

The computational power of execution bounded chemical reaction networks

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UC DAVIS
COMPUTER SCIENCE



Acknowledgments

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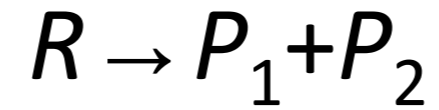


For teaching us about
"Theorems of the Alternative"

Chemical reaction networks

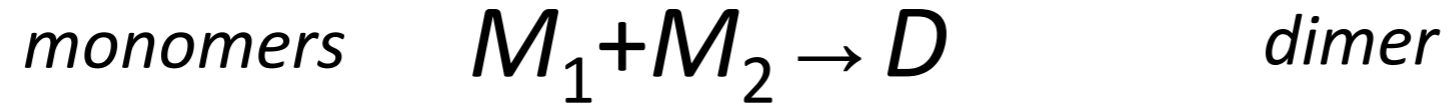
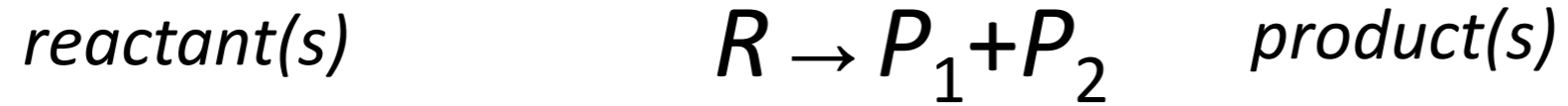
Chemical reaction networks

reactant(s)



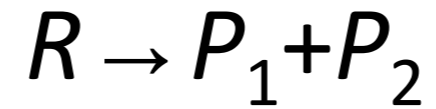
product(s)

Chemical reaction networks



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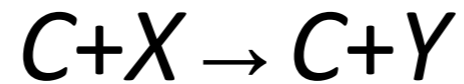
product(s)

monomers

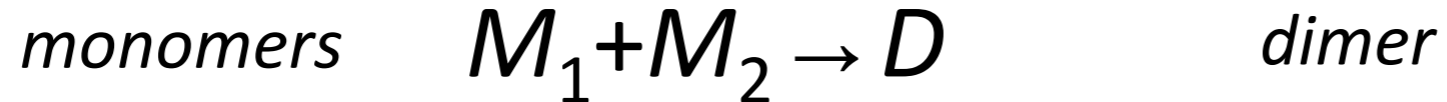
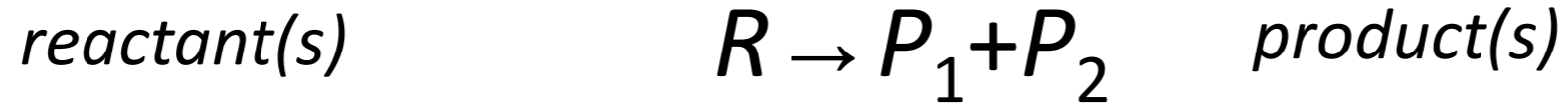


dimer

catalyst



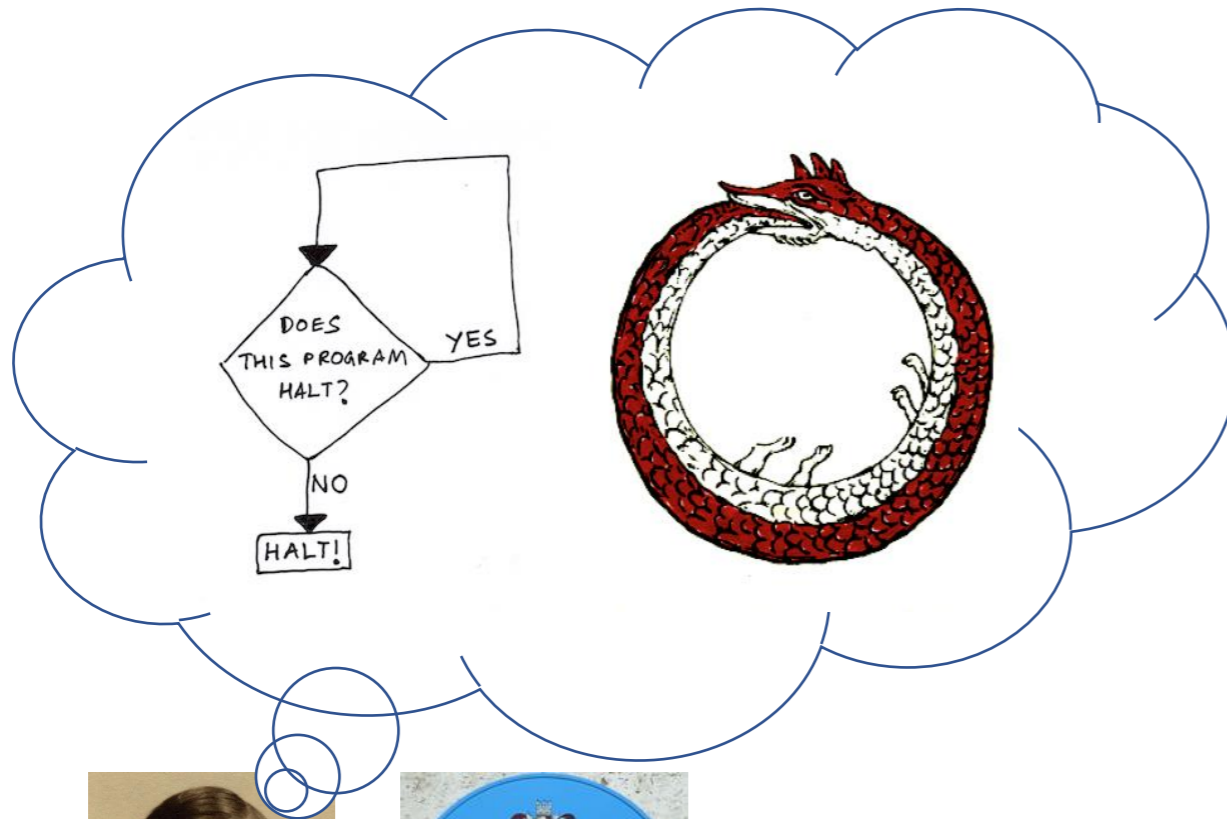
Chemical reaction networks



Traditionally a descriptive **modeling** language...

