ is the final state produced by a protocol P for a fixed CRN $\mathcal{C} = (\mathcal{A}, \mathcal{R})$:

$$\llbracket x \rrbracket^\rho = \rho(x)$$

$$\llbracket x_0, V, T \rrbracket^\rho = (x_0, V, T)$$

$$\llbracket Mix(P_1, P_2) \rrbracket^\rho =$$

$$let (x_0^1, V_1, T_1) = \llbracket P_1 \rrbracket^\rho$$

$$let (x_0^2, V_2, T_2) = \llbracket P_2 \rrbracket^\rho$$

$$\left(\frac{x_0^1 V_1 + x_0^2 V_2}{V_1 + V_2}, V_1 + V_2, \frac{T_1 V_1 + T_2 V_2}{V_1 + V_2} \right)$$

$$\llbracket let x = P_1 in P_2 \rrbracket^\rho =$$

$$let (x_0, V, T) = \llbracket P_1 \rrbracket^\rho$$

$$let \rho_1 = \rho\{x \leftarrow (x_0, V, T)\}$$

$$\llbracket P_2 \rrbracket^{\rho_1}$$

$$\llbracket let x, y = Split(P_1, p) in P_2 \rrbracket^\rho =$$

$$let (x_0, V, T) = \llbracket P_1 \rrbracket^\rho$$

$$let \rho_1 = \rho\{x \leftarrow (x_0, V \cdot p, T), y \leftarrow (x_0, V \cdot (1 - p), T)\}$$

$$\llbracket P_2 \rrbracket^{\rho_1}$$

$$\llbracket Equilibrate(P, t) \rrbracket^\rho =$$

$$let (x_0, V, T) = \llbracket P \rrbracket^\rho$$

$$\llbracket ((\mathcal{A}, \mathcal{R}, x_0), V, T) \rrbracket(H)(t)$$

$$\llbracket Dispose(P) \rrbracket^\rho = (0^{|\mathcal{A}|}, 0, 0),$$

State produced by CRN $\mathcal{C} = (\mathcal{A}, \mathcal{R})$ with flux F at time t :

$$\llbracket ((\mathcal{A}, \mathcal{R}, x_0), V, T) \rrbracket(H)(t) =$$

$$let G : [0..H] \rightarrow \mathbb{R}^{|\mathcal{A}|} \text{ be the solution of } G(t') = x_0 + \int_0^{t'} F(V, T)(G(s)) ds$$

$$(G(t), V, T)$$

A Joint Semantics

This semantics gives us a *joint simulation algorithm*, connecting chemical simulation with protocol simulation.

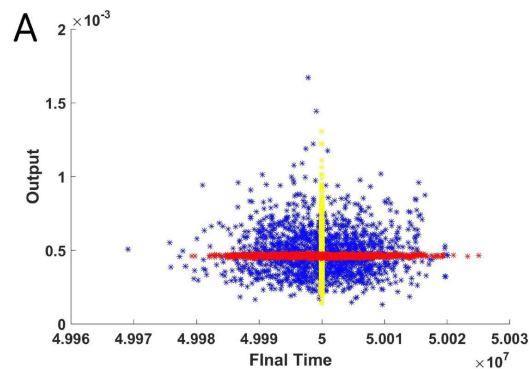
In this presentation everything is *deterministic*. The state of the protocol is passed to the chemical simulator, which computes a new state that it passes to the protocol simulator, and so on.

Kaemika uses such a joint simulation algorithm for *stochastic* simulation, passing also variance information back and forth between chemical and protocol simulation.

This requires an extension of the above semantics using the *Linear Noise Approximation* of chemical kinetics, which computes mean and variance of concentrations (both by ODEs, not e.g. by Gillespie algorithm), and a similar extension of the protocol operations.

Stochastic Analysis

- We can ask: what is the probability of a certain outcome given uncertainties in *both the protocol and the model*?
- Conversely: which parameters of *both the protocol and the model* best fit the observed result?
- Also, we can use Statistical Modelchecking:



1500 executions including protocol uncertainty due timing and pipetting errors (red).

1500 executions including only model uncertainty about rates of the CRN (yellow).

1500 executions including both sources of uncertainty (blue).

We may estimate by Statistic Model Checking, e.g. the probability that Output will fall in a certain range, given distributions over uncertain model and protocol parameters.

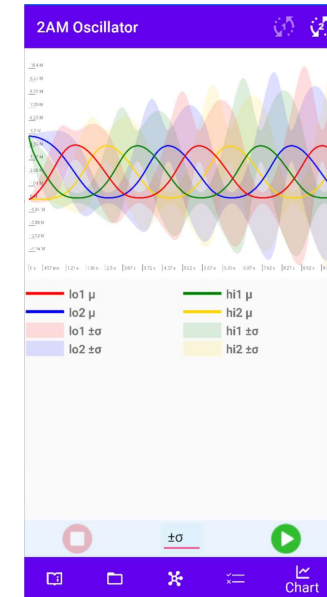
Simulating Reaction Networks together with Digital Protocols

Kaemika

- A prototype language for chemical models & protocols

- <http://lucacardelli.name/kaemika.html>

- Search "Kaemika" in the App stores



- CRN simulation
- Microfluidics simulation
- Reaction graphs
- ODE equations
- Stochastic noise (LNA)

Main features

- *Species and reactions*
 - Characterized by initial values and rates
- *"Samples" (compartments) and Protocols*
 - Isolate species and reactions in a compartment, and mix compartments
- *Kinetics (simulation)*
 - Deterministic (ODE) or stochastic (LNA) for chemical models
 - Digital microfluidics for chemical protocols
- *Programming abstractions*
 - Assemble models and protocols as compositions of modules

Species and Reactions

```
//=====
// Lotka 1920, Volterra 1926
// (simplified with all rates = 1)
//=====
```

