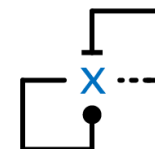
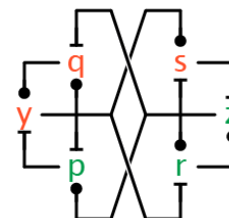


Morphisms of Reaction Networks

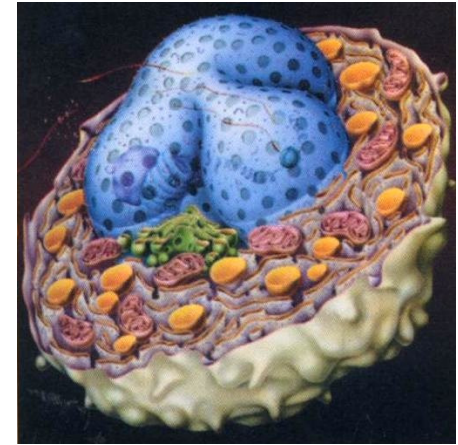
Luca Cardelli, Microsoft Research & Oxford University

IMT Lucca, 2014-07-08



Motivation

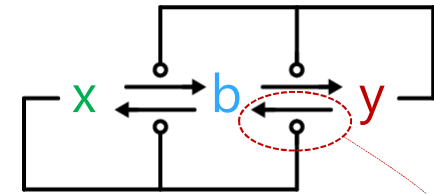
- Give substance to the claim that “cells compute”
 - Yes, but *what* do they compute?
- Catch nature red-handed in the act of running a computational task
 - Something that a computer scientist would recognize as an *algorithm*



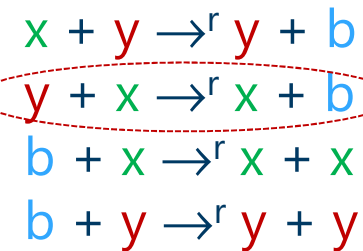
A Consensus Algorithm

- Population Protocols
 - 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)
- A Population Consensus Problem
 - Find which state x or y is in majority in the population
 - By converting the *whole* population to x or y
- Approximate Majority (AM) Algorithm
 - Uses a 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 is $O(n \log n) \Rightarrow$ fast (optimal)
 - Correct outcome if the initial disparity is $\omega(\sqrt{n} \log n) \Rightarrow$ robust
 - In parallel time, converges in $O(\log n)$

catalysis \odot



chemical reaction network

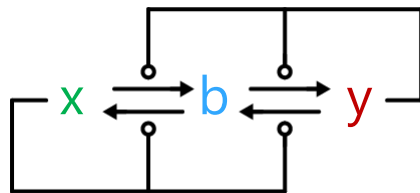


Dana Angluin · James Aspnes · David Eisenstat

A Simple Population Protocol for Fast Robust Approximate Majority

A Biological Implementation

Approximate Majority (AM)



Bistable
Even when $x=y$ (stochastically)

Fast
 $O(\log n)$ convergence time

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

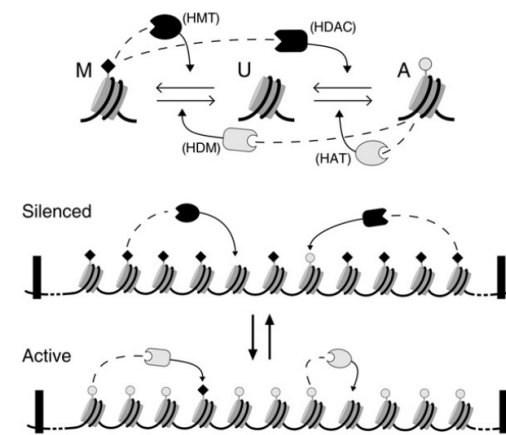


Figure 1. Basic Ingredients of the Model

Theory

Theoretical Analysis of Epigenetic
Cell Memory by Nucleosome Modification

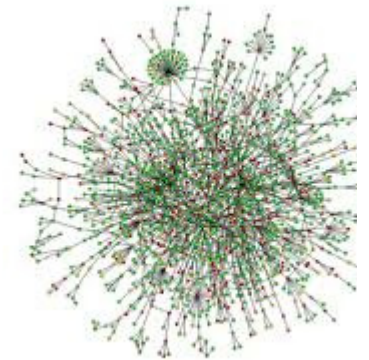
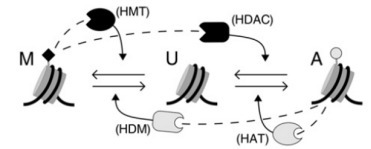
Jan B. Dückel,^{1,2} Mikha A. Mochlyakov,¹ Kim Sjögreen,^{1,2} and Genevieve Thori
¹Center for Molecular Life, Niels Bohr Institute, Copenhagen Ø, Denmark
²Department of Molecular and Biomedical Science, University of Adelaide, SA 5005, Australia
³Department of Molecular Biology, University of Copenhagen, Copenhagen N, Denmark
Correspondence: jduckel@nbi.dk
DOI: 10.1016/j.cel.2007.02.012

Cell

2007

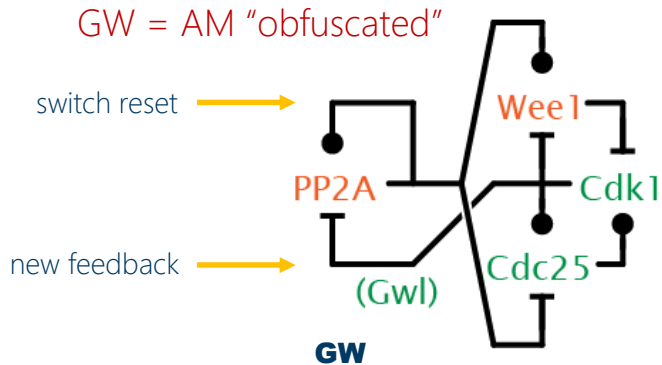
Motivation (cont'd)

- 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 convoluted and... approximate
 - Like finding an algorithm in a haystack...



Obfuscated Implementations

- GW is a better cell cycle switch than [the traditional switch]



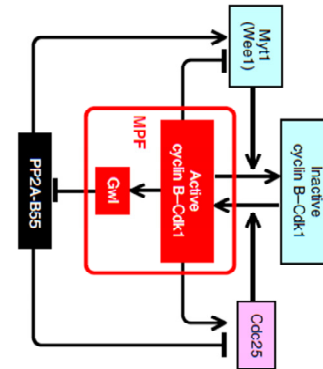
The Cell Cycle Switch Computes Approximate Majority

SUBJECT AREAS:
COMPUTATIONAL
BIOLOGY

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

Sep 2012

- GW is how the cell cycle switch "really works"



ARTICLE

Received 6 Jul 2012 | Accepted 14 Aug 2012 | Published 11 Sep 2012

DOI:10.1038/ncomms2062

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

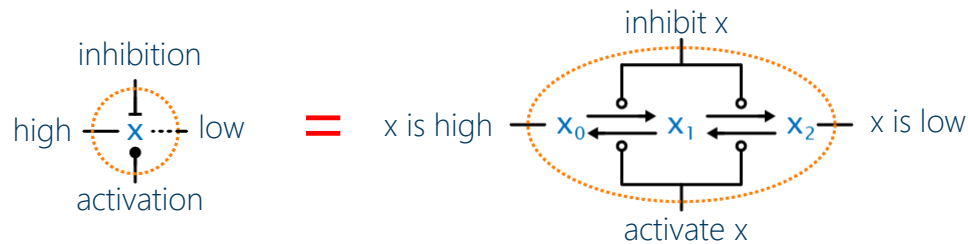
Masatoshi Hara^{1,†}, Yusuke Abe^{1,†}, Toshiaki Tanaka², Takayoshi Yamamoto^{1,†}, Eiichi Okumura¹ & Takeo Kishimoto¹

Sep 2012

Motivation (cont'd)

- When does a biologically messy network X “implement” some ideal algorithm Y?
 - Pushed coauthors into thinking about approximate stochastic bisimulation metrics for CTMCs
 - But they didn’t come back...
- Some networks behave similarly because “their ODEs are just equivalent” [David S.]
 - Epiphany! Forget stochastic! Forget approximate! When are CRNs “deterministically equivalent”?
 - Or better, when do trajectories of one CRN “collapse” into trajectories of another?
 - Much simpler! And can be solved on the *static structure* of CRNs as opposed to their kinetics.
 - Independently on rates and initial conditions (of one of the two networks).