Let's instead use it as a prescriptive **programming** language

Theoretical Computer Science Approach



What computation is possible and what is not?
(Computability theory)

NP

NP-complete

- protein folding*
- Boolean satisfiability*
- Hamiltonian path*

integer factoring

P

- DNA sequence alignment*
- polynomial factoring*
- integer multiplication*
- shortest path*

What computations necessarily take a long time and what can be done quickly?
(Computational complexity theory)

Outline

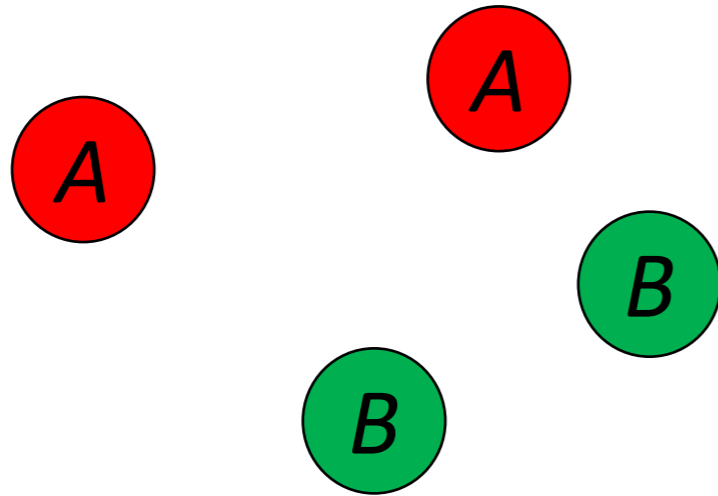
- **Formal definition of chemical reaction networks**
- Execution bounded chemical reaction networks and linear potential functions
- What is “computation” with chemical reactions?
- Limitations of computation with execution bounded chemical reaction networks
- Possibilities of computation with execution bounded chemical reaction networks

Chemical Reaction Network (CRN)

- finite set of d species $\Lambda = \{ A, B, C, D, \dots \}$
- finite set of reactions: *e.g.*
 $A + B \rightarrow A + C$
 $C \rightarrow A + A$
 $C + B \rightarrow C$
- state $\mathbf{x} \in \mathbb{N}^d$: molecular counts of each species

What is **possible**:

Example execution (reaction sequence)



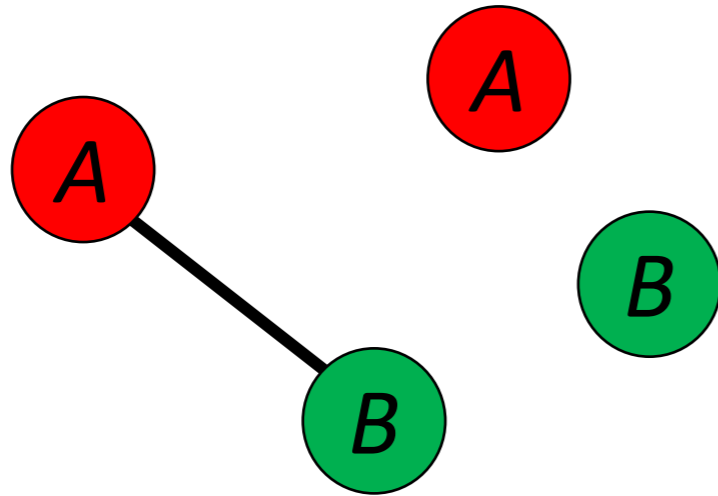
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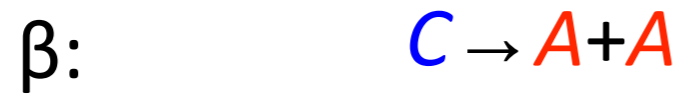
A B C

$x = (2, 2, 0)$



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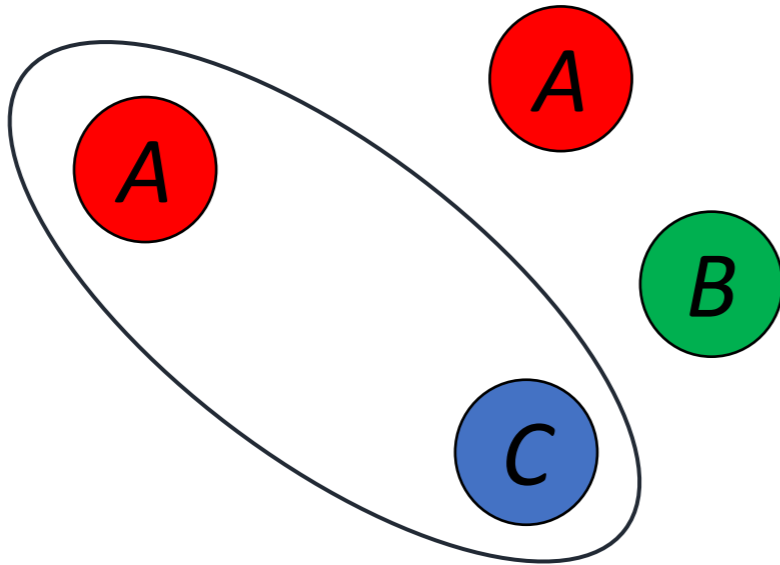


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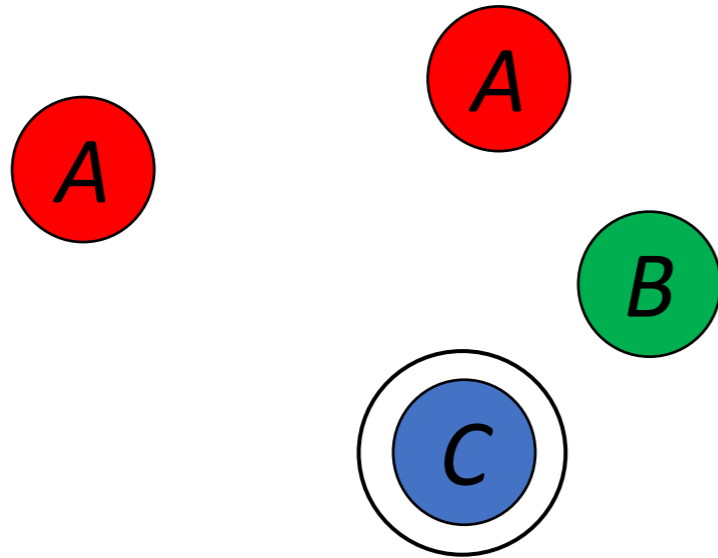
$\alpha \Downarrow$