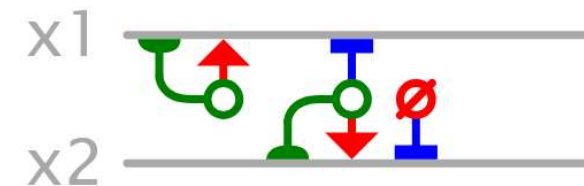
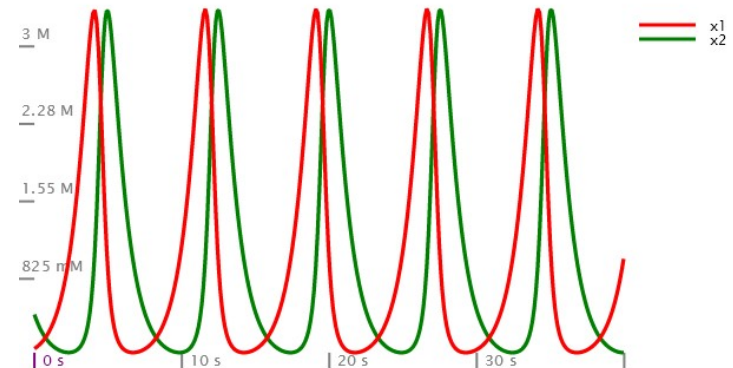
```
number x1_0 <- uniform(0,1) // random x1_0
number x2_0 <- uniform(0,1) // random x2_0
```

```
species x1 @ x1_0 M // prey
species x2 @ x2_0 M // predator
```

```
x1 -> x1 + x1 {1} // prey reproduces
x1 + x2 -> x2 + x2 {1} // predator eats prey
x2 -> ∅ {1} // predator dies
```

```
equilibrate for 40
```

UNDAMPED OSCILLATIONS, ETC. 1595
 UNDAMPED OSCILLATIONS DERIVED FROM THE LAW OF MASS ACTION.
 BY ALFRED J. LOTKA.
 Received June 2, 1920.



Stochastic (LNA) simulation

- For *all* programs (any CRN, any Protocol)

2AM Oscillator

$$\begin{aligned}\partial lo1 &= -hi1 \cdot lo1 - 0.5 \cdot hi2 \cdot lo1 + lo1 \cdot md + 0.5 \cdot lo2 \cdot md \\ \partial hi2 &= -0.5 \cdot hi1 \cdot hi2 - hi2 \cdot lo2 + hi2 \cdot md_{\gg 1} + 0.5 \cdot lo1 \cdot md_{\gg 1} \\ \partial lo2 &= 0.5 \cdot hi1 \cdot md_{\gg 1} - hi2 \cdot lo2 - 0.5 \cdot lo1 \cdot lo2 + lo2 \cdot md_{\gg 1} \\ \partial hi1 &= -hi1 \cdot lo1 - 0.5 \cdot hi1 \cdot lo2 + hi1 \cdot md + 0.5 \cdot hi2 \cdot md \\ \partial md &= 2 \cdot hi1 \cdot lo1 + 0.5 \cdot hi1 \cdot lo2 + 0.5 \cdot hi2 \cdot lo1 - hi1 \cdot md - 0.5 \cdot hi2 \cdot md - lo1 \cdot md - 0.5 \cdot lo2 \cdot md \\ \partial md_{\gg 1} &= 0.5 \cdot hi1 \cdot hi2 - 0.5 \cdot hi1 \cdot md_{\gg 1} + 2 \cdot hi2 \cdot lo2 + 0.5 \cdot lo1 \cdot lo2 - hi2 \cdot md_{\gg 1} - 0.5 \cdot lo1 \cdot md_{\gg 1} - lo2 \cdot md_{\gg 1}\end{aligned}$$

$$\begin{aligned}\partial \text{var}(lo1) &= -\text{cov}(hi1, lo1) \cdot lo1 - 0.5 \cdot \text{cov}(hi2, lo1) \cdot lo1 - \text{cov}(lo1, hi1) \cdot lo1 - 0.5 \cdot \text{cov}(lo1, hi2) \cdot lo1 + \text{cov}(lo1, md) \cdot lo1 + hi1 \cdot lo1 + 0.5 \cdot hi2 \cdot lo1 + 0.5 \cdot \text{cov}(lo1, md) \cdot lo2 + \text{cov}(md, lo1) \cdot lo1 + 0.5 \cdot \\ &\text{cov}(md, lo1) \cdot lo2 + 0.5 \cdot \text{cov}(lo1, lo2) \cdot md + 0.5 \cdot \text{cov}(lo2, lo1) \cdot md + lo1 \cdot md + 0.5 \cdot lo2 \cdot md - 2 \cdot hi1 \cdot \text{var}(lo1) - hi2 \cdot \text{var}(lo1) + 2 \cdot md \cdot \text{var}(lo1)\end{aligned}$$

$$\begin{aligned}\partial \text{cov}(lo1, hi2) &= \text{cov}(lo1, md_{\gg 1}) \cdot hi2 - 0.5 \cdot \text{cov}(lo1, hi1) \cdot hi2 - \text{cov}(hi1, hi2) \cdot lo1 - 1.5 \cdot \text{cov}(lo1, hi2) \cdot hi1 - 0.5 \cdot \text{cov}(lo1, hi2) \cdot hi2 - \text{cov}(lo1, lo2) \cdot hi2 + 0.5 \cdot \text{cov}(lo1, md_{\gg 1}) \cdot lo1 + \text{cov}(md, hi2) \cdot lo1 - \\ &\text{cov}(lo1, hi2) \cdot lo2 + 0.5 \cdot \text{cov}(md, hi2) \cdot lo2 + \text{cov}(lo1, hi2) \cdot md + 0.5 \cdot \text{cov}(lo2, hi2) \cdot md + \text{cov}(lo1, hi2) \cdot md_{\gg 1} - 0.5 \cdot lo1 \cdot \text{var}(hi2) + 0.5 \cdot md_{\gg 1} \cdot \text{var}(lo1)\end{aligned}$$

