

The Cell Cycle Switch Computes Approximate Majority

Luca Cardelli, Microsoft Research & Oxford University

Mestre, 2015-04-08

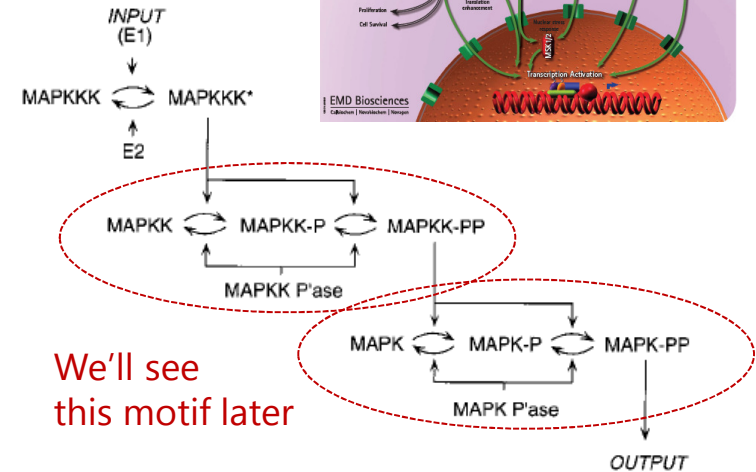
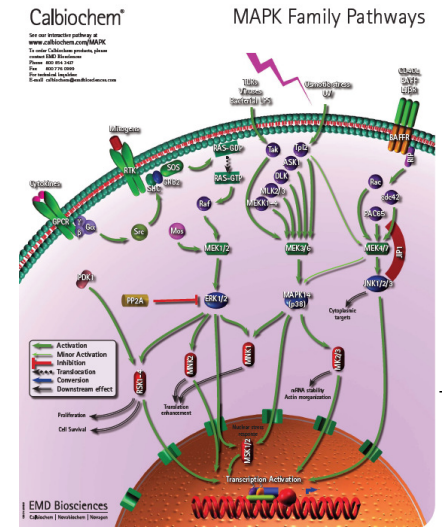
Outline

- Cellular Computation
 - Computational capabilities of biochemical mechanisms that may (or may not) be used by biological entities
- Chemical Algorithms
 - Specific instances of (bio-)chemical computation
 - Particularly, *consensus* and the cell cycle switch
- Obfuscation
 - How to hide a simple algorithm in a complex network
 - How to understand a complex network by a simple algorithm (de-obfuscation)

Cellular Computation

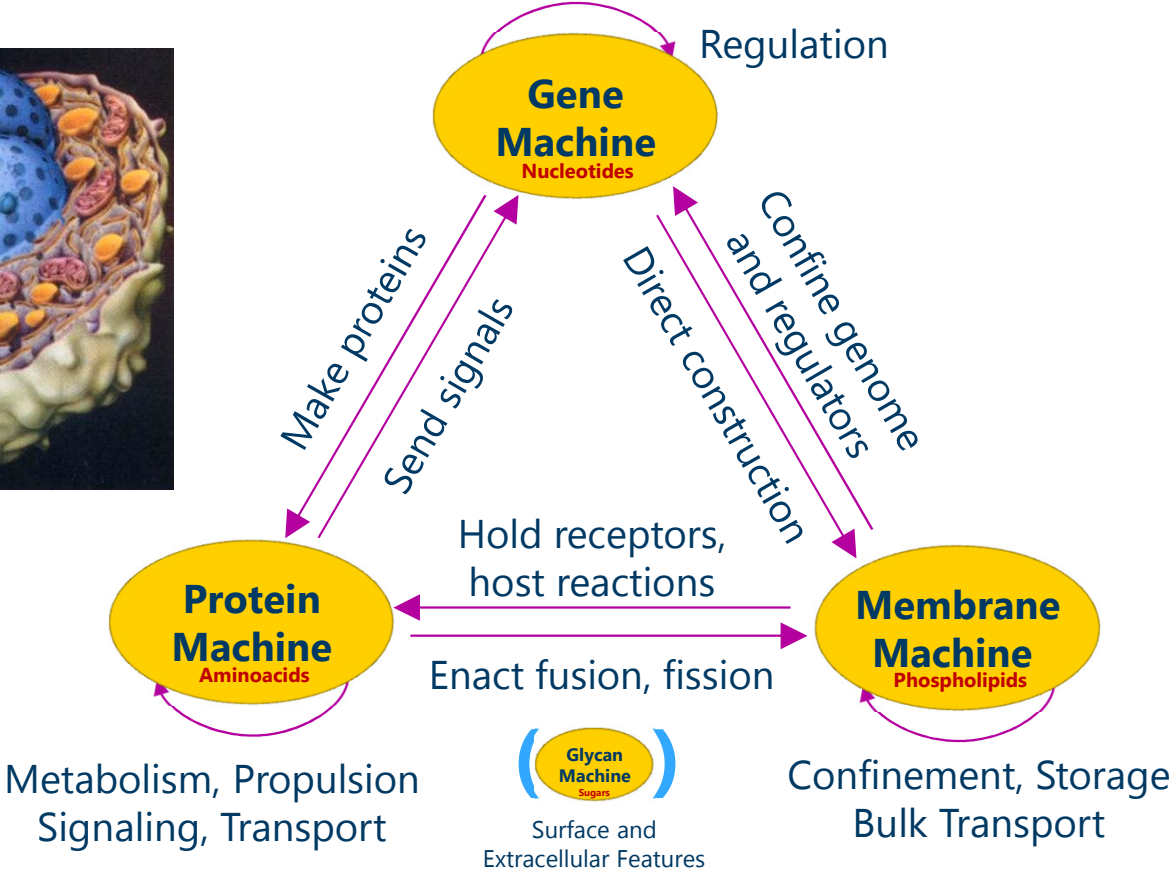
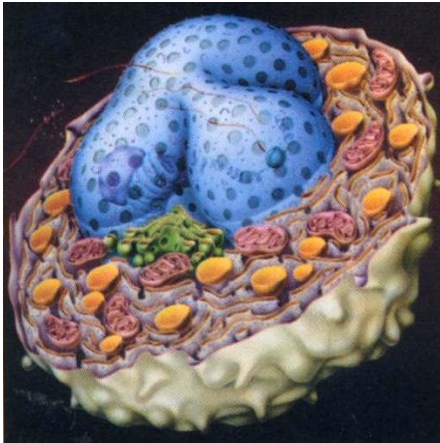
Cellular Computation

- No survival without computation!
 - Finding food
 - Avoiding predators
- How do cells compute?
 - *Clearly* doing “information processing”
 - What are their computational primitives?

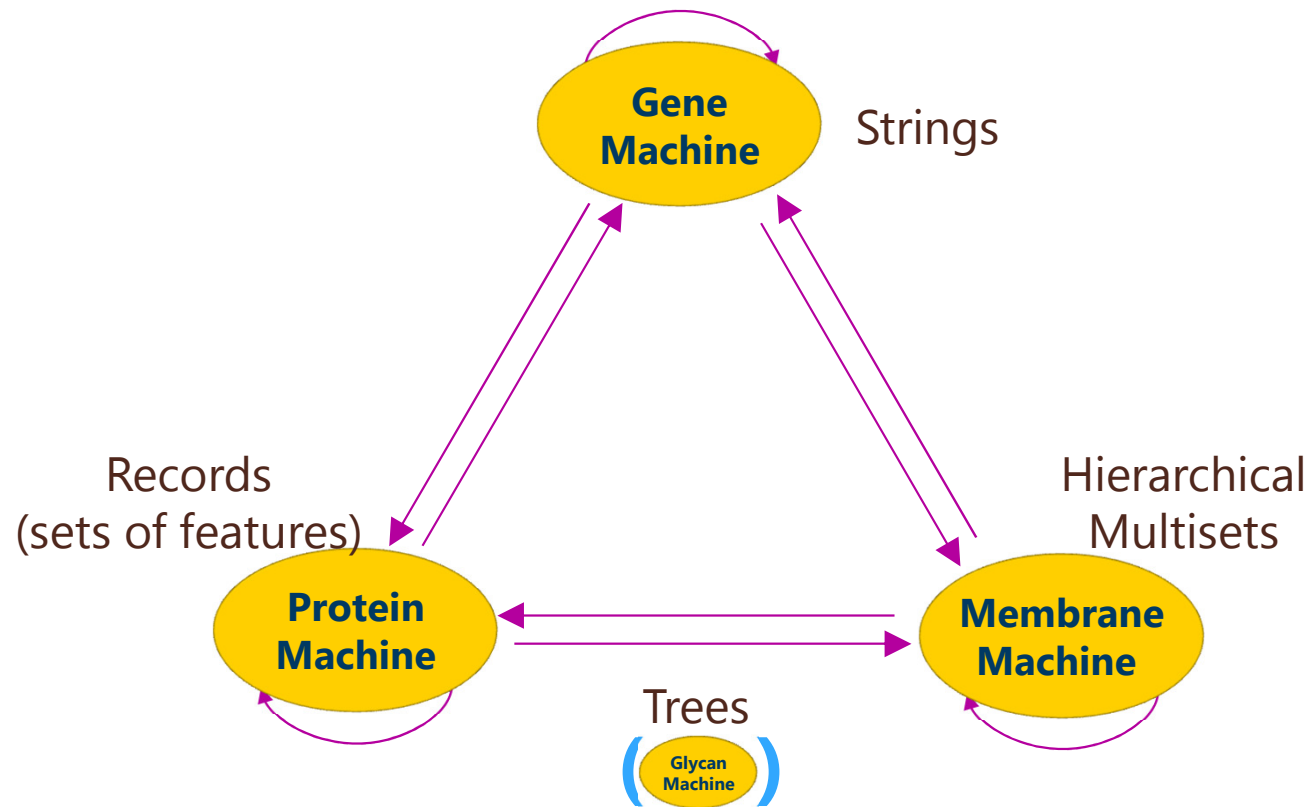


[Ultrasensitivity in the mitogen-activated protein cascade](#), Chi-Ying F. Huang and James E. Ferrell, Jr., 1996, *Proc. Natl. Acad. Sci. USA*, 93, 10078-10083.

