

The Cell Cycle Switch Computes Approximate Majority

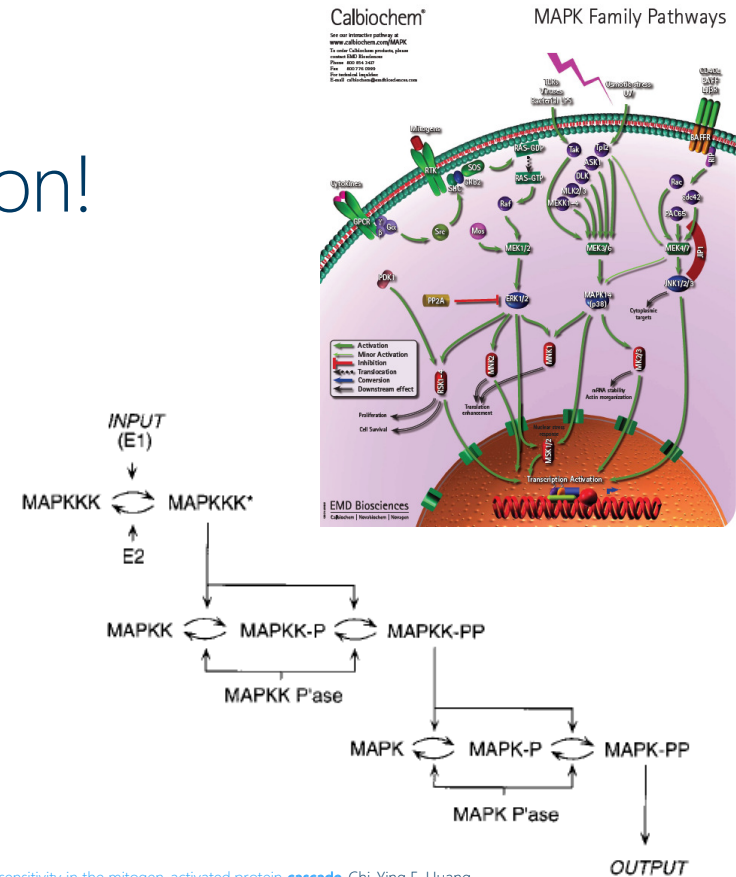
Luca Cardelli, Microsoft Research & Oxford University

Joint work with Attila Csikász-Nagy, CoSBI & King's College London

Warwick, 2014-01-09

Cells Compute

- No survival without computation!
 - Finding food
 - Avoiding predators
- How do they compute?
 - Clearly doing “information processing”
 - Based on complex, higher-order interactions
 - **MAPKKK** = MAP Kinase Kinase Kinase = *that which operates on that which operates on that which operates on protein.*
 - How ‘sophisticated’ are natural algorithms?



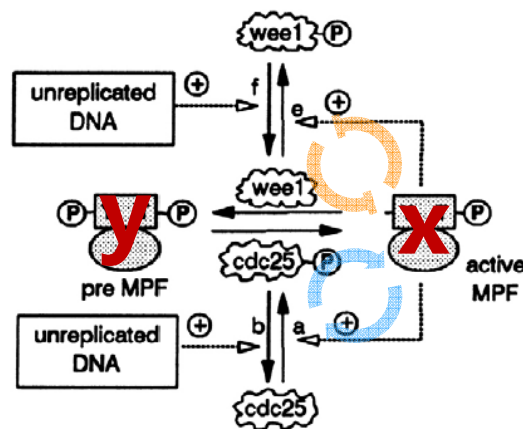
Ultrasensitivity in the mitogen-activated protein cascade, Chi-Ying F. Huang and James E. Ferrell, Jr., 1996, *Proc. Natl. Acad. Sci. USA*, 93, 10078-10083.

Outline

- Analyzing biomolecular networks
 - Try do understand the function of a network
 - But also try to understand its *structure*, and what determines it
- The Cell-Cycle Switches
 - Some of the best studied molecular networks
 - Important because of their fundamental function (cell division) and the stability of the network across evolution
- We ask:
 - What does the cell cycles switch compute?
 - How does it compute it?

The Cell Cycle Switch

- This network is **universal in all Eukaryotes** [P. Nurse]
 - I.e., the **network** at the core of cell division is *the same* from yeast to us
 - *Not the components of the network, nor the rates*



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

Bela Novak* and John J. Tyson†

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†Author for correspondence

Double positive feedback on x
 Double negative feedback on x
 No feedback on y
 What on earth ... ???

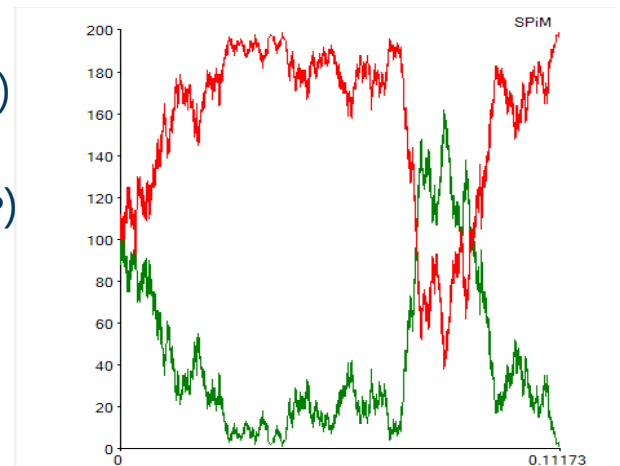
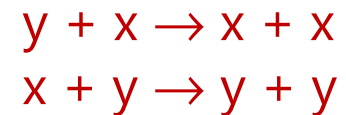
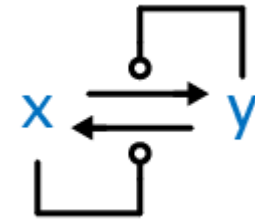
- The function is very well-studied. But why this structure?
- I.e., **why this algorithm?**

How to Build a Good Switch

- What is 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
- “Population” Switches
 - Populations of identical agents (molecules) with the whole population switching from one state to another as a whole
 - Highly concurrent (**stochastic**)

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*)



A Very Good Algorithm

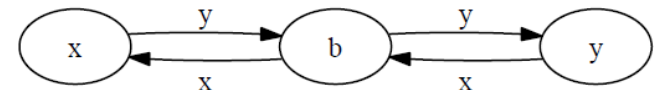
- Approximate Majority (AM)
 - Decide which of two populations is in majority
- A fundamental 'population protocol'
 - Agents in a population start in state x or state y
 - A pair of agents is chosen randomly at each step, they interact ('collide') and change state
 - The whole population must eventually agree on a majority value (all- x or all- y) with probability 1

Dana Angluin · James Aspnes · David Eisenstat

A Simple Population Protocol for Fast Robust Approximate Majority

We analyze the behavior of the following population protocol with states $Q = \{b, x, y\}$. The state b is the **blank** state. Row labels give the initiator's state and column labels the responder's state.

	x	b	y
x	(x, x)	(x, x)	(x, b)
b	(b, x)	(b, b)	(b, y)
y	(y, b)	(y, y)	(y, y)



Third 'undecided' state

- 1) Disagreements cause agents to become undecided
- 2) Undecided agents believe any non-undecided agent they meet

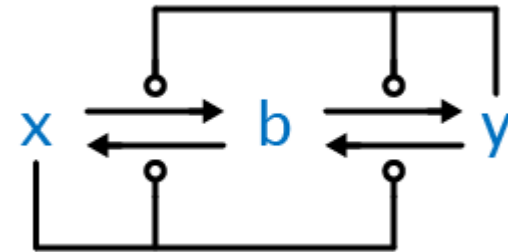
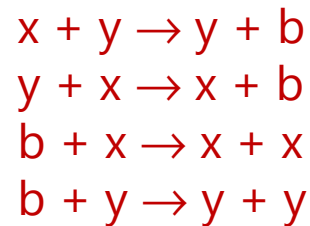
Properties

[Angluin et al., <http://www.cs.yale.edu/homes/aspnes/papers/disc2007-eisenstat-slides.pdf>]

- With high probability, for n agents
 - The total number of interactions before converging is $O(n \log n)$
⇒ fast
 - The final outcome is correct if the initial disparity is $\omega(\sqrt{n} \log n)$
⇒ 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 $O(\log n)$

Chemical Implementation

Chemistry as a programming language for population algorithms!