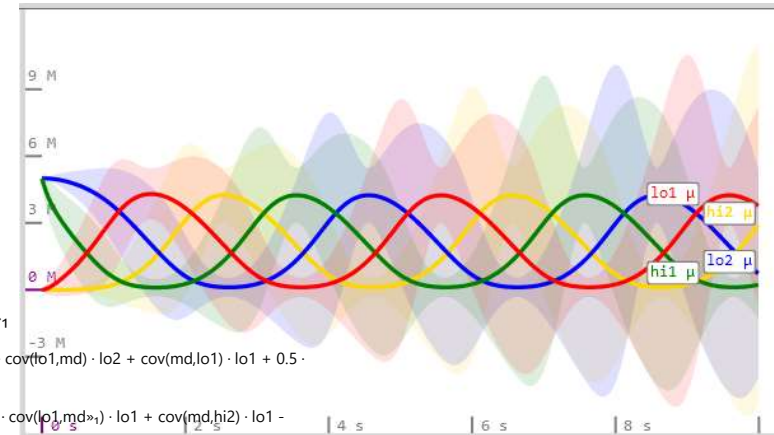
$$\begin{aligned}\partial \text{cov}(lo1, lo2) &= 0.5 \cdot \text{cov}(lo1, md_{\gg 1}) \cdot hi1 - \text{cov}(hi1, lo2) \cdot lo1 - 0.5 \cdot \text{cov}(hi2, lo2) \cdot lo1 + \text{cov}(lo1, md_{\gg 1}) \cdot lo2 + \text{cov}(md, lo2) \cdot lo1 + 0.5 \cdot \text{cov}(md, lo2) \cdot lo2 + 0.5 \cdot \text{cov}(lo1, hi1) \cdot md_{\gg 1} - 0.5 \cdot \text{cov}(lo1, lo2) \cdot \\ &lo1 - \text{cov}(lo1, hi2) \cdot lo2 - \text{cov}(lo1, lo2) \cdot hi1 - 1.5 \cdot \text{cov}(lo1, lo2) \cdot hi2 + \text{cov}(lo1, lo2) \cdot md + \text{cov}(lo1, lo2) \cdot md_{\gg 1} - 0.5 \cdot lo2 \cdot \text{var}(lo1) + 0.5 \cdot md \cdot \text{var}(lo2)\end{aligned}$$

$$\begin{aligned}\partial \text{cov}(lo1, hi1) &= \text{cov}(lo1, md) \cdot hi1 + 0.5 \cdot \text{cov}(lo1, md) \cdot hi2 - \text{cov}(lo1, hi1) \cdot lo1 + \text{cov}(md, hi1) \cdot lo1 - 0.5 \cdot \text{cov}(lo1, hi1) \cdot lo2 - 0.5 \cdot \text{cov}(lo1, lo2) \cdot hi1 - 0.5 \cdot \text{cov}(hi2, hi1) \cdot lo1 - \text{cov}(lo1, hi1) \cdot hi1 - 0.5 \cdot \\ &\text{cov}(lo1, hi1) \cdot hi2 + 0.5 \cdot \text{cov}(md, hi1) \cdot lo2 + 2 \cdot \text{cov}(lo1, hi1) \cdot md + 0.5 \cdot \text{cov}(lo1, hi2) \cdot md + 0.5 \cdot \text{cov}(lo2, hi1) \cdot md - lo1 \cdot \text{var}(hi1) - hi1 \cdot \text{var}(lo1)\end{aligned}$$

$$\begin{aligned}\partial \text{cov}(lo1, md) &= 2 \cdot \text{cov}(lo1, hi1) \cdot lo1 - \text{cov}(hi1, md) \cdot lo1 - \text{cov}(lo1, md) \cdot lo1 - hi1 \cdot lo1 - 0.5 \cdot hi2 \cdot lo1 + 0.5 \cdot \text{cov}(lo1, hi1) \cdot lo2 - 0.5 \cdot \text{cov}(lo1, md) \cdot lo2 - \text{cov}(lo1, hi1) \cdot md + 0.5 \cdot \text{cov}(lo1, lo2) \cdot hi1 - 0.5 \cdot \\ &\text{cov}(hi2, md) \cdot lo1 + 0.5 \cdot \text{cov}(lo1, hi2) \cdot lo1 - 0.5 \cdot \text{cov}(lo1, hi2) \cdot md - 0.5 \cdot \text{cov}(lo1, lo2) \cdot md - 2 \cdot \text{cov}(lo1, md) \cdot hi1 - \text{cov}(lo1, md) \cdot hi2 + \text{cov}(lo1, md) \cdot md + 0.5 \cdot \text{cov}(lo2, md) \cdot md - lo1 \cdot md - 0.5 \cdot lo2 \\ &\cdot md + 2 \cdot hi1 \cdot \text{var}(lo1) + 0.5 \cdot hi2 \cdot \text{var}(lo1) + lo1 \cdot \text{var}(md) + 0.5 \cdot lo2 \cdot \text{var}(md) - md \cdot \text{var}(lo1)\end{aligned}$$