$$(2, 1, 1)$$



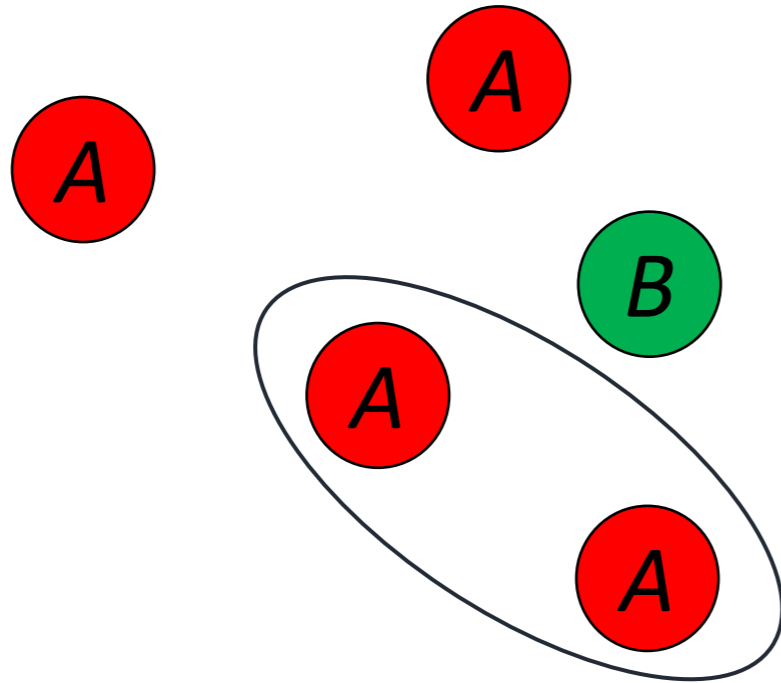
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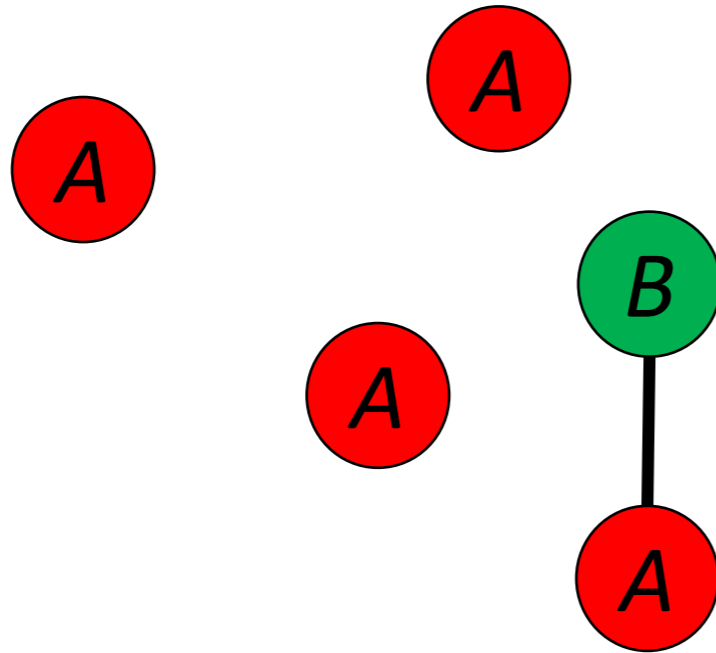
$\beta \Downarrow$

$$(4, 1, 0) \dots$$

$\Downarrow \alpha$

What is possible:

Example execution (reaction sequence)



$A \quad B \quad C$

$$\mathbf{x} = (2, 2, 0)$$

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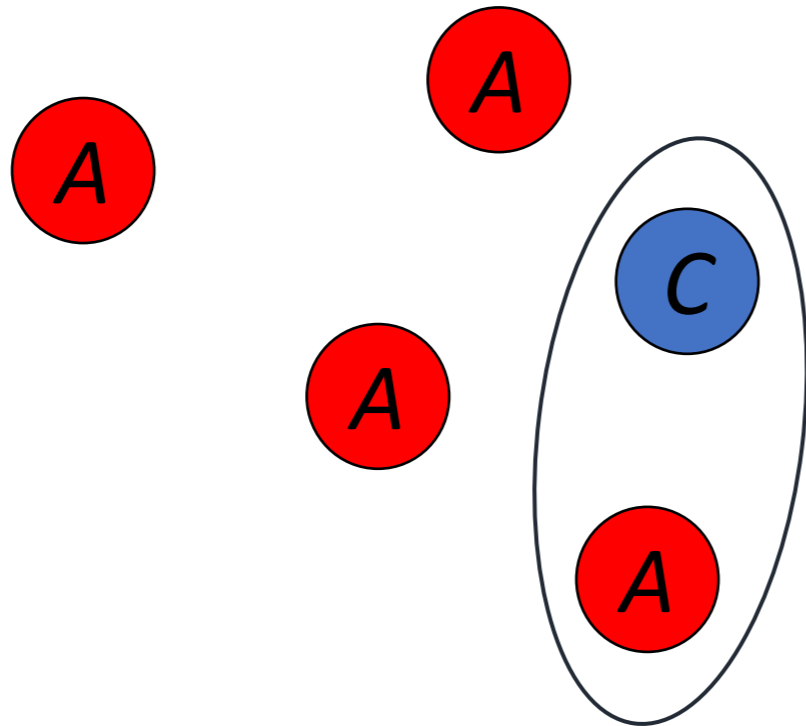
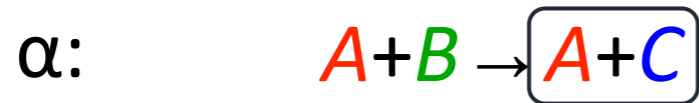
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$(4, 0, 1)$

\dots

Key property of reachability: additivity

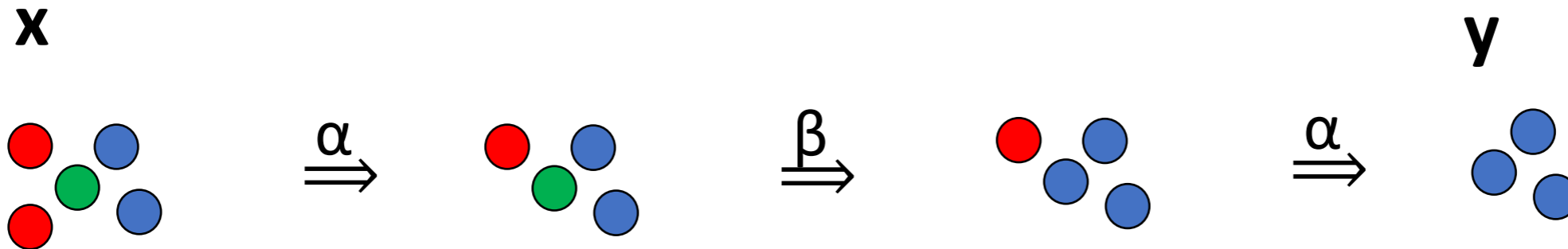
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The presence of extra molecules (represented by \mathbf{c}) cannot *prevent* reactions from occurring.

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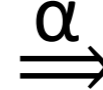
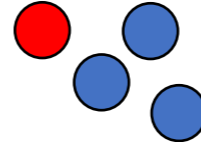
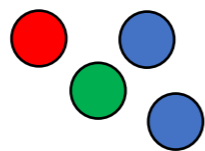
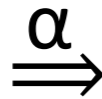
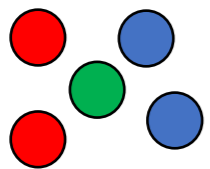
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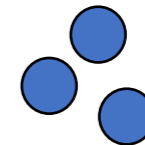
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\mathbf{x}