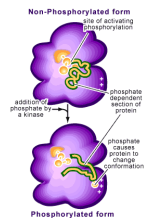
Influence Networks



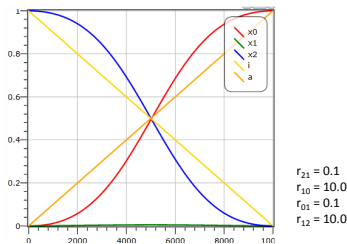
triplet motif

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

They actually implement a
Hill function of coefficient 2:



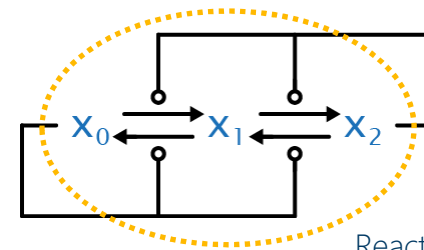
biological mechanism:
(e.g.:) multisite
phosphorylation



$\Gamma_{21} = 0.1$
 $\Gamma_{10} = 10.0$
 $\Gamma_{01} = 0.1$
 $\Gamma_{12} = 10.0$

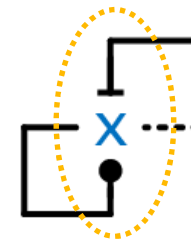
activation ●
inhibition T
catalysis ○

Approximate Majority



Reaction Network

=

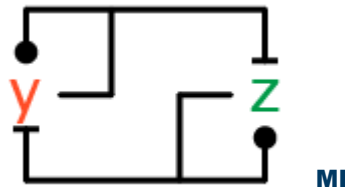


Influence Network

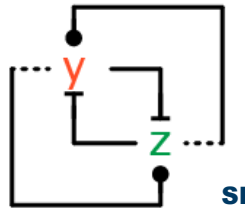
Biological Influence Networks

activation ●
inhibition ⊣

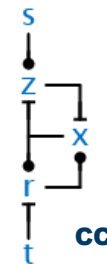
Mutual Inhibition & Self Activation



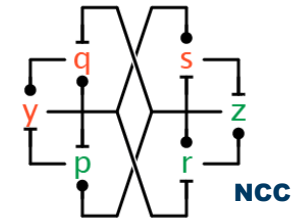
Mutual Inhibition & Mutual Anti-activation



Cell Cycle Switching



Better Switching



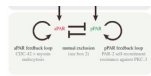
Cell cycle transitions

Molecular mechanisms creating bistable switches at cell cycle transitions
Amal Vengalil, P. K. Sirock, J. Tyson and Bela Novak
Open Access 2013, 15(1):74. Published 15 March 2013



Polarity establishment

PHILOSOPHICAL TRANSACTIONS OF THE ROYAL SOCIETY
The PAR network redundancy and robustness in a symmetry-breaking system
Forrás Máté^{1,2} and Gábor Szabó¹
¹Department of Biophysics, Theoretical Physics and Informatics, Institute of Physical Sciences, National University of Science and Technology, H-1525 Budapest, Hungary
²Department of Biology, University of Cambridge, Cambridge CB2 3EJ, UK



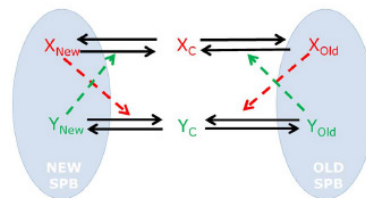
Gene networks

Construction of a genetic toggle switch in *Escherichia coli*

Timothy S. Gardner^{1,2}, Charles R. Cantor^{1,2} & James J. Collins^{1,2}



Septation Initiation



Dynamics of SIN Asymmetry Establishment

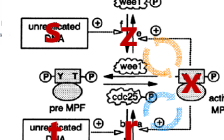
Anshu Rajan¹, Arno Paschke¹, Jun-Sung Cha², Daniel McCollum¹, Massimo Saito^{1,3}, Roland E. Cross^{1,2,3}, Ashwin Vignani¹, Gerd A. Blobel¹, Armin Eichler^{1,2,3}

The G₂/M cell cycle switch

Journal of Cell Science 116, 1033-1041 (2003)
Printed in Great Britain © The Company of Biologists Limited 2003

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

Bela Novak¹ and John J. Tyson²
¹Department of Biology, Virginia Polytechnic Institute
²Permanent address: Department of Agricultural Chemistry, Author for correspondence



Novak 144, 501 - 508 (5 April 1995), doi:10.1046/j.1365-3113.1995.144501.x

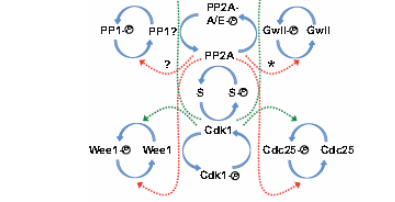
Universal control mechanism regulating onset of M-phase

PAUL NASEVIC
MCF Cell Cycle Drive, Department of Biochemistry, University of Oxford, South Parks Road, Oxford OX1 3YU, UK

The "new" cell cycle switch

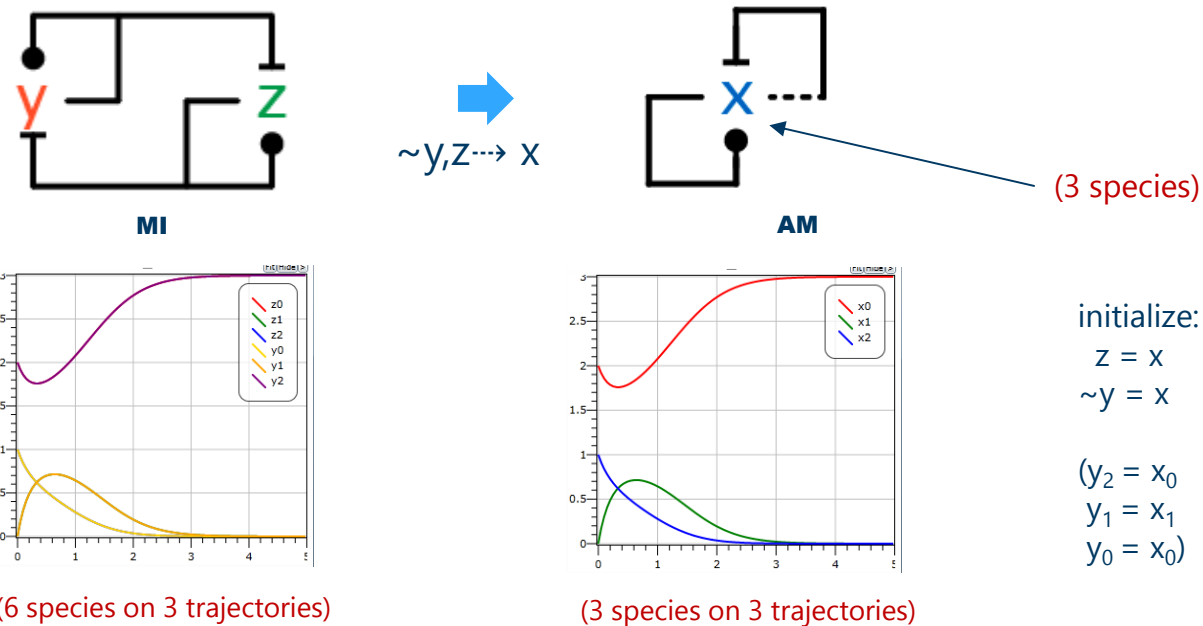
Phosphorylation network dynamics in the control of cell cycle transitions

Daniel Fisher^{1,2}, Liliana Krasinska^{1,2}, Damien Coudreuse^{1,2} and Bela Novak^{1,2}
¹Unité de Génétique Moléculaire de Montpellier, UMRI 1025, CNRS, IRSM, Université Montpellier I and II, 34293 Montpellier, France
²National Centre for Integrative Systems Biology, Department of Biochemistry, University of Oxford, South Parks Road, Oxford OX1 3YU, UK
These authors contributed equally to this work



Network Emulation: MI emulates AM

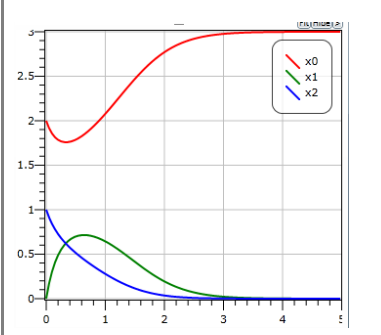
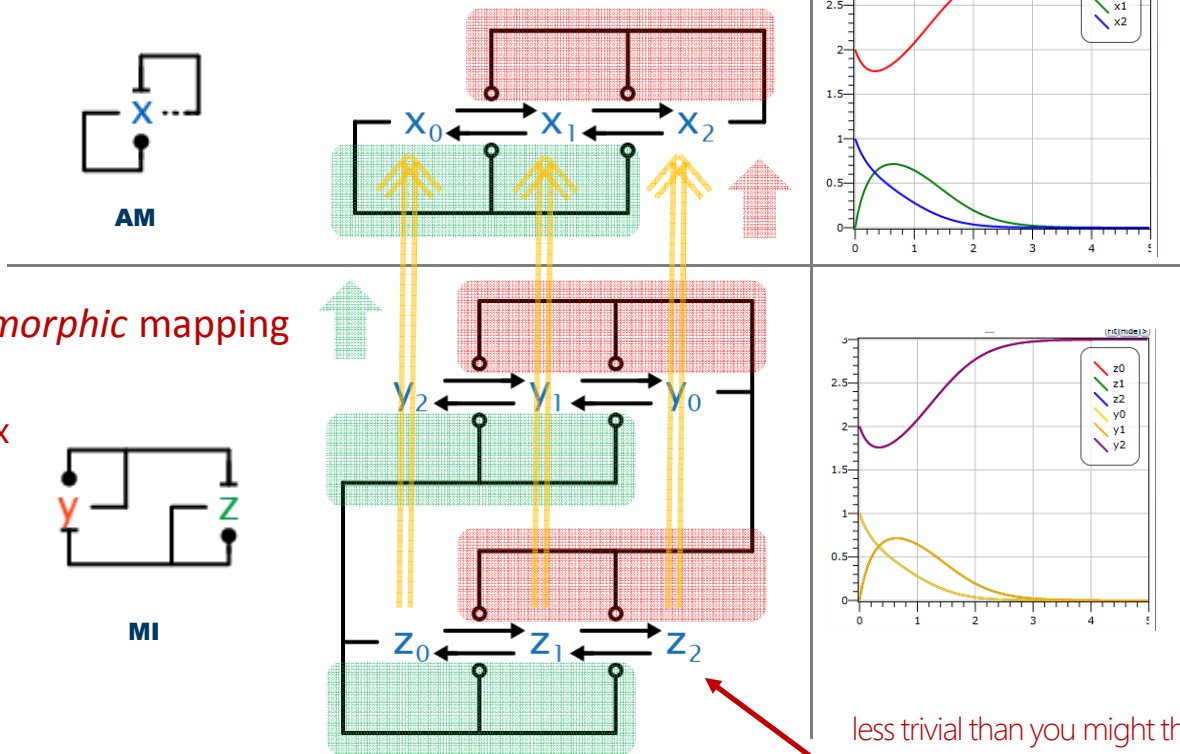
- For *any* rates and initial conditions of AM, we can find *some* rates and initial conditions of MI such that the (6) trajectories of MI retrace those (3) of AM:



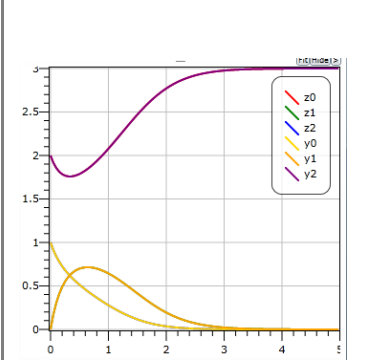
- How do we find these matching parameters? By a **network morphism!**

MI to AM Emulation: Network Morphism

A mapping of species and reactions



any initial conditions



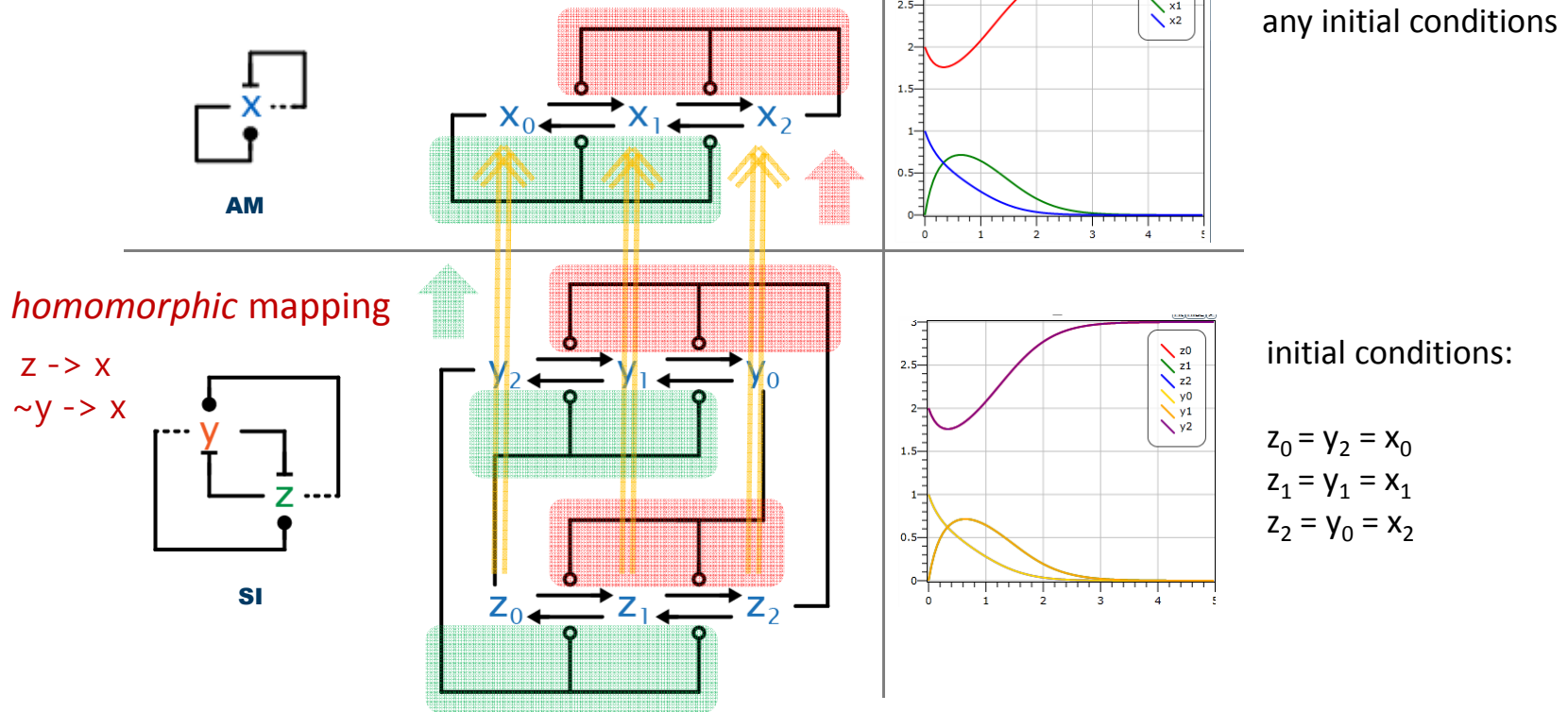
initial conditions:

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

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

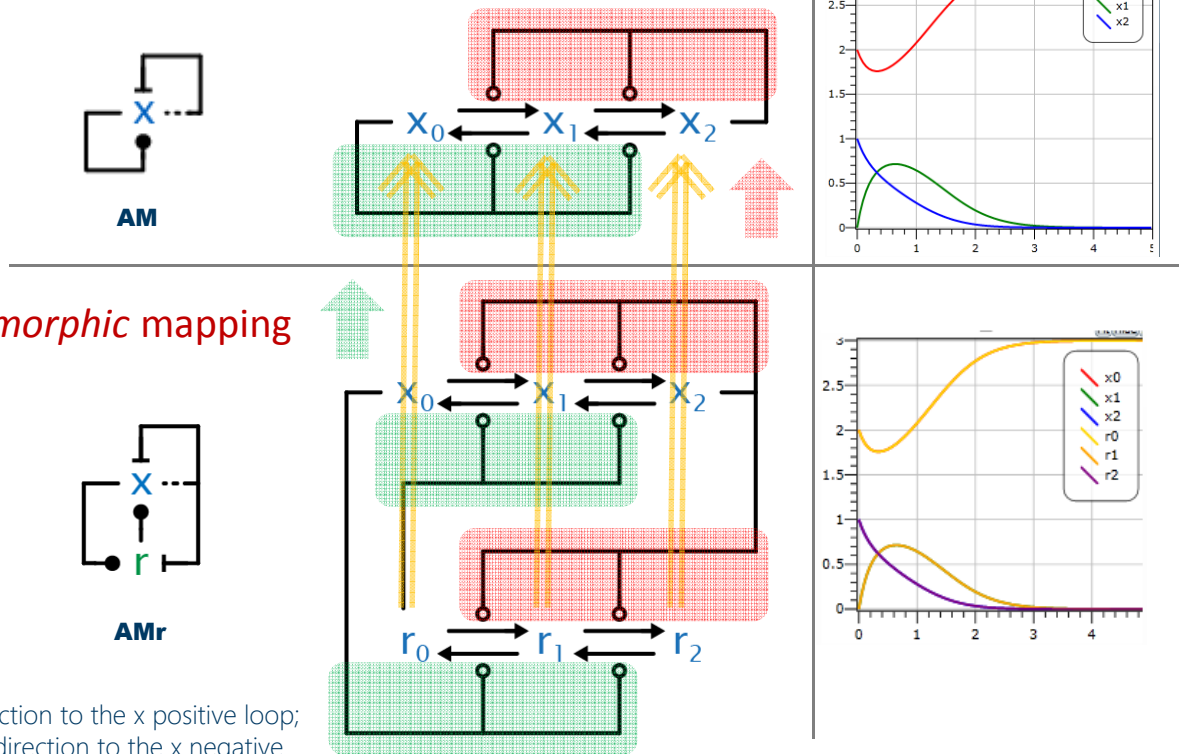
SI to AM Emulation: Network Morphism

A mapping of species and reactions



AMr to AM Emulation: Network Morphism

A mapping of species and reactions



any initial conditions

initial conditions:

$$r_0 = x_0 = x_0^{AM}$$

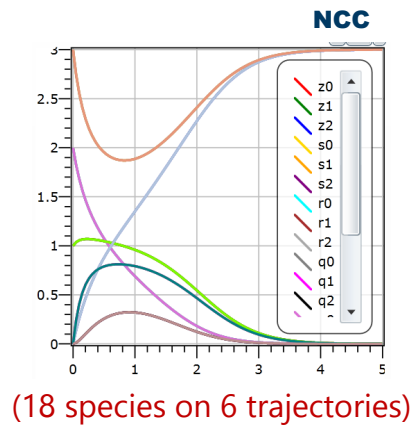
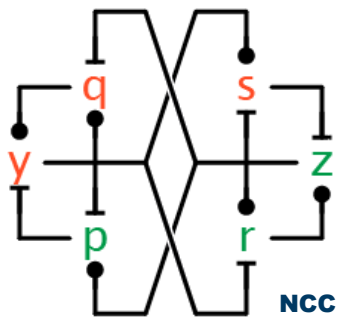
$$r_1 = x_1 = x_1^{AM}$$

$$r_2 = x_2 = x_2^{AM}$$

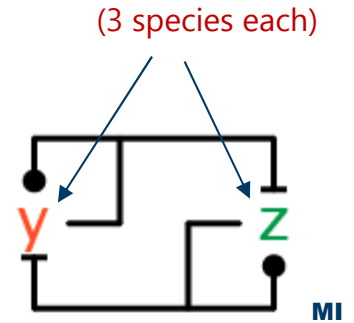
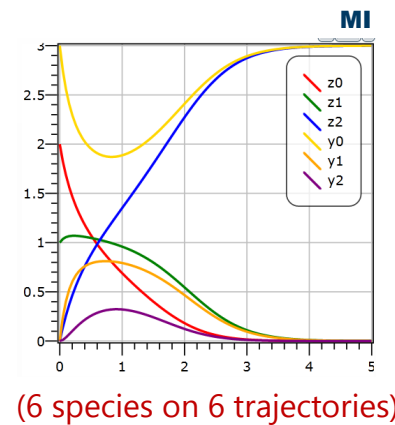
(AMr adds an indirection to the x positive loop; if we also add an indirection to the x negative loop, we obtain a prototypical cell cycle switch that also emulates AM: CCR)