$$\begin{aligned}\partial \text{cov}(lo1, md_{\gg 1}) &= 0.5 \cdot \text{cov}(lo1, hi1) \cdot hi2 - \text{cov}(hi1, md_{\gg 1}) \cdot lo1 - 0.5 \cdot \text{cov}(lo1, md_{\gg 1}) \cdot lo1 - \text{cov}(lo1, md_{\gg 1}) \cdot lo2 + \text{cov}(md, md_{\gg 1}) \cdot lo1 + 0.5 \cdot \text{cov}(md, md_{\gg 1}) \cdot lo2 - 0.5 \cdot \text{cov}(lo1, hi1) \cdot md_{\gg 1} + 0.5 \cdot \\ &\text{cov}(lo1, hi2) \cdot hi1 + 2 \cdot \text{cov}(lo1, lo2) \cdot hi2 - 0.5 \cdot \text{cov}(hi2, md_{\gg 1}) \cdot lo1 + 0.5 \cdot \text{cov}(lo1, lo2) \cdot lo1 + 2 \cdot \text{cov}(lo1, hi2) \cdot lo2 - \text{cov}(lo1, lo2) \cdot md_{\gg 1} + 0.5 \cdot \text{cov}(lo2, md_{\gg 1}) \cdot md - \text{cov}(lo1, hi2) \cdot md_{\gg 1} - 1.5 \cdot \\ &\text{cov}(lo1, md_{\gg 1}) \cdot hi1 - 1.5 \cdot \text{cov}(lo1, md_{\gg 1}) \cdot hi2 + \text{cov}(lo1, md_{\gg 1}) \cdot md + 0.5 \cdot lo2 \cdot \text{var}(lo1) - 0.5 \cdot md_{\gg 1} \cdot \text{var}(lo1)\end{aligned}$$

...



Writing Models Compositionally

- Embedded chemical notation

Programs freely contain both chemical reactions and control flow
Can generate unbounded-size reaction networks

- Rich data types

numbers, species, functions, networks, lists, flows (time-courses)

flows are composable functions of time used in *rates, plotting, and observation*

- Modern abstractions

Functional: programs take *data* as parameters and produce *data* as results

Monadic: programs also produce *effects (species, reactions, liquid handling)*

Nominal: *lexically scoped* chemical species (species are not "strings")

Ex: Predatorial

```

function Predatorial(number n) {
  if n = 0 then
    define species prey @ 1 M
    prey -> 2 prey // prey reproduces
    report prey
    yield prey
  else
    define species predator @ 1/n M
    species prey = Predatorial(n-1)
    prey + predator -> {n} 2 predator // predator eats
    predator -> Ø // predator dies
    report predator
    yield predator
  end
}

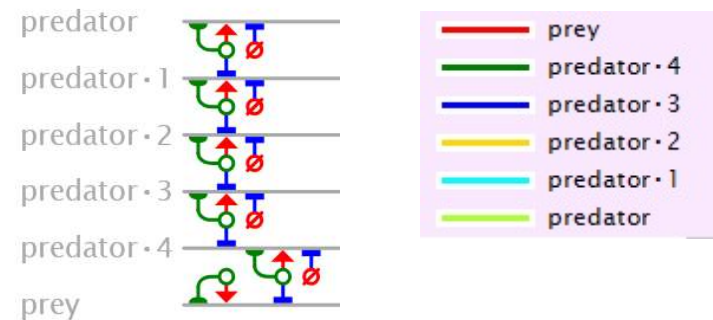
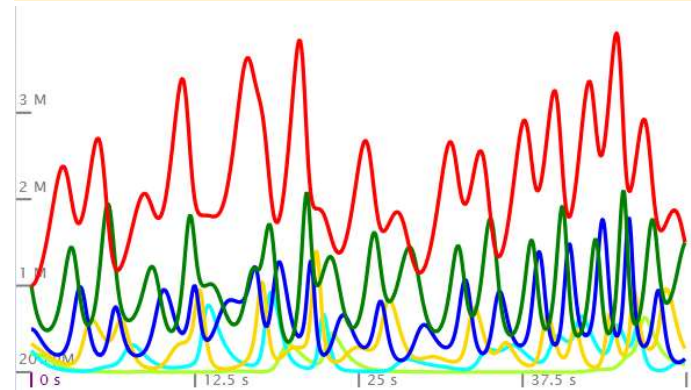
species apexPredator = Predatorial(5)
equilibrate for 50

```

```

//=====
// Creates a stack of predator-prey
// relationships in Lotka-Volterra style,
// and returns the apex predator.
//=====

```



Describing a Protocol

- *Samples* (e.g., test tubes)
 - Are characterized by a volume and a temperature
 - Contain a specified set of species
 - Evolve according to reactions that operates on those species
 - Isolate species and reactions
- *Protocol Operations* (e.g., liquid handling)
 - Accept and produce samples
 - Accepted samples are *used up* (they can only be operated-on once)

Samples

- Samples contain concentrations of species, acted over by reactions.
- Each sample has a fixed volume and a fixed temperature through its evolution.
- Sample concentrations are in units of molarity $M = \text{mol/L}$.
- The default implicit sample is called the **vessel** {1 mL, 20 C}

```
species {c} // a species for multiple samples

sample A {1μL, 20C} // volume and temperature
species a @ 10mM in A // species local to A
amount c @ 1mM in A // amount of c in A
a + c -> a + a

sample B {1μL, 20C}
species b @ 10mM in B // species local to B
amount c @ 1mM in B // amount of c in B
b + c -> c + c
```

An amount can also be given in grams (if molar mass is specified). The resulting concentration is then relative to sample volume.

```
species {NaCl#58.44}

sample C {1mL, 20C}
amount NaCl @ 8g in C
```

Reactions can be specified with Arrhenius parameters {collision frequency, activation energy}. The reaction kinetics is then relative to sample temperature T.

```
a + c ->{2, 5} a + a
// rate is  $2 \cdot e^{(-5/(R \cdot T))}$ 
```

<= Demo: MixAndSplit

Ex: Serial Dilution (recursive protocol)

```
network SerialDilution(number count, sample s, network f) {
  if count > 0 then
    sample solvent {9*observe(volume,s) L, observe(kelvin,s) K}
    mix s = s, solvent
    split s, dilution = s by 0.1, 0.9
    f(dilution)
    SerialDilution(count-1, s, f)
  end
}

//initial sample to be diluted:

sample init {1mL, 25C}
species A @ 1M in init
species B @ 1M in init
A + B ->{20} A
A -> Ø

//apply this network to each dilution;
//note that this invokes a simulation
//each time in each solution

network test(sample s) {
  equilibrate s for 10
  dispose s
}

//dilute 4 times

SerialDilution(4, init, test)
```

