

# Biochemical Switching Algorithms

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CoSBi

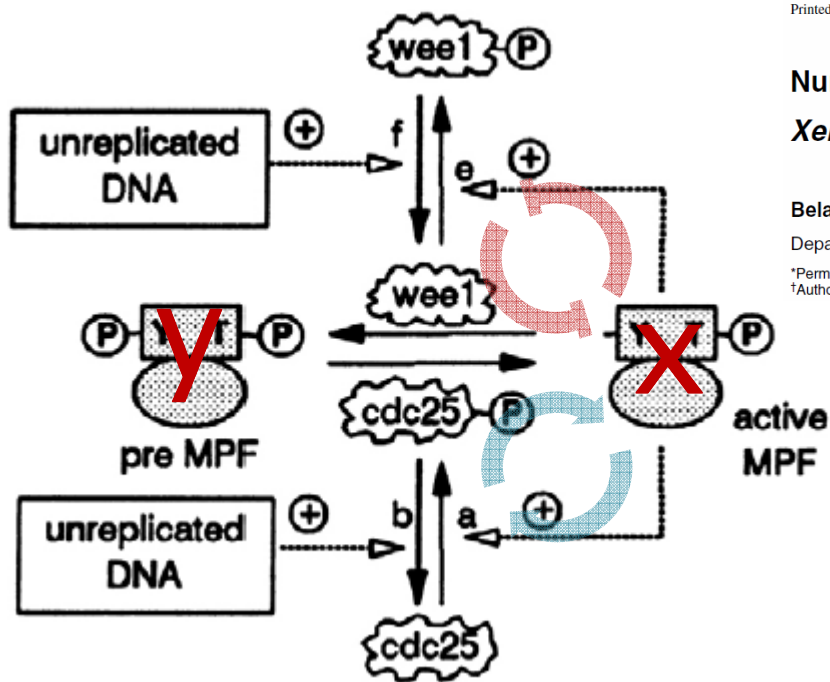
Newton Institute Lecture, Liverpool, 2012-07-16  
<http://lucacardelli.name>

# Outline

- **Analyzing molecular networks**
  - Various biochemical/bioinformatic techniques can tell us something about network structures.
  - We try to discover the function of the network, or to verify hypotheses about its function.
  - We try to understand how the structure is dictated by the function and other natural constraints.
- **The Cell–Cycle Switches and Oscillators**
  - Some of the best studied molecular networks.
  - Important because of their fundamental function (cell division) and preservation across evolution.

# The Cell Cycle Switch

- At the core of the cell-cycled oscillator.
  - This network is universal in all Eukaryotes [P. Nurse].



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

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†Author for correspondence

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

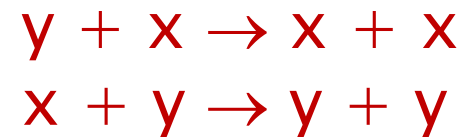
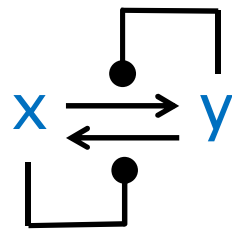
- Well studied. But *why this structure?*

# How to Build a Switch

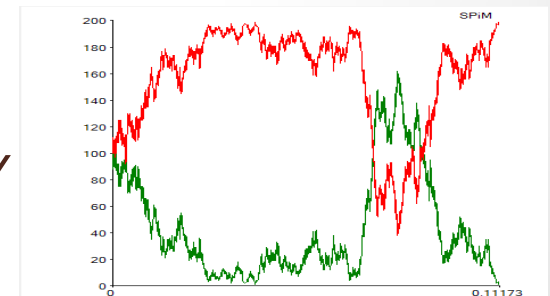
- What is a “good” switch?
  - We need first a *bistable* system: one that has two *distinct* and *stable* states. I.e., given *any* initial state the system must *settle* into one of two states.
  - The settling must be *fast* (not get stuck in the middle for too long) and *robust* (must not spontaneously switch back).
  - Finally, we need to be able to *flip* the switch: drive the transitions by external inputs.
- “Population” Switches
  - Populations of identical agents (molecules) that switch from one state to another *as a whole*.
  - Highly concurrent (stochastic).

# A Bad Algorithm

- Direct x-y competition
  - x catalyzes the transformation of y into x
  - y catalyzes the transformation of x into y



- This system is bistable, but
  - Convergence to a stable state is *slow* (a random walk).
  - *Any* perturbation of a stable state can initiate a random walk to the other stable state.



```
stochastic sample 0.0002
1000
stochastic plot x(), y(), 30
width = 100.0
time axis 0.0002 steps
time axis 0.11173 time
set x() =
  do 'x catalyzes y'
or 'y catalyzes x'
end y() =
  do 'y catalyzes x'
or 'x catalyzes y'
run 1000 of x()
run 1000 of y()
```

# A Very Good Algorithm

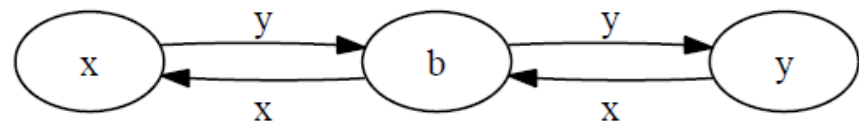
- Approximate Majority
  - Decide which of two populations is in majority
- A fundamental ‘population protocol’
  - Agents in a population start in state  $x$  or state  $y$ .
  - A pair of agents is chosen randomly at each step, they interact ("collide") and change state.
  - The whole population must 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.

# Properties

- With high probability, for  $n$  agents

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

- The number of state changes before converging is  $O(n \log n)$
- The total number of interactions before converging is  $O(n \log n)$
- The final outcome is correct if the initial disparity is  $\omega(\sqrt{n} \log n)$

- The algorithm is the fastest possible

- Must wait  $\Omega(n \log n)$  steps in expectation for all agents to interact

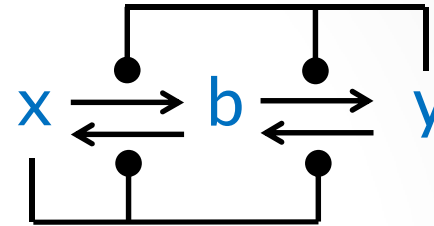
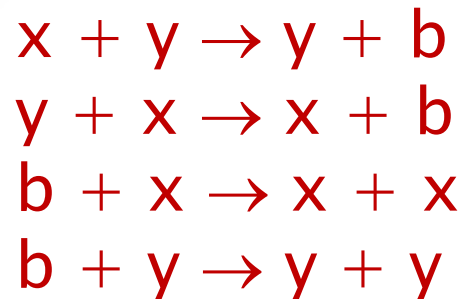
- Logarithmic time bound

- Parallel time is the number of steps divided by the number of agents.
- In parallel time the algorithm converges with high probability in  $O(\log n)$ .
- That is true for any initial conditions, even  $x=y!$

“Although we have described the population protocol model in a sequential light, in which each step is a single pairwise interaction, interactions between pairs involving different agents are independent and may be thought of as occurring in parallel. In measuring the speed of population protocols, then, we define 1 unit of parallel time to be  $\sum_j j$  steps. The rationale is that in expectation, each agent initiates 1 interaction per parallel time unit; this corresponds to the chemists’ idealized assumption of a well-mixed solution.”  
Distributed Computing 21(2):87-102.

# Chemical Implementation

A programming language for population algorithms!



Worse case test: start with  $x=y$ .