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



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

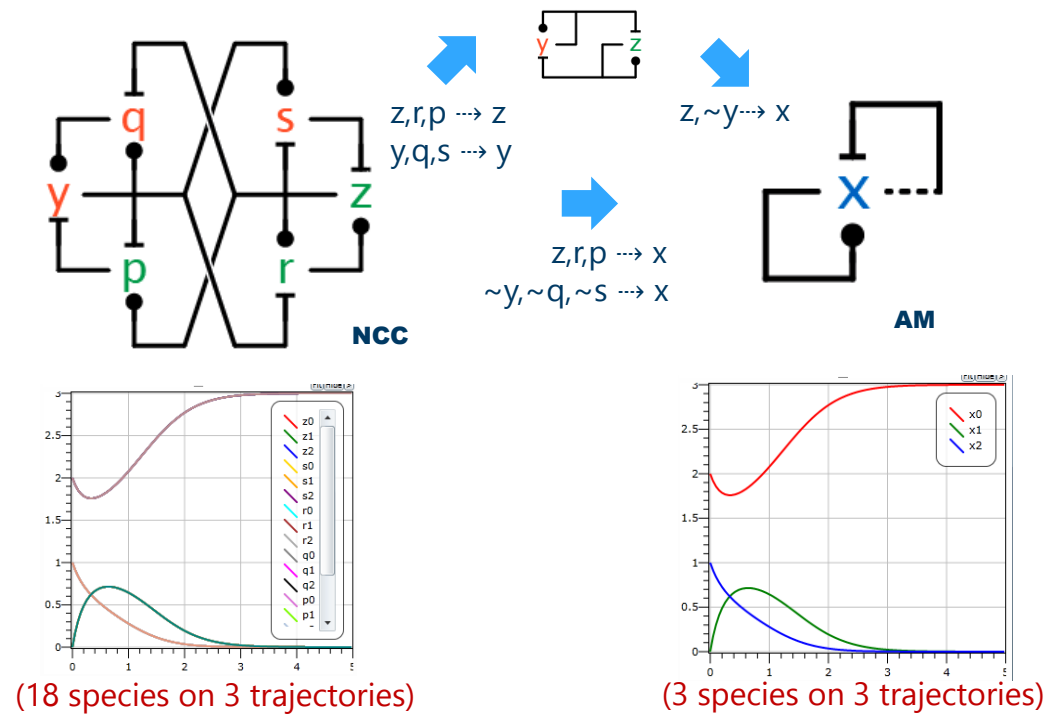


initialize
 $z, r, p = z$
 $y, q, s = y$

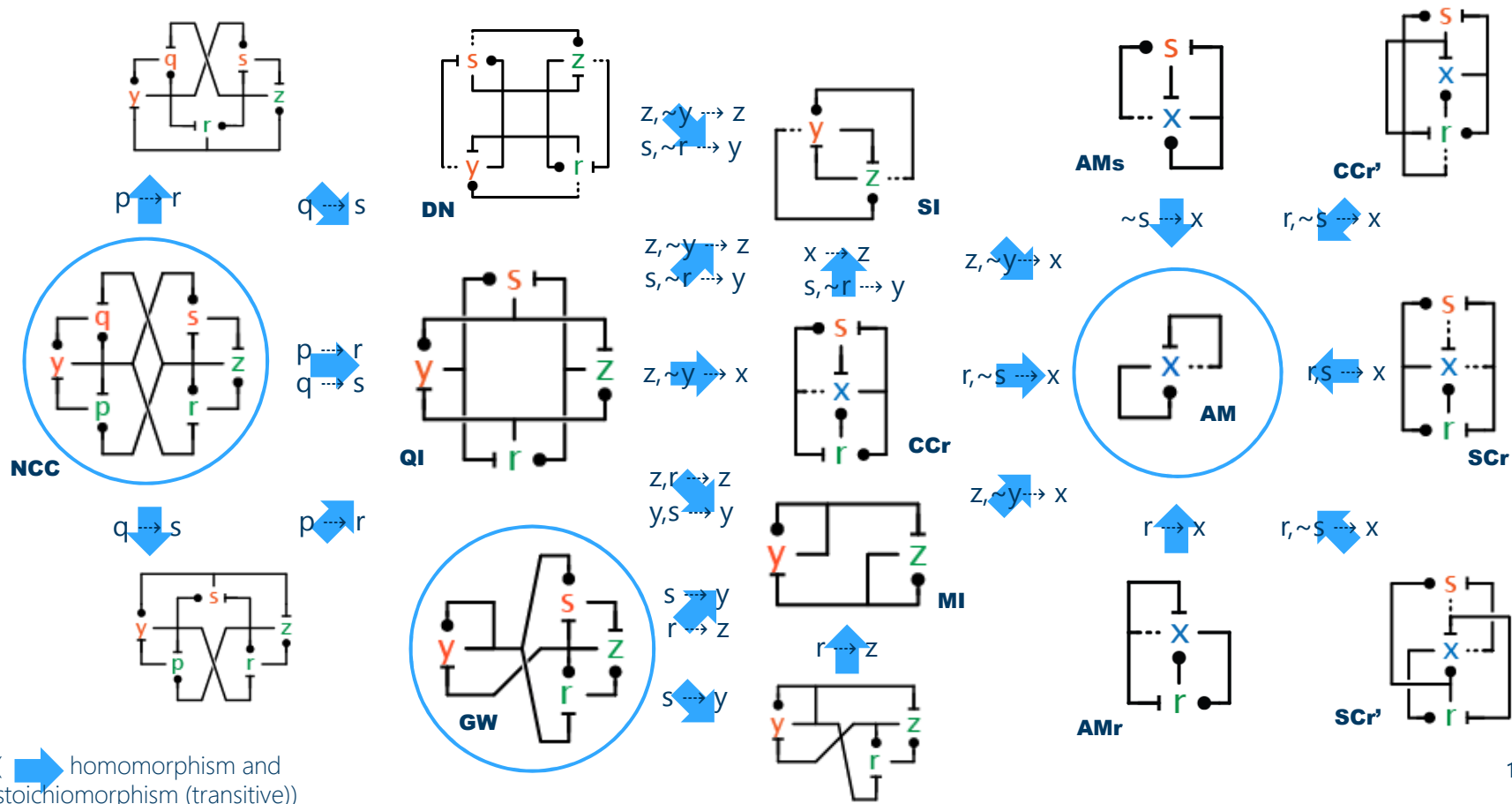
- Why does this work so well?

Emulations Compose: NCC emulates AM

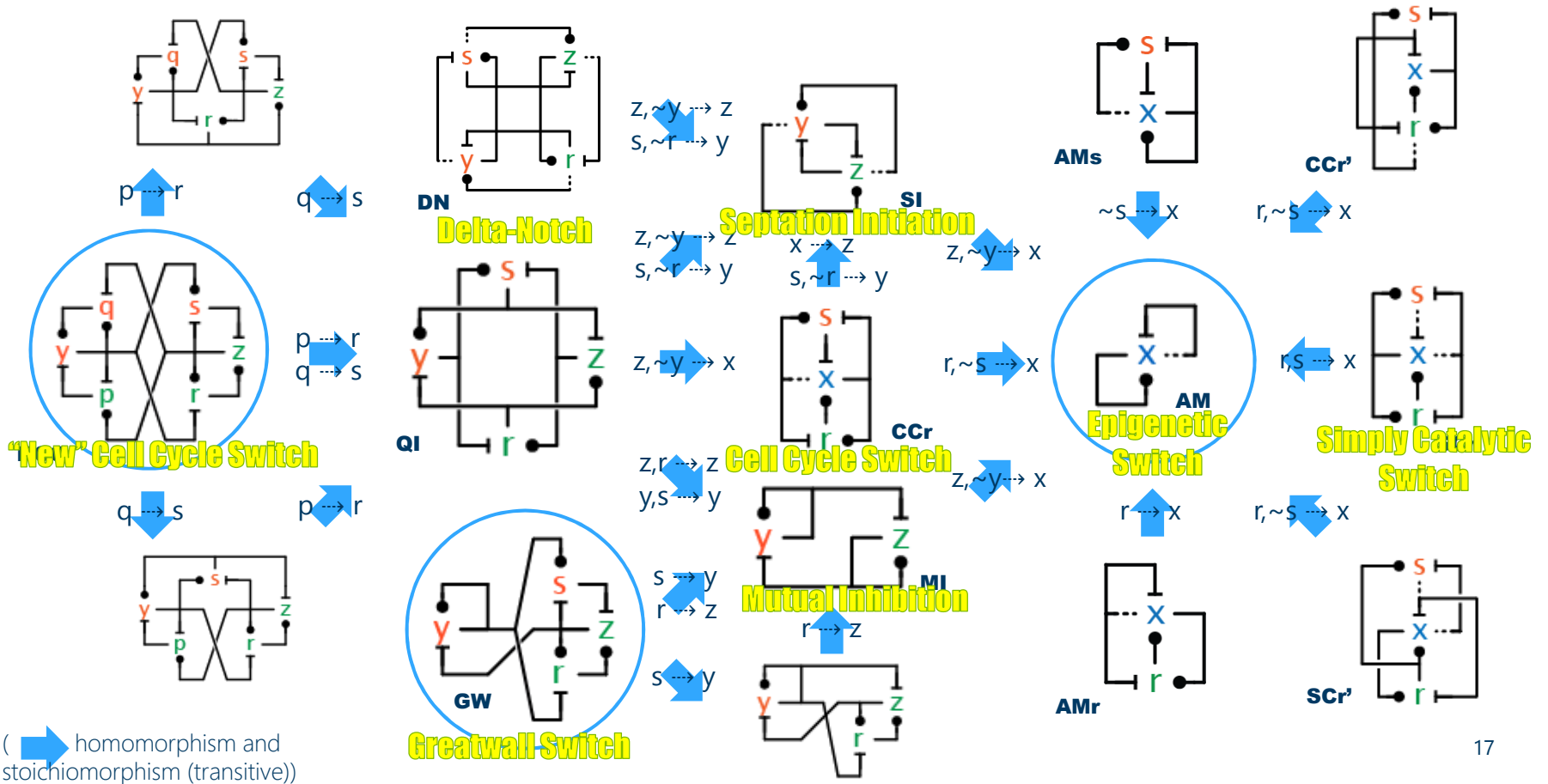
- The (18) trajectories NCC can *always* retrace those (3) of AM