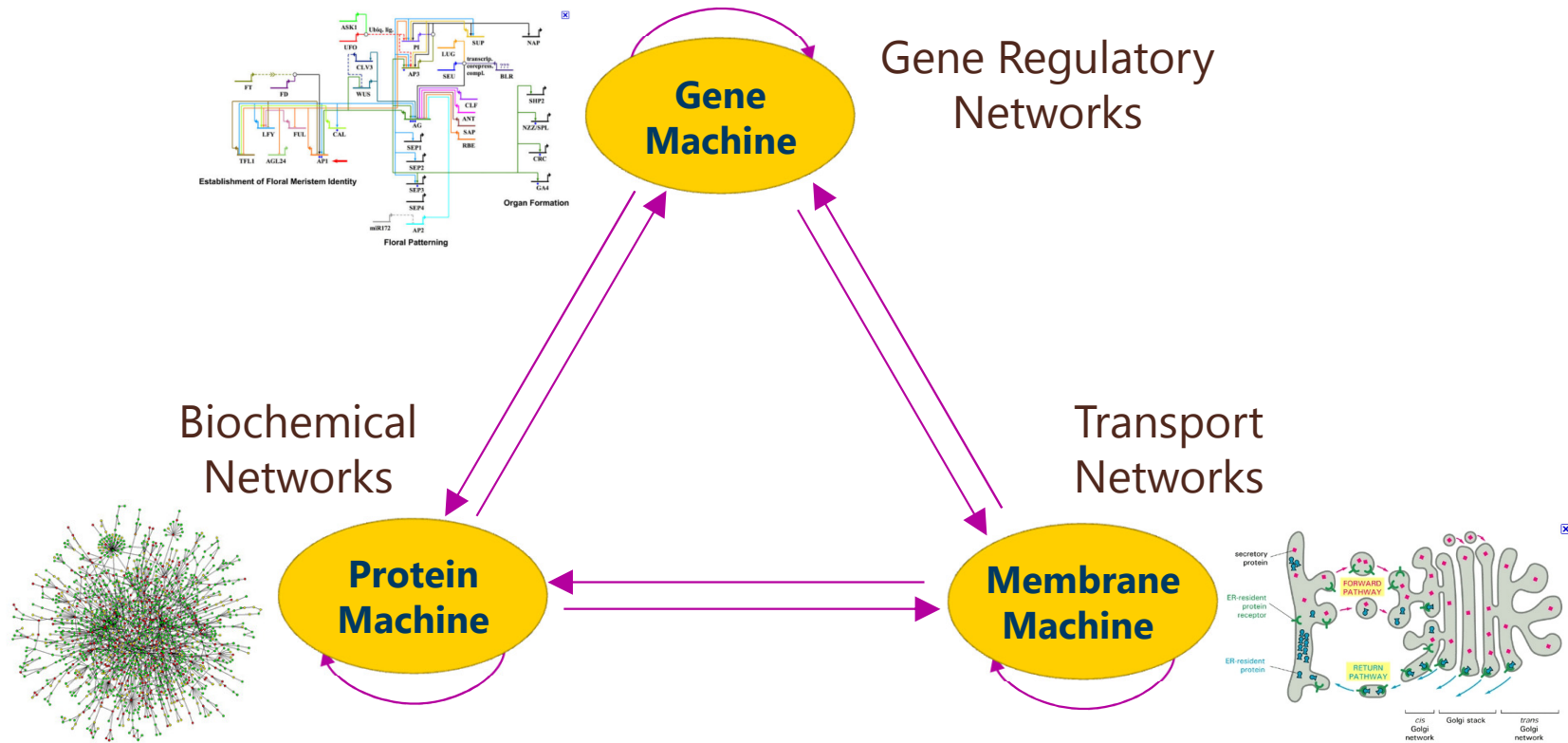
Abstract Machines of Biochemistry



Bioinformatics View (Data Structures)

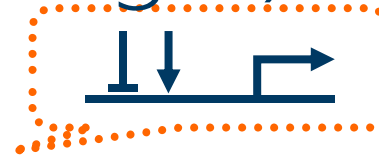
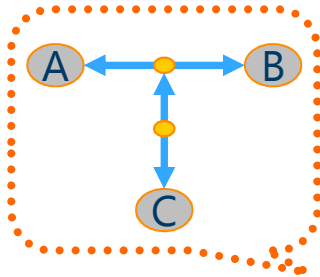


Systems Biology View (Networks)



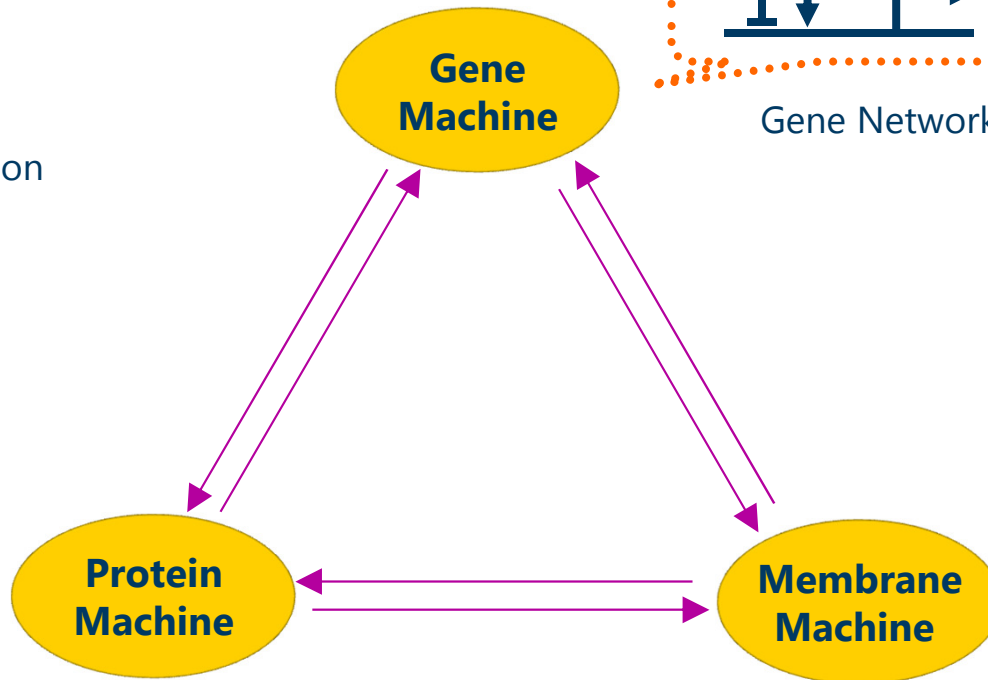
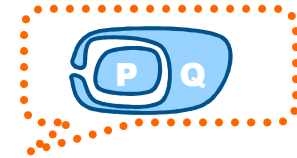
Algorithmic View (Languages)

Molecular Interaction Maps



Gene Networks

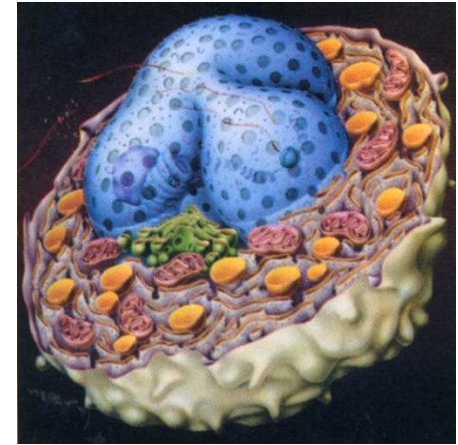
Transport Networks



**These 3 machines
are Turing powerful!**

More concretely

- Give substance to the claim that “cells compute”
 - Yes, but *what* do they compute?
- Catch nature red-handed in the act of running a computational task
 - Something that a computer scientist would recognize as an *algorithm*



Chemical Algorithms

Can *Chemistry* Compute?

- If we believe that biology can do computation...
 - It must be somehow based on chemistry
- So, can chemistry compute, and how?
 - That is in itself a very interesting question with non-trivial answers

Chemical Programming Examples

spec



program



(extra mass comes from "somewhere")

Advanced Programming Examples

spec

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

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

program

$X1 + X2 \rightarrow Y$

$X1 \rightarrow L1 + Y$

$X2 \rightarrow L2 + Y$

$L1 + L2 \rightarrow K$

$Y + K \rightarrow 0$

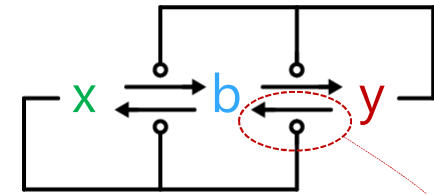
$\max(X1, X2) =$
 $(X1 + X2) - \min(X1, X2)$

(but is not computed
"sequentially": it is a form
of concurrent computation)

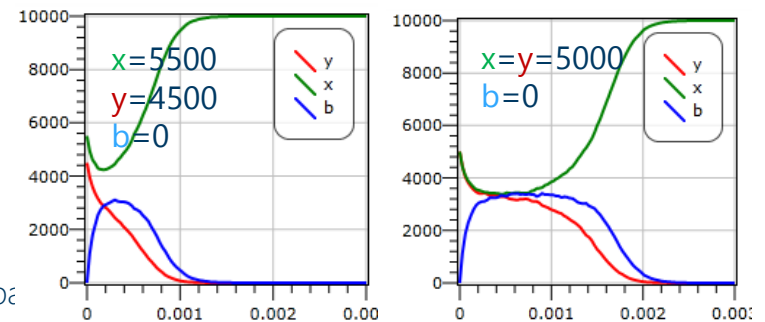
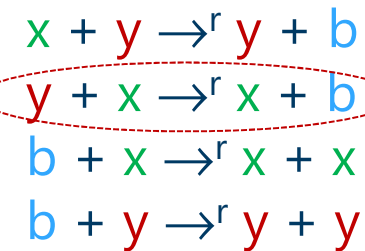
A Consensus Algorithm