**Bistable**

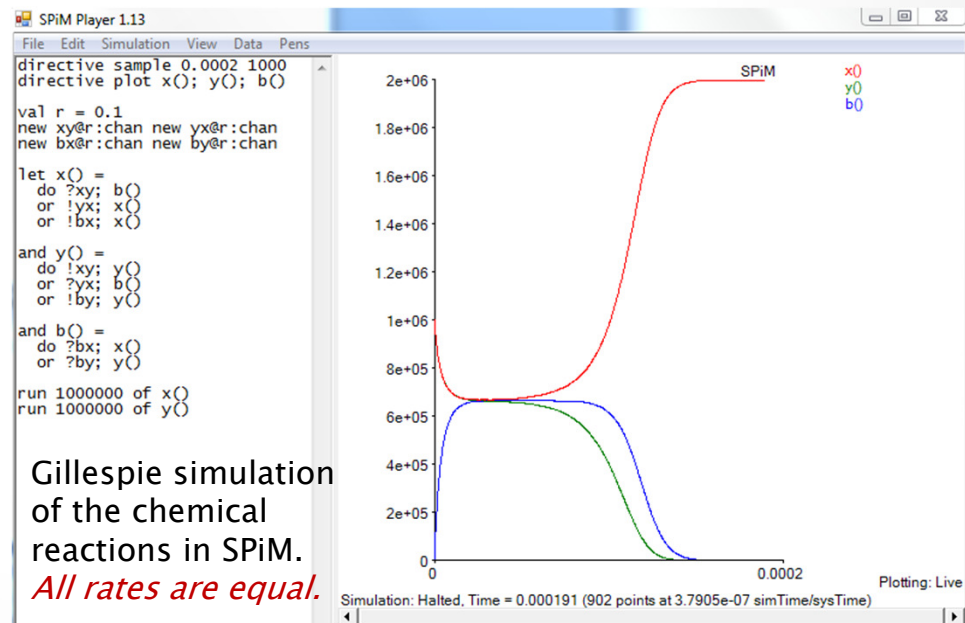
Even when  $x=y$ ! (stochastically)

**Fast**

$O(\log n)$  convergence time

**Robust**

$\omega(\sqrt{n \log n})$  majority wins whp



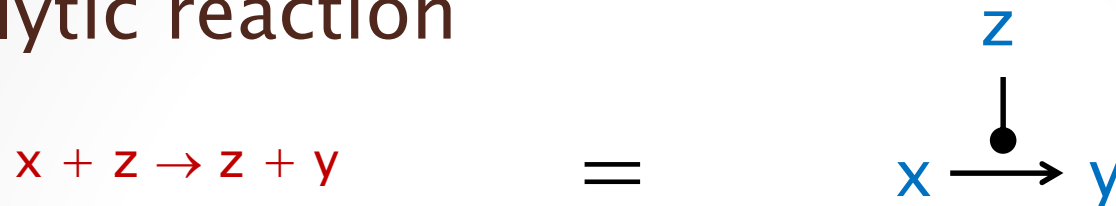


# Back to the Cell Cycle

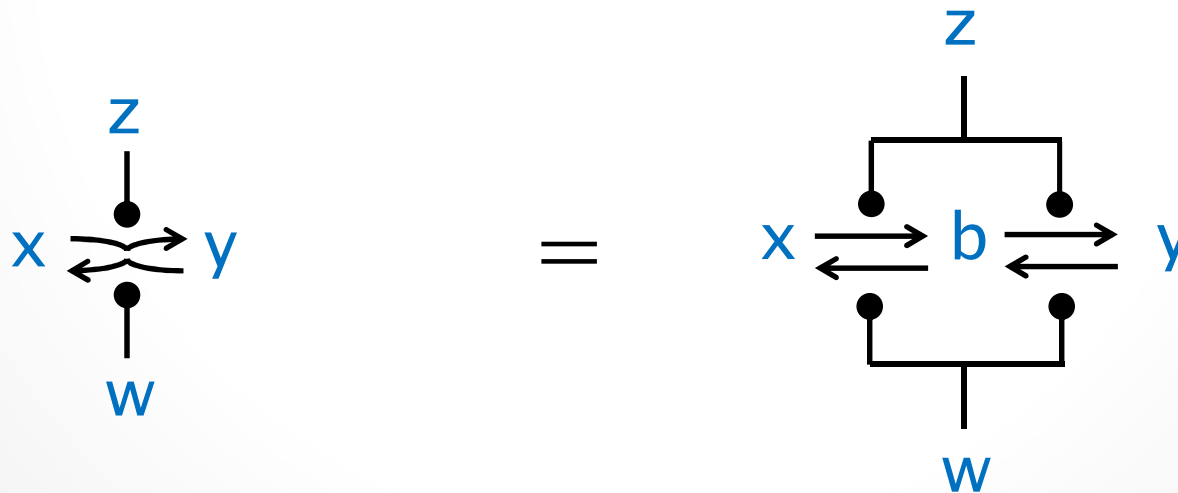
- The AM algorithm has great properties for settling a population into one of two states.
- But that is not what the cell cycle uses to switch its populations of molecules.
- Or is it?

# Some Notation

- Catalytic reaction

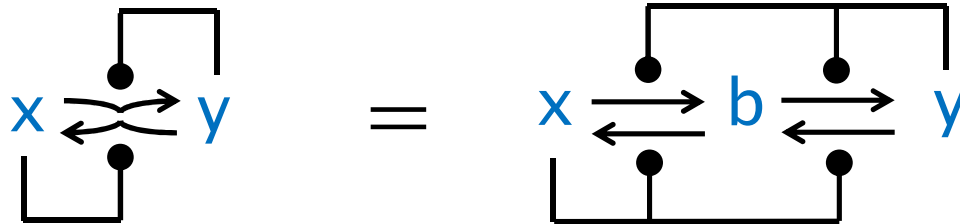


- Double 'kinase-phosphatase' reactions



# Step 1: the AM Network

*Abbreviated notation:*

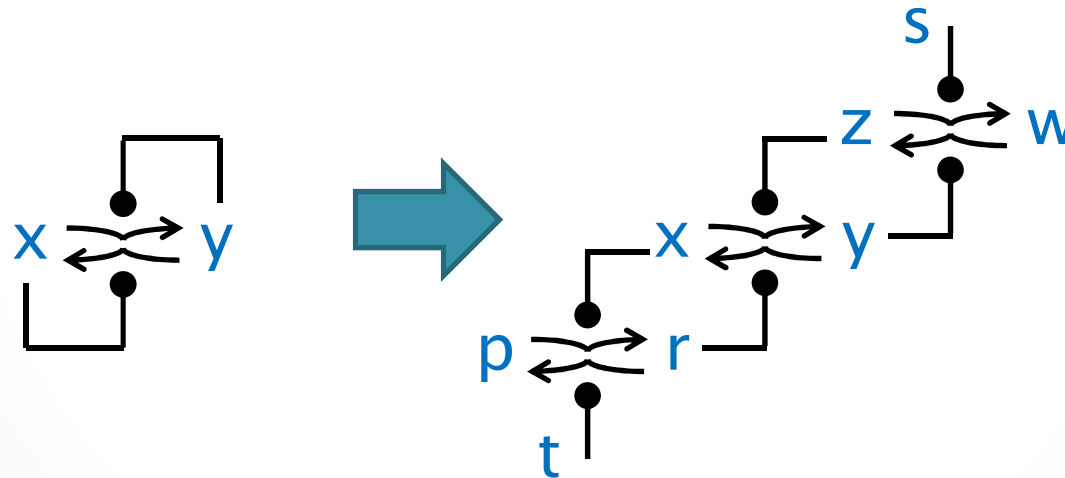


- CONSTRAINT: Autocatalysis, and especially intricate autocatalysis, is not commonly seen in nature.