Bistable

Even when $x=y$! (stochastically)

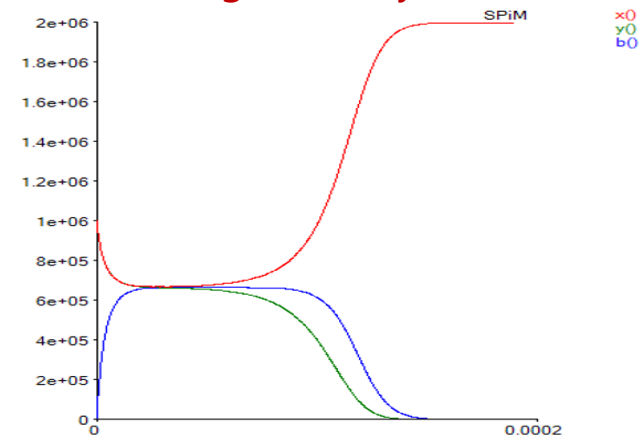
Fast

$O(\log n)$ convergence time

Robust to perturbation

above a threshold, initial majority wins *whp*

Worse-case scenario example, starting with $x=y$, $b=0$:



A Biological Implementation

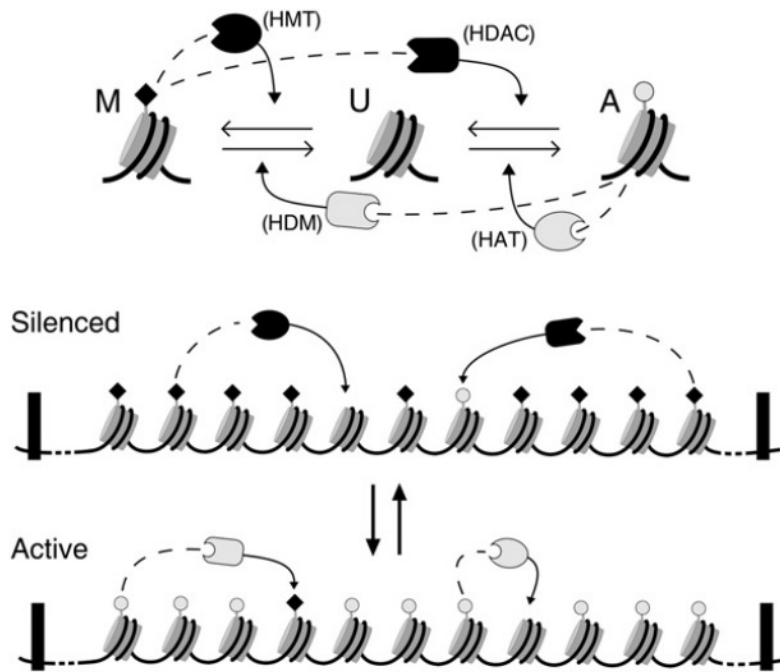
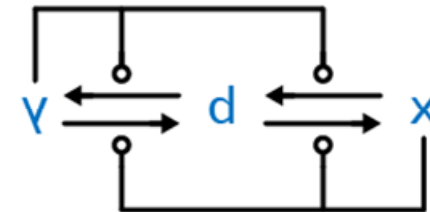


Figure 1. Basic Ingredients of the Model



Theory

Theoretical Analysis of Epigenetic Cell Memory by Nucleosome Modification

Ian B. Dodd,^{1,2} Mille A. Michelsen,¹ Kim Sneppen,^{1*} and Genevieve Thon³
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 DOI: 10.1016/j.cell.2007.02.053

Cell

Back to the Cell Cycle

- The AM algorithm has ideal properties for settling a population into one of two states
- But that is not what the cell cycle uses
- Or is it?

Influence Network Notation

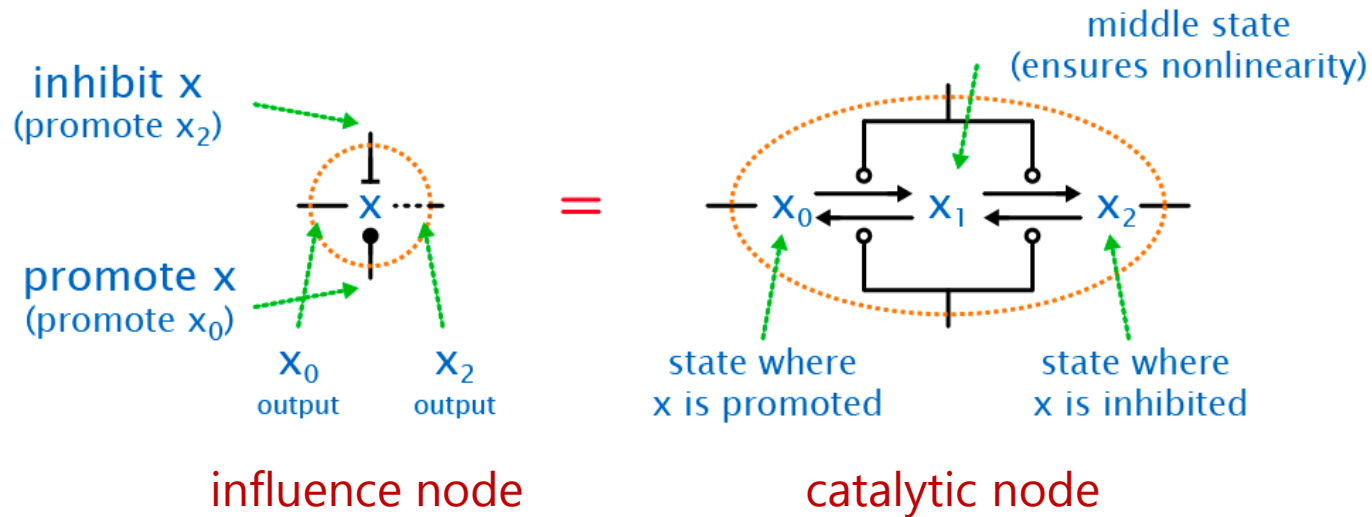
- Catalytic reaction



z is the catalyst



- 'Double kinase-phosphatase' motif



Influence Network Duality

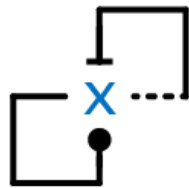
- Let $\sim X$ be the species such that

$$(\sim X)_0 = X_2, \quad (\sim X)_1 = X_1, \quad (\sim X)_2 = X_0$$

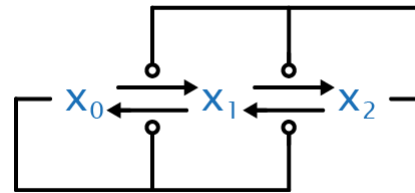
so that promoting x is the same as inhibiting $\sim x$ etc. Then:



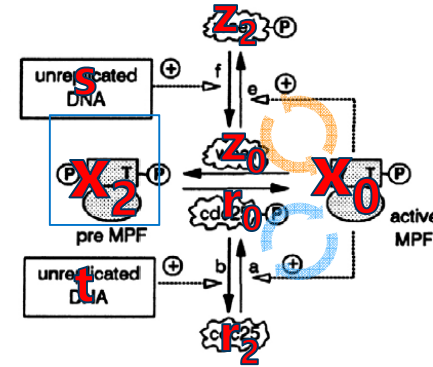
AM and CC Influence Networks



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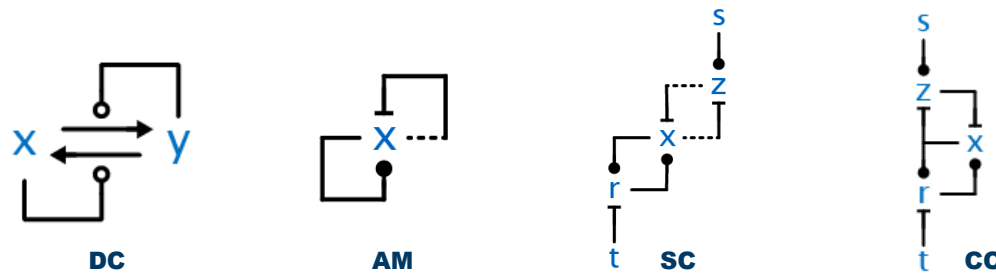


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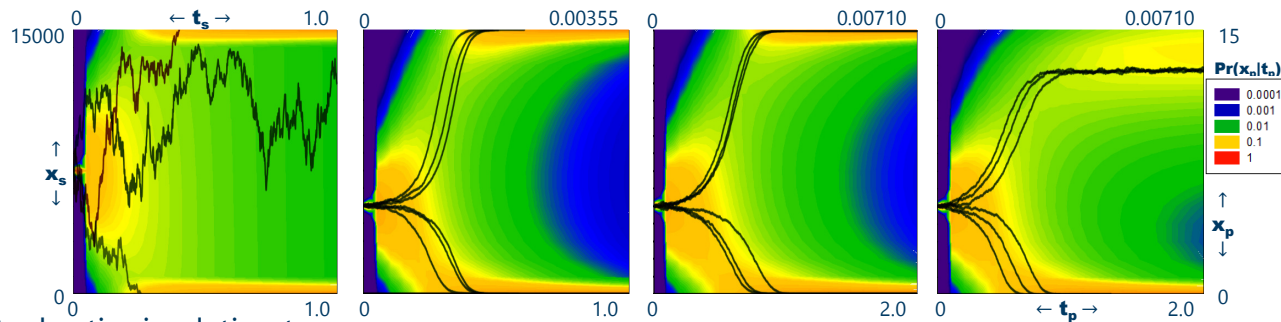


Convergence Analysis

- Switches as computational systems



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

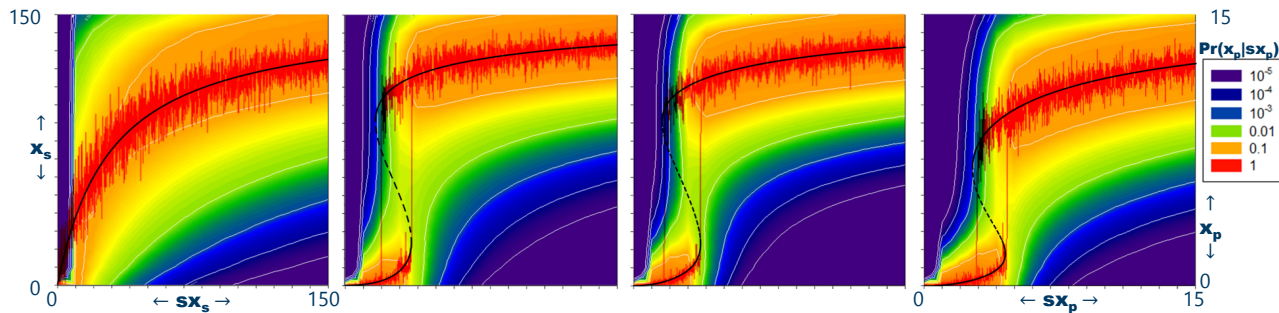
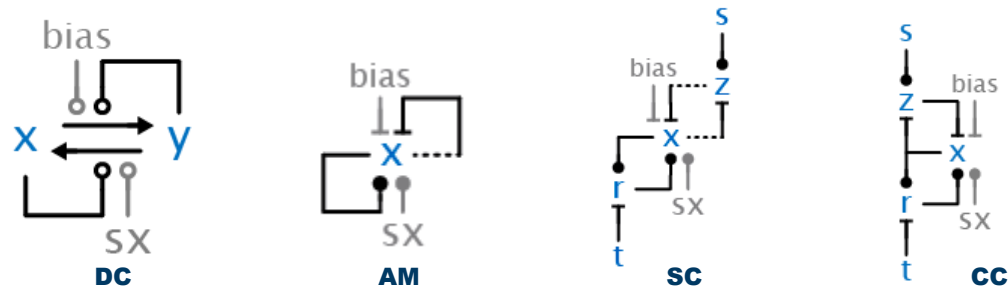