- A Population Consensus Problem
 - Given two populations of x and y "agents" (entities/molecules)
 - we want them to "reach consensus"
 - by converting *all* agents to x or to y depending on which population was in majority initially
- Approximate Majority (AM) Algorithm
 - Uses a third "undecided" population b
 - Disagreements cause agents to become undecided
 - Undecided agents agree with any non-undecided agent
- Population Protocols Model
 - Finite-state identity-free agents (molecules) interact in randomly chosen pairs
 - Each interaction (collision) can result in state changes
 - Complete connectivity, no centralized control (well-mixed solution)

catalysis 



chemical
reaction
network

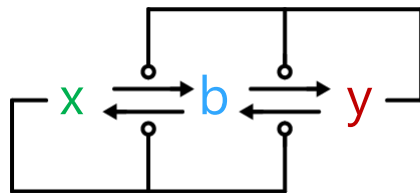


Dana Angluin · James Aspnes · David Eisenstat

A Simple Population Protocol for Fast Robust Approximate Majority

A Biological Implementation

Approximate Majority (AM)



- 1) **Bistable**
Even when initially $x=y$ (stochastically)
- 2) **Fast (asymptotically optimal)**
 $O(\log n)$ convergence time
- 3) **Robust to perturbation**
above a threshold, initial majority wins *whp*

Dana Angluin · James Aspnes · David Eisenstat

A Simple Population Protocol for Fast Robust Approximate Majority

2007

Epigenetic Switch

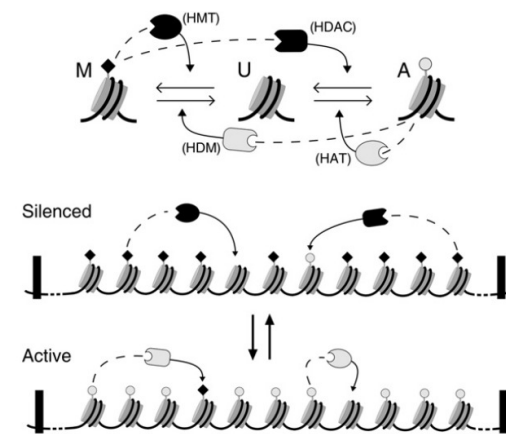


Figure 1. Basic Ingredients of the Model

Theory

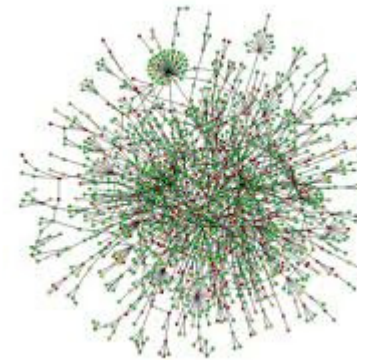
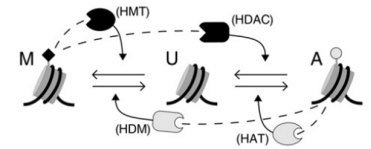
Theoretical Analysis of Epigenetic Cell Memory by Nucleosome Modification

Jan B. Dückel,^{1,2} Mikha A. Mikhaylovskiy,¹ Kim Sjögreen,^{1,2} and Genevieve Thoriin¹
¹Center for Molecular Life Mechanics Institute, Biogenetics IT, DK-2200, Copenhagen N, Denmark
²Department of Molecular and Biomedical Science, Biochemistry, University of Adelaide, SA 5005, Australia
³Department of Molecular Biology, University of Copenhagen, Biocenter, Ole Høvels Vej 5, DK-2200 Copenhagen N, Denmark
 Correspondence: jueduckel@bionet.au.dk
 DOI: 10.1101/041207 (2007)

2007

Here We Got Lucky

- We can claim that the epigenetic switch is a *direct* biological implementation of an algorithm
 - Although we may have to qualify that with some notion of approximation of the (enzymatic) kinetics
- In most cases the biological implementation seems more *indirect* or *obfuscated*
 - "Nature is subtle but not malicious - Einstein" Ha! think again!
 - Other implementations of Approximate Majority seem more convoluted and... approximate

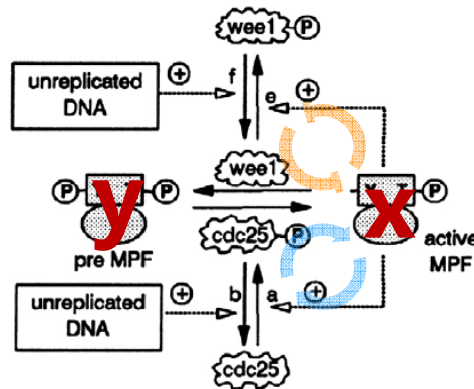


The Cell Cycle Switch

Universal control mechanism regulating onset of M-phase

Paul Nurse

- This basic network is **universal in Eukaryotes** [P. Nurse]
 - The *switching function* and the *basic network* is **the same** from yeast to us.
 - In particular detail, in frog eggs:



Journal of Cell Science 106, 1153-1168 (1993)
Printed in Great Britain © The Company of Biologists Limited 1993

Numerical analysis of a comprehensive model of M-phase control in *Xenopus* oocyte extracts and intact embryos

Bela Novak* and John J. Tyson†

Department of Biology, Virginia Polytechnic Institute and State University, Blacksburg, Virginia 24060-0406, USA

*Permanent address: Department of Agricultural Chemical Technology, Technical University of Budapest, 1521 Budapest Gellert Ter 4, Hungary

†Author for correspondence

- The function is very well-studied. But why this network structure?
- That is, **why this peculiar algorithm?**

How to Build a Good Switch

- We need first a **bistable** system: one that has two *distinct* and *stable* states. I.e., given any initial state the system must settle into one of two states
- The settling must be **fast** (not get stuck in the middle for too long) and **robust** (must not spontaneously switch back)
- Finally, we need to be able to **flip** the switch by external inputs

A Bad Algorithm

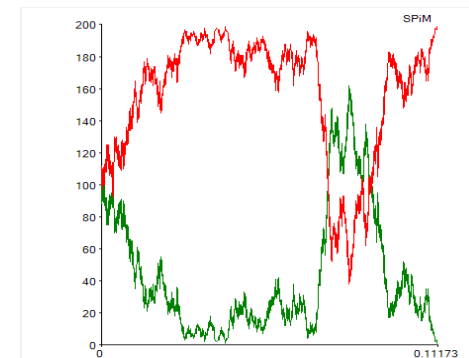
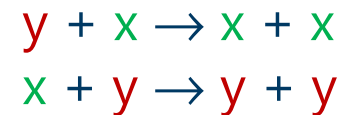
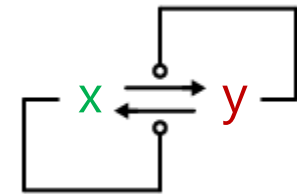
- Direct Competition

- x catalyzes the transformation of y into x
- y catalyzes the transformation of x into y
- when all-x or all-y, it stops

- This system has two end states, but

- Convergence to an end state is slow (a random walk)
- Any perturbation of an end state can start a random walk to the other end state (hence not really *bistable*)

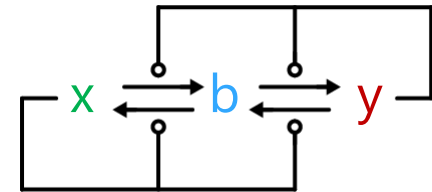
catalysis 



A Good Algorithm

- Approximate Majority (AM)
 - Third, undecided, state b
 - Disagreements cause agents to become undecided
 - Undecided agents believe any non-undecided agent
- With high probability, for n agents
 - The total number of interactions before converging is $O(n \log n)$
 \Rightarrow fast (optimal)
 - The final outcome is correct if the initial disparity is $\omega(\sqrt{n} \log n)$
 \Rightarrow solution states are robust to perturbations
- Logarithmic time bound in parallel time
 - *Parallel time* is the number of steps divided by the number of agents
 - In parallel time the algorithm converges with high probability in $O(\log n)$

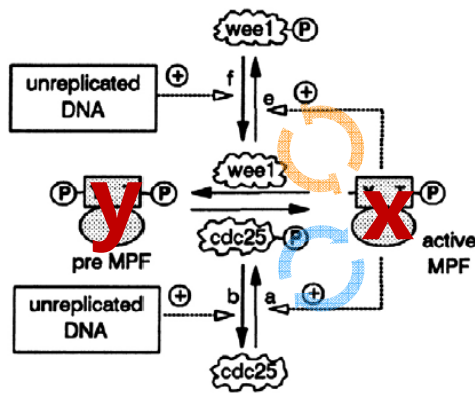
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Dana Angluin · James Aspnes · David Eisenstat

A Simple Population Protocol for Fast Robust Approximate Majority

An "Ugly" Algorithm: Cell Cycle Switch



Nobel-prize winning network

Variation on a distributed algorithm?



Need to explain this network notation!

Double positive feedback on x
 Double negative feedback on x
 No feedback on y. Why ???

- Is it a good algorithm? Is it bad?
- Is it optimal or suboptimal?

How to model "Influence"

"True" molecular interactions.

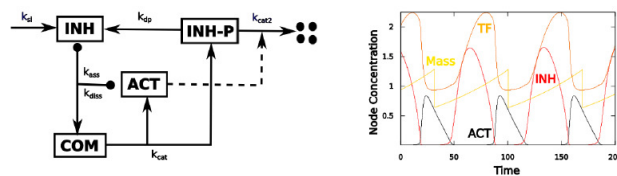


Figure 3: a) Schematic diagram of a simplified SIMM model [17]. The activa-

"Equivalent" influence interactions.

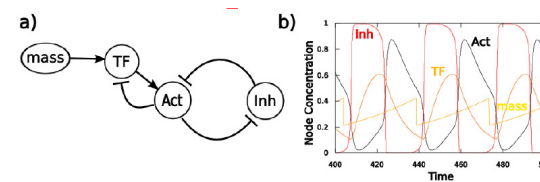


Figure 4: a) Schematic diagram of a primitive cell cycle in the reinitz framework.

Chemical Reaction Network



Influence Network

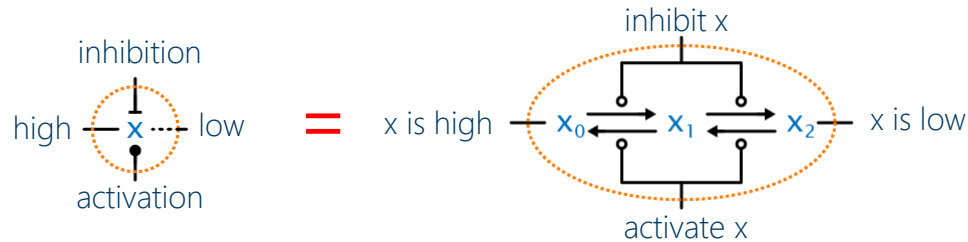
Evolving a Primitive Eukaryotic Cell Cycle Model

Malte Lücken, Jotun Hein, Bela Novak

Instead of modeling basic interactions, such as binding, synthesis, and degradation of molecular components, this framework models interactions simply as activation or inhibition. This approach also reduces the number of nodes necessary in the network, as e.g. the inhibitor binding tightly to the activator to form a complex, which produces phosphorylated inhibitor to be degraded under catalysis by the activator, is now simply a double negative feedback loop shown in Figure 1. This type of interaction is the basis of both aforementioned molecular model, therefore they can both be summarized in a single Reinitz model.