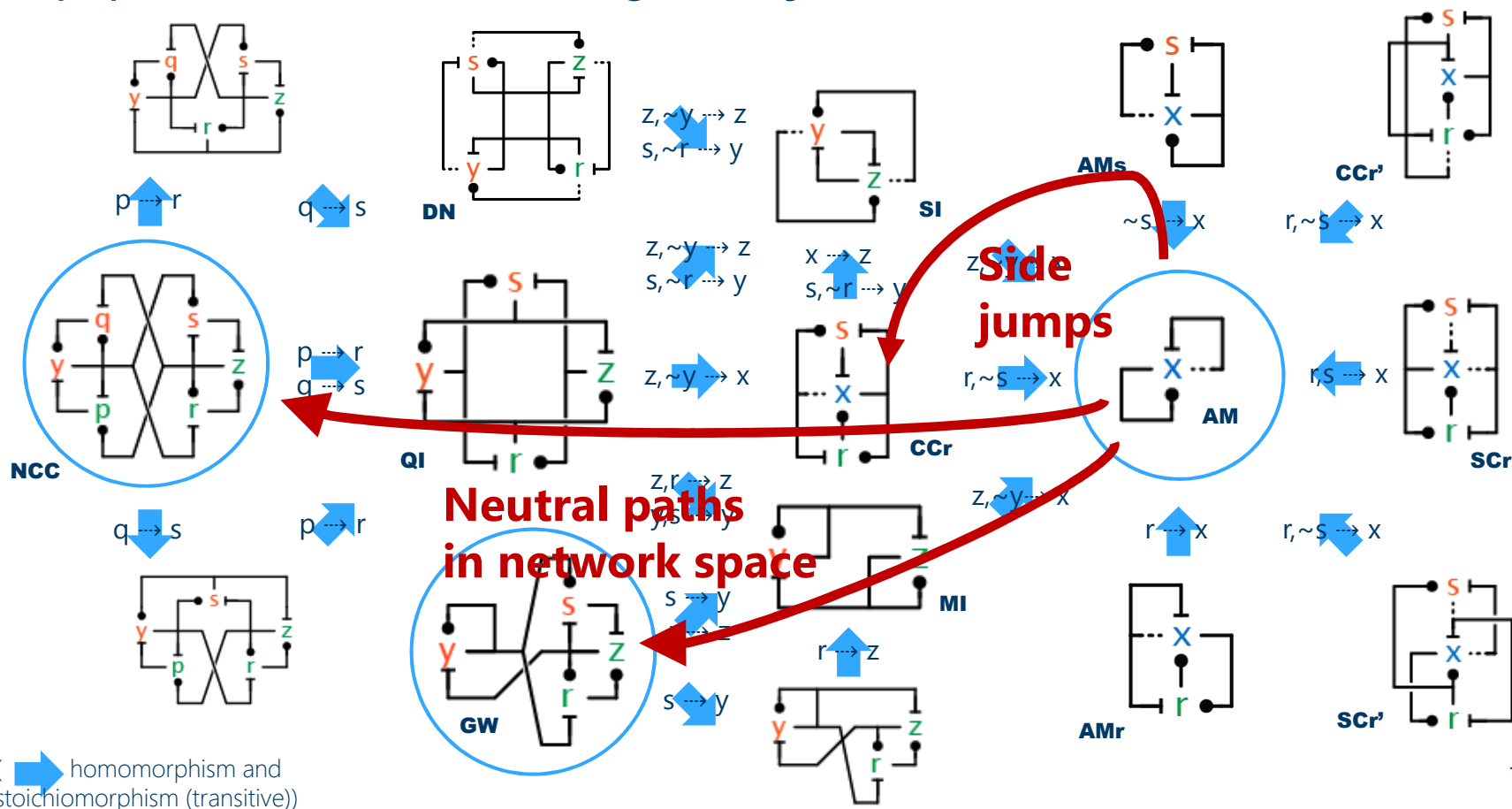
Approximate Majority Emulation Zoo



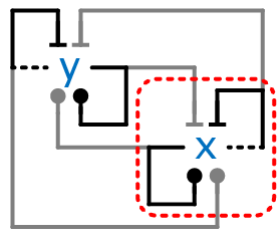
Approximate Majority Emulation Zoo



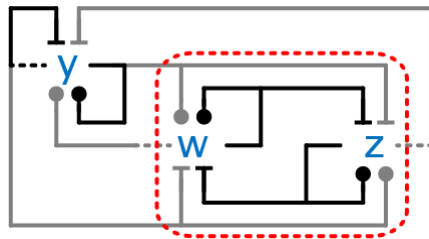
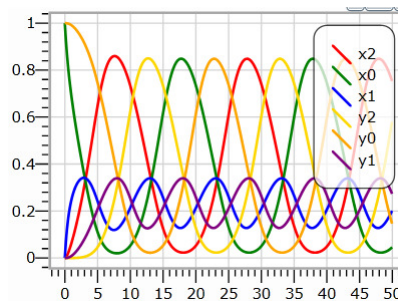
Approximate Majority Emulation Zoo



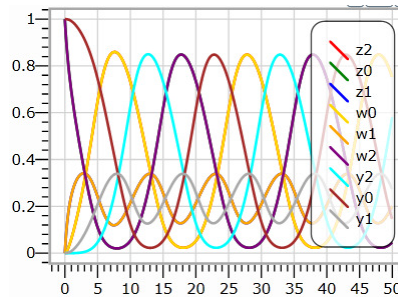
Emulation in Context



AM-AM Oscillator



AM-MI Oscillator



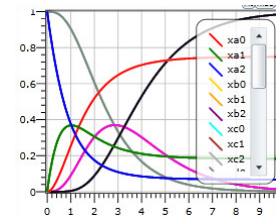
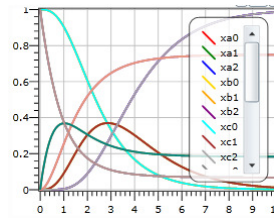
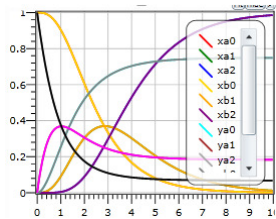
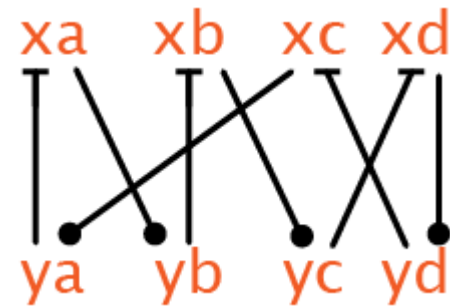
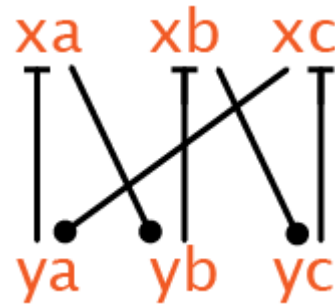
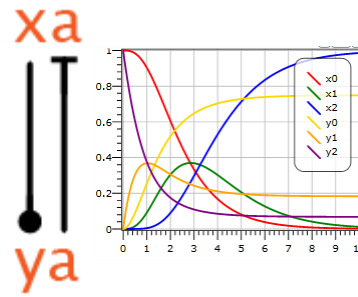
$m \in \text{MI} \rightarrow \text{AM}$ is an emulation:
it maps $z \rightarrow x$ and $\sim w \rightarrow x$

We can replace AM with MI in a context. The mapping m tells us how to wire MI to obtain an overall emulation:

Each influence crossing the dashed lines into x is replaced by a similar influence into *both* z and $\sim w$. The latter is the same as an opposite influence into w (shown).

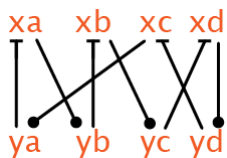
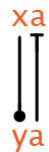
Each influence crossing the dashed lines out of x is replaced by a similar influence from the same side of *either* z or $\sim w$. The latter is the same as a similar influence from the opposite side of w (shown), and the same as an opposite influence from the same side of w .