Black lines: several stochastic simulation traces
Color: full probability distribution of small-size system

NEW!
CC appears to converge in log time

Steady State Analysis

- Switches as dynamical systems

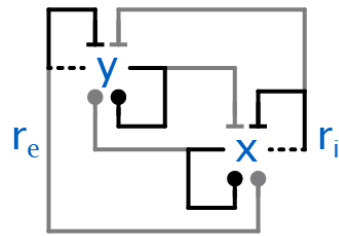


Black lines: deterministic ODE bifurcation diagrams
 Red lines: noisy stochastic simulations
 Color: full probability distribution of small-size system

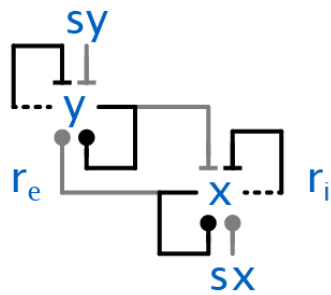
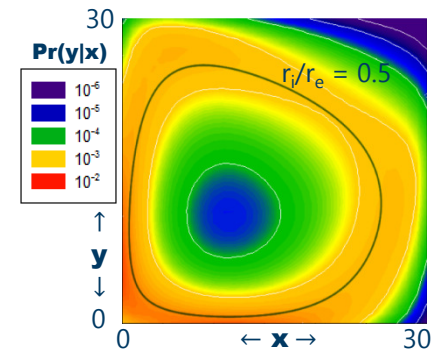
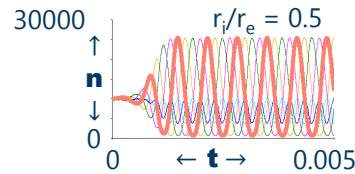
NEW!
 AM shows hysteresis

Contextual Analysis

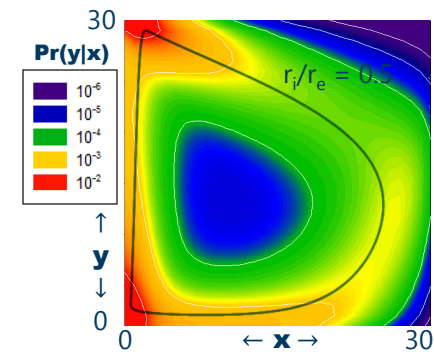
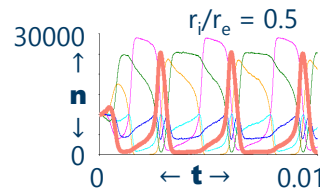
- AM switches in the context of oscillators



Trammel

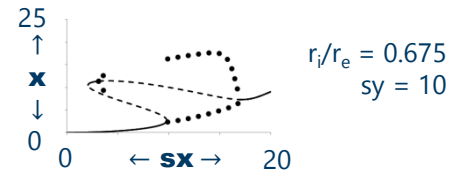
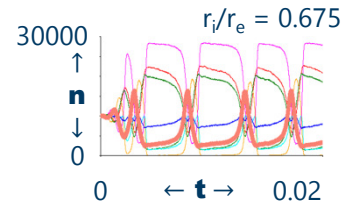
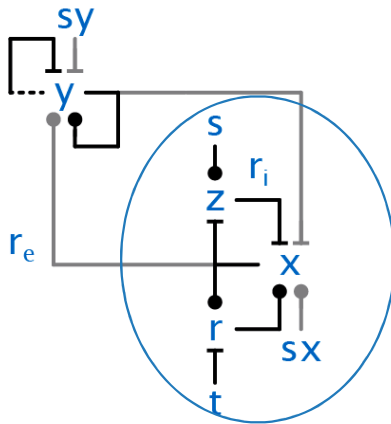
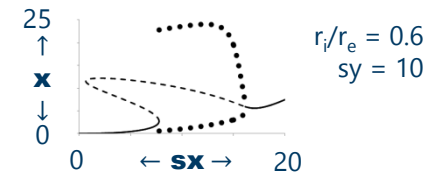
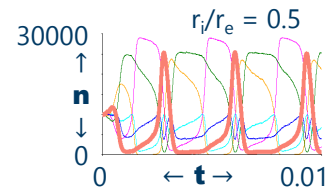
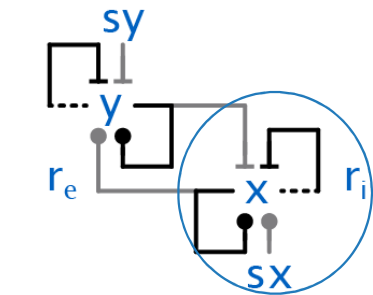


Shishi Odoshi



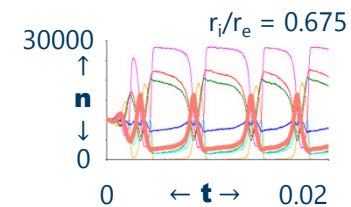
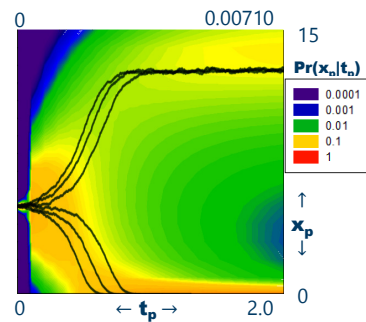
Modularity Analysis

- CC swapped in for AM



Evidence that CC is 'similar' to AM

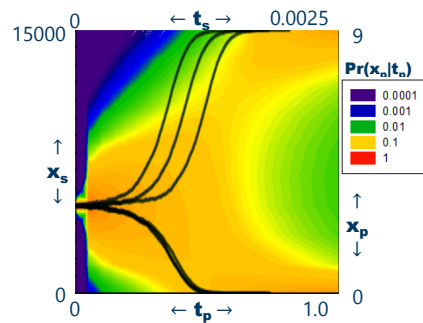
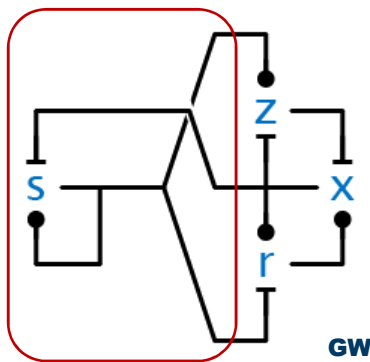
- But there is a difference
 - The classical cell cycle switch, CC, works ok but never as well as AM
 - The output of CC does not go 'fully on':



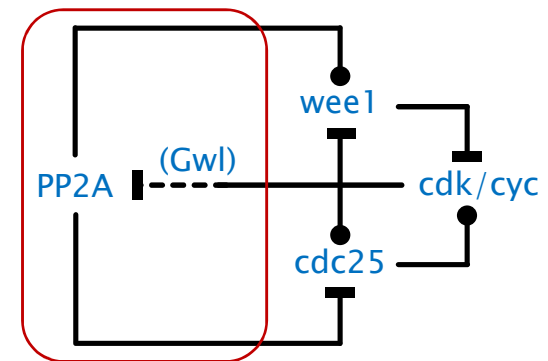
- Because s continuously inhibits x through z , so that x cannot fully express
- Q: Why didn't nature do better than that?

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



Full activation!



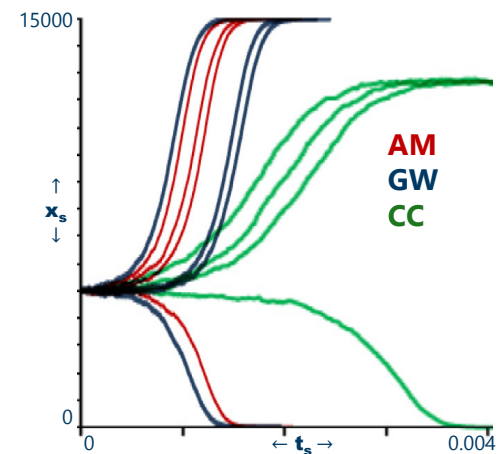
- s and x now are antagonists: they are the two halves of the switch, mutually inhibiting each other (through intermediaries).

More surprisingly

- Made 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 (in our published paper):
Nature is trying as hard as it can to implement an AM-class algorithm!

The "classical" cell cycle switch does not appear to be the full picture: the extra feedback completes it algorithmically.