The Triplet Model of Influence

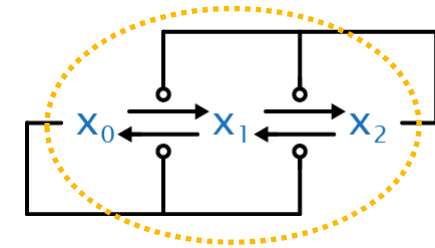
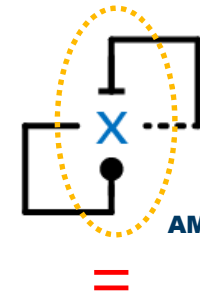
activation ●
inhibition T
catalysis ○



triplet motif

We model them by
4 mass action reactions over
3 species x_0, x_1, x_2

For example:



Approximate Majority

Usually modeled by
sigmoid (e.g. Hill or
Reinitz) functions

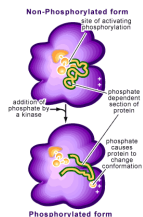


Functional Motifs in
Biochemical Reaction
Networks
John J. Tyson¹ and Bela Novák²

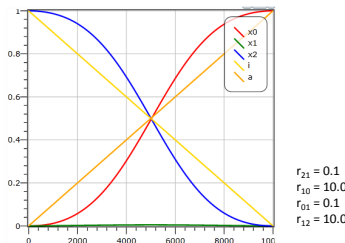
$$\frac{dx_i}{dt} = \gamma_i \frac{[A_i(1-x_i) - B_i x_i]}{A_i + B_i}, \quad i = 1, \dots, N.$$

$$A_i = \exp\left\{\alpha_i \left(\alpha_{i0} + \sum_{j=1}^N \alpha_{ij} X_j\right)\right\}, \quad B_i = \exp\left\{\beta_i \left(\beta_{i0} + \sum_{j=1}^N \beta_{ij} X_j\right)\right\}.$$

biological mechanism:
(e.g.): multisite
phosphorylation

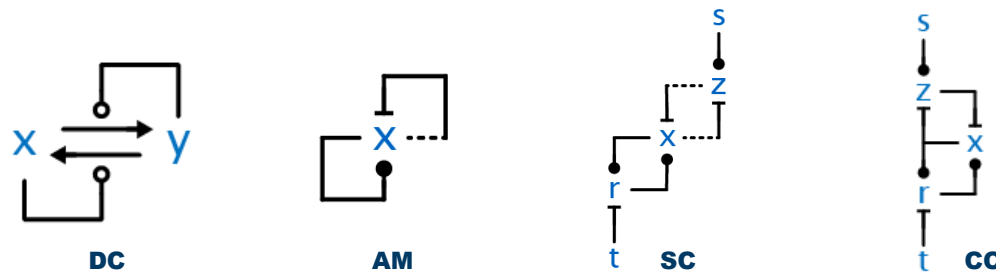


They actually implement a
Hill function of coefficient 2:

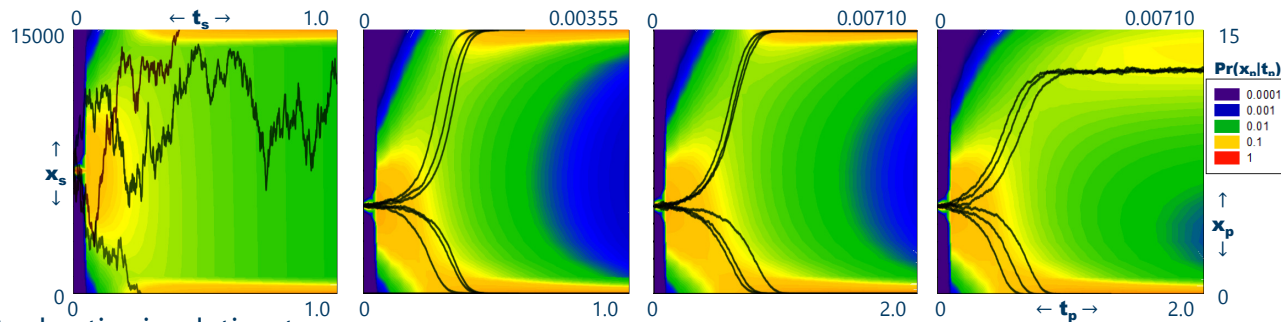


Convergence Analysis

- Switches as **computational systems** CC converges in $O(\log n)$ time (like AM) (but 2x slower than AM, and does not fully switch)



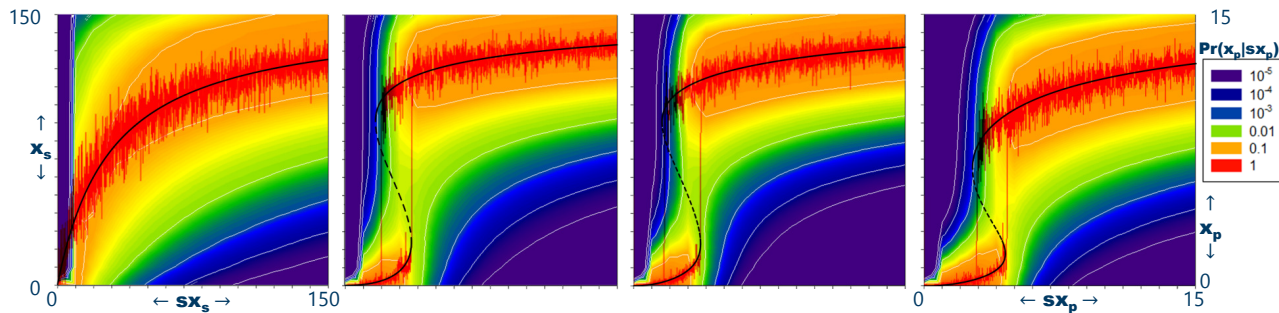
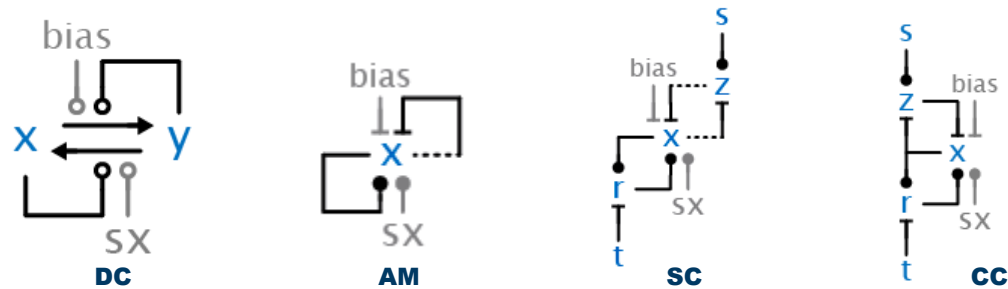
Start symmetrical
($x_0 = x_1 = x_2$ etc.)



Black lines: several stochastic simulation traces
Color: full probability distribution of small-size system

Steady State Analysis

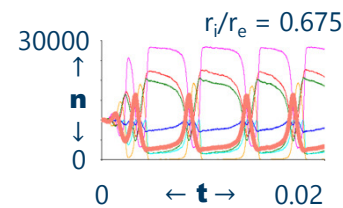
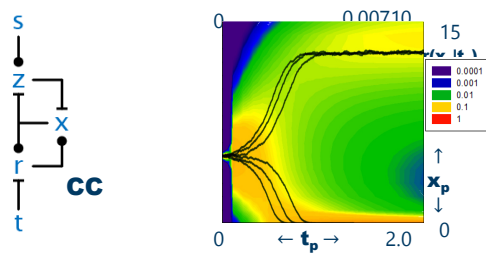
- Switches as dynamical systems



Black lines: deterministic ODE bifurcation diagrams
 Red lines: noisy stochastic simulations
 Color: full probability distribution of small-size system

Why is CC worse than AM?

- The classical CC has an algorithmic “bug”
 - It works ok but never as well as AM
 - Because s continuously inhibits x through z , so that x cannot fully express

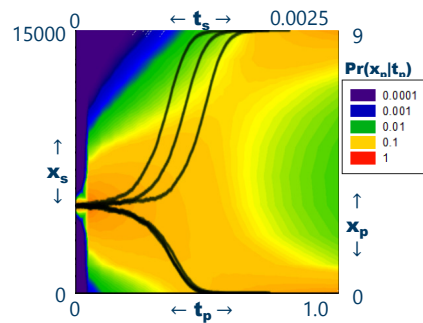
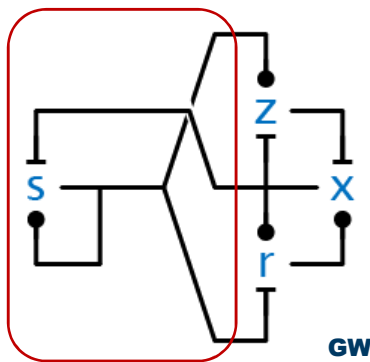


The corresponding cell cycle oscillator is also depressed

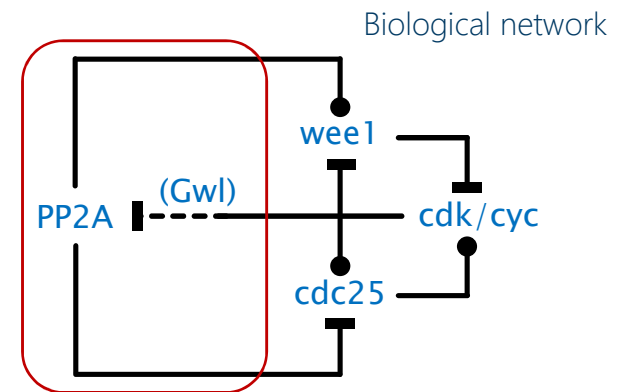
- So let's fix the bug!
 - Easy: let x inhibit s and t “in retaliation”
 - Q: Why didn't nature fix it?