\mathbf{y}



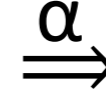
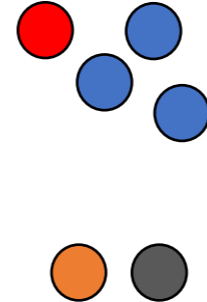
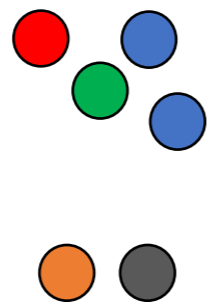
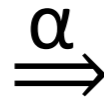
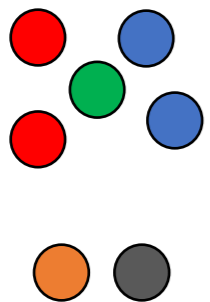
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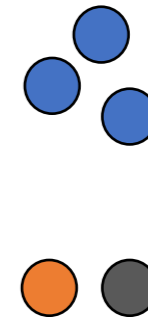
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$\mathbf{x} + \mathbf{c}$



$\mathbf{y} + \mathbf{c}$



Notation

- For vectors $\mathbf{x}, \mathbf{y} \in \mathbb{N}^d$
 - $\mathbf{x} \preceq \mathbf{y}$: $\mathbf{x}(i) \leq \mathbf{y}(i)$ for $1 \leq i \leq d$ $(1,2) \preceq (1,2)$
 - $\mathbf{x} \leq \mathbf{y}$: $\mathbf{x} \preceq \mathbf{y}$ and $\mathbf{x} \neq \mathbf{y}$ $(1,2) \leq (1,4)$
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 - If $\mathbf{x} \preceq \mathbf{0}$, \mathbf{x} is **nonnegative**. $(0,0)$
 - If $\mathbf{x} \geq \mathbf{0}$, \mathbf{x} is **semipositive**. $(0,1)$
 - If $\mathbf{x} > \mathbf{0}$, \mathbf{x} is **positive**. $(1,1)$

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Execution bounded CRNs

- Definition: A CRN C is **execution bounded** from state \mathbf{x} if all executions starting at \mathbf{x} are finite.

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- Why prefer execution bounded CRNs?
 - Wet lab implementations of CRNs use up “fuel” to execute reactions; execution bounded CRNs limit the amount of fuel needed
 - Easier to reason about: as long as reactions keep happening, they make “progress” towards reaching a final state.

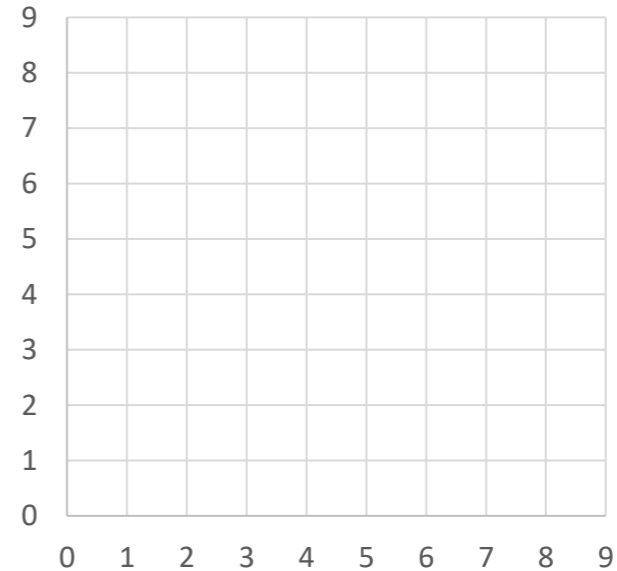
Self-covering executions

Easy Lemma: CRN C is not execution bounded from \mathbf{x}_0 if and only if there is an execution $(\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \dots)$ that is **self-covering**: $\mathbf{x}_i \preceq \mathbf{x}_k$ for some $i < k$.