The Greatwall Kinase

- Our paper appeared:
 - Suggesting GW is a better switch than CC
- Another paper the 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'

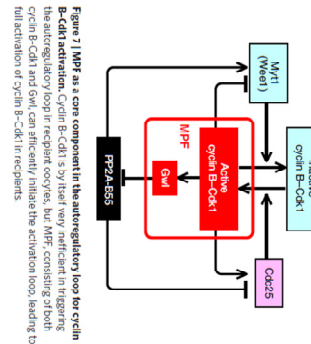
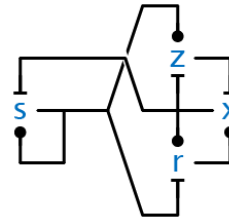


Figure 7 | MPF as a core component in the autoregulatory loop for cyclin B-Cdk1 activation. Cyclin B-Cdk1 is by itself very inefficient in triggering the autoregulatory loop in recipient oocytes, but MPF, consisting of both cyclin B-Cdk1 and Gwl, can efficiently initiate the activation loop, leading 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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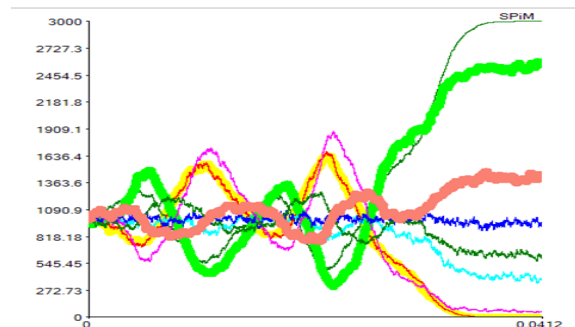
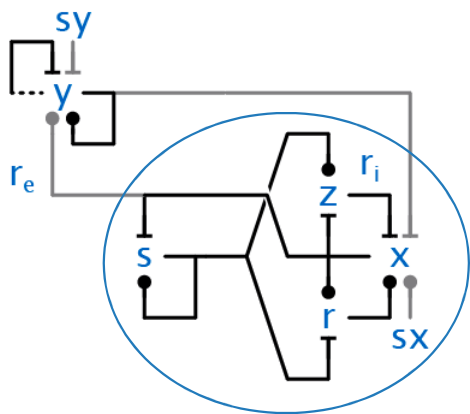
DOI:10.1038/ncomms2062

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

A new cell cycle switch candidate: GW

- Will it work in the normally-wired oscillator?

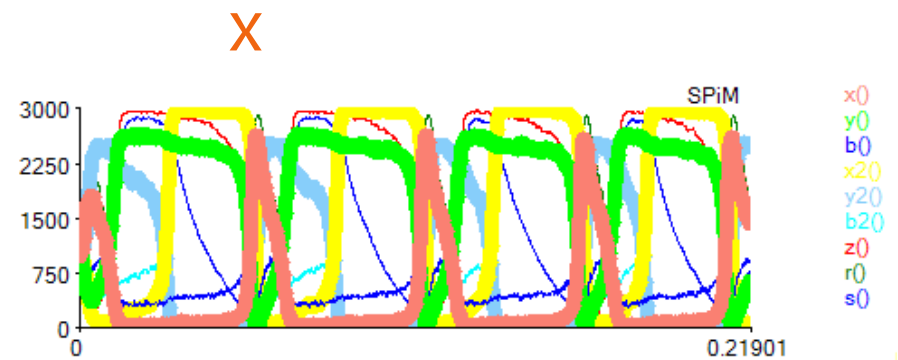
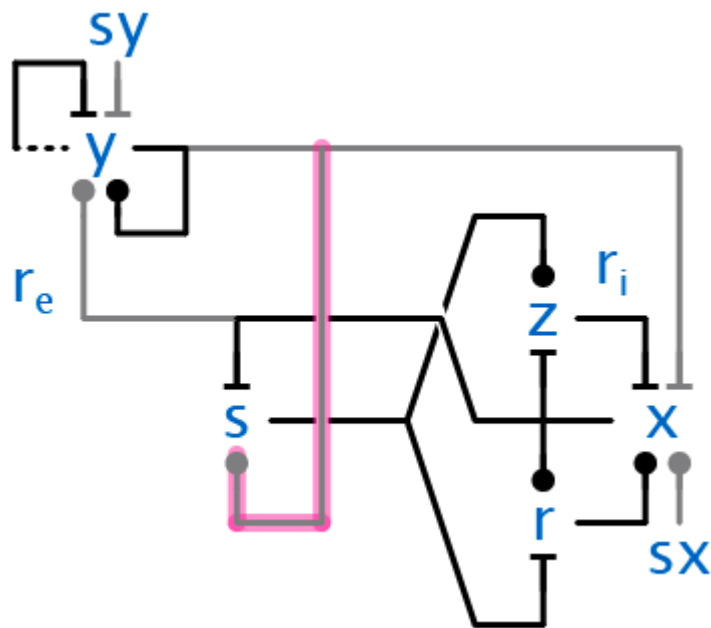


- Absolutely not! ☹️

- The x stable state is just too strong: a high x will shut down s completely; which means that r will be fully on, and it in turn will reinforce x fully. And y can never be strong enough to push down x when x - r are in such a strong mutual feedback. No amount of fiddling seems to give enough control on that situation.

However this will

- Put s under control of y so it can undermine x

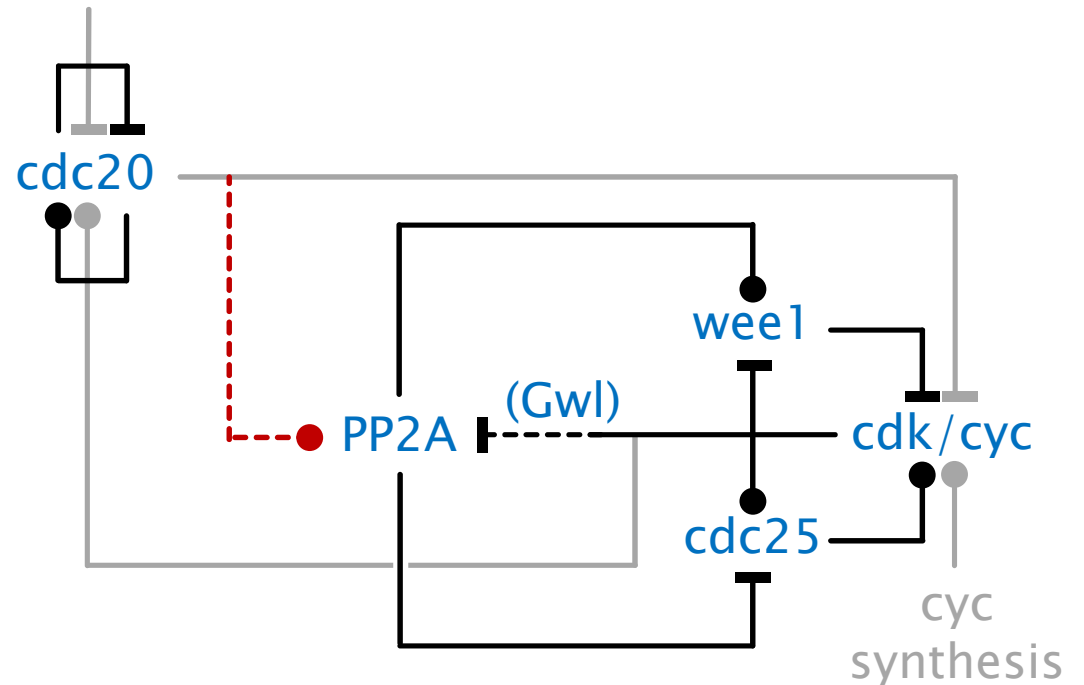


Robust full-on oscillation with all-default parameters
(all black rates 1.0, all gray rates 0.5, all initial quantities equal)

Suggests a new interaction

- Either Gwl or PP2A or something along that path must be under control of cdc20.
- There are some hints in the literature that this may be the case, but no direct experimental validation.

checkpoint



Part II: What is network structure
really telling us about kinetics?

An Analytical Theory of Network Emulation (with thanks to David Soloveichik)

- So far, our evidence is empirical
 - Although based on numerical simulations and covering both kinetic and steady state behavior
- Analytical evidence is harder to obtain
 - The proof techniques 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)

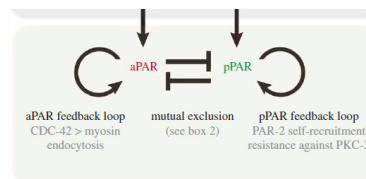
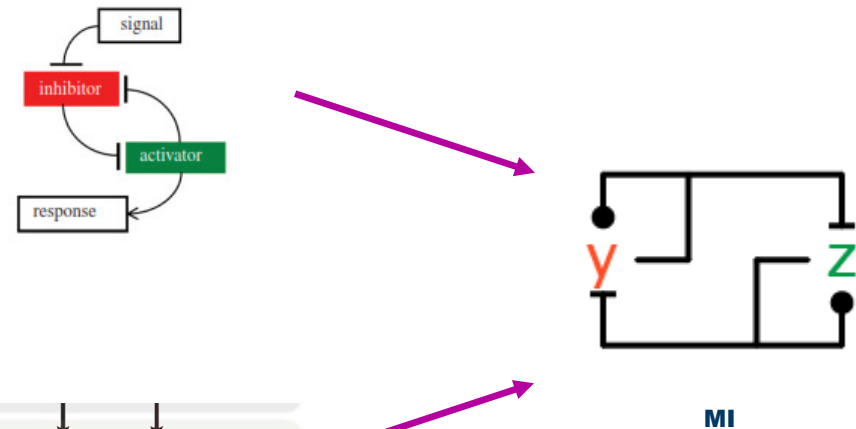
Mutual Inhibition

- A recent paper suggests that 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):