Nature fixed it!

- There is another known feedback loop
 - By which x suppresses s "in retaliation" via the so-called **Greatwall** loop
 - Also, s and t happen to be the same molecule ($=s$)



Full activation!



- s and x now are antagonists: they are **the two halves of the switch**, mutually inhibiting each other (through intermediaries).

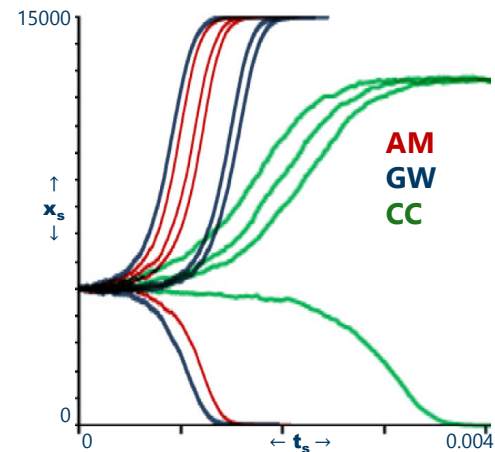
More surprisingly

- The fix makes it faster too!
 - The extra feedback also speeds up the decision time of the switch, making it about as good as the 'optimal' AM switch:

Conclusion:

Nature is trying as hard as it can to implement an AM-class algorithm!

The "classical" cell cycle switch is only half of the picture: the extra feedback completes it *algorithmically*.



Publications

- Our paper appeared:
 - Suggesting GW is a better switch than CC. *September 2012*
- Another paper that same week:
 - Showing experimentally that the Greatwall loop is a **necessary** component of the switch, i.e. the not-as-good-as-AM network has been 'refuted'

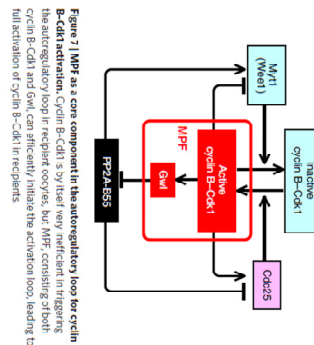
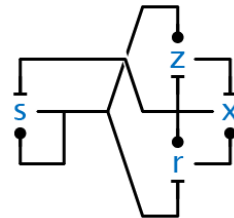


Figure 7 | MPF as a core component in the autoregulatory loop for cyclin B-Cdk1 activation. Cyclin B-Cdk1 is by itself very inefficient in triggering the autoregulatory loop in recipient oocytes, but MPF, consisting of both cyclin B-Cdk1 and GW, can efficiently initiate the activation loop, leading to full activation of cyclin B-Cdk1 in recipients.



The Cell Cycle Switch Computes Approximate Majority

SUBJECT AREAS:
COMPUTATIONAL
BIOLOGY

Luca Cardelli¹ & Attila Csikász-Nagy^{2,3}



ARTICLE

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Greatwall kinase and cyclin B-Cdk1 are both critical constituents of M-phase-promoting factor

Masatoshi Hara^{1,1}, Yusuke Abe^{1,1}, Toshiaki Tanaka², Takayoshi Yamamoto^{1,1}, Eiichi Okumura¹ & Takeo Kishimoto¹

What we learned

- The network structure of AM implements an input-driven switching function (in addition to the known majority function).
- The network structure of CC/GW implements a input-less majority function (in addition to the known switching function).
- The behavior of AM and CC/GW in isolation are related.
- The behavior of AM and CC/GW in oscillator contexts are related (not shown).
- A refinement (GW) of the core CC network, known to occur in nature, improves its switching performance and brings it in line with AM performance.

But again, is CC (or GW) the “same” as AM?

- Our evidence for computational content of biochemical networks is so far
 - Quantitative, covering both kinetic and steady state behavior of *what* networks do
 - But empirical (based on simulations/numerical solutions)
 - And it does not yet explain *how* the CC/GW network relates to the AM network, that is, how each *piece* of CC/GW corresponds to each *piece* of AM
- Analytical evidence is harder to obtain
 - The proofs of the computational properties (optimality etc.) for the AM algorithm are hard and do not generalize easily to more complex networks
 - Quantitative theories of behavioral equivalence and behavioral approximation, e.g. in process algebra, are still lacking (although rich qualitative theories exist)

Obfuscation

When does a (complex) network
implement a (simpler) algorithm?

Antagonistic Networks

- Let's generalize:
 - AM is based on antagonism between two species (inside the triplet)
 - So (essentially) is GW
 - So (essentially) are many standard biological networks
- Are they somehow related?
 - We could try the same empirical analysis as for CC/AM
 - But we can do better

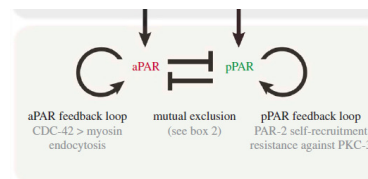
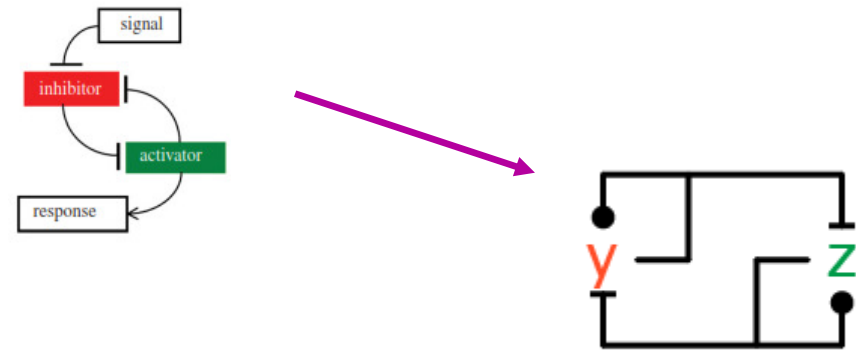
Mutual Inhibition (1 vs. 1)

- “All” cellular switches in all phases of the cell cycle follow (abstractly) a mutual inhibition pattern:

Molecular mechanisms creating bistable switches at cell cycle transitions

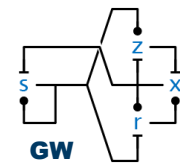
Anael Verdugo, P. K. Vinod, John J. Tyson and Bela Novak
Open Biol. 2013 3, 120179, published 13 March 2013

- Also found in other areas (cell polarity establishment):



MI

cf.:



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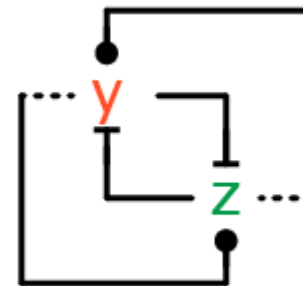
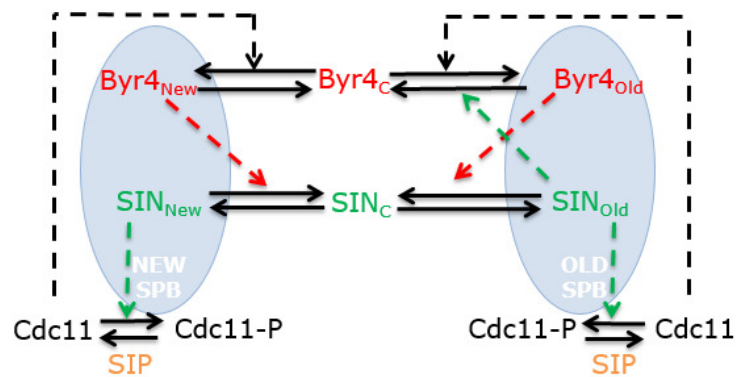
The PAR network: redundancy and robustness in a symmetry-breaking system

Fumio Motegi^{1,2,3} and Geraldine Seydoux⁴

¹Temasek Life Sciences Laboratory, ²Mechanobiology Institute, and ³Department of Biological Sciences, National University of Singapore, 1 Research Link, Singapore 117604, Republic of Singapore
⁴Department of Molecular Biology and Genetics and HHMI, Johns Hopkins University School of Medicine, Baltimore, MD 21205, USA

Septation Initiation (1 vs. 1)

- Other (inherently different) biological networks are based on mutual inhibition, and share characteristics with AM



SIN inhibiting Byr4,
 absence of SIN promoting Byr4
 Byr4 inhibiting SIN,
 absence of Byr4 promoting SIN

Dynamics of SIN Asymmetry Establishment

Archana Bajpai¹, Anna Feoktistova², Jun-Song Chen², Dannel McCollum³, Masamitsu Sato^{4,5}, Rafael E. Carazo-Salas⁶, Kathleen L. Gould², Attila Csikász-Nagy^{1,7,8*}

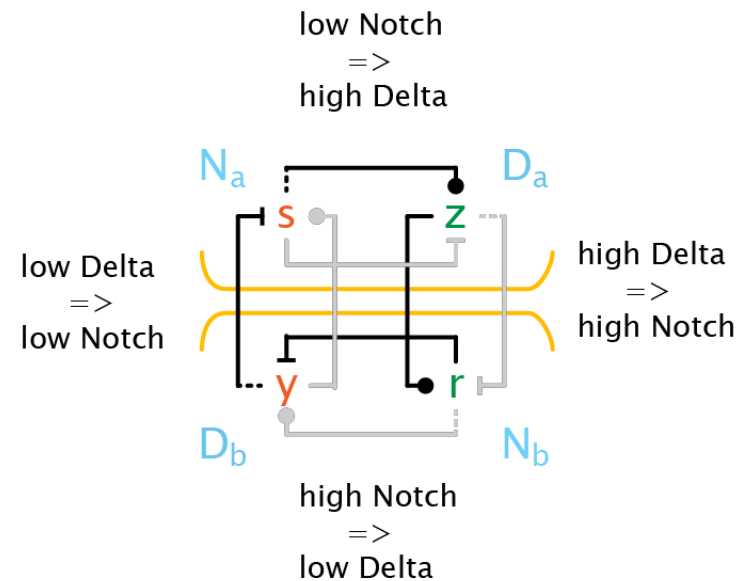
Delta-Notch (2 vs. 2)