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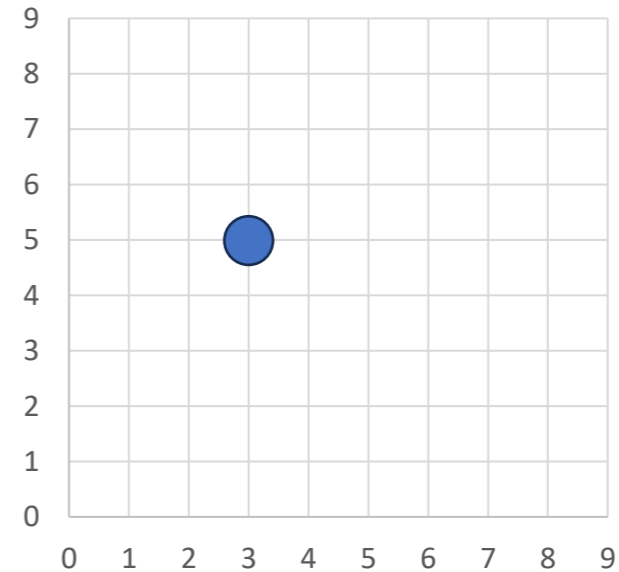
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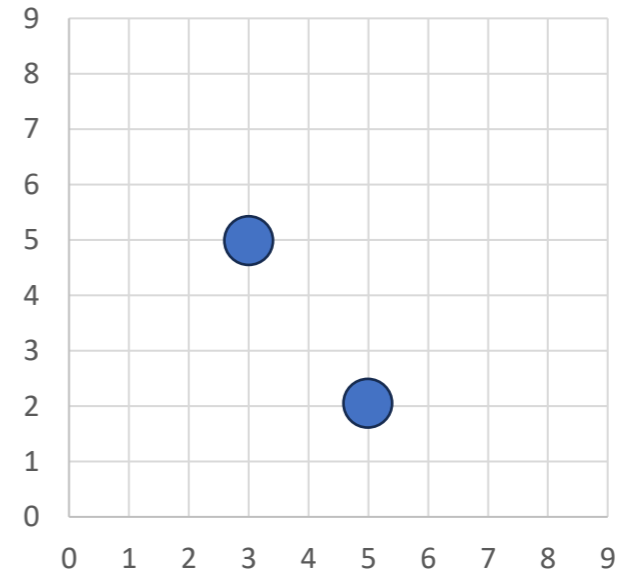
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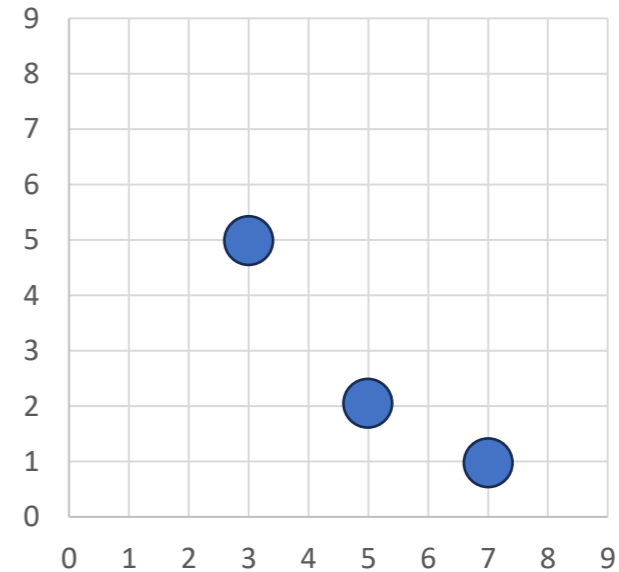
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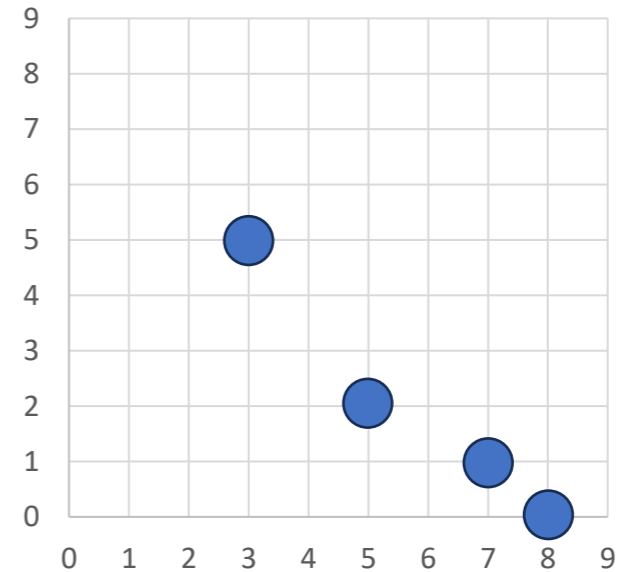
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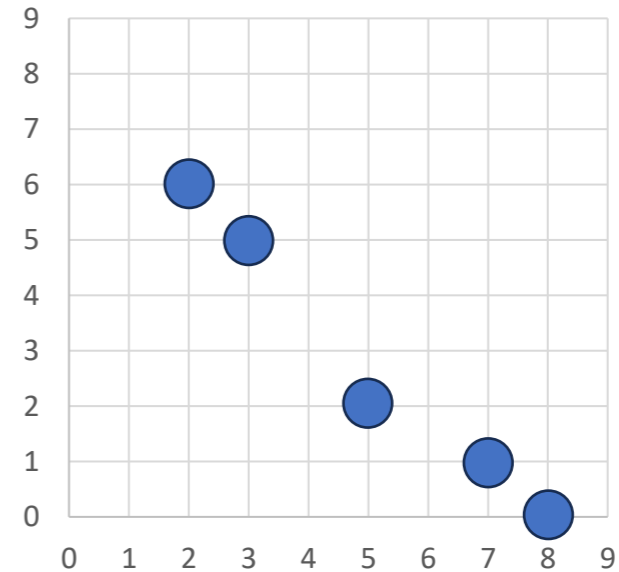
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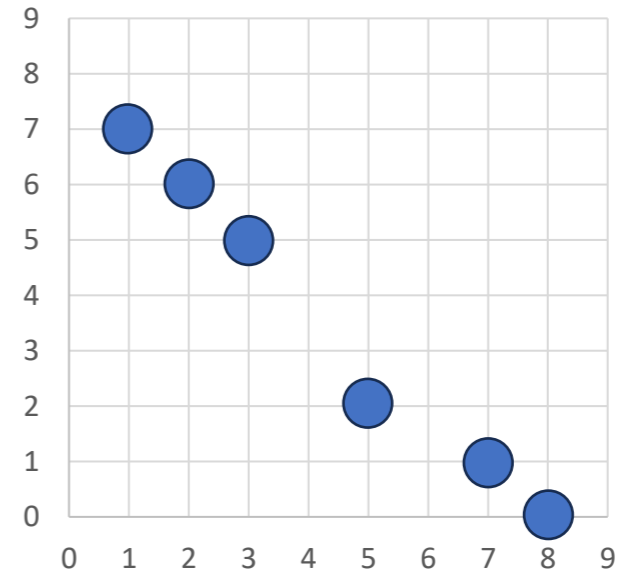
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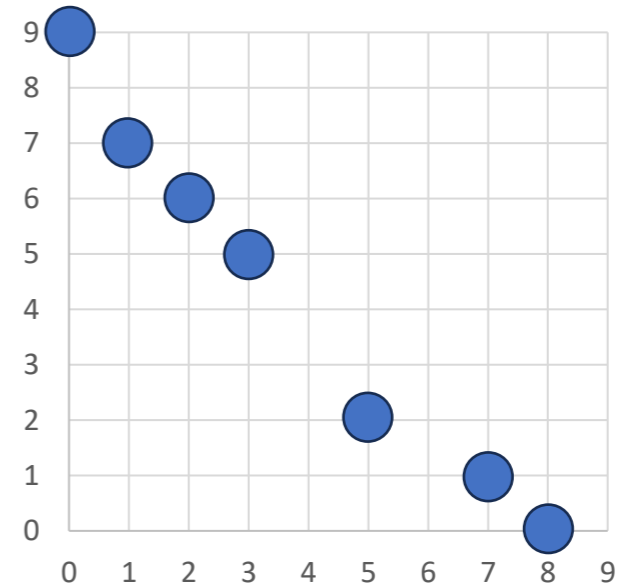
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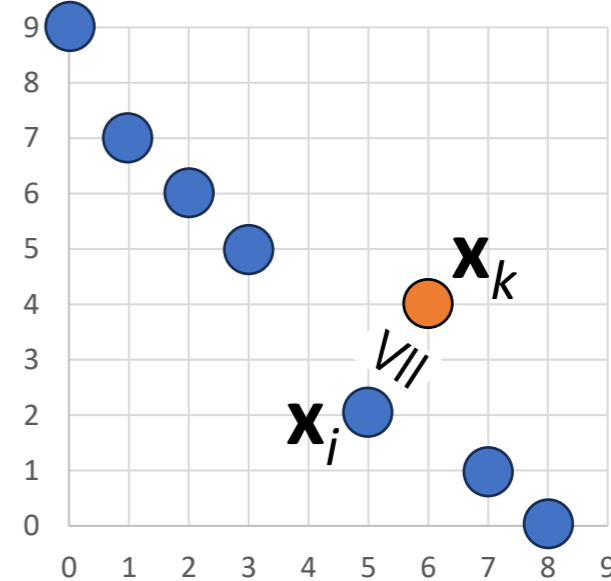
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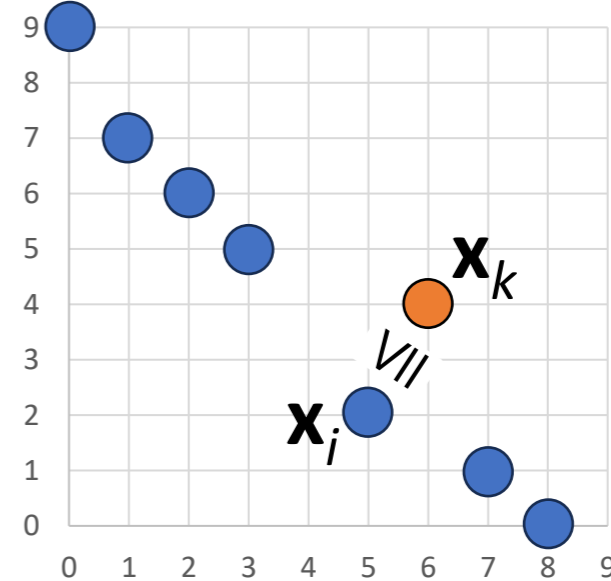