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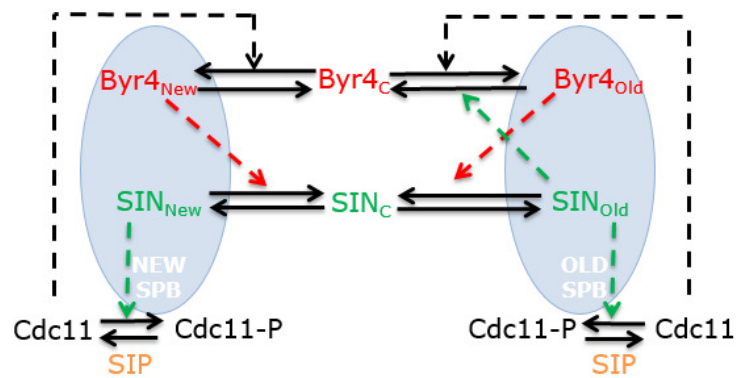
The PAR network: redundancy and robustness in a symmetry-breaking system

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

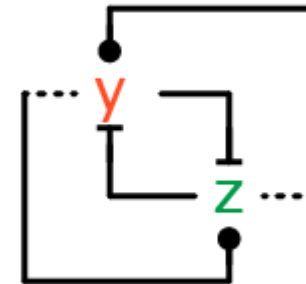
¹Temasek LifeSciences Laboratory, ²Mechanobiology Institute, and ³Department of Biological Sciences, National University of Singapore, 1 Research Link, Singapore 117604, Republic of Singapore
⁴Department of Molecular Biology and Genetics and HHMI, Johns Hopkins University School of Medicine, Baltimore,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

Septation Initiation

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



SIN inhibiting Byr4,
absence of SIN promoting Byr4



New Cell Cycle Network

- A recent paper presents a more complete view of the cell cycle switch
- N.B. “phosphorylation network dynamics” 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

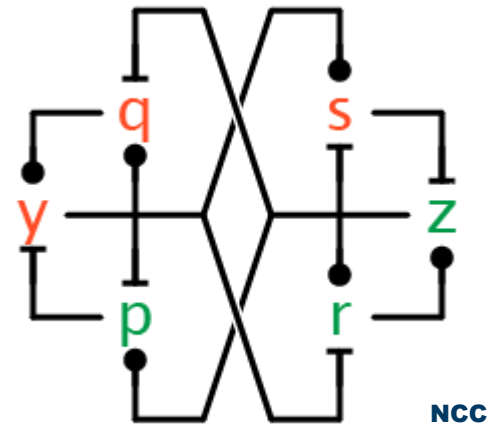
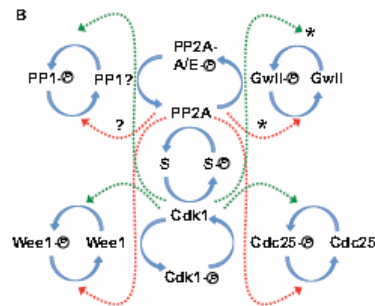
²Institute of Genetics and Development of Rennes, CNRS UMR 6290, 35043 Rennes, France

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

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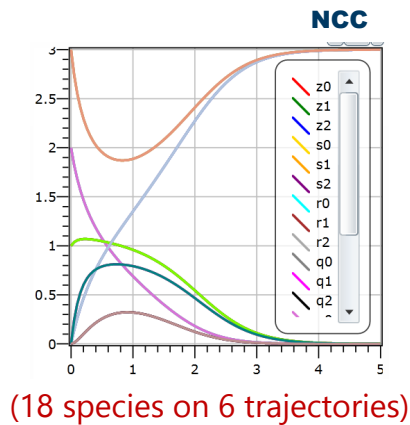
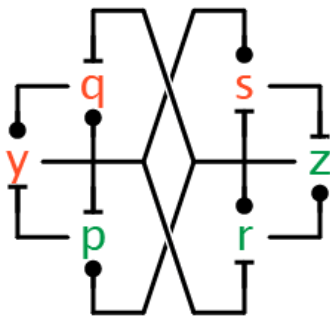
[†]These authors contributed equally to this work.

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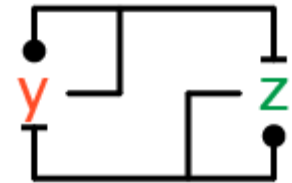
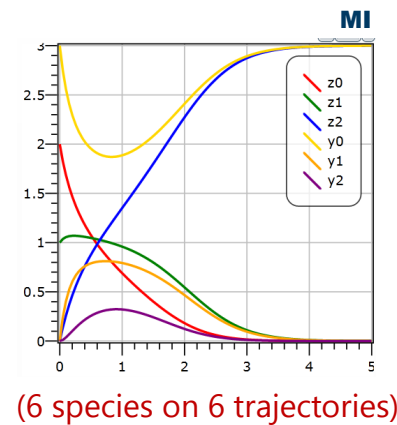


Network Emulation: NCC to MI

- For *any* initial state of MI we can find *some* initial state of NCC (actually by *copying* the state of MI) such that NCC *exactly* emulates MI



z,r,p \rightarrow z
y,q,s \rightarrow y



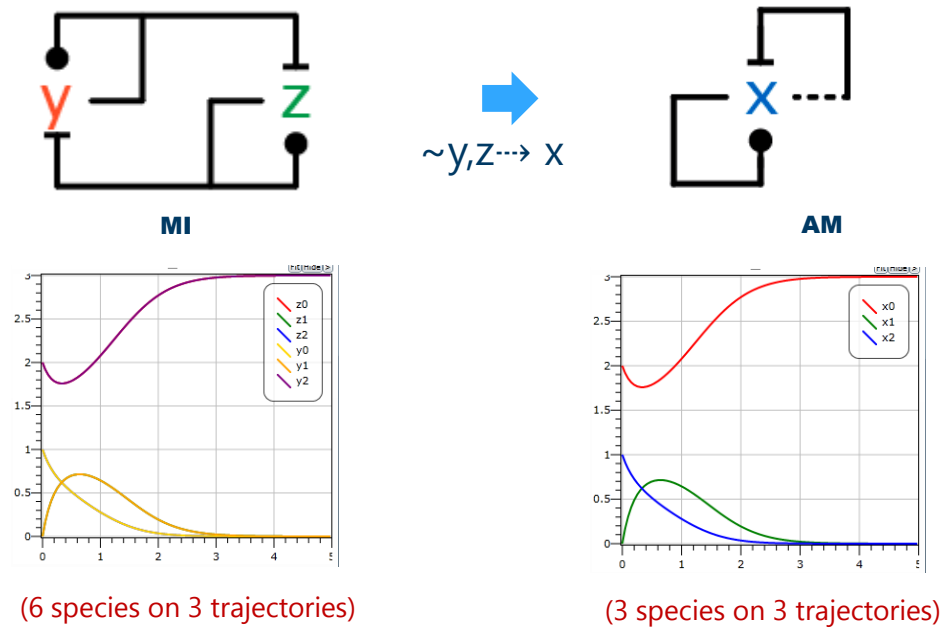
initialize z,r,p,
identically to z;

initialize y,q,s
identically to y

- Why does this work so well?

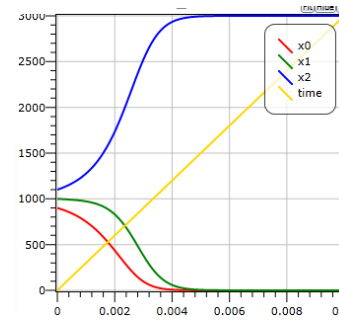
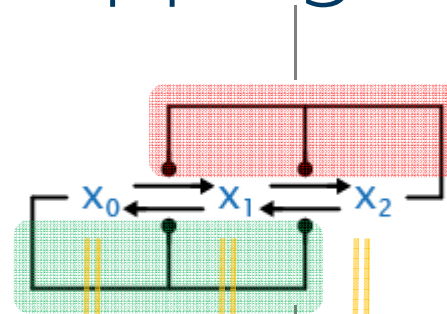
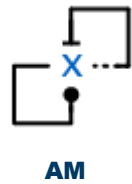
Network Emulation: MI to AM

- For chosen initial conditions of MI, the (6) trajectories of MI emulate those (3) of AM:



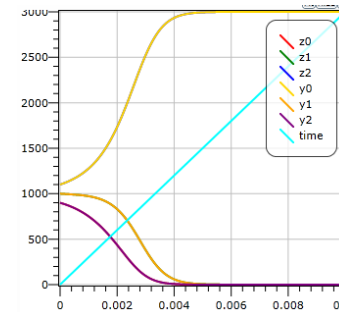
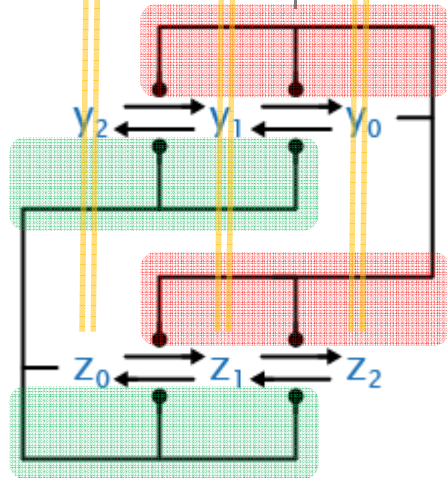
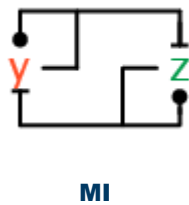
initialize $\sim y, z$,
identically to x