- A mutual inhibition pattern
 - Involving *two* species in each cell
- In two cells a,b
 - D_a, N_b antagonize D_b, N_a

Lateral Inhibition through Delta-Notch Signaling: A Piecewise Affine Hybrid Model*

Ronojoy Ghosh and Claire J. Tomlin

M.D. Di Benedetto, A. Sangiovanni-Vincentelli (Eds.): HSCC 2001, LNCS 2034, pp. 232-246, 2001.
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New Cell Cycle Network (3 vs. 3)

- A recent paper presents a more complete view of the cell cycle switch
- N.B. “phosphorylation network dynamics” here is the same as our x_0 - x_1 - x_2 motif

Phosphorylation network dynamics in the control of cell cycle transitions

Daniel Fisher^{1*}, Lillana Krasinska^{1,2}, Damien Coudreuse^{2,3} and Béla Novák^{3,2}

¹Institut de Génétique Moléculaire de Montpellier, IGMM, CNRS UMR 5535, Université Montpellier I and II, 34293 Montpellier, France

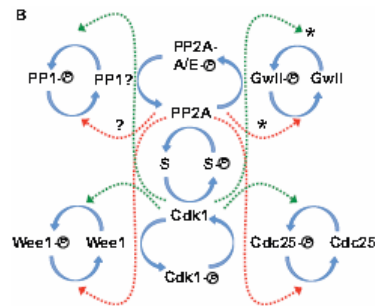
²Institute of Genetics and Development of Rennes, CNRS UMR 6290, 35043 Rennes, France

³Oxford Centre for Integrative Systems Biology, Department of Biochemistry, University of Oxford, South Parks Road, Oxford OX1 3OU, UK

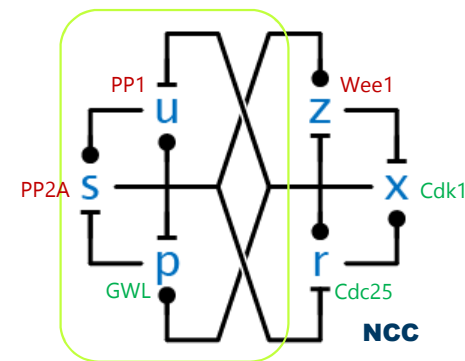
*Author for correspondence (daniel.fisher@igmm.cnrs.fr)

[†]These authors contributed equally to this work.

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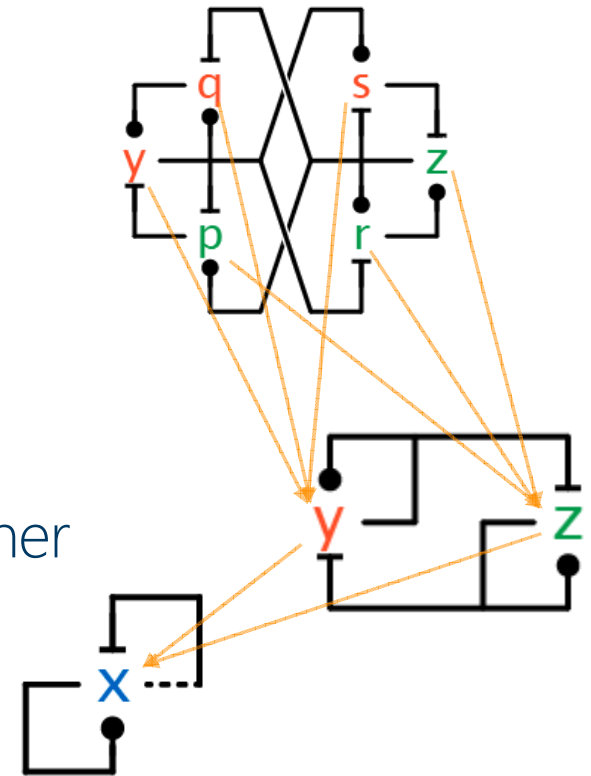


Mutual inhibition between *three* species each



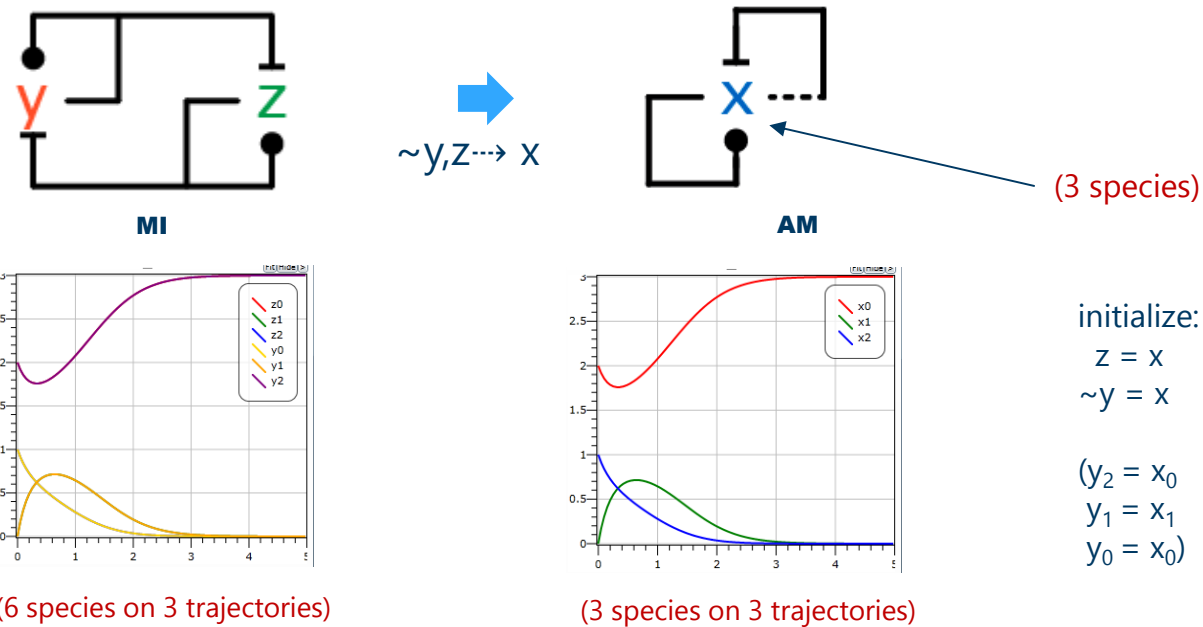
Comparing networks

- How can we compare different networks?
 - Different number of species
 - Different number of reactions
 - Apparently unrelated connectivity
- So that we can compare their function?
 - Does antagonism (in network structure) guarantee bistability (in function)?
- We do it by *mapping* networks onto one another so that they *emulate* each other



Network Emulation: MI emulates AM

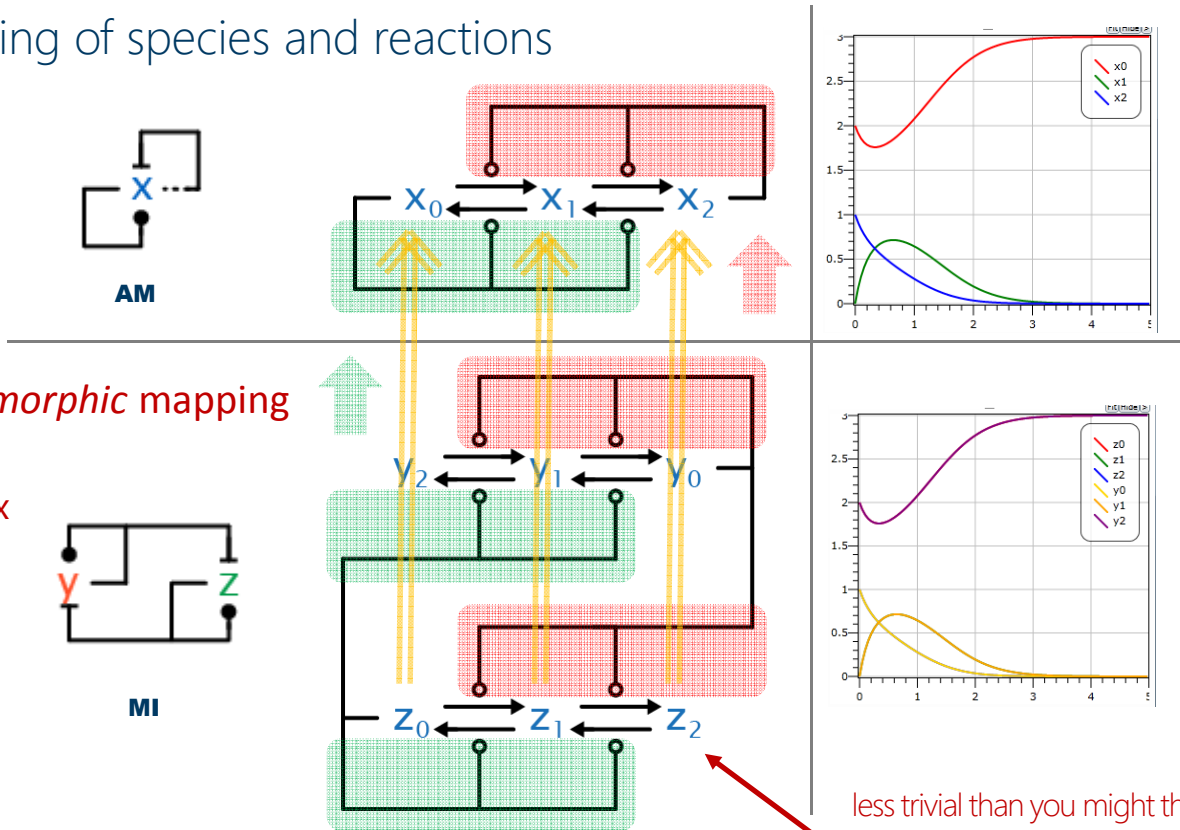
- For **any rates and initial conditions** of AM, we can find *some* rates and initial conditions of MI such that the (6) trajectories of MI retrace those (3) of AM:



- How do we find these matching parameters? By a **network morphism!**

Network Emulation: MI emulates AM

A mapping of species and reactions



any initial conditions

initial conditions:

$$z_0 = y_2 = x_0$$

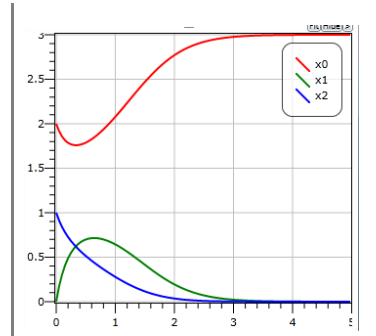
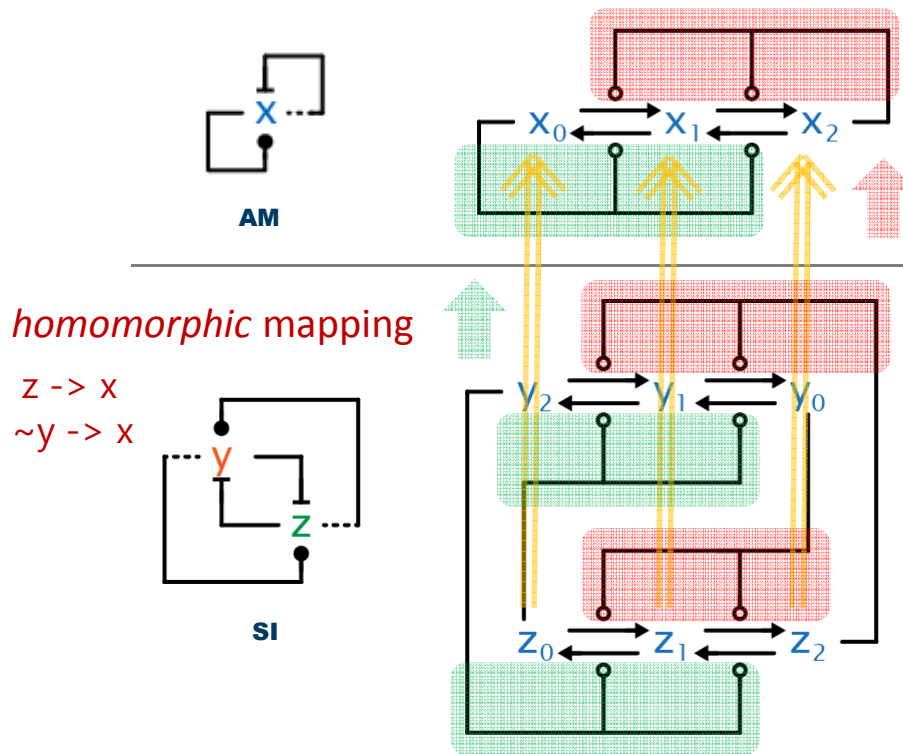
$$z_1 = y_1 = x_1$$

$$z_2 = y_0 = x_2$$

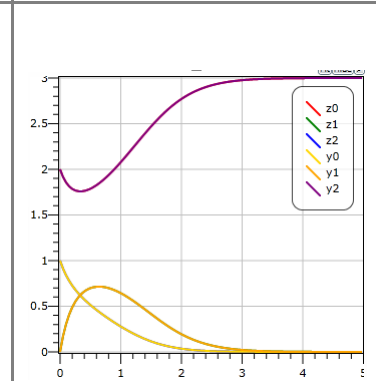
less trivial than you might think:
it need not preserve the out-degree of a node!

Network Emulation: SI emulates AM

A mapping of species and reactions



any initial conditions



initial conditions:

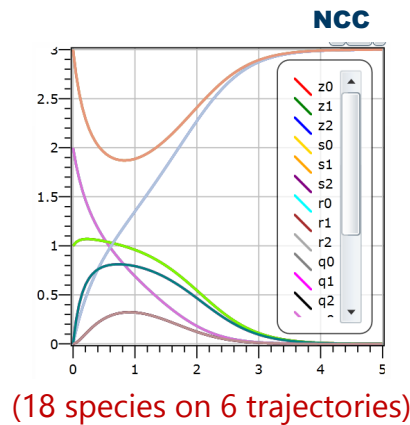
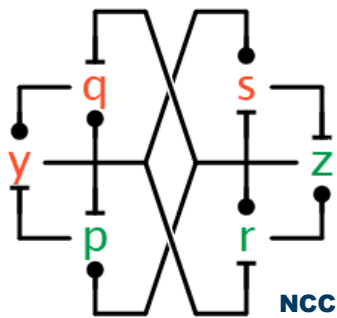
$$z_0 = y_2 = x_0$$

$$z_1 = y_1 = x_1$$

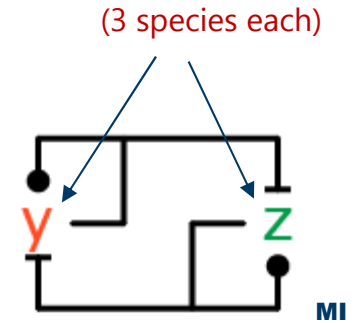
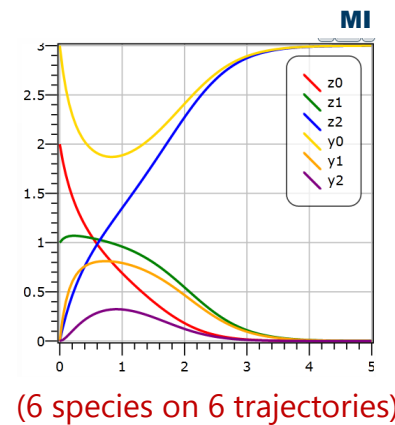
$$z_2 = y_0 = x_2$$

Network Emulation: NCC emulates MI

- For *any* rates and initial conditions of MI we can find *some* rates and initial conditions of NCC such that the (18) trajectories of NCC retrace those (6) of MI



$z, r, p \rightsquigarrow z$
 $y, q, s \rightsquigarrow y$

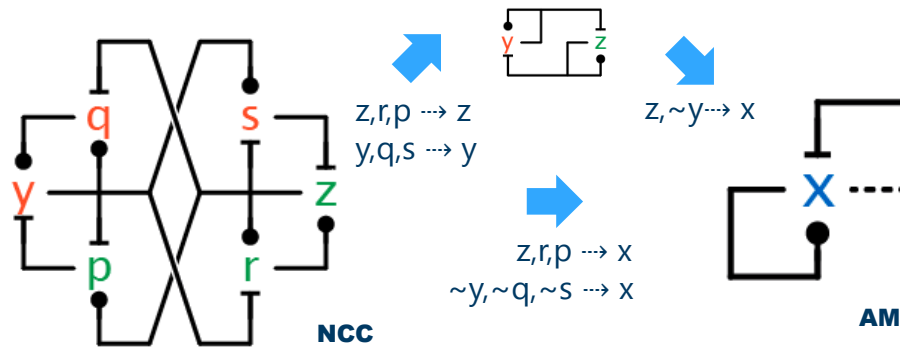


initialize
 $z, r, p = z$
 $y, q, s = y$

- Why does this work so well?

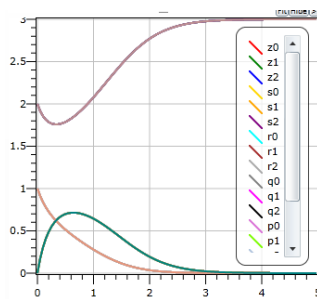
Emulations Compose: NCC emulates AM

- The (18) trajectories NCC can *always* retrace those (3) of AM

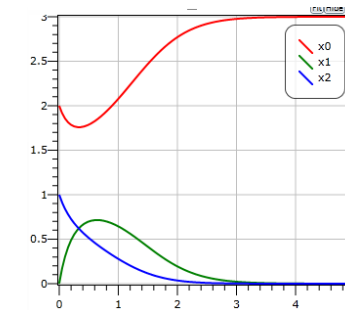


The new cell cycle switch can emulate AM *exactly*.
 For *any* initial conditions of AM.

And for *any* rates of AM.

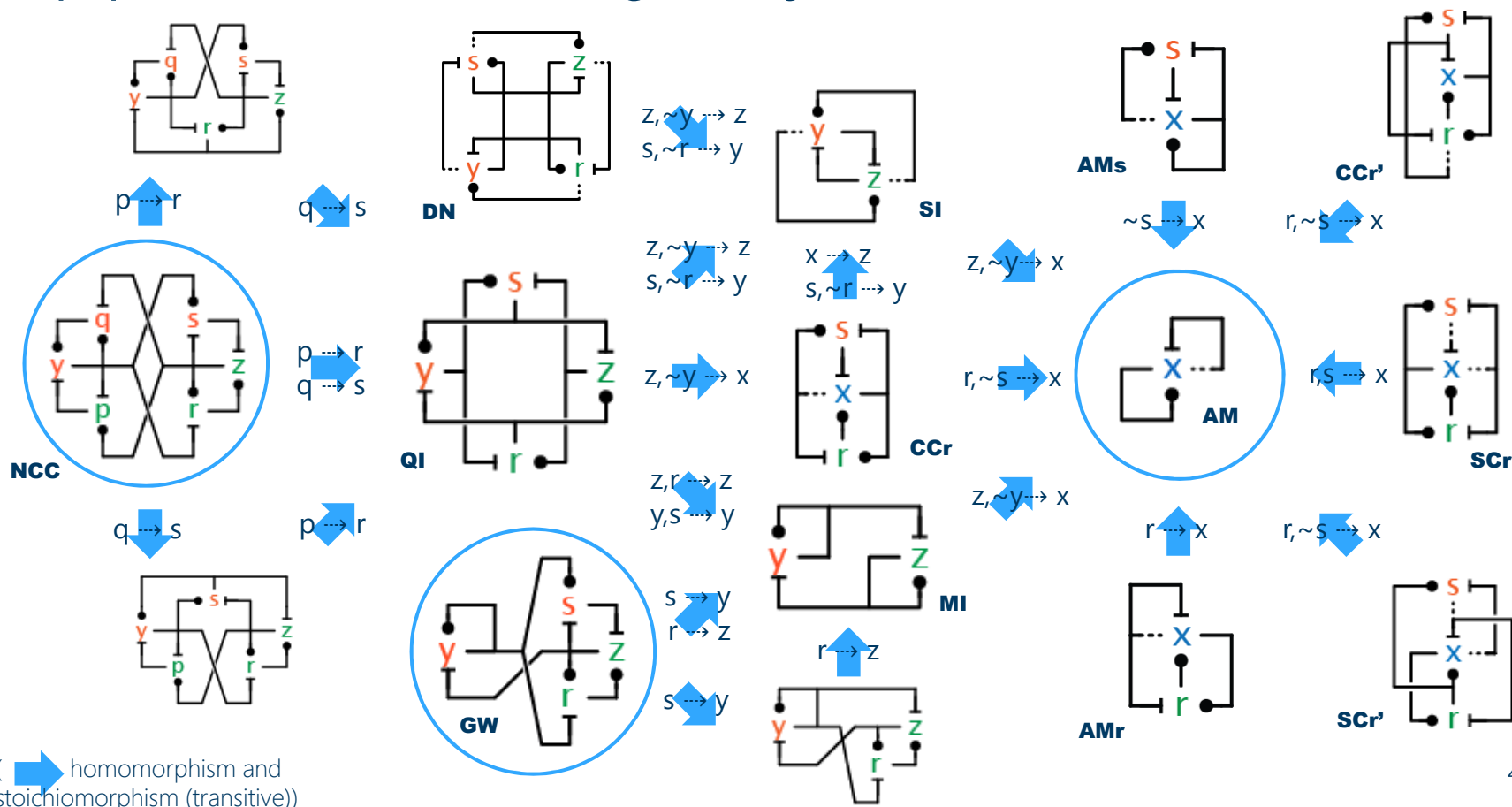


(18 species on 3 trajectories)