Another Zoo



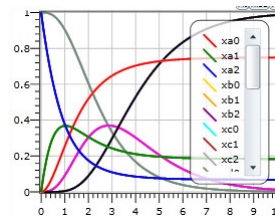
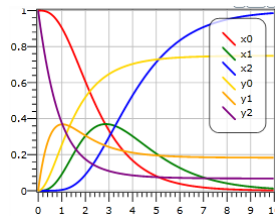
Network Perturbations

Network

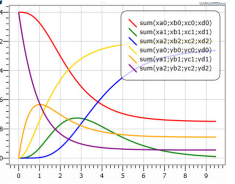
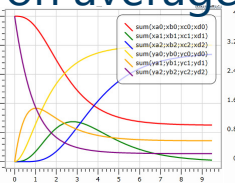
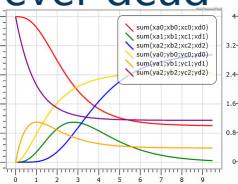
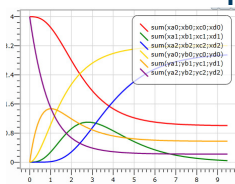
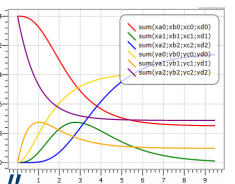
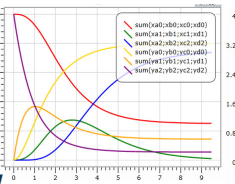
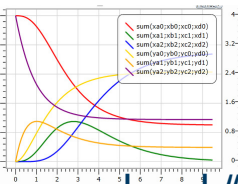
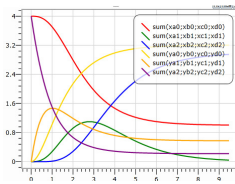
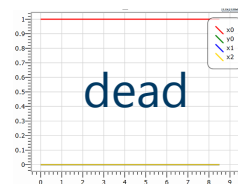
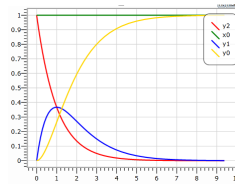


A complex but robust implementation of the simple network

Normal Behavior



Removing each link in turn



never dead "on average"

Morphisms of CRNs

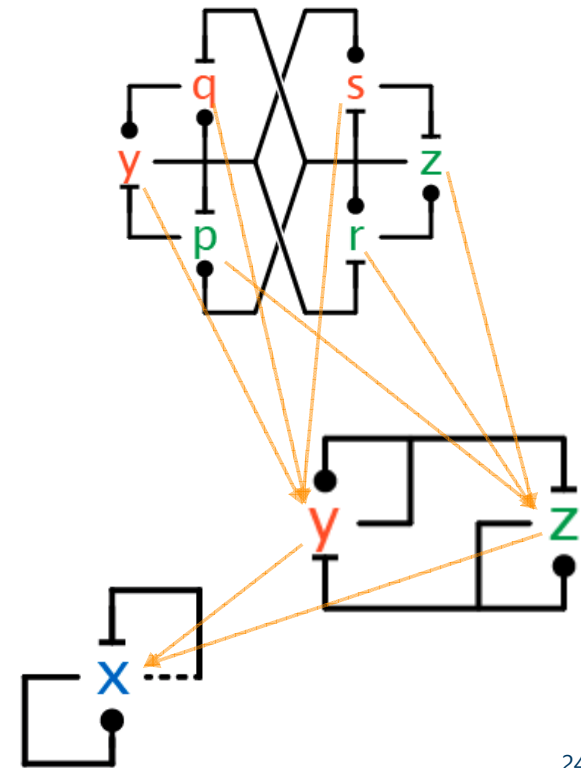
A Theory of Network Emulation

(with thanks to David Soloveichik)

- So far, evidence is empirical
 - Specific simulations based on a choice of parameters
- But indeed...
 - We can show analytically that, GW, NCC, etc. are *exactly* and *always* as good as AM
 - Where *exactly* means *numerically* as good, not just in the same complexity class
 - And *always* means for *any* choice of rates and initial conditions (of the target network)
- A network *emulates* another network:
 - When it can *exactly* reproduce the kinetics of another network for *any* choice of rates and initial conditions (of the other network)
 - We aim to show that e.g. the cell cycle switch can emulate AM in that sense
 - And moreover that the emulation is *algorithmic*: it is determined by static network *structure* (including rate constants and stoichiometric constants), not by random kinetic

When can a Network Emulate Another?

- What kind of morphisms guarantee emulation?
 - do they preserve network structure?
 - do they preserve stoichiometry?

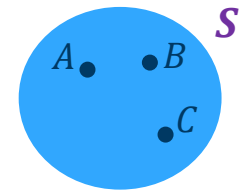


Chemical Reaction Networks

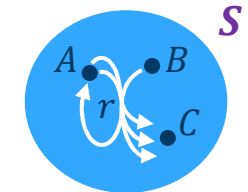
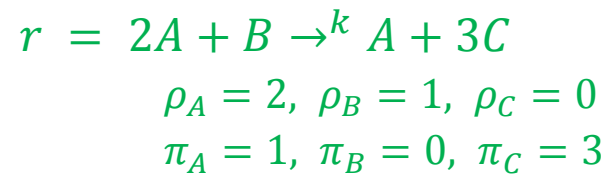
- A CRN is a pair (S, R) where
 - $S = \{s_1, \dots, s_n\}$ a finite set of *species*
 - $R = \{r_1, \dots, r_m\}$ a finite set of *reactions*^(*)

$$S = \{A, B, C\}$$

$$R = \{r\}$$



- Reactions $r = \rho \rightarrow^k \pi \in R$
with *complexes* $\rho, \pi \in \mathbb{N}^S$
stoichiometric numbers ρ_s, π_s for $s \in S$
and *rate constants* $k > 0$



- The *stoichiometry* of s in $\rho \rightarrow^k \pi$ is:

$$\eta(s, \rho \rightarrow^k \pi) = \pi_s - \rho_s$$

$$\varphi(s, \rho \rightarrow^k \pi) = k \cdot (\pi_s - \rho_s)$$

$$\eta(A, r) = -1 \quad \text{net stoichiometry}$$

$$\varphi(A, r) = -k \quad \text{(instantaneous) stoichiometry}$$

$$(*) \rho \rightarrow^k \pi, \rho \rightarrow^{k'} \pi \in R \Rightarrow k = k'$$

CRN Morphisms

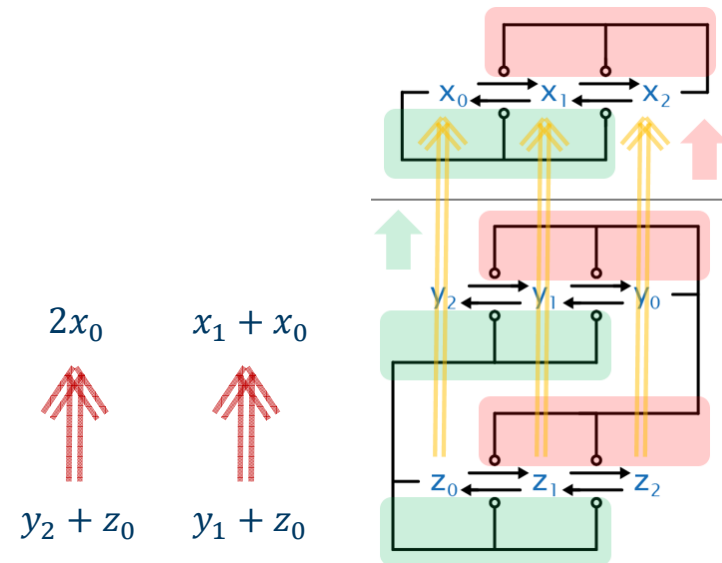
A *CRN morphism* from (S, R) to (\hat{S}, \hat{R})
written $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$

is a pair of maps $m = (m_S, m_R)$
a species map $m_S \in S \rightarrow \hat{S}$
a reaction map $m_R \in R \rightarrow \hat{R}$

extended to a complex map $m_S \in \mathbb{N}^S \rightarrow \mathbb{N}^{\hat{S}}$
linearly: $m_S(\rho)_{\hat{s}} = \sum_{s \in m_S^{-1}(\hat{s})} \rho_s$

(sometimes omitting the subscripts on m)

Mappings (symmetries)
between two networks



3 Key Morphisms

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

$$m_{\mathcal{R}}(\rho \xrightarrow{k} \pi) = m_{\mathcal{S}}(\rho) \xrightarrow{k} m_{\mathcal{S}}(\pi) \quad \Rightarrow \quad m_{\mathcal{S}}^T \cdot \varphi = \hat{\varphi} \cdot m_{\mathcal{R}}^T$$

- a *CRN reactant morphism* if $m_{\mathcal{R}}$ is determined by $m_{\mathcal{S}}$ on reactants. $\exists \hat{k}, \hat{\pi}$:

$$m_{\mathcal{R}}(\rho \xrightarrow{k} \pi) = m_{\mathcal{S}}(\rho) \xrightarrow{\hat{k}} \hat{\pi} \quad \Leftrightarrow \quad m_{\mathcal{S}}^T \cdot \rho = \hat{\rho} \cdot m_{\mathcal{R}}^T$$

- a *CRN stoichiomorphism* if:

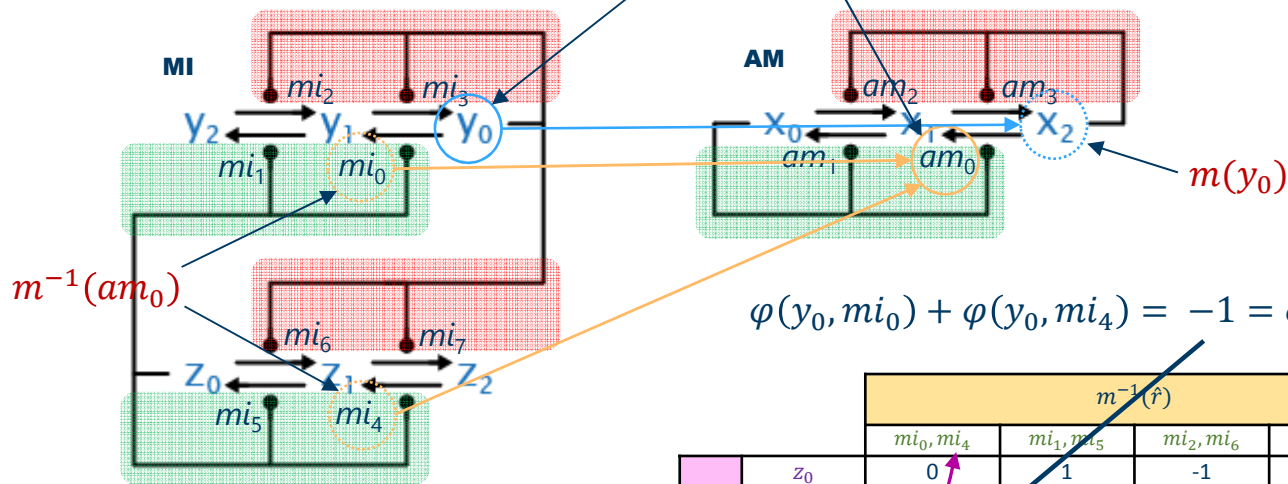
def. $\varphi \cdot m_{\mathcal{R}} = m_{\mathcal{S}} \cdot \hat{\varphi}$

$\varphi, \hat{\varphi}$ are the respective stoichiometric matrices
 $\rho, \hat{\rho}$ are the respective reactant matrices
 $m_{\mathcal{S}}, m_{\mathcal{R}}$ are the characteristic 0-1 matrices of $m_{\mathcal{S}}, m_{\mathcal{R}}$
 $m_{\mathcal{S}}(s, \hat{s}) = 1$ if $m_{\mathcal{S}}(s) = \hat{s}$ else 0

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 (sufficient because of another theorem)

This is both a homomorphism and a stoichiomorphism

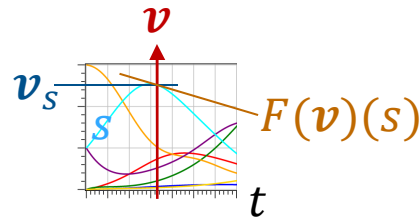
		$m^{-1}(\hat{r})$				$m(s)$
		mi_0, mi_4	mi_1, mi_5	mi_2, mi_6	mi_3, mi_7	
$\forall s \in \text{MI}$	z_0	0	1	-1	0	x_0
	z_1	1	-1	1	-1	x_1
	z_2	-1	0	0	1	x_2
	y_0	-1	0	0	1	x_2
	y_1	1	-1	1	-1	x_1
	y_2	0	1	-1	0	x_0
		am_0	am_1	am_2	am_3	
		$\forall \hat{r} \in \text{AM}$				

CRN Kinetics

A *state* of a CRN (S, R) is a $\mathbf{v} \in \mathbb{R}_+^S$

a vector of concentrations for each species

The *differential system* of a CRN (S, R) , $F \in \mathbb{R}_+^S \rightarrow \mathbb{R}^S$



$F(\mathbf{v})(s)$ gives the instantaneous change of concentration of a species in a given state

Given by the *law of mass action*:

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

sum over all reactions of the stoichiometry of the species in the reaction times the mass action of the reaction in the state

Usually written as a system of coupled concentration ODEs, integrated over time:

$$\frac{d\mathbf{v}_s}{dt} = F(\mathbf{v})(s)$$

the mass action of a reaction in state is the product of reagent concentrations according to their stoichiometric numbers:

$$[\rho \rightarrow^k \pi]_{\mathbf{v}} = \mathbf{v}^\rho = \prod_{s \in S} \mathbf{v}_s^{\rho_s} \quad 29$$

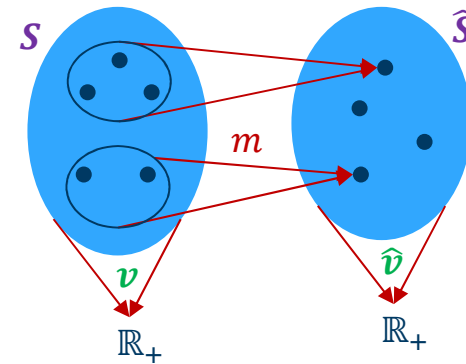
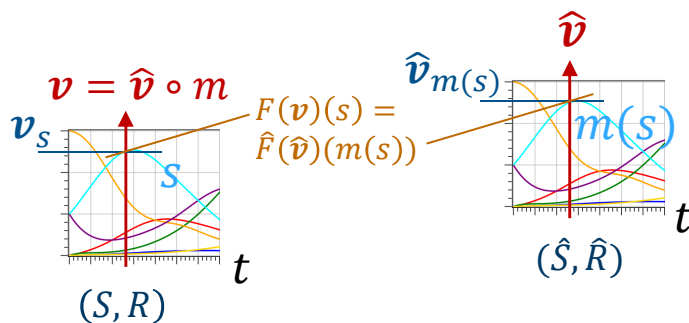
Kinetic Emulation

A morphism $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a *CRN emulation* if for the respective differential systems F, \hat{F} , $\forall \hat{v} \in \mathbb{R}_+^{\hat{S}}$:

$$F(\hat{v} \circ m) = \hat{F}(\hat{v}) \circ m$$

$$\begin{array}{ccc} \hat{v} \circ m & \xrightarrow{F} & \mathbb{R}^S \\ \uparrow - \circ m & & \uparrow - \circ m \\ \hat{v} & \xrightarrow{\hat{F}} & \mathbb{R}^{\hat{S}} \end{array}$$

That is: $\forall s \in S. F(\hat{v} \circ m)(s) = \hat{F}(\hat{v})(m(s))$



if the derivative of s (in state $\hat{v} \circ m$) equals the derivative of $m(s)$ (in state \hat{v})

if we start the two systems in states $v = \hat{v} \circ m$ (which is a copy of \hat{v} according to m) and \hat{v} resp., for each s the solutions are equal and the derivatives are equal, hence they will have identical trajectories by determinism

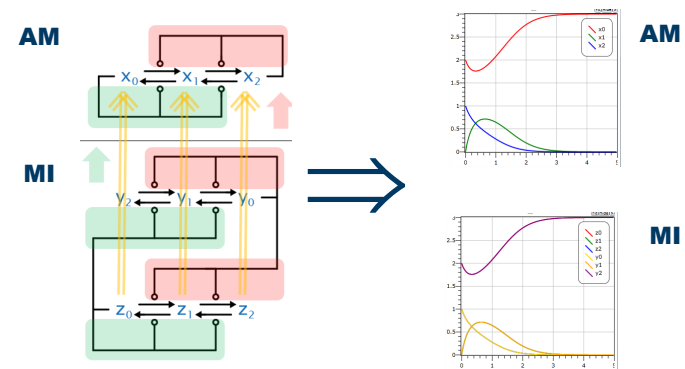
Emulation Theorem

Theorem: If $m \in (S, R) \rightarrow (\hat{S}, \hat{R})$ is a CRN reactant morphism and stoichiomorphism then it is a CRN emulation

reactant morphism $\mathbf{m}_S^T \cdot \boldsymbol{\rho} = \hat{\boldsymbol{\rho}} \cdot \mathbf{m}_R^T$ preserve enough network structure
 stoichiomorphism $\boldsymbol{\varphi} \cdot \mathbf{m}_R = \mathbf{m}_S \cdot \hat{\boldsymbol{\varphi}}$ preserve enough chemical stoichiometry
 \Downarrow
 emulation $\forall \hat{\mathbf{v}}. F(\hat{\mathbf{v}} \circ \mathbf{m}_S) = \hat{F}(\hat{\mathbf{v}}) \circ \mathbf{m}_S$ preserve derivatives

F is the differential system of (S, R) , given by the law of mass action, $\hat{\mathbf{v}}$ is a state of (\hat{S}, \hat{R}) . $\boldsymbol{\varphi}$ is the stoichiometric matrix and $\boldsymbol{\rho}$ is the related reactant matrix. \mathbf{m}_S and \mathbf{m}_R are the characteristic 0-1 matrices of the morphism maps \mathbf{m}_S (on species) and \mathbf{m}_R (on reactions). $-^T$ is transpose. Homomorphism implies reactant morphism.

Thus, for *any initial conditions* of (\hat{S}, \hat{R}) we can initialize (S, R) to match its trajectories. And also (another theorem), for *any rates* of (\hat{S}, \hat{R}) we can choose rates of (S, R) that lead to emulation.



Change of Rates Theorem

A *change of rates* for (S, R) is morphism $\iota \in (S, R) \rightarrow (S, R')$ such that $\iota(S)$ is the identity and $\iota(\rho, \pi, k) = (\rho, \pi, k')$.