MI to AM mapping in detail



any initial conditions

homomorphic mapping

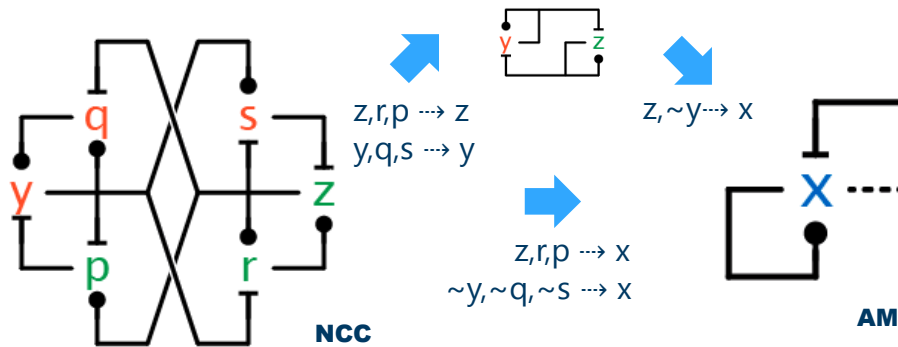


initial conditions:

$$\begin{aligned} z_0 &= y_2 = x_0 \\ z_1 &= y_1 = x_1 \\ z_2 &= y_0 = x_2 \end{aligned}$$

Network Emulation Composes: NCC to AM

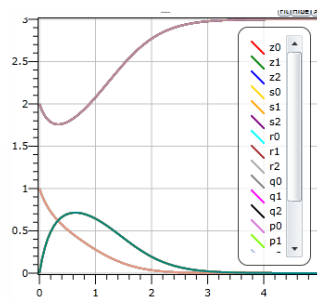
- For chosen initial conditions of NCC, the (18) trajectories of NCC emulate those (3) of AM



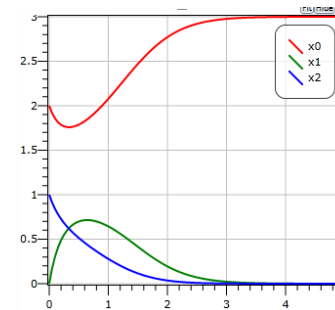
This works also for GW, but not for the original CC.

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

And for *any* rates of AM.
Why?



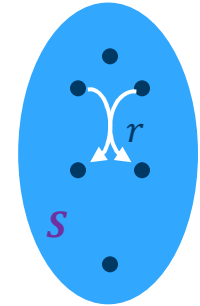
(18 species on 3 trajectories)



(3 species on 3 trajectories)

Chemical Reaction Networks

- A CRN is a pair (S, R) where
 - $S = \{s_1, \dots, s_n\}$ is a finite set of *species*
 - $R = \{r_1, \dots, r_m\}$ is a finite set of *reactions* over S
- Reactions $r = (\rho, \pi, k)$ written $\sum_{S \in S} \rho_S \cdot S \xrightarrow{k} \sum_{S \in S} \pi_S \cdot S$



- Ex.: $r = 2A + B \xrightarrow{k} A + 3C$
- with $\rho_A = 2, \rho_B = 1, \rho_C = 0$ *reactant stoichiometric numbers*
 $\pi_A = 1, \pi_B = 0, \pi_C = 3$ *product stoichiometric numbers*

- The *stoichiometry* of a species s in a reaction r is:

$$\eta(s, (\rho, \pi, k)) = \pi_s - \rho_s \quad \text{net stoichiometry} \quad \eta(A, r) = -1$$

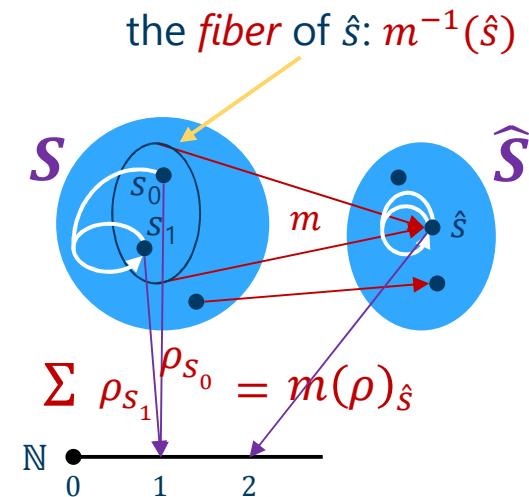
$$\varphi(s, (\rho, \pi, k)) = k \cdot (\pi_s - \rho_s) \quad \text{(instantaneous) stoichiometry} \quad \varphi(A, r) = -k$$

Species Maps and Reaction Maps

- A *species map* is a map $m \in S \rightarrow \hat{S}$
 - Ex: $m(s_0) = m(s_1) = \hat{s}$
- It induces a canonical *reaction map* $R \rightarrow \hat{R}$
 - Ex: $m(s_0 + s_1 \rightarrow^1 s_1) = 2\hat{s} \rightarrow^1 \hat{s}$
- Where $m(\rho, \pi, k) = (m(\rho), m(\pi), k)$
- And $m(\rho)$ (similarly $m(\pi)$) is the sum over fibers:

$$m(\rho)_{\hat{s}} = \sum_{s \in m^{-1}(\hat{s})} \rho_s$$

in case two species in the same reaction are mapped to the same species.



CRN Isomorphisms

- A *CRN morphism* is a map $m \in (S, R) \rightarrow (\hat{S}, \hat{R}) = (m_S, m_R)$ with $m_S \in S \rightarrow \hat{S}$ and $m_R \in R \rightarrow \hat{R}$.
- A *CRN isomorphism* $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a morphism made of two bijections on S and R that *agree on stoichiometric numbers and rate*:

$$m_R(\rho, \pi, k) = (m_S(\rho), m_S(\pi), k)$$

- As a consequence they also *agree on stoichiometry*:

$$\varphi(s, r) = \varphi(m_S(s), m_R(r))$$

- But what if m is not injective or surjective on species or reactions?
 - We need to generalize "agreement on stoichiometry" to such cases.

CRN Homomorphisms

- $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a *CRN homomorphism* if $m_{\mathcal{R}}$ is determined by m_S :

$$m_{\mathcal{R}}(\rho, \pi, k) = (m_S(\rho), m_S(\pi), k)$$

- Ex:

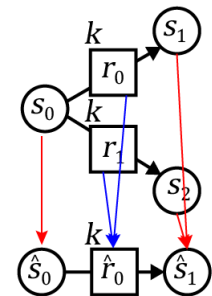
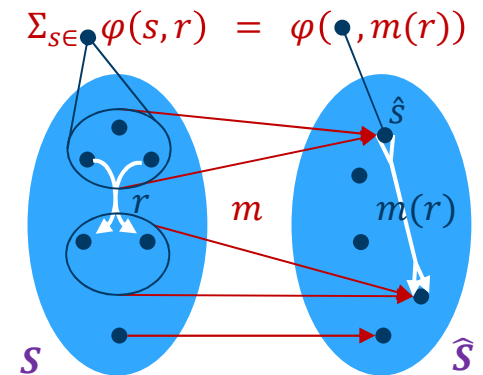
$$r_0: m_{\mathcal{R}}(s_0, s_1, k) = (\hat{s}_0, \hat{s}_1, k) = (m_S(s_0), m_S(s_1), k)$$

$$r_1: m_{\mathcal{R}}(s_0, s_2, k) = (\hat{s}_0, \hat{s}_1, k) = (m_S(s_0), m_S(s_2), k)$$

- It implies that for each reaction it preserves stoichiometry summed over species fibers

$$\forall \hat{s} \in \hat{S}. \forall r \in R. \sum_{s \in m^{-1}(\hat{s})} \varphi(s, r) = \varphi(\hat{s}, m(r))$$

- But $\varphi(s_0, r_0) + \varphi(s_0, r_1) = -2k \neq -1k = \varphi(\hat{s}_0, \hat{r}_0)$ (see next slide)



Homomorphism
(but not stoichiomorphism)

CRN Stoichiomorphisms

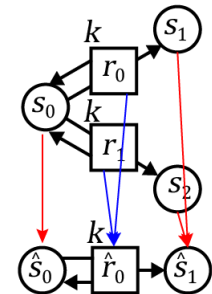
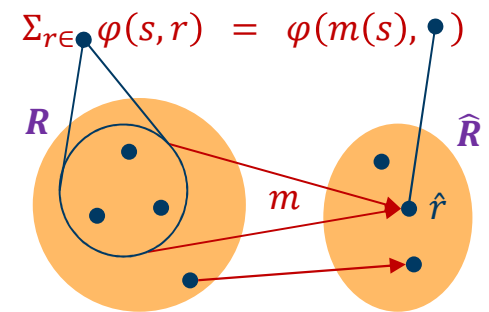
- $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a *CRN stoichiomorphism* if for each species it preserves stoichiometry summed over reaction fibers

$$\forall s \in S. \forall \hat{r} \in \hat{R}. \sum_{r \in m^{-1}(\hat{r})} \varphi(s, r) = \varphi(m(s), \hat{r})$$

- This condition can be checked over the *syntax* of CRNs, without any consideration of their kinetics
- Ex:

$$\begin{aligned} s_0, \hat{r}_0: \quad & \varphi(s_0, r_0) + \varphi(s_0, r_1) = 0 = \varphi(\hat{s}_0, \hat{r}_0) \\ s_1, \hat{r}_0: \quad & \varphi(s_1, r_0) + \varphi(s_1, r_1) = 1k = \varphi(\hat{s}_1, \hat{r}_0) \\ s_2, \hat{r}_0: \quad & \varphi(s_2, r_0) + \varphi(s_2, r_1) = 1k = \varphi(\hat{s}_1, \hat{r}_0) \end{aligned}$$

- We will show that existence of a stoichiomorphism implies *identical network kinetics* (in certain conditions).



Homomorphism and stoichiomorphism.