# Step 2: remove auto-catalysis

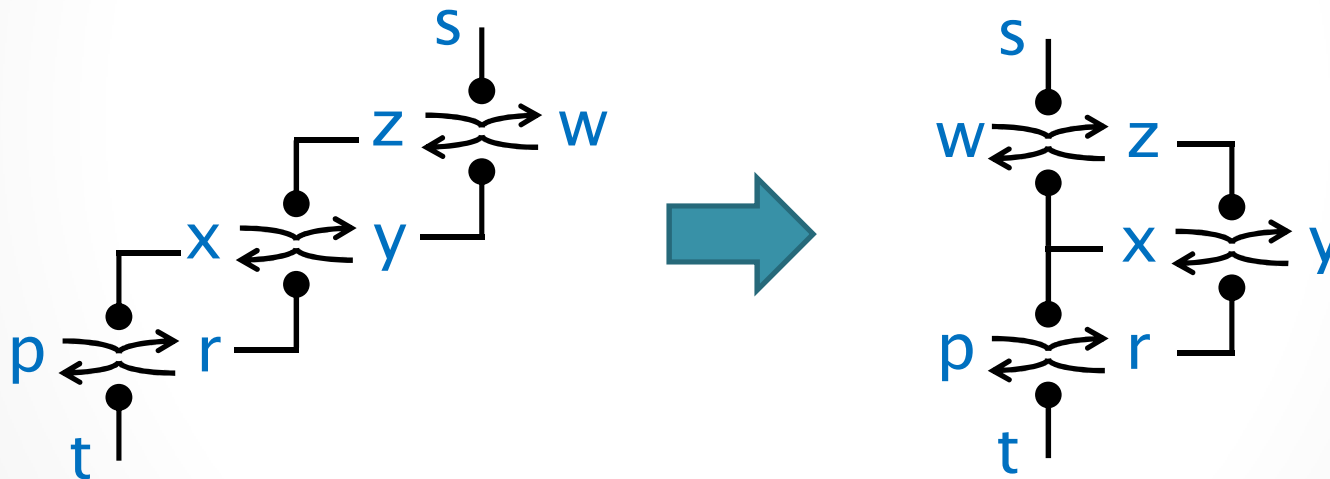
- Replace autocatalysis by mutual (simple) catalysis, introducing intermediate species z, r.
  - Here z breaks the y auto-catalysis, and r breaks the x auto-catalysis, while preserving the feedbacks.
  - z and r need to 'relax back' (to w and p) when they are not catalyzed: s and t provide the back pressure.



- **CONSTRAINT:** x and y (two states of the same molecule) are distinct active catalysts: that is not common in nature.

# Step 3: only one active state

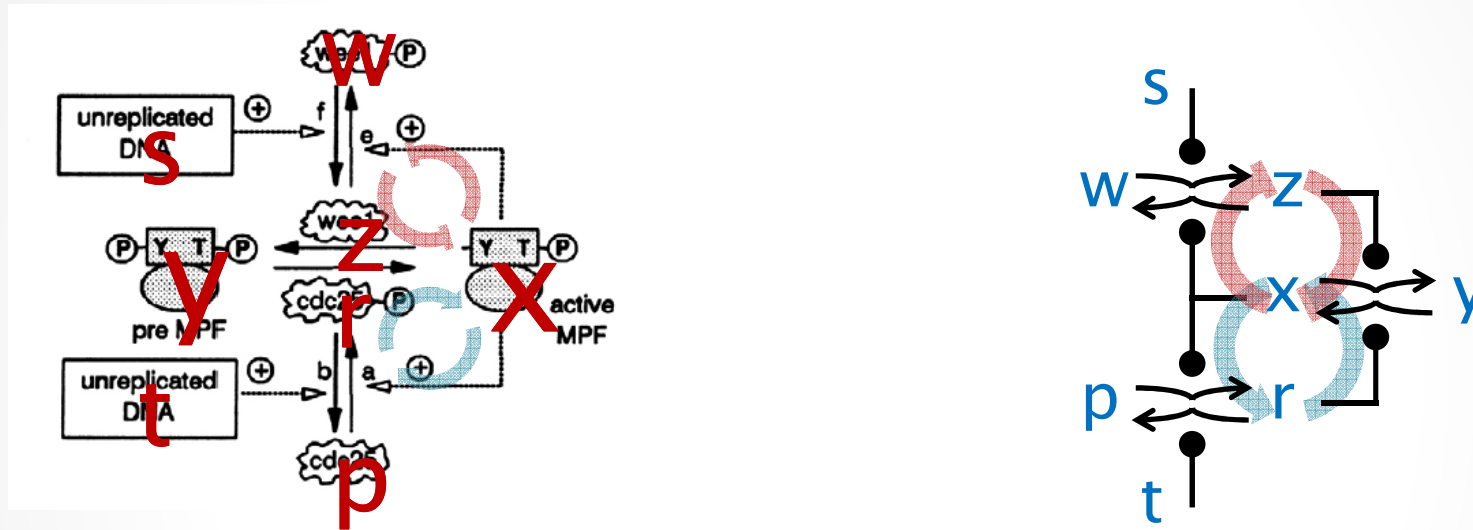
- Remove the catalytic activity of  $y$ .
  - Instead of  $y$  activating itself through  $z$ , we are left with  $z$  activating  $y$  (which remains passive). Hence, to deactivate  $y$  we now need to deactivate  $z$ . Since  $x$  'wants' to deactivate  $y$ , we make  $x$  deactivate  $z$ .



- All species now have one active ( $x, z, r$ ) and one inactive ( $y, w, p$ ) form. This is 'normal'.

# Network Structure

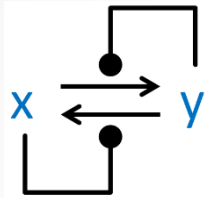
- ... and that *is* the cell-cycle switch!



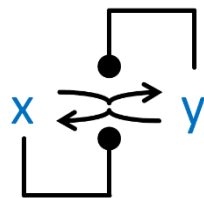
- The question is: did we preserve the *AM function* through our *network transformations*?
  - Ideally: prove either that the networks are ‘contextually equivalent’ or that the transformations are ‘correct’.
  - Practically: compare their ‘typical’ behavior.