Prepare a series of increasingly diluted solutions and apply a network *f* to each (*f* can add species and reactions to the solutions)

RESULT:

```
sample init {1mL, 298.2K} {A = 1M, B = 1M}
sample s2 {1mL, 298.2K} {A = 100mM, B = 100mM}
sample s4 {1mL, 298.2K} {A = 10mM, B = 10mM}
sample s7 {1mL, 298.2K} {A = 1mM, B = 1mM}
sample s10 {1mL, 298.2K} {A = 100uM, B = 100uM}
```

Digital Microfluidics Compiler

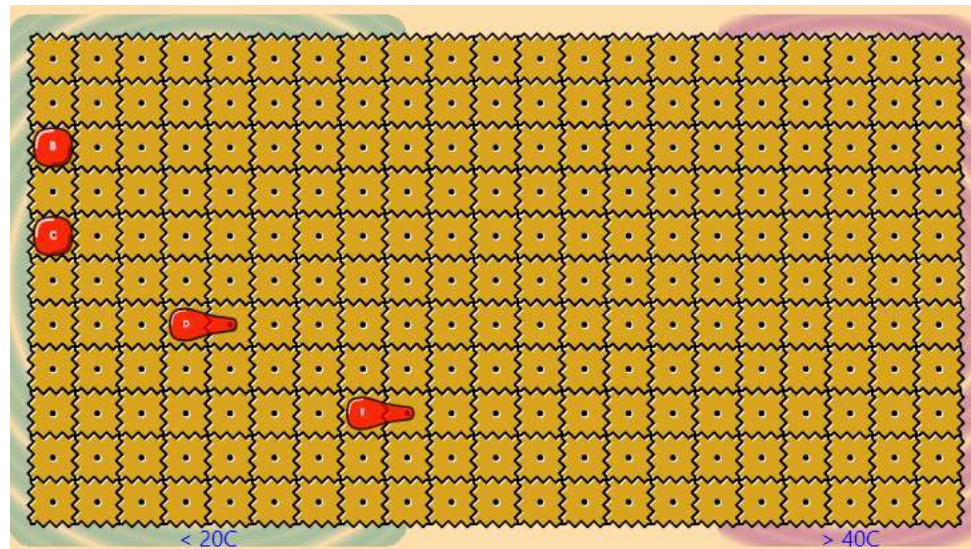
- Mix, split, equilibrate, dispose
- Automatic routing – no geometrical information
- Hot/cold zones

sample A {3 μ L, 20C}

split B,C,D,E = A

mix F = E,C,B,D

dispose F



<= Demo: MixAndSplit

Extracting the Model and the Protocol

From the script

```

species {c}

sample A
species a @ 1M in A
amount c @ 0.1M in A
a + c -> a + a
equilibrate A1 = A for 1

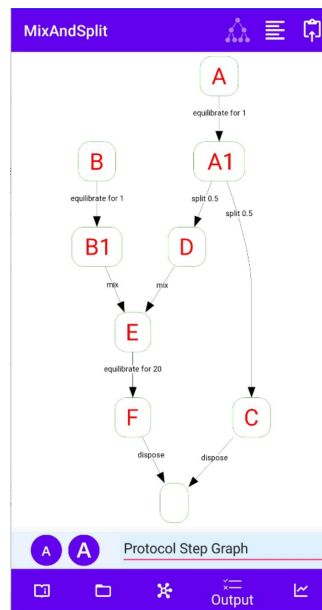
sample B
species b @ 1M in B
amount c @ 0.1M in B
b + c -> c + c
equilibrate B1 = B for 1

split C,D = A1 by 0.5
dispose C

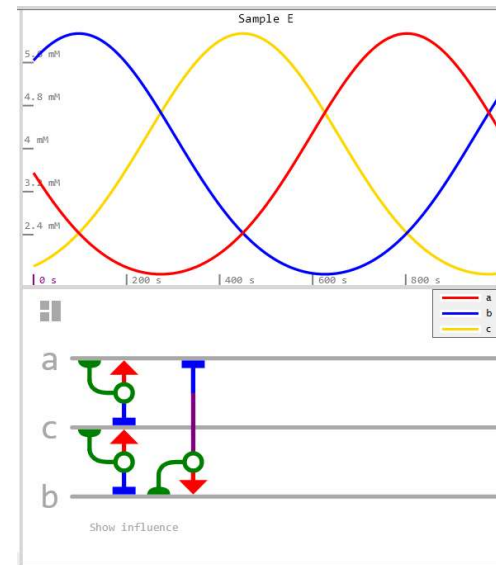
mix E = D with B1
a + b -> b + b

equilibrate F = E for 20
dispose F
    
```

The protocol



The (final) model (sample E)



```

STATE_5
sample E {1.5mL, 293.2K} {
  a = 354.5mM
  c = 178mM
  b = 0.5674M
  consumed
  a + c -> a + a
  b + c -> c + c
  a + b -> b + b
}
    
```

```

KINETICS for STATE_5 (sample E) for 20 time units:
∂a = a * c - a * b
∂c = c * b - a * c
∂b = a * b - c * b
    
```