CRN Morphism Conditions

- Homomorphism consequence:

$$\forall \hat{s} \in \hat{S}. \forall r \in R. \sum_{s \in m^{-1}(\hat{s})} \varphi(s, r) = \varphi(\hat{s}, m(r))$$

- Stoichiomorphism condition:

$$\forall s \in S. \forall \hat{r} \in \hat{R}. \sum_{r \in m^{-1}(\hat{r})} \varphi(s, r) = \varphi(m(s), \hat{r})$$

- If m is an isomorphism (injective and surjective, with singleton fibers) then they both reduce to the isomorphism consequence:

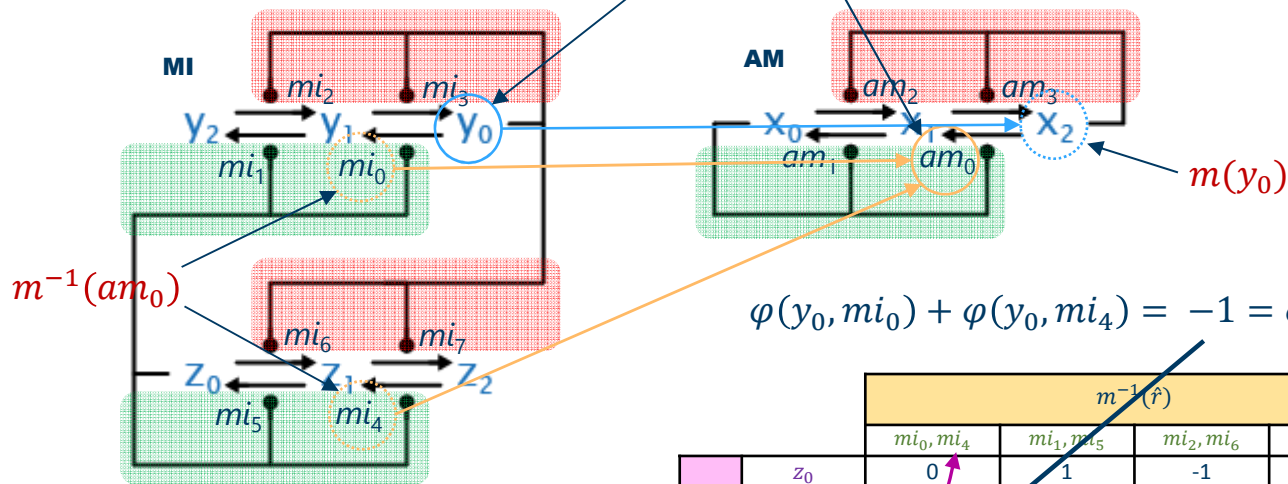
$$\forall s \in S. \forall r \in R. \varphi(s, r) = \varphi(m(s), m(r))$$

- But we will be typically interested in mappings that “simplify” networks and that are at least not injective.

Checking the Stoichiomorphism Condition

$m \in \text{MI} \rightarrow \text{AM}$

$$\forall s \in S. \forall \hat{r} \in \hat{R}. \sum_{r \in m^{-1}(\hat{r})} \varphi(s, r) = \varphi(m(s), \hat{r})$$



All unit rates (for simplicity)

This is both a homomorphism and a stoichiomorphism

$$\varphi(y_0, mi_0) + \varphi(y_0, mi_4) = -1 = \varphi(x_2, am_0)$$

		$m^{-1}(\hat{r})$				$m(s)$
		mi_0, mi_4	mi_1, mi_5	mi_2, mi_6	mi_3, mi_7	
$\forall s \in \text{MI}$	z_0	0	1	-1	0	x_0
	z_1	1	-1	1	-1	x_1
	z_2	-1	0	0	1	x_2
	y_0	-1	0	0	1	x_2
	y_1	1	-1	1	-1	x_1
	y_2	0	1	-1	0	x_0
		am_0	am_1	am_2	am_3	
		$\forall \hat{r} \in \text{AM}$				

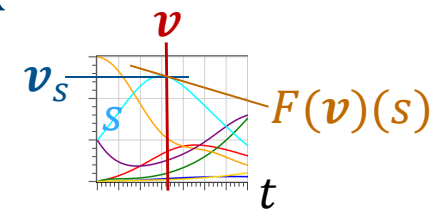
CRN Kinetics

- A *state* of a CRN (S, R) is a vector of concentrations for each species: $\mathbf{v} \in \mathbb{R}^{+S}$.
- The *mass action* $[r] \in \mathbb{R}^{+S} \rightarrow \mathbb{R}^+$ of a reaction $r \in R$ is:

$$[r]_{\mathbf{v}} = [(\rho, \pi, k)]_{\mathbf{v}} = \prod_{s \in S} v_s^{\rho_s} = \mathbf{v}^\rho$$

- The *differential system* of a CRN (S, R) is the map $F \in \mathbb{R}^{+S} \rightarrow \mathbb{R}^S$
(for each state, gives the differential of concentration for each species):

$$F(\mathbf{v})(s) = \sum_{r \in R} \varphi(s, r) \cdot [r]_{\mathbf{v}}$$



- Normally written as a system of concentration ODEs, integrated over time:

$$\frac{dv_s}{dt} = F(\mathbf{v})(s) = \sum_{(\rho, \pi, k) \in R} k \cdot (\pi_s - \rho_s) \cdot \mathbf{v}^\rho$$

Kinetic Emulation

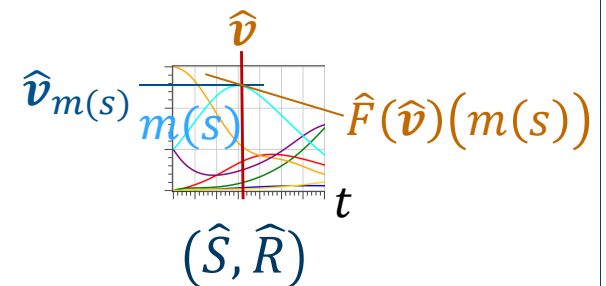
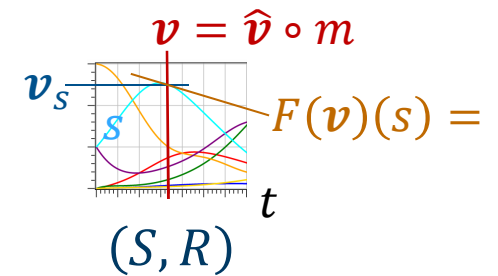
- A map $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a *CRN emulation* if the following holds for the respective differential systems F, \hat{F} :

$$\forall \hat{v} \in \mathbb{R}^{+\hat{S}}. \forall s \in S. F(\hat{v} \circ m)(s) = \hat{F}(\hat{v})(m(s))$$

(the derivative of s in state $\hat{v} \circ m$ is equal to the derivative of $m(s)$ in state \hat{v})

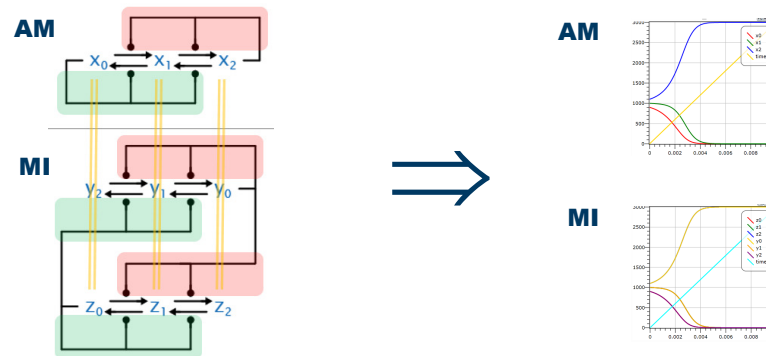
- It follows that for *any* initial state \hat{v} of (\hat{S}, \hat{R}) there is an initial state $v (= \hat{v} \circ m)$ of (S, R) such that the trajectory of any s in (S, R) is identical to (*emulates*) the trajectory of $m(s)$ in (\hat{S}, \hat{R}) .

(With minor caveats if m is not surjective.)



Emulation Theorem

- Theorem: If m is a CRN homomorphism and stoichiomorphism then it is a CRN emulation.



that is, for *any initial conditions* we can match trajectories.

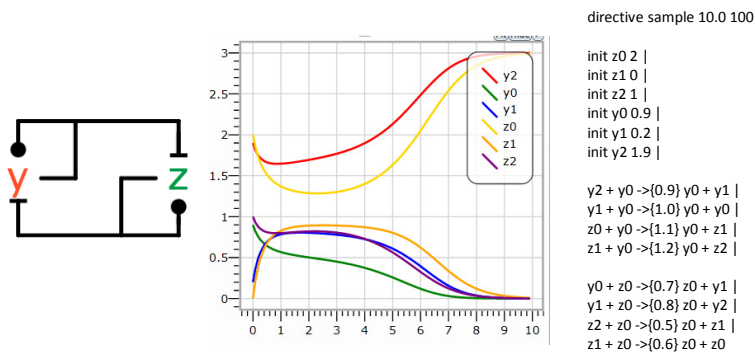
- Actually, m need not be a homomorphism for this to hold: it is enough for m to be a *reactant morphism* and a stoichiomorphism. A reactant morphism agrees with the species map on the reactant species, but allows rates and product species to disagree. This allows a wider range of network mappings that preserve kinetics.