# Convergence Analysis

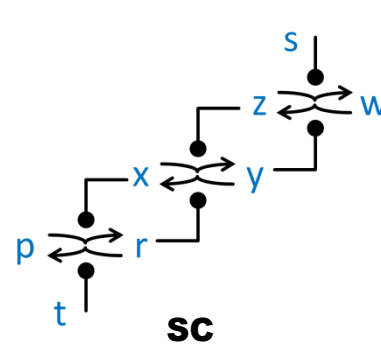
## Switches as Computational Systems – Convergence



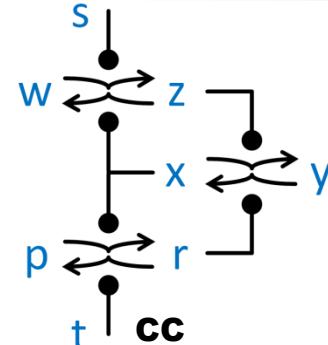
**DC**



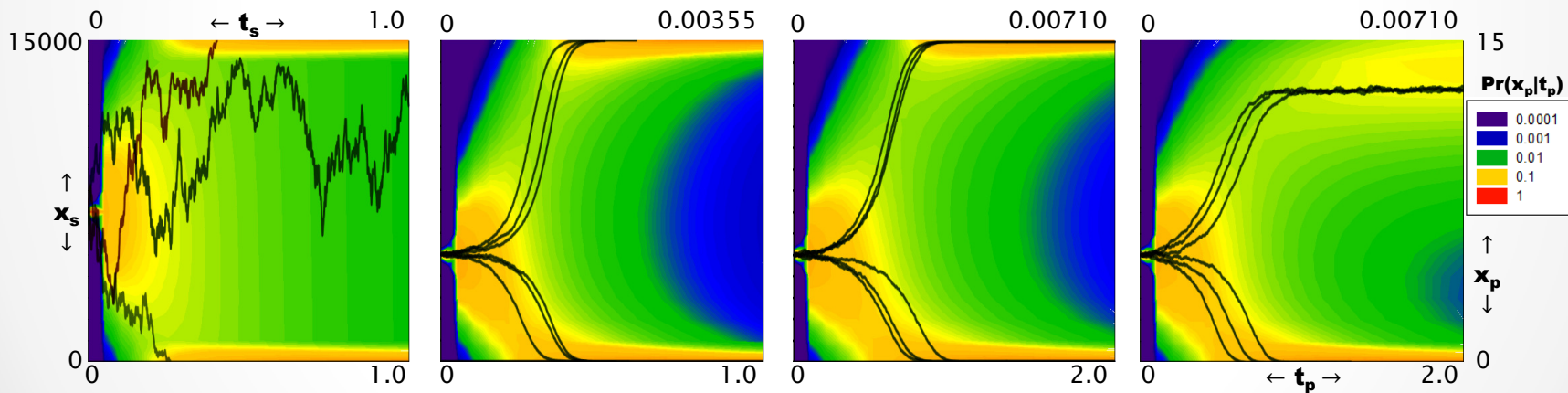
**AM**



**SC**



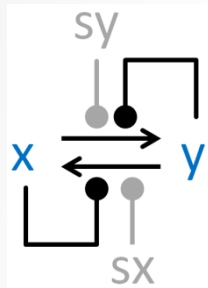
**CC**



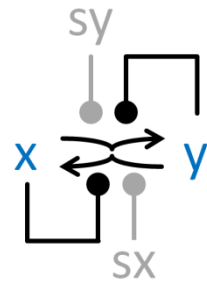
**NEW!**  
CC  
converges  
in log time

# Steady State Analysis

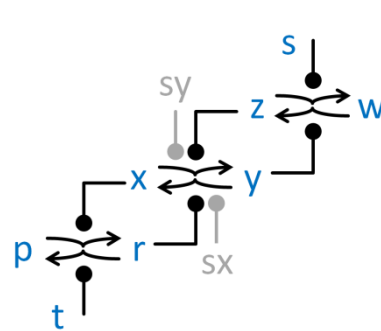
Switches as **Dynamical Systems** – Steady State Response



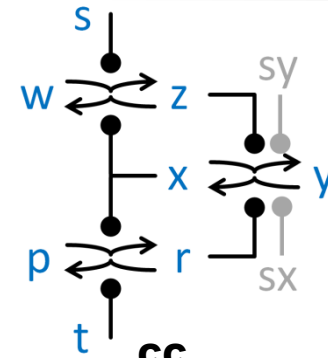
**DC**



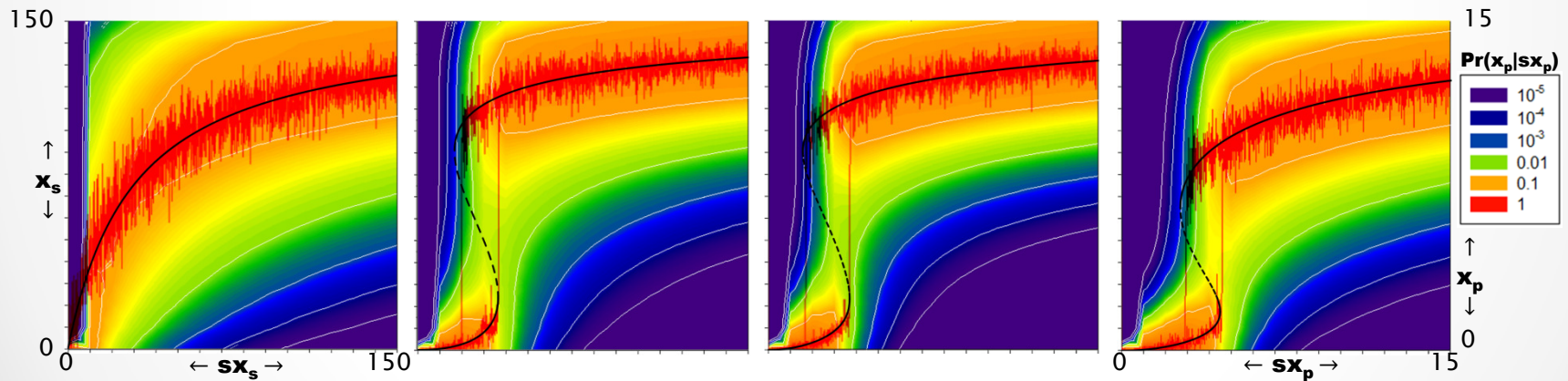
**AM**



**SC**



**CC**



**NEW!**  
AM shows  
hysteresis



# The Argument So Far

- Relating dynamical and computational systems in isolation (as *closed systems*)
  - The AM algorithm (network) implements an input-driven switching function (in addition to the known majority function).
  - The CC algorithm implements a input-less majority function (in addition to the known switching function).
  - The structures of AM and CC are related, and an intermediate network shares some properties of both.
- But what about the context?
  - Will AM and CC behave similarly in any context (as *open systems*)?
  - That's a hard question, so we look at their intended context: implementing oscillators.

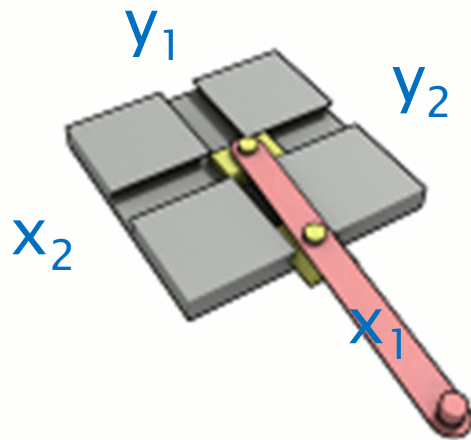
# Oscillators

- Basic in **Physics**, studied by simple *phenomenological* (not structural) ODE models.
- Non-trivial in **Chemistry**: it was only discovered in the 20's (Lotka) that chemical systems can oscillate: before it was thought impossible in closed systems. Shown experimentally only in the 50's.
- **Mechanics** (since antiquity) and modern **Electronics** (as well as Chemistry) must **engineer** the *network structure* of oscillators.
- **Biology**: all natural cycles are oscillators. Here we must **reverse engineer** their network structure.
- **Computing**: how can populations of agents (read: molecules) **interact** (network) to achieve oscillations?