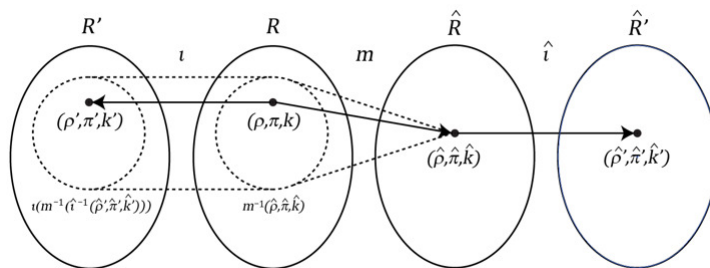
a morphism that modifies rates only

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.

thus, for *any rates* of (\hat{S}, \hat{R}) we can match trajectories

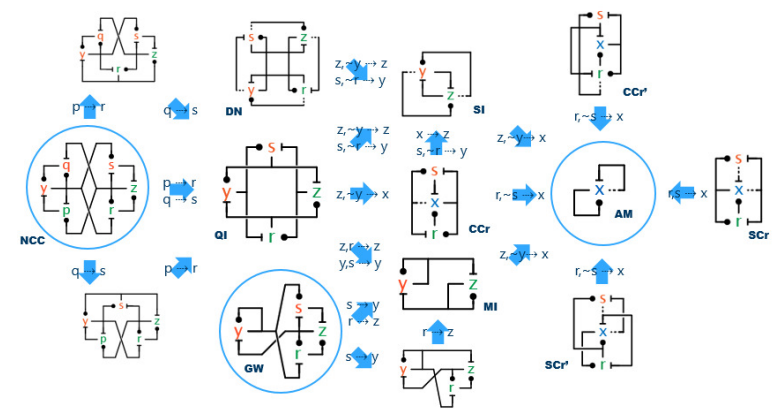
In fact, ι changes rates by the ratio with which $\hat{\iota}$ changes rates:

$$\iota(\rho, \pi, k) = \left(\rho, \pi, k \cdot \frac{\hat{k}'}{\hat{k}}\right) \text{ where } m(\rho, \pi, k) = (\hat{\rho}, \hat{\pi}, \hat{k}) \text{ and } \hat{\iota}(\hat{\rho}, \hat{\pi}, \hat{k}) = (\hat{\rho}', \hat{\pi}', \hat{k}').$$

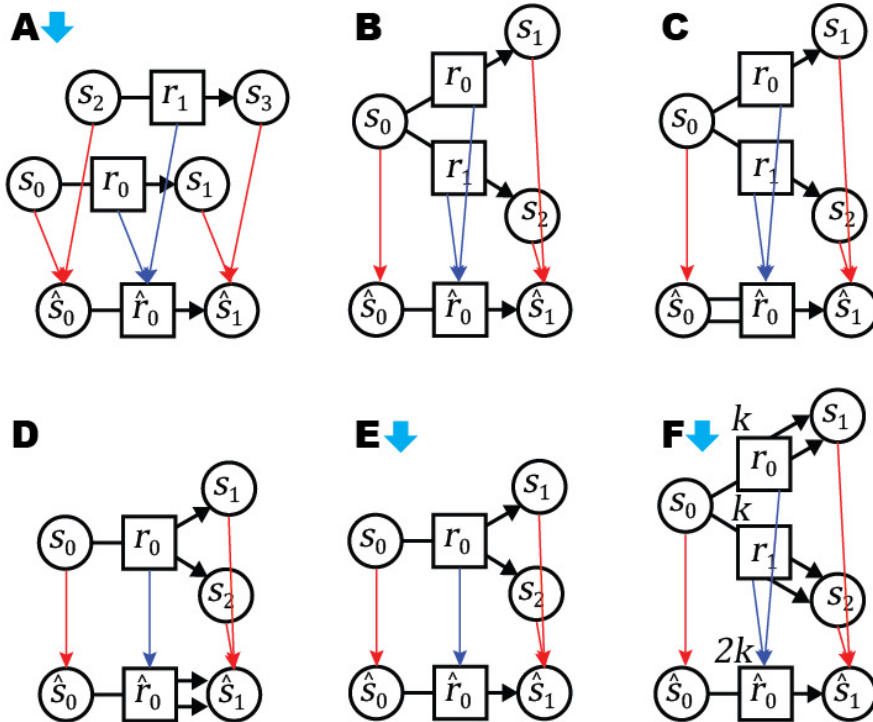


Corollaries

- By checking only static network and morphism properties we can learn that:
 - All these networks are (at least) bistable
 - (We do not have to reanalyze the steady states of all these dynamical systems)
 - All these networks can perform *exactly* as fast as AM
 - (We do not have to reprove the complexity bounds for all these networks)



Examples of CRN morphisms



Circles are species and squares are reactions. Red arrows are species mappings m_S and blue arrows are reaction mappings m_R . Solid arrows indicate morphisms that are emulations.

(A) A simple stoichiomorphism: the species in the source reactions are distinct. In general, multiple separate copies of a system will map to it via a trivial map that is a homomorphism and stoichiomorphism.

(B) This is a homomorphism, but is not a stoichiomorphism. For s_0, \hat{r}_0 : $\sum_{r \in m_R^{-1}(\hat{r}_0)} \varphi(s_0, r) = -2 \neq -1 = \varphi(m_S(s_0), \hat{r}_0)$.

(C) This is a stoichiomorphism, but is not a homomorphism or a reactant morphism. $r_0 = \rho \rightarrow \pi$ with $\rho_{s_0} = 1$ but $m_R(r_0) = \hat{r}_0 = \hat{\rho} \rightarrow \hat{\pi}$ with $\hat{\rho}_{m_S(s_0)} = \hat{\rho}_{\hat{s}_0} = 2$, so $\hat{\rho} \neq m_S(\rho)$ and $m_R(r_0) \neq m_S(\rho) \rightarrow \hat{\pi}$.

(D) This is a homomorphism but not a stoichiomorphism. For s_1, \hat{r}_0 : $\sum_{r \in m_R^{-1}(\hat{r}_0)} \varphi(s_1, r) = 1 \neq 2 = \varphi(m_S(s_1), \hat{r}_0)$.

(E) This stoichiomorphism is not a homomorphism, but is a reactant morphism. $r_0 = \rho \rightarrow \pi$ and $m_R(r_0) = \hat{r}_0 = \hat{\rho} \rightarrow \hat{\pi}$ with $\hat{\rho} = m_S(\rho)$ and $m_R(r_0) = m_S(\rho) \rightarrow \hat{\pi}$.

(F) This reactant morphism is not a homomorphism but is a stoichiomorphism. E.g., for s_1, \hat{r}_0 : $\sum_{r \in m_R^{-1}(\hat{r}_0)} \varphi(s_1, r) = \varphi(s_1, r_0) + \varphi(s_1, r_1) = 2 \cdot k + 0 \cdot k = 1 \cdot 2k = \varphi(m_S(s_1), \hat{r}_0)$.

Conclusions

Network Emulation Morphisms

- What guarantees emulation?
 - Reactant morphism + stoichiomorphism: static, state-independent (*structural*) conditions
- How do you find them?
 - Emulation Theorem => they do not depend on initial conditions
 - Change of Rates Theorem => can look for rate-1 morphisms
 - E.g. test all possible rate-1 homomorphism between two networks to see if they are stoichiomorphisms
- How common are they?
 - Likely relatively rare, but still many useful ones => richness of networks space
- How useful are they?
 - Establish structural, algorithmic, (non-accidental) *reasons* for kinetic similarity
 - Explain simple behavior “facets” of complicated networks
 - Investigate evolutionary paths (maybe)
- How brittle are they?
 - Will a perturbed trajectory of the source network converge to a trajectory of the target network?
 - What about other reaction kinetics?
- What about stochastic?
 - Is there a CME Emulation Theorem?

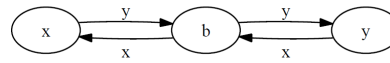
Population Majority

2004: **Computation in networks of passively mobile finite-state sensors.** Dana Angluin, James Aspnes, Zoë DiMadi, Michael J. Fischer, René Peralta. PODC'04.

Majority.
The value of the majority function is 1 if there are more 1's than 0's in the input; otherwise, it is 0.
The states of our protocol consist of a live bit and a counter with values in the set $\{-1, 0, 1\}$. Initially, the live

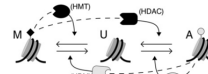
Exact Majority - 6-state
Nondeterministic.
(population protocol)

2007: **A Simple Population Protocol for Fast Robust Approximate Majority.** Dana Angluin, James Aspnes, David Eisenstat. DISC'07.



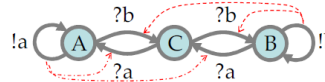
Approximate Majority - 3-state
Stochastic, discrete time
(DTMC) Fundamental results.

2007: **Theoretical Analysis of Epigenetic Cell Memory by Nucleosome Modification.** Ian B. Dodd, Mille A. Micheelsen, Kim Sneppen, Genevieve Thon. Cell.



Approximate Majority - 3-state
Stochastic, discrete time
(ad-hoc)

2009. **Artificial Biochemistry.** Luca Cardelli. Algorithmic Bioprocesses, Springer.



Approximate Majority - 3-state
Stochastic, **continuous time**
(CTMC). Simulations.

2009: **Robust Stochastic Chemical Reaction Networks and Bounded Tau Leaping (Appendix 4).** David Solov'evichik. J.Comput.Biol.

Transfer complexity results from discrete time population protocols to continuous time **stochastic chemical reaction networks.**

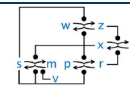
2009. **Using Three States for Binary Consensus on Complete Graphs.** Etienne Perron, Dinkar Vasudevan, and Milan Vojnovic. IEEE Infocom.

Approximate Majority - 3-state
Stochastic, **continuous time**
(CTMC) Fundamental results.

2010: **Convergence Speed of Binary Interval Consensus.** Moez Draief, Milan Vojnovic. Infocom'10.

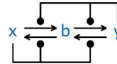
Exact Majority - 4-state
Stochastic, **continuous time.**
(similar to 2004 paper)

2012: **The Cell Cycle Switch Computes Approximate Majority.** Luca Cardelli, Attila Csikász-Nagy. Scientific Reports.



The biological cell cycle switch is a (non-obvious) implementation of approximate majority. Simulations.

2014: **Morphisms of Reaction Networks that Couple Structure to Function.** Luca Cardelli.



Approximate Majority - 3-state
Continuous space, continuous time
(Deterministic ODE). Emulation theorem.