Extracting the Hybrid Transition System

From the script

The full story (Hybrid system)

species {c}

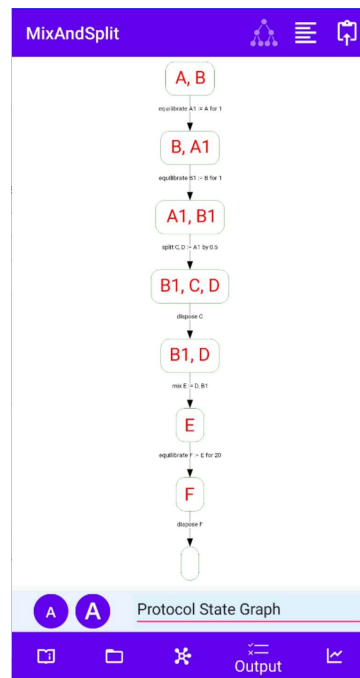
sample A
 species a @ 1M in A
 amount c @ 0.1M in A
 $a + c \rightarrow a + a$
 equilibrate A1 = A for 1

sample B
 species b @ 1M in B
 amount c @ 0.1M in B
 $b + c \rightarrow c + c$
 equilibrate B1 = B for 1

split C,D = A1 by 0.5
 dispose C

mix E = D with B1
 $a + b \rightarrow b + b$

equilibrate F = E for 20
 dispose F



```

MixAndSplit

STATE_0
sample A {1mL, 293.2K} {
  a = 1M
  c = 100mM
  consumed
  a + c -> a + a
},
sample B {1mL, 293.2K} {
  b = 1M
  c = 100mM
  consumed
  b + c -> c + c
}

KINETICS for STATE_0 (sample A) for 1 time units:
d_a = a * c
d_c = - a * c

TRANSITION
[STATE_0 (equilibrate A1 := A for 1)=> STATE_1]

STATE_1
sample B {1mL, 293.2K} {
  b = 1M
  c = 100mM
  consumed
  b + c -> c + c
},
sample A1 {1mL, 293.2K} {
  a = 1.064M
  c = 36.38mM
  consumed
  a + c -> a + a
}

KINETICS for STATE_1 (sample B) for 1 time units:
d_b = - b * c
d_c = b * c
    
```

```

MixAndSplit

TRANSITION
[STATE_1 (equilibrate B1 := B for 1)=> STATE_2]

STATE_2
sample A1 {1mL, 293.2K} {
  a = 1.064M
  c = 36.38mM
  consumed
  a + c -> a + a
},
sample B1 {1mL, 293.2K} {
  b = 0.8512M
  c = 248.8mM
  consumed
  b + c -> c + c
}

TRANSITION
[STATE_2 (split C, D := A1 by 0.5)=> STATE_3]

STATE_3
sample B1 {1mL, 293.2K} {
  b = 0.8512M
  c = 248.8mM
  consumed
  b + c -> c + c
},
sample C {500uL, 293.2K} {
  a = 1.064M
  c = 36.38mM
  consumed
  a + c -> a + a
}
    
```

```

MixAndSplit

TRANSITION
[STATE_3 (dispose C)=> STATE_4]

STATE_4
sample B1 {1mL, 293.2K} {
  b = 0.8512M
  c = 248.8mM
  consumed
  b + c -> c + c
},
sample D {500uL, 293.2K} {
  a = 1.064M
  c = 36.38mM
  consumed
  a + c -> a + a
}

TRANSITION
[STATE_4 (mix E := D, B1)=> STATE_5]

STATE_5
sample E {1.5mL, 293.2K} {
  a = 354.0mM
  c = 179mM
  b = 0.5574M
  consumed
  a + c -> a + a
  b + c -> c + c
  a + b -> b + b
}

KINETICS for STATE_5 (sample E) for 20 time units:
d_a = a * c - a * b
d_c = c * b - a + c
d_b = a * b - c + b

TRANSITION
[STATE_5 (equilibrate F := E for 20)=> STATE_6]

STATE_6
    
```

```

STATE_6
sample F {1.5mL, 293.2K} {
  a = 0.5267M
  c = 167.6mM
  b = 485.7mM
  consumed
  a + c -> a + a
  b + c -> c + c
  a + b -> b + b
}

TRANSITION
[STATE_6 (dispose F)=> STATE_7]

STATE_7
    
```

Kaemika: Extra features

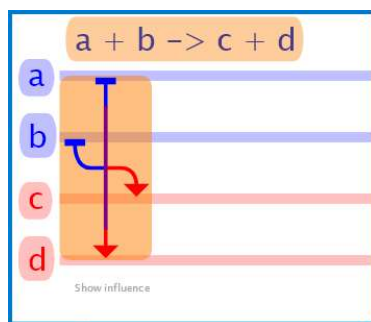
Extra features

- General kinetic rates
 - Fractions, rational powers, exponentials, trigonometry. E.g., $x \rightarrow y \{ \{ 1/x \} \}$
 - Work with both deterministic and stochastic simulation and equation-extraction
 - Even triggers (discontinuous waveforms)
- Direct ODE notation
 - Instead of a reaction, just write an ODE like $\partial x = s \cdot y - s \cdot x$
 - This is translated to the reaction $\emptyset \rightarrow x \{ \{ s \cdot y - s \cdot x \} \}$ using general kinetic rates
- Timeflows (trajectories as first-class values)
 - Programmable plot reports (e.g., $\text{var}(2 \cdot a - 3 \cdot b)$)
 - Capture timeflow outputs to combine (e.g., avg) and re-plot/export them later
- Mass action compiler
 - Turn *any* elementary ODE system (with fractions, rational powers, exponentials, trigonometry) into an equivalent system of pure mass action reactions.
- Programmable random numbers and distributions
 - As in MIT's Omega probabilistic language, with rejection sampling.
- Export
 - SBML, ODE, Bitmap, SVG, GraphViz

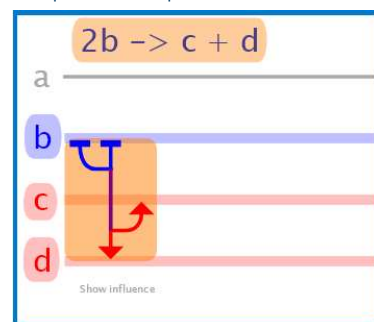
Reaction scores (graphical representation of reaction networks)

Horizontal lines: *species*. Vertical stripes: *reactions*. Blue: reagents. Red: products. Green: catalysts.