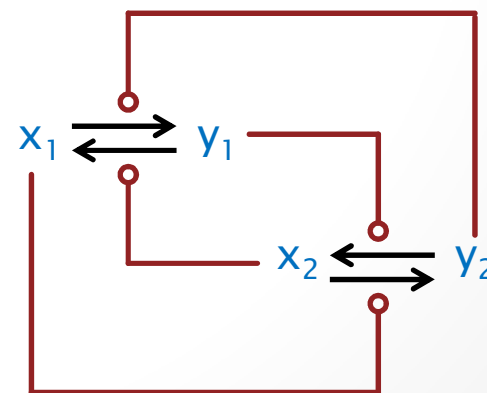
# The Trammel of Archimedes

- A device to draw ellipses
  - Two interconnected switches.
  - When one switch is on (off) it flips the other switch on (off).  
When the other switch is on (off) it flips the first switch off (on).
  - The amplitude is kept constant by mechanical constraints.

The function

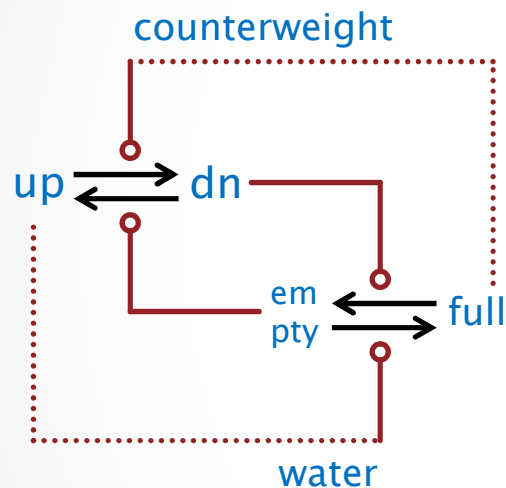


The network

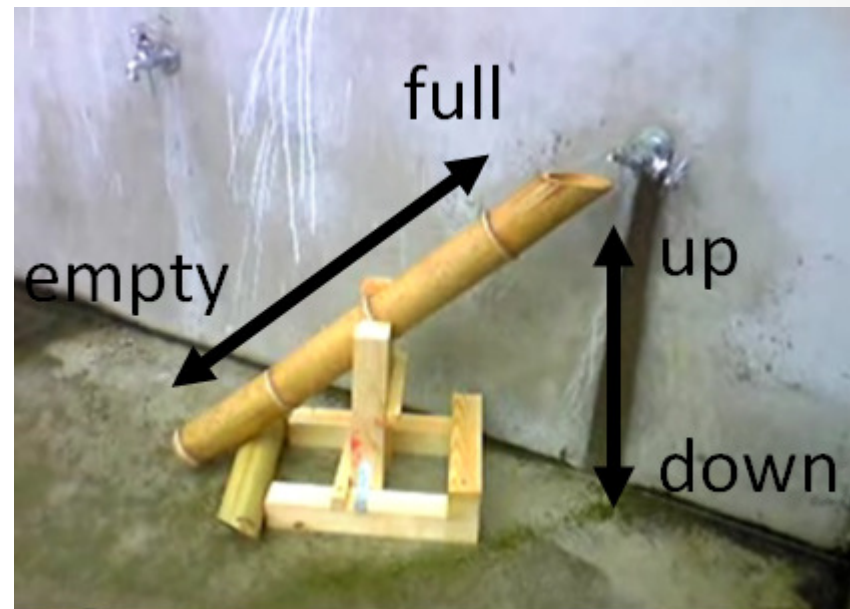


# The Shishi Odoshi

- A Japanese scarecrow (*lit.* scare-deer)
  - Used by Bela Novak to illustrate the cell cycle switch.



empty + up  $\rightarrow$  up + full  
up + full  $\rightarrow$  full + dn  
full + dn  $\rightarrow$  dn + empty  
dn + empty  $\rightarrow$  empty + up

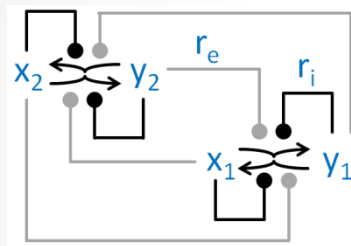


<http://www.youtube.com/watch?v=VbvecTlftcE&NR=1&feature=fwp>

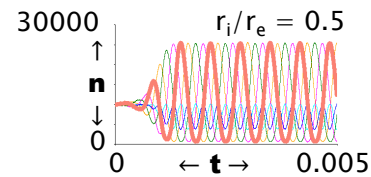
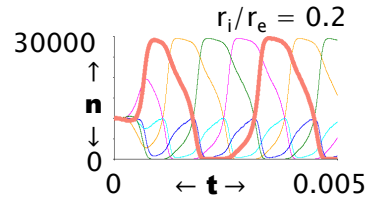
Outer switched connections replaced by constant influxes: tap water and gravity.

# Contextual Analysis

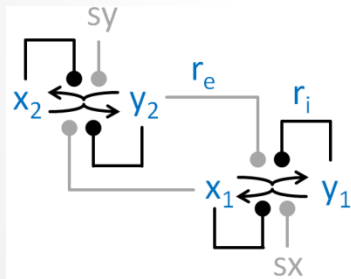
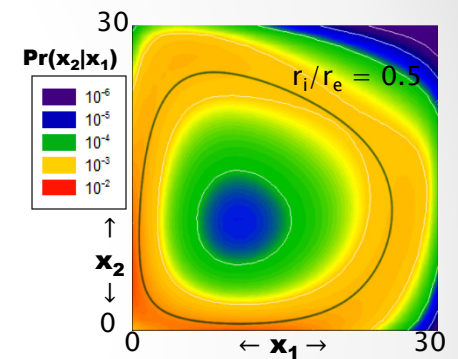
AM switches in the context of larger networks (oscillators).



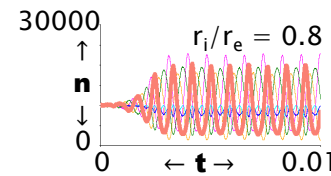
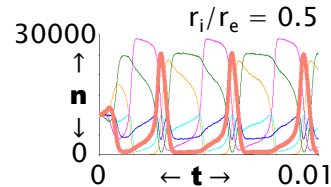
**Trammel**



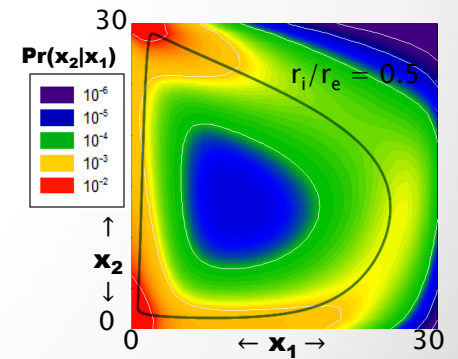
x1  
y1  
b1  
x2  
y2  
b2



**Shishi Odoshi**

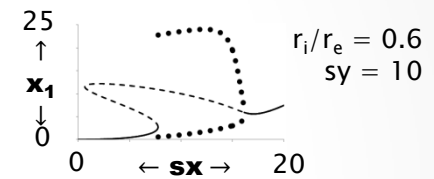
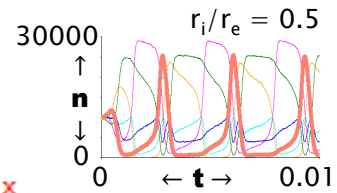
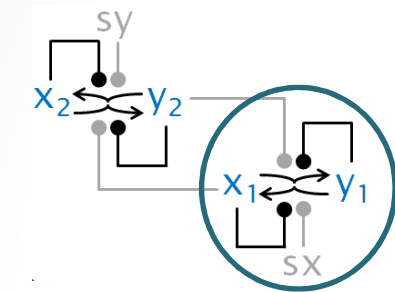


x1  
y1  
b1  
x2  
y2  
b2

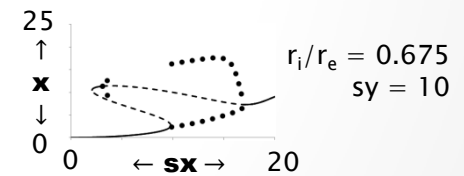
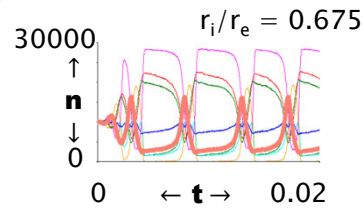


# Modularity Analysis

CC can be swapped in for AM.