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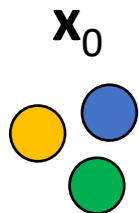
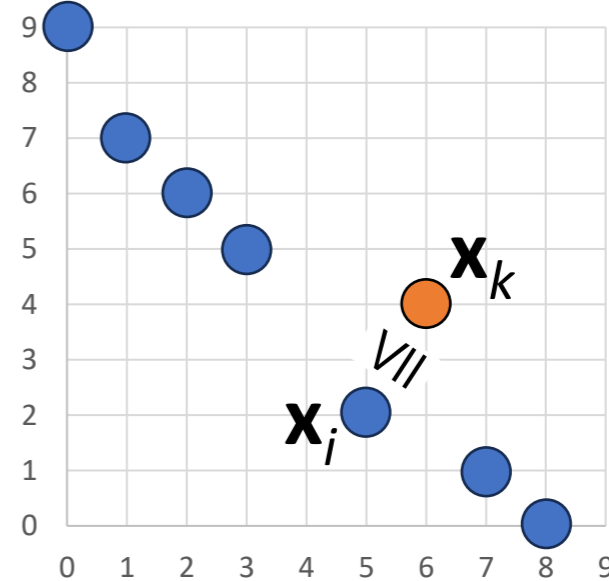


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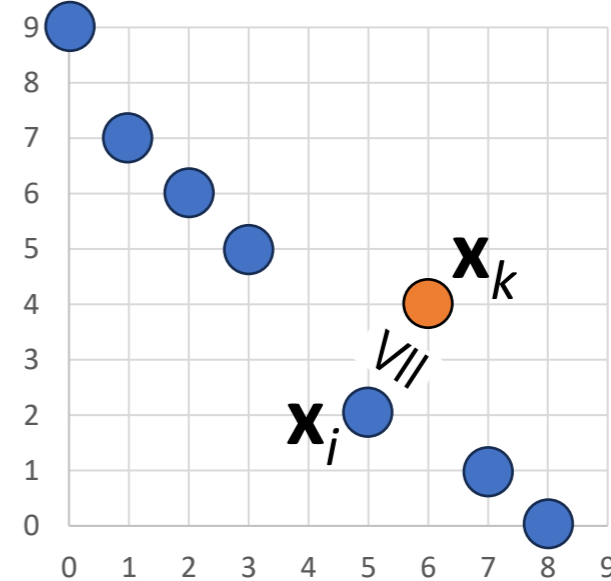
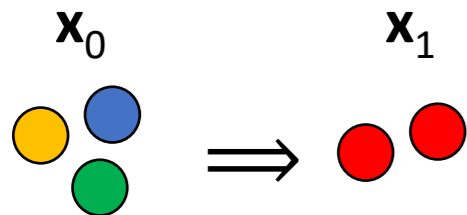


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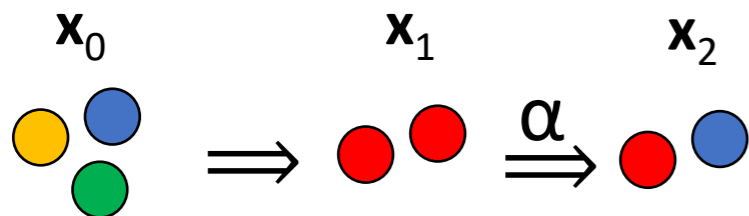
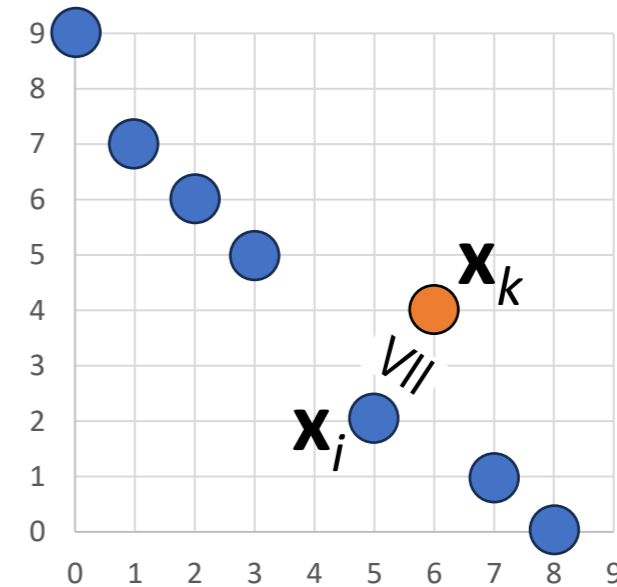


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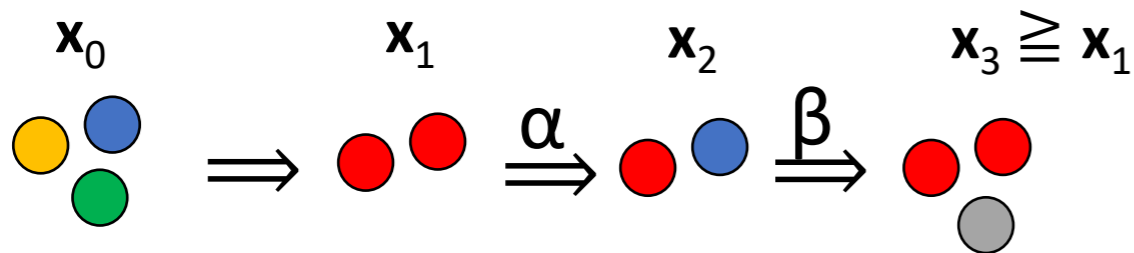
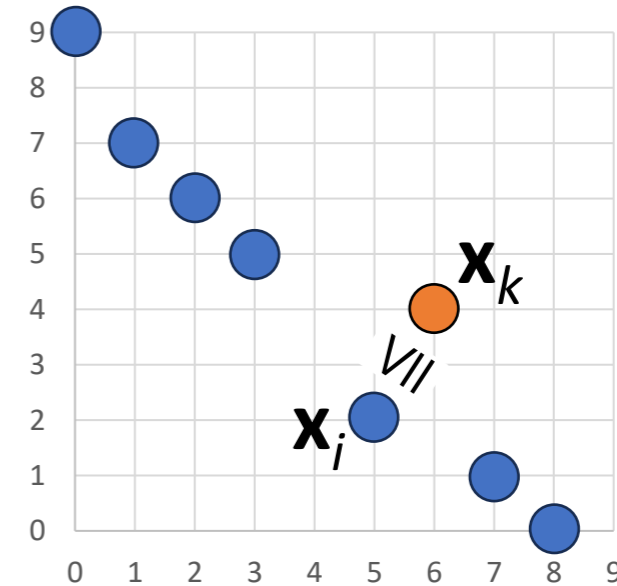