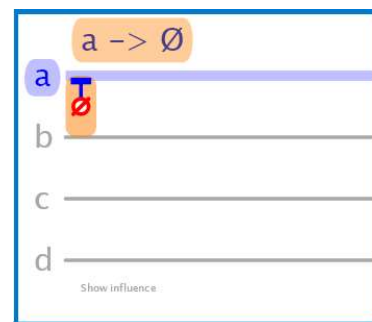
Reactants and products



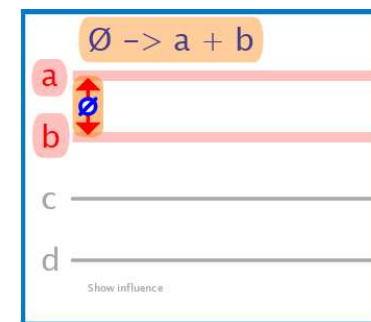
Repeated species



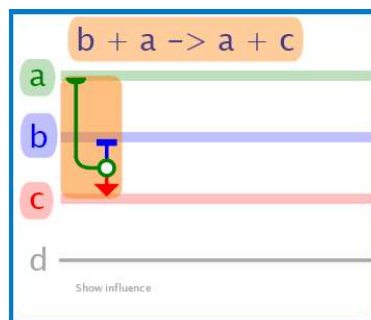
Reactants but no products



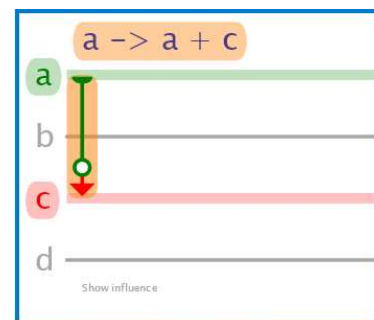
Products but no reactants



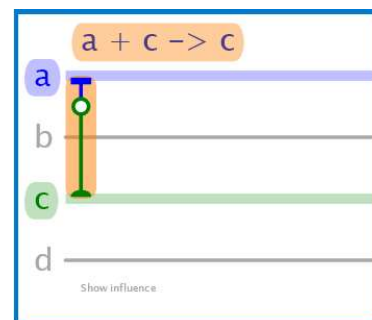
Catalyst



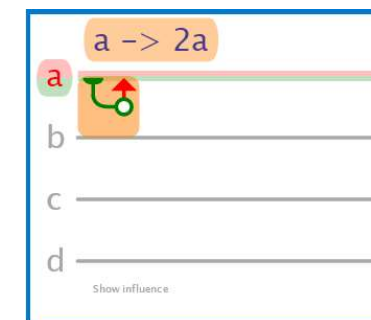
Catalyst but no reactants



Catalyst but no products



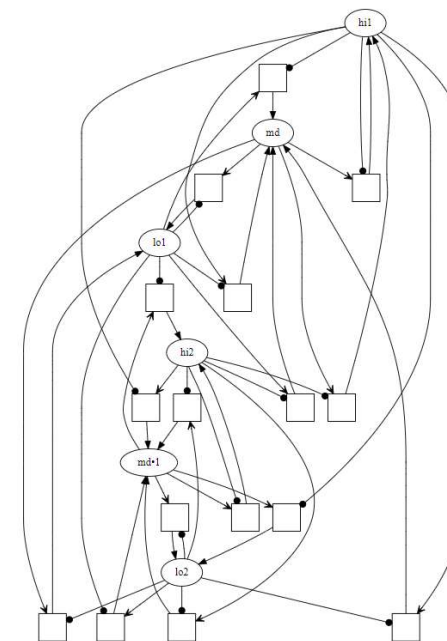
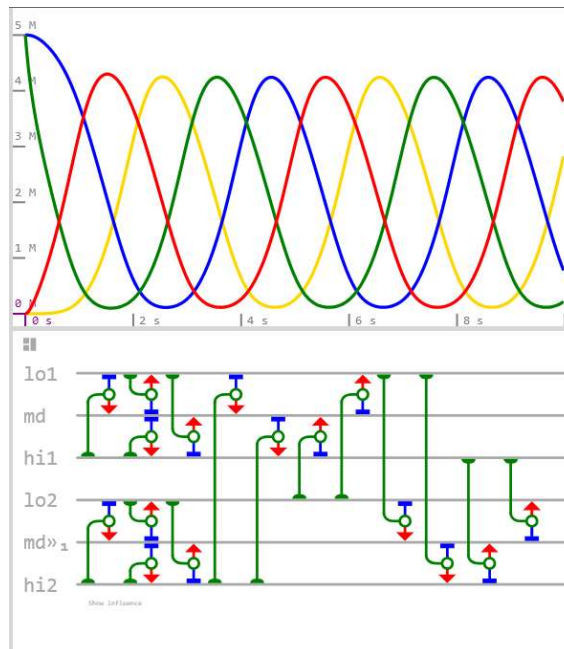
Autocatalyst



Reaction Scores vs. Reaction Graphs

- 2AM Oscillator

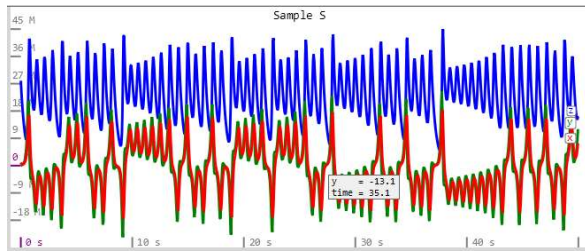
```
hi1 + md -> 2hi1
lo1 + hi1 -> lo1 + md
lo1 + md -> 2lo1
hi2 + lo1 -> hi2 + md {0.5}
hi2 + md -> hi2 + hi1 {0.5}
lo2 + hi1 -> lo2 + md {0.5}
lo2 + md -> lo2 + lo1 {0.5}
hi2 + lo2 -> hi2 + md»1
hi2 + md»1 -> 2hi2
lo2 + hi2 -> lo2 + md»1
lo2 + md»1 -> 2lo2
lo1 + lo2 -> lo1 + md»1 {0.5}
lo1 + md»1 -> lo1 + hi2 {0.5}
hi1 + hi2 -> hi1 + md»1 {0.5}
hi1 + md»1 -> hi1 + lo2 {0.5}
```