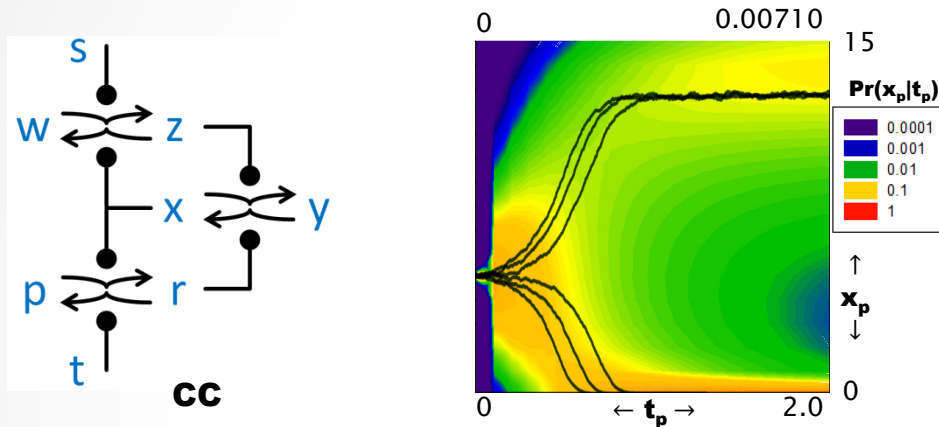


x  
y  
b  
x2  
y2  
b2  
z  
r



# CC does not “fully switch”

We have seen that the output of CC does not go ‘fully on’ like AM:

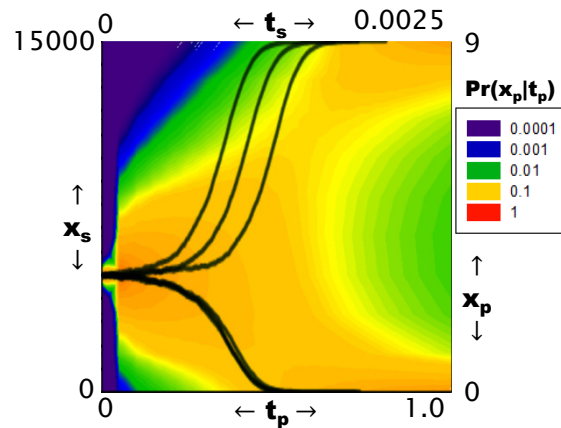
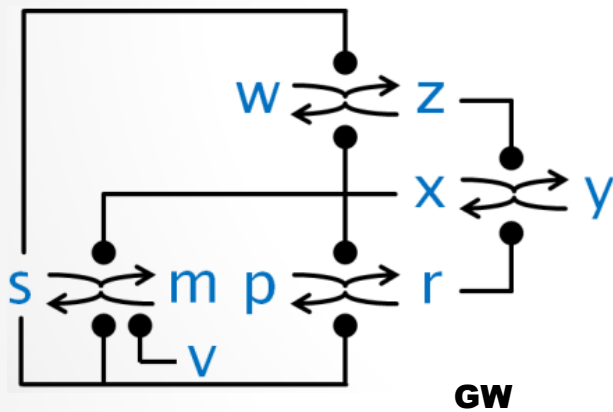


because  $s$  continuously inhibits  $s$  so that  $x$  cannot fully express.  
This could be solved if  $x$  would inhibit  $s$  in retaliation.

Q: How would *you* fix this problem?

# Nature fixed it!

There is another known feedback loop in real cell cycle switches by which  $x$  suppresses  $s$ :



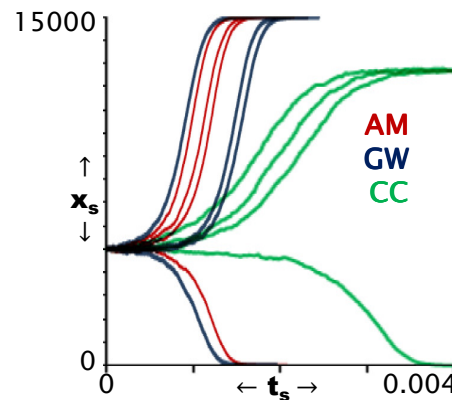
Full activation!

(Also,  $s$  and  $t$  happen to be the same molecule)



# And made it fast too!

More surprising: the extra feedback also speeds up the decision time of the switch, making it about as good as the 'optimal' AM switch:



**Conclusion:**

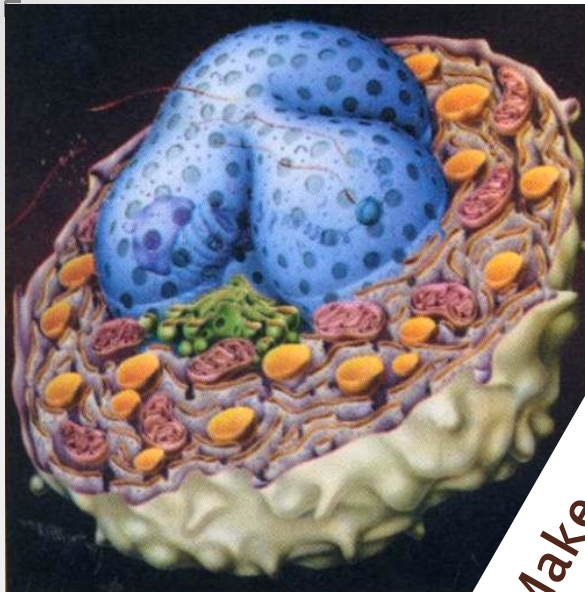
Nature is trying as hard as it can to implement an AM-class algorithm!