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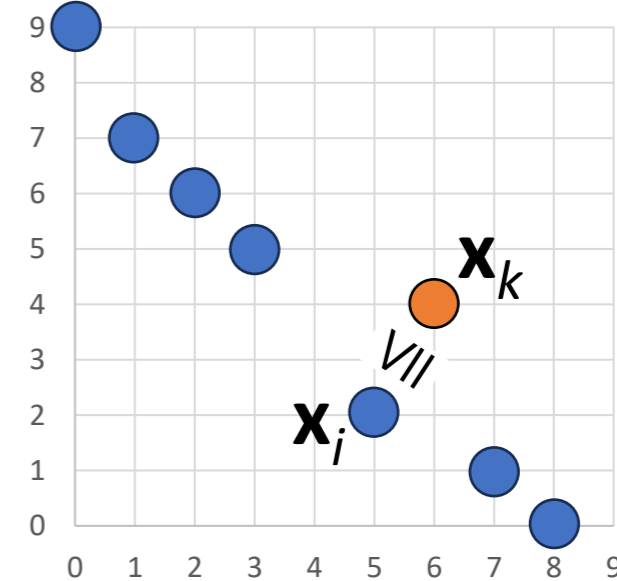
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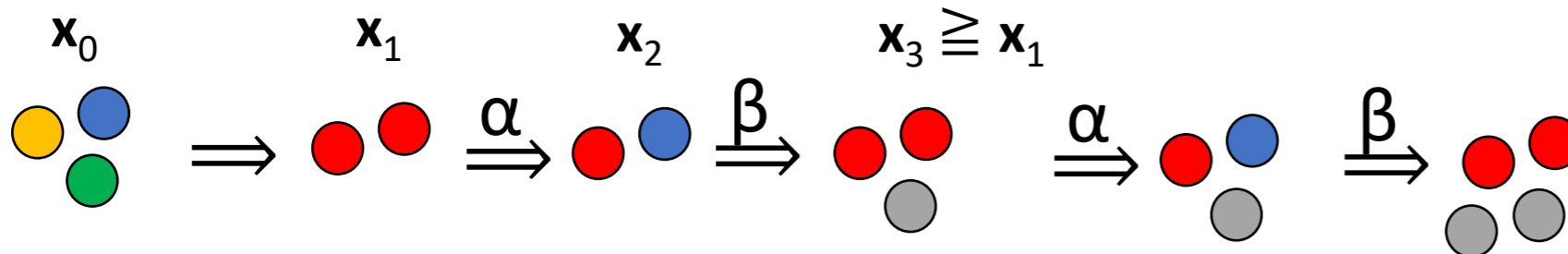
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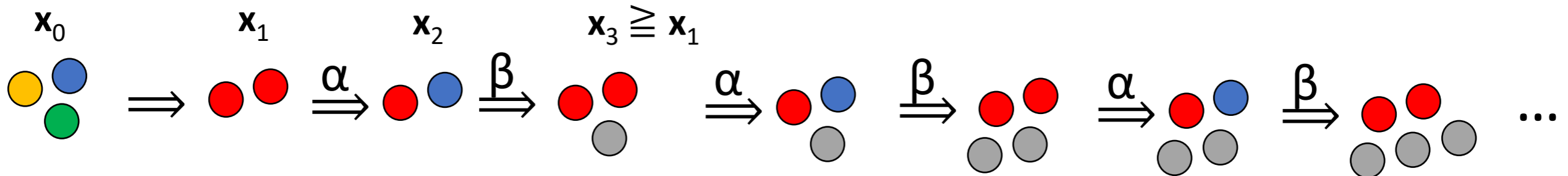
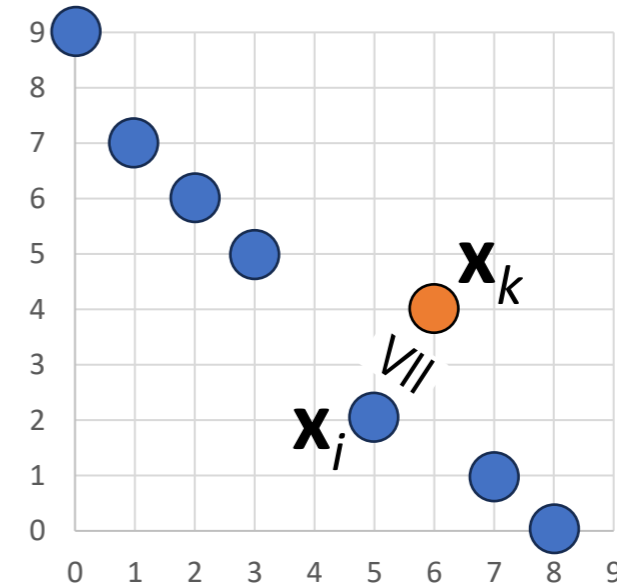


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- Definition: $\Phi: \mathbb{N}^d \rightarrow \mathbb{R}_{\geq 0}$ is a **linear potential function** for a CRN if it is a nonnegative linear function of states that every reaction strictly decreases.

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- By clearing denominators, we can assume each v_S is an integer, so each reaction decreases Φ by at least 1.

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Theorem: A CRN has a linear potential function if and only if it is execution bounded from every state.

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Forward direction is easy: Since each reaction reduces Φ by at least 1, from any state \mathbf{x} , at most $\Phi(\mathbf{x})$ reactions are possible.

Key technical tool for reverse direction

Theorem: (Gale 1960) “*Theorem of the Alternative*” (similar to Farkas’ Lemma):
Let \mathbf{M} be a matrix. Then exactly one of the following statements is true:

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[David Gale. The Theory of Linear Economic Models. University of Chicago press, 1960.]

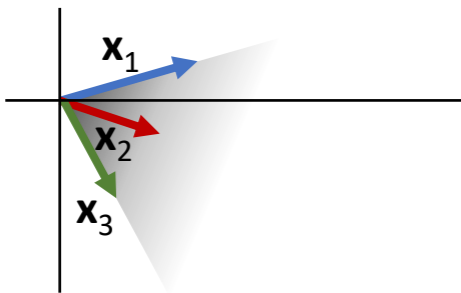
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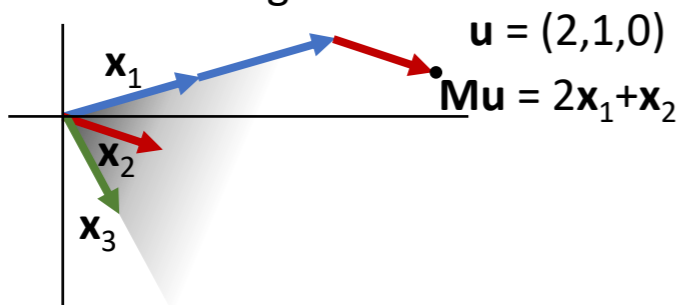
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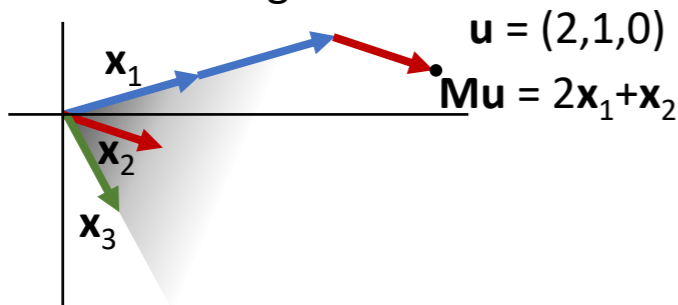
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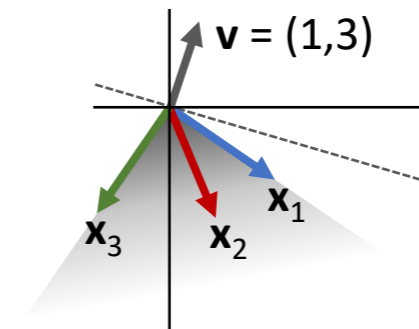
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2. Or it doesn't, and then some hyperplane (dashed line) separates that cone from the nonnegative orthant:

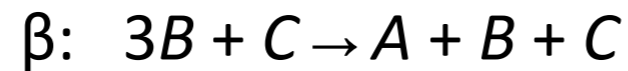
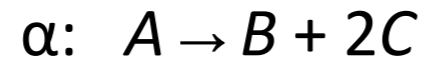


$$\mathbf{v}\mathbf{M} < \mathbf{0} \Rightarrow (\forall i) \mathbf{v} \cdot \mathbf{x}_i < 0$$

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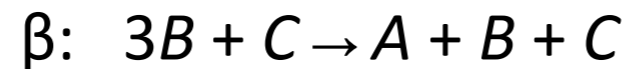
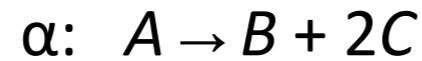
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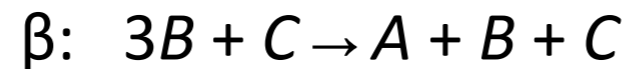
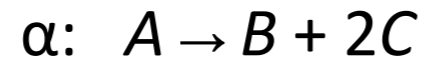


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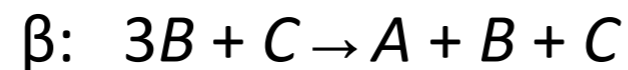
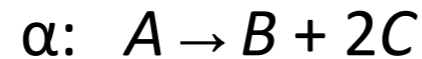
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Then there is a vector $\mathbf{v} \geq \mathbf{0}$ such that $\mathbf{v}\mathbf{M} < \mathbf{0}$. Let \mathbf{v} be the coefficients of a linear function $\Phi(\mathbf{x}) = \mathbf{v} \cdot \mathbf{x}$. Then $\mathbf{v}\mathbf{M} < \mathbf{0}$ means each reaction decreases Φ : it is a linear potential function. QED

Outline

- Formal definition of chemical reaction networks
- Execution bounded chemical reaction networks and linear potential functions
- **What is “computation” with chemical reactions?**
- Limitations of computation with execution bounded chemical reaction networks
- Possibilities of computation with execution bounded chemical reaction networks

Can we compute with chemistry?

“Not every chemical reaction network describes real chemicals!”, i.e. “where’s the *compiler*?”

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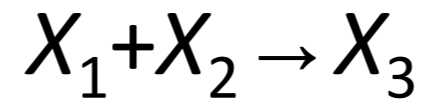
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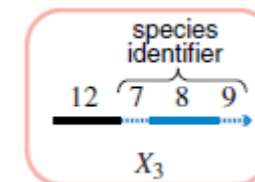
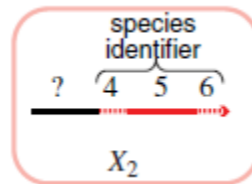
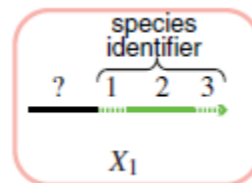
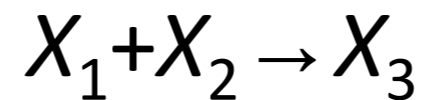
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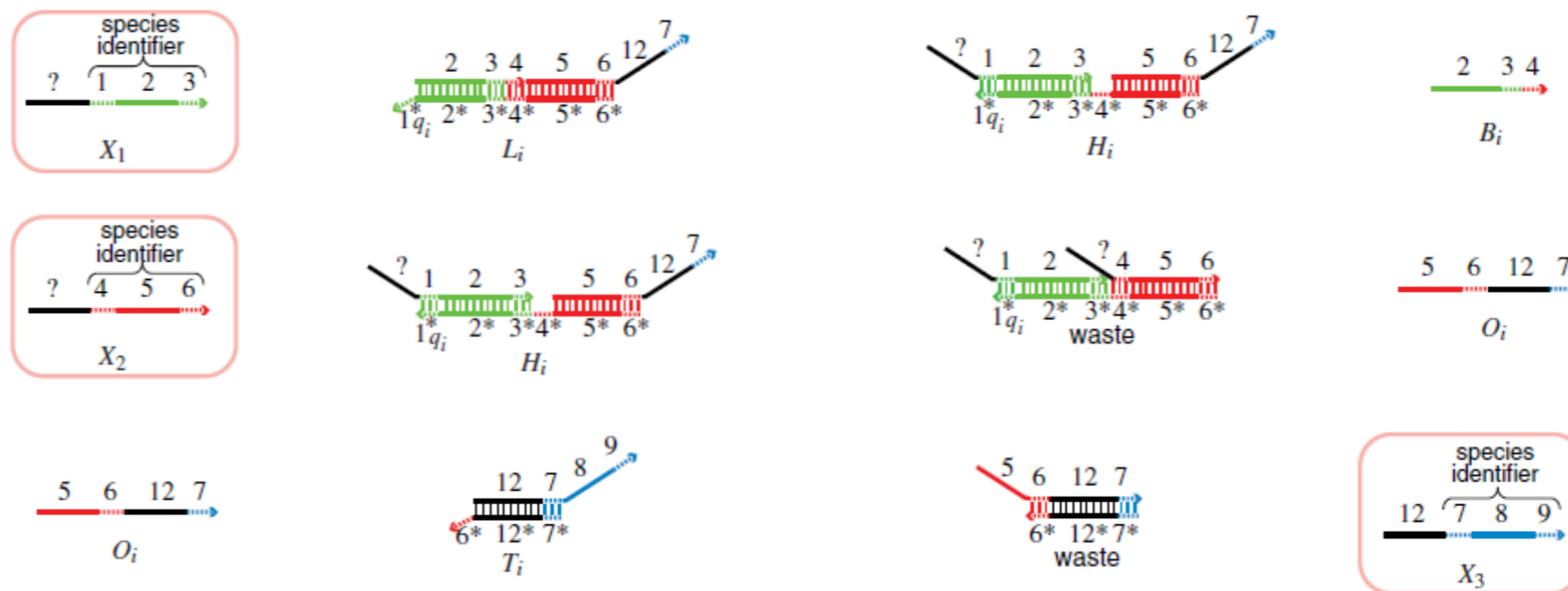
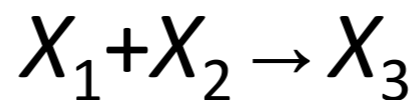
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