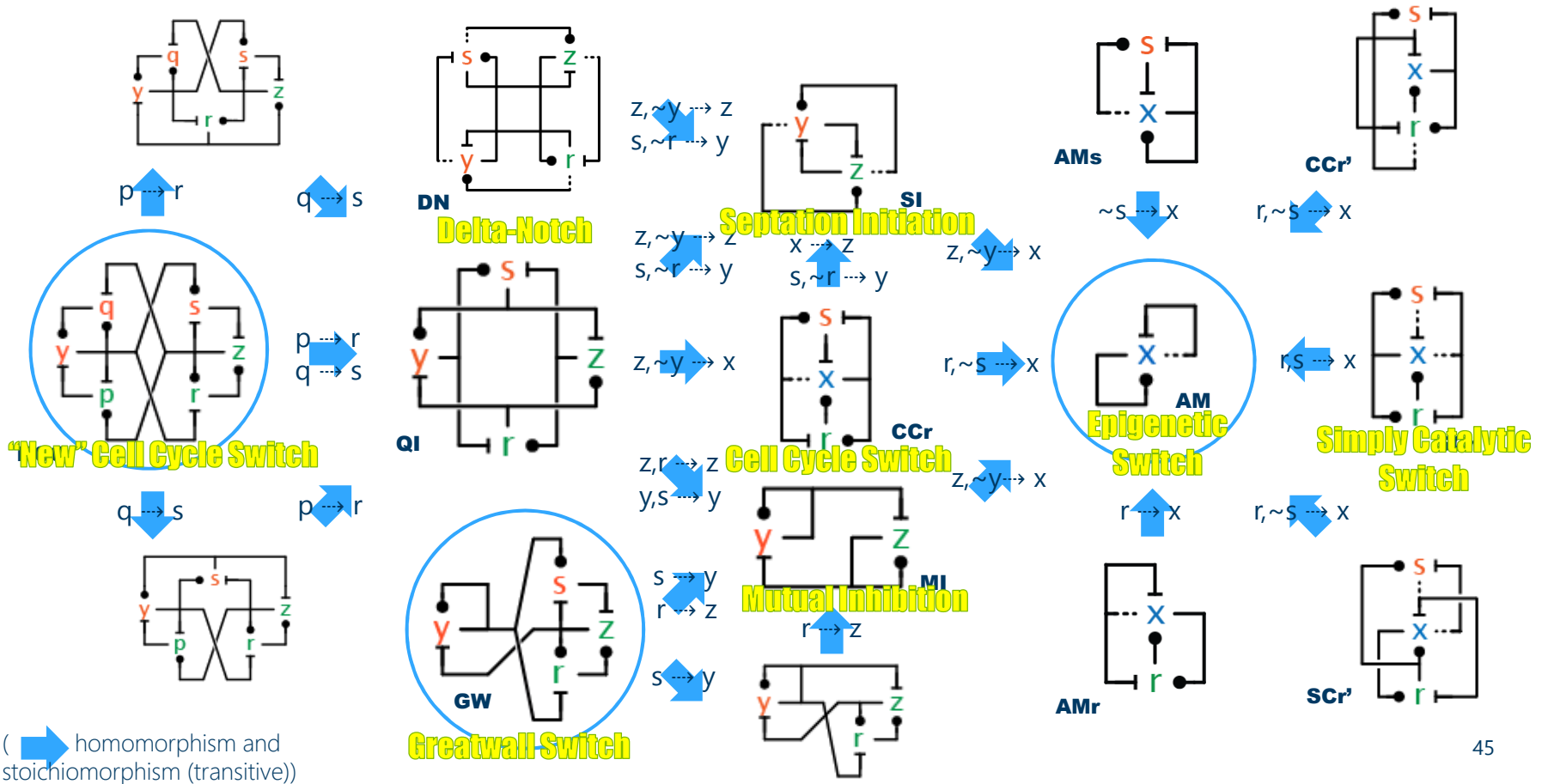


(3 species on 3 trajectories)

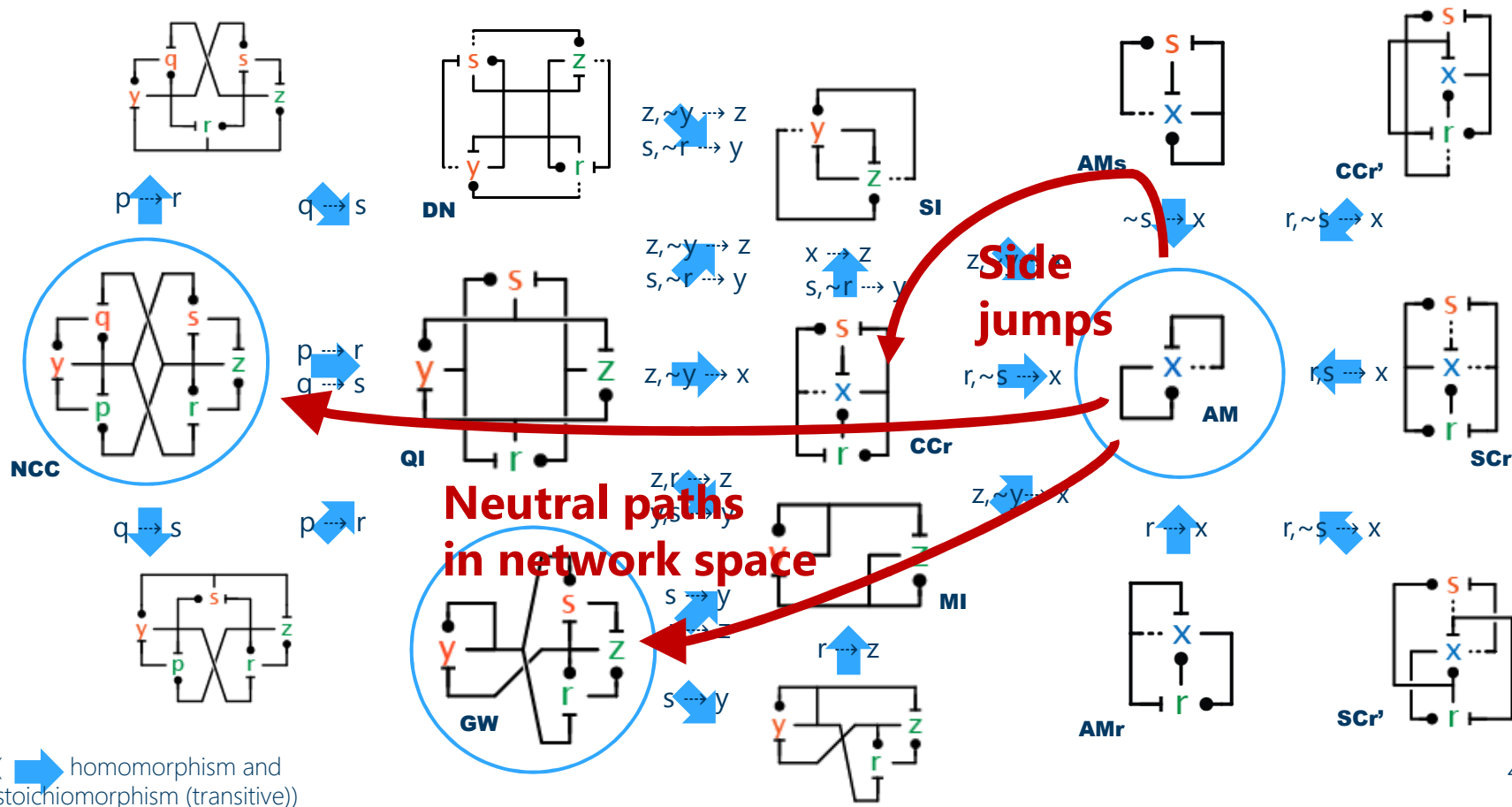
Approximate Majority Emulation Zoo



Approximate Majority Emulation Zoo



Approximate Majority Emulation Zoo



Conclusions

Relating Networks

- Real biological networks
 - Are of course much more complex than these simple patterns
 - How much of that is obfuscation and how much is functional?
- Network emulation can be checked *statically*
 - By stoichiometric/reaction-rate (*structural*) properties
 - That is, no need to compare ODE (*functional*) properties
 - For *any* initial conditions and rates of (one of) the networks
- Efficient algorithms can find emulations
 - Automatic model reduction of large networks

Computational Approach

- Q (traditional): What kind of **dynamical system** is the cell-cycle switch?
- A (traditional): Bistability – ultrasensitivity – hysteresis ...
 - Focused on how sub-populations change over time.
- Q (computational): What kind of **algorithmic system** is the cell-cycle switch?
- A (computational): Interaction – complexity - convergence ...
 - Focused on how individual molecules interact as algorithmic components.
- Leading to a better understanding of not just the *function* but also the *network* (algorithm).
 - If there is some clever population algorithm in nature that we have not invented yet (unlike AM) how shall we recognize it?