# Summary

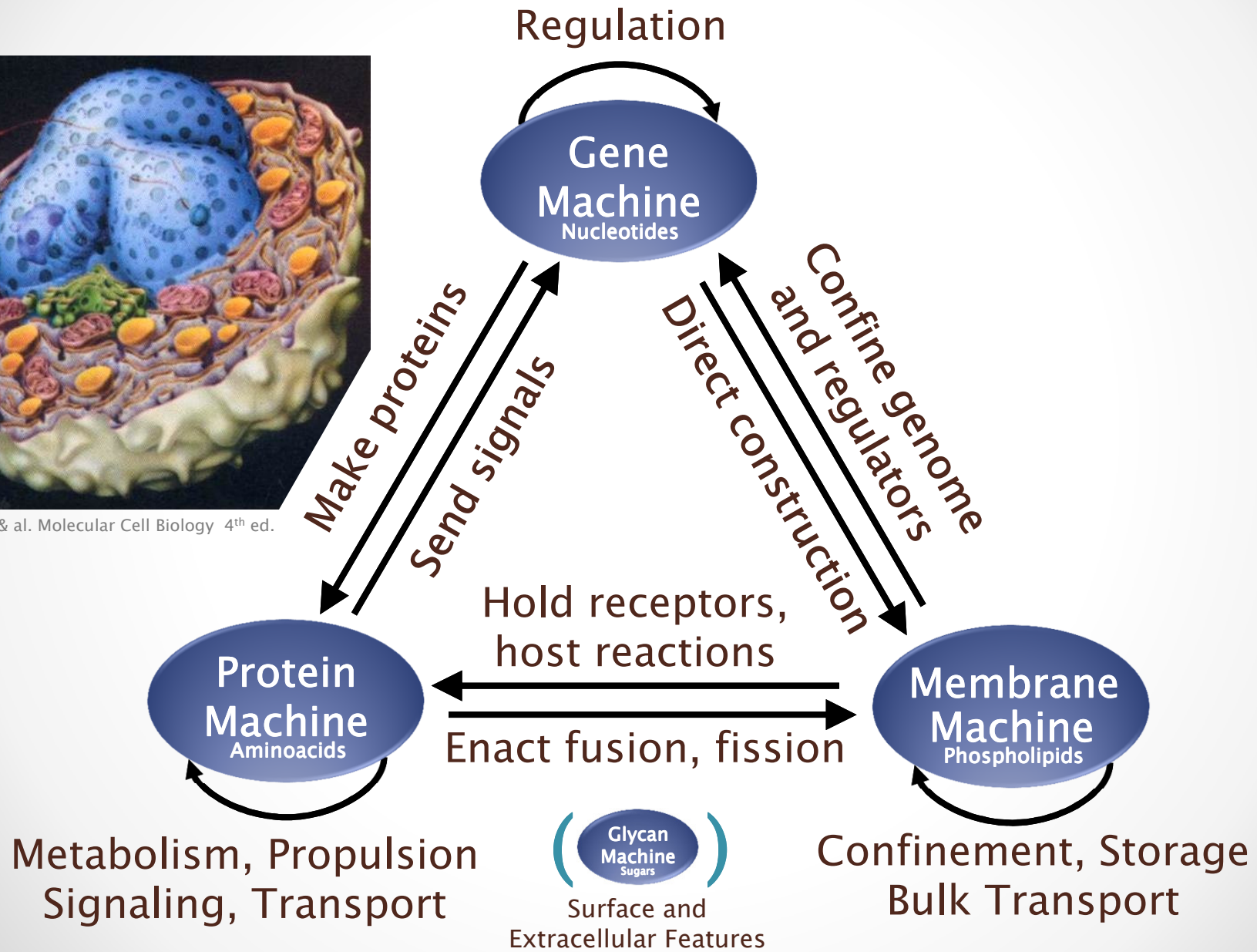
- The structure of AM implements an input-driven switching function (in addition to the known majority function).
- The structure of CC implements a input-less majority function (in addition to the known switching function).
- The structures of AM and CC are related, and an intermediate network shares the properties of both.
- The behaviors of AM and CC in isolation are related.
- The behaviors of AM and CC in oscillator contexts are related.
- A refinement of the core CC network, known to occur in nature, improves switching performance and brings it in line with AM performance.

# Computational Outlook

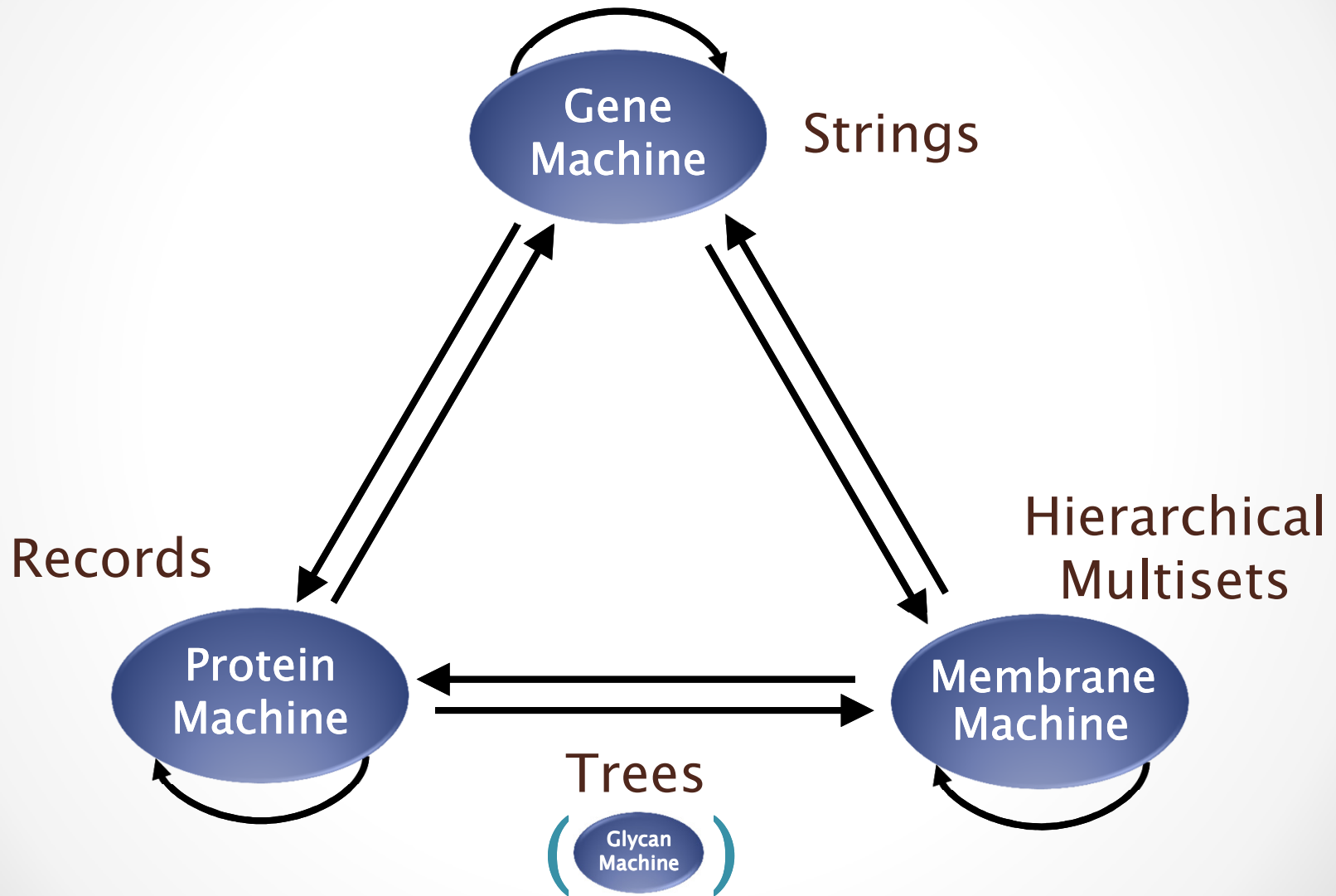
# Abstract Machines of Biochemistry



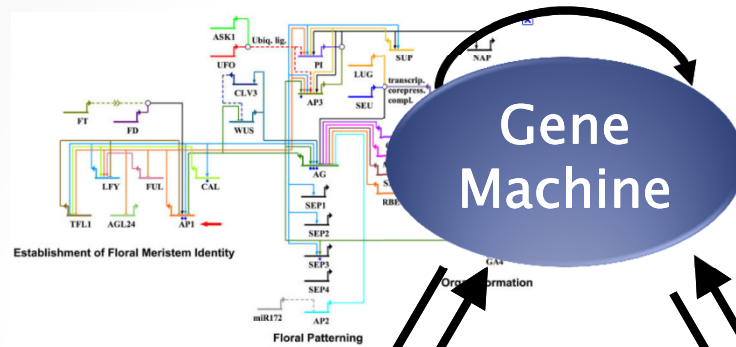
H.Lodish & al. Molecular Cell Biology 4<sup>th</sup> ed.