Change of Rates Theorem

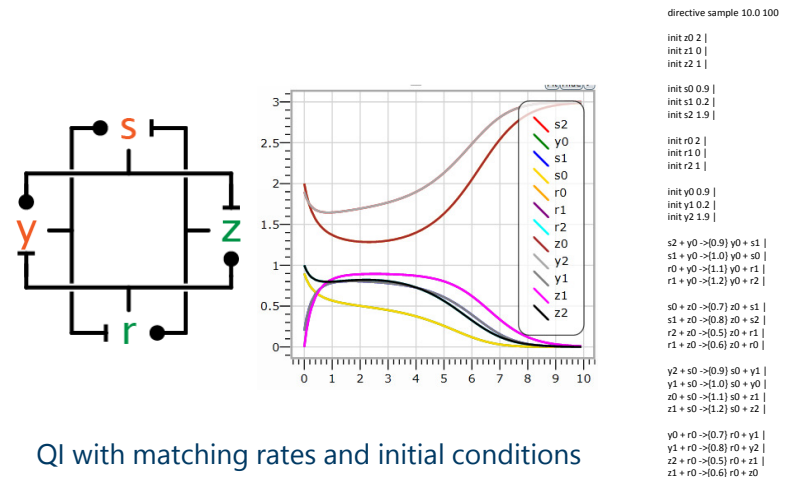
- A *change of rates* for (S, R) is bijection $\iota \in (S, R) \rightarrow (S, R')$ such that $\iota(\rho, \pi, k) = (\rho, \pi, k')$.
- Theorem: If $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a stoichiomorphism, then for *any* change of rates $\hat{\iota}$ of (\hat{S}, \hat{R}) there is a change of rates ι of (S, R) such that $\hat{\iota} \circ m \circ \iota^{-1}$ is a stoichiomorphism.
 - In fact, ι changes rates by the ratio with which $\hat{\iota}$ changes rates:
$$\iota(\rho, \pi, k) = \left(\rho, \pi, k \cdot \frac{\hat{k}'}{\hat{k}}\right)$$
 where $m(\rho, \pi, k) = (\hat{\rho}, \hat{\pi}, \hat{k})$ and $\hat{\iota}(\hat{\rho}, \hat{\pi}, \hat{k}) = (\hat{\rho}, \hat{\pi}, \hat{k}')$.
- Corollary: If $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a stoichiomorphism and homomorphism, then for *any change of rates* $\hat{\iota}$ of (\hat{S}, \hat{R}) there is a change of rates ι of (S, R) such that $\hat{\iota} \circ m \circ \iota^{-1}$ is an emulation.

Any Rates, Any Initial Conditions

- A stoichiomorphism $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ that is also a homomorphism, determines an emulation for any choice of rates of (\hat{S}, \hat{R}) .
- Those emulations can match any initial conditions of any choice of rates of (\hat{S}, \hat{R}) with some initial conditions of some choice of rates of (S, R) .

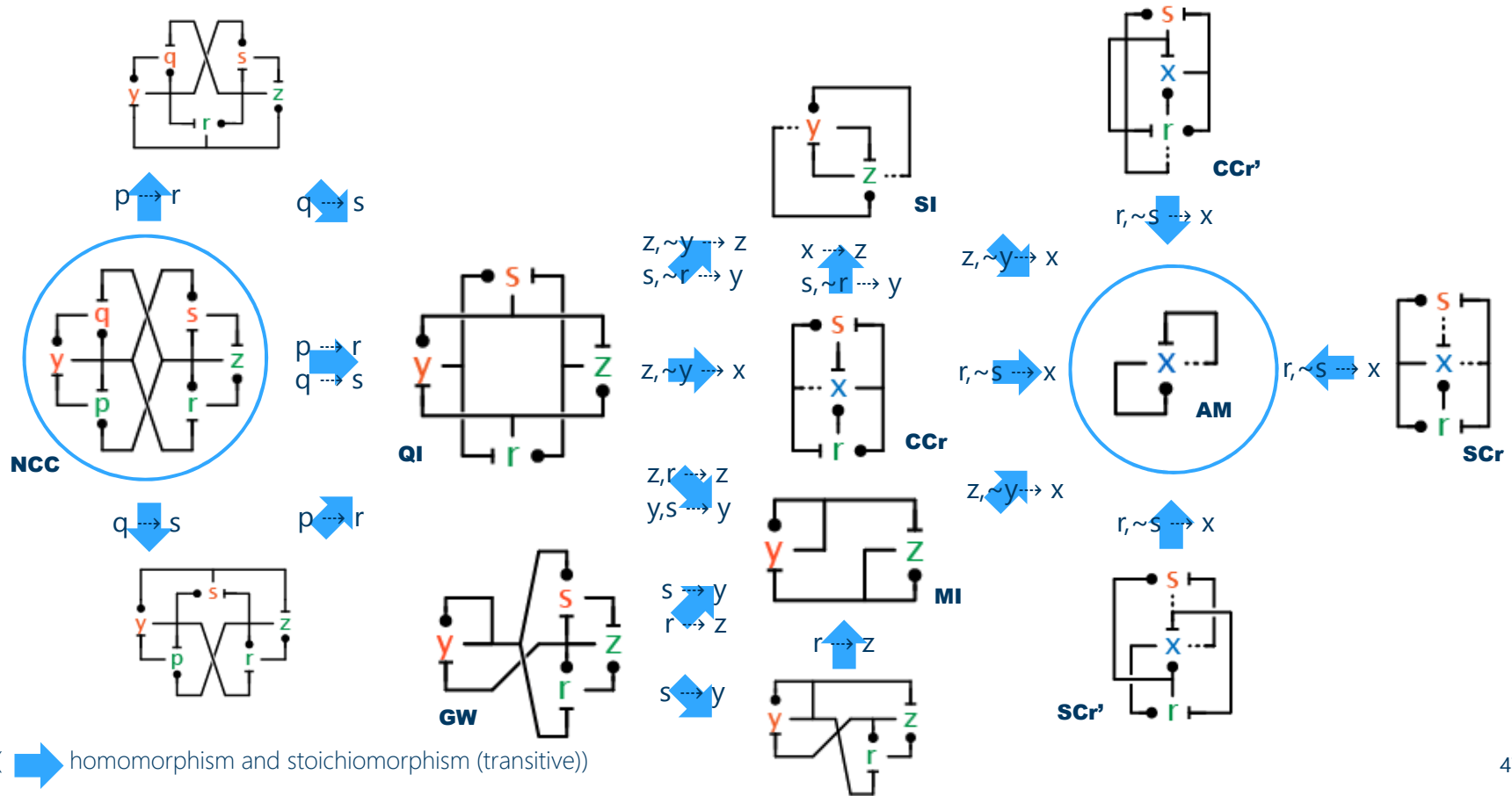


MI with completely heterogeneous rates and initial conditions

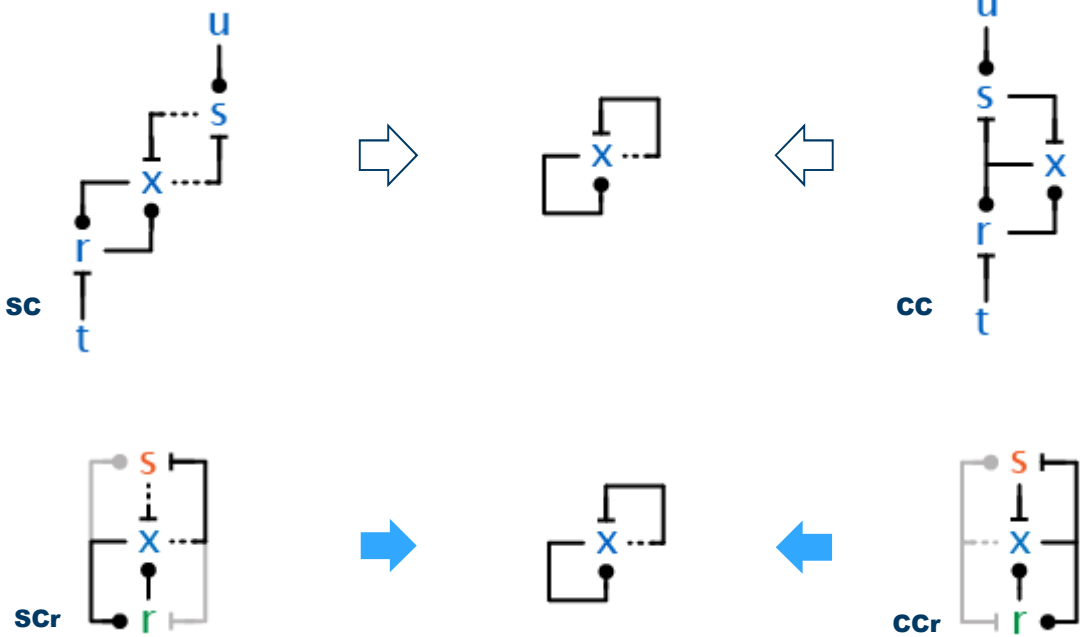


QI with matching rates and initial conditions

Stoichiomorphism Zoo



Old Friends

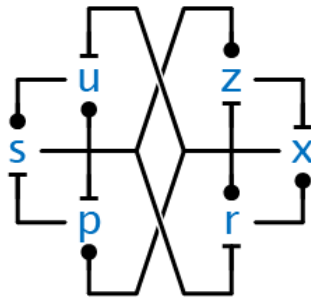


Interpretation of Stoichiomorphism

- Ignorance about initial conditions
 - We may not know the concentrations of species in the more complex network, but at least we know that if they satisfy certain conditions, then it behaves like the simpler network.
- Neutral paths in network space (evolution)
 - If an evolutionary event happens to be a stoichiomorphism, or close to it, it will not be immediately selected against, because it is “kinetically neutral”.
 - This allows the network to increase its complexity without kinetic penalty.
 - Later, the extra degrees of freedom can lead to kinetic differentiation.
 - But meanwhile, the organism can explore variations of network structure.
- Relationship to abstraction / coarse-graining
 - Stoichiomorphism are not about abstractions that preserve behavior, on the contrary, they are about *concretions* that preserve behavior.

Conclusions

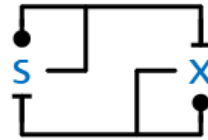
- The cell cycle switch *can exactly* emulate AM



NCC

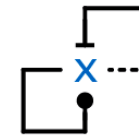
(New) cell cycle switch

emulates:



MI

emulates:



AM

Approximate majority
algorithm

- Nature likes a good algorithm!