GraphViz

Mass Action Compiler

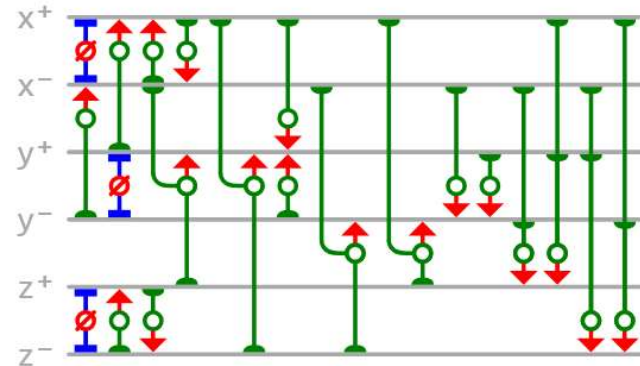
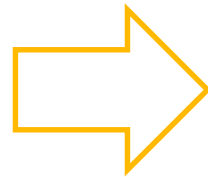
- Lorenz chaotic attractor



not mass action

$$\begin{aligned} \partial x &= s \cdot y - s \cdot x \\ \partial y &= r \cdot x - x \cdot z - y \\ \partial z &= x \cdot y - b \cdot z \end{aligned}$$

$$\begin{aligned} s &= 10 \\ b &= 8/3 \\ r &= 28 \\ x_0 &= 1 \\ y_0 &= 0 \\ z_0 &= 28 \end{aligned}$$



```

x+ + x- -> Ø
y+ -> y+ + x+ {10}
x- -> x- + x+ {10}
y- -> y- + x- {10}
x+ -> x+ + x- {10}
y+ + y- -> Ø
z+ + x- -> z+ + x- + y+
z- + x+ -> z- + x+ + y-
x+ -> x+ + y+ {28}
y- -> y- + y+
z+ + x- -> z+ + x- + y-
z- + x+ -> z- + x+ + y-
x- -> x- + y- {28}
y+ -> y+ + y-
z+ + z- -> Ø
y+ + x- -> y+ + x- + z+
y- + x+ -> y- + x+ + z-
z+ -> z+ + z+ {2.667}
y+ + x- -> y+ + x- + z-
y- + x+ -> y- + x+ + z+
z- -> z- + z- {2.667}
    
```

```

Initial:
x+ = 1
x- = 0
y+ = 0
y- = 0
z+ = 28
z- = 0
    
```

<= Demo: LorenzAttractor

Global Sensitivity Analysis (of a Lotka-Volterra system)

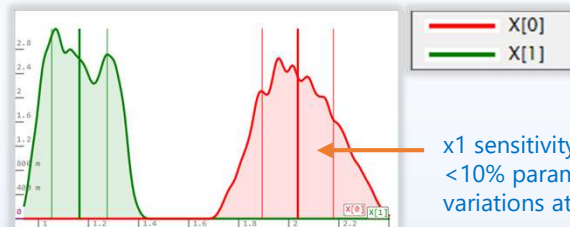
```
function f(number r1 r2 r3) {
  define
  sample S
  species x1 @ 0.66 M in S
  species x2 @ 0.44 M in S
  x1 -> x1 + x1      {r1}
  x1 + x2 -> x2 + x2 {r2}
  x2 -> Ø            {r3}
  equilibrate S for 2.5
  yield [observe(x1,S), observe(x2,S)]
}
```

- <- A function f to run one simulation (ri are the input parameters to be perturbed)
- <- define D yield E returns the value of E after executing the statements D
- <- Make a new sample S to contain species and reactions for simulation
- <- Lotka-Volterra prey species x1 (initial conditions could be a parameter as well)
- <- Lotka-Volterra predator species x2
- <- Prey reproduces, with perturbed rate r1
- <- Predator eats prey, with perturbed rate r2
- <- Predator dies, with perturbed rate r3
- <- Simulate the system up to time 2.5 (first peak of the oscillation)
- <- Return the output concentrations of x1,x2 from S at time 2.5 as pairs

```
random X(omega w) {
  f(1+(w(0)-0.5)/10, 1+(w(1)-0.5)/10, 1+(w(2)-0.5)/10)
}
```

- <- Create a bivariate random variable X over uniform[0..1] sample spaces w(i)
- <- producing random instances $f(1+e_1, 1+e_2, 1+e_3) = [x_1, x_2]_{e_1, e_2, e_3, t=2.5}$ with e_1, e_2, e_3 being 10% independent perturbations of the parameters

draw 2000 from X



x1 sensitivity to random < 10% parameter variations at time 2.5

- <- Produce a density plot of 2000 instances drawn from X i.e. a plot of the distributions of $X[0]=x_1$ and $X[1]=x_2$ at time 2.5 vertical bars are mean and standard deviation

N.B., consider also exporting your Kaemika model to SBML and use the Sobol' method of global sensitivity analysis in e.g. Copasi.

Conclusions

Integrated modeling

Of chemical reaction networks and protocols

How the Kaemika app supports it

Why it needs a *new language* for smooth integration

Closed-loop modeling, experimentation and analysis

For complete lab automation

To “scale up” the scientific method

Experimental biological protocols with formal semantics

Alessandro Abate, Luca Cardelli, Marta Kwiatkowska,
Luca Laurenti, Boyan Yordanov. CMSB 2018.

Kaemika app - Integrating protocols and chemical simulation

Luca Cardelli. CMSB 2020.

Kaemika User Manual

<http://lucacardelli.name/Papers/Kaemika%20User%20Manual.pdf>

Thanks to:

Gold (parser generator)

OSLO (ODE simulator)

C#/Xamarin (IDE)

App store reviewers

NO thanks to:

XAML (general obfuscator)

App store certificates

Dark mode support