# Bioinformatic View (Data Structures)

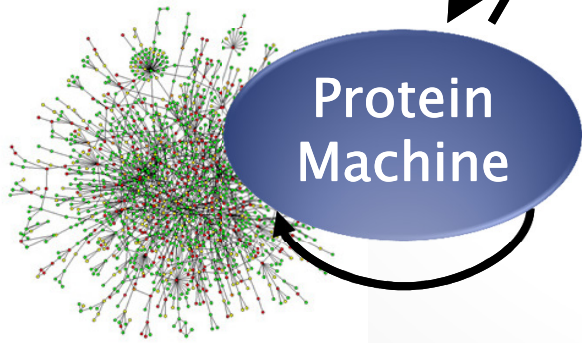


# Systems Biology View (Networks)

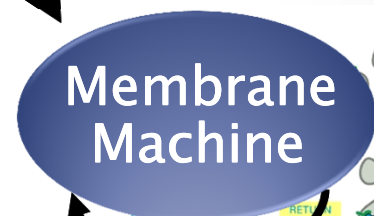
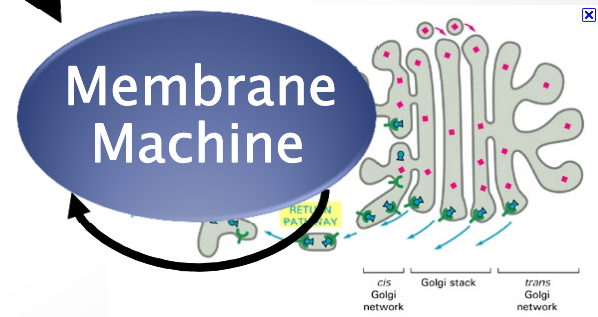


Gene Regulatory Networks

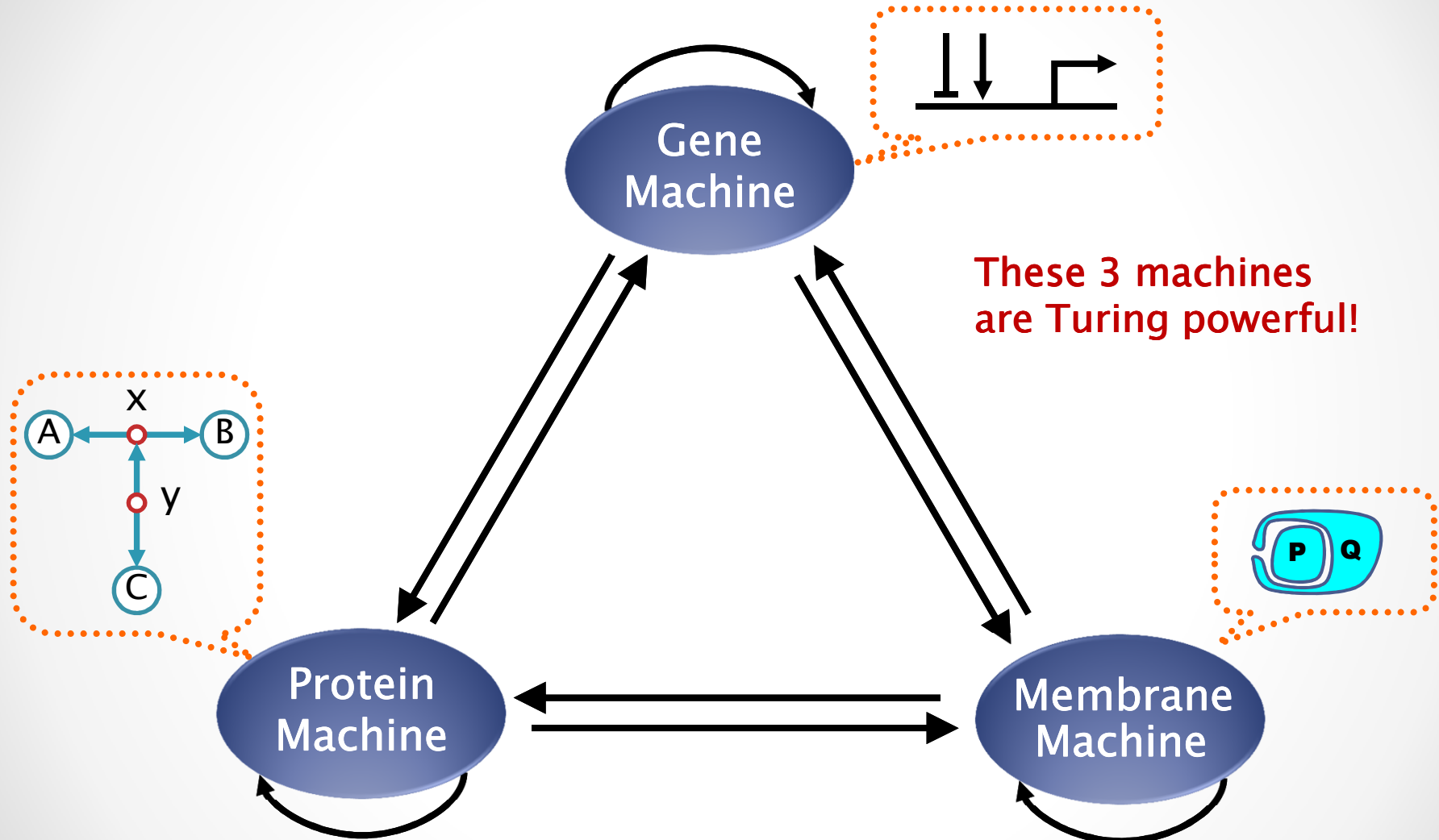
Biochemical Networks



Transport Networks



# Algorithmic View



# Computational viewpoint

- Cells are computational engines
  - Their *primary* function is information processing
    - Which controls feeding, escape, and reproduction.
    - Without properly processing information cells soon die (by starvation or predation).
    - Hence a strong pressure to process information better.
  - Which *happens* to be implemented by chemistry
    - Fundamental is not the ‘hardware’ (proteins etc.) which easily varies between organisms but the ‘software’ the runs on the hardware.
- So, what algorithms do they run?



# Reverse Engineering

- Q (traditional): What kind of dynamical system is the cell-cycle switch?
- A (traditional): Bistability – ultrasensitivity – hysteresis ...  
Focused on how unstructured sub-populations change over time.
  
- Q: What kind of algorithmic system is the cell-cycle switch?
- A: Interaction – complexity – convergence ...  
Focused on individual molecules as programmable, structured, algorithmic entities.
  
- Leading to a better understanding of not just the *function* but also the *network* (algorithm).

# Direct Engineering

- The AM algorithm was not learned from nature
  - CC was invented ~2.7 billions years ago.
  - AM was invented ~6 years ago (but independently).
- But nature may have more tricks
  - If there is some clever population algorithm out there, how will we recognize it?
  - We need to understand how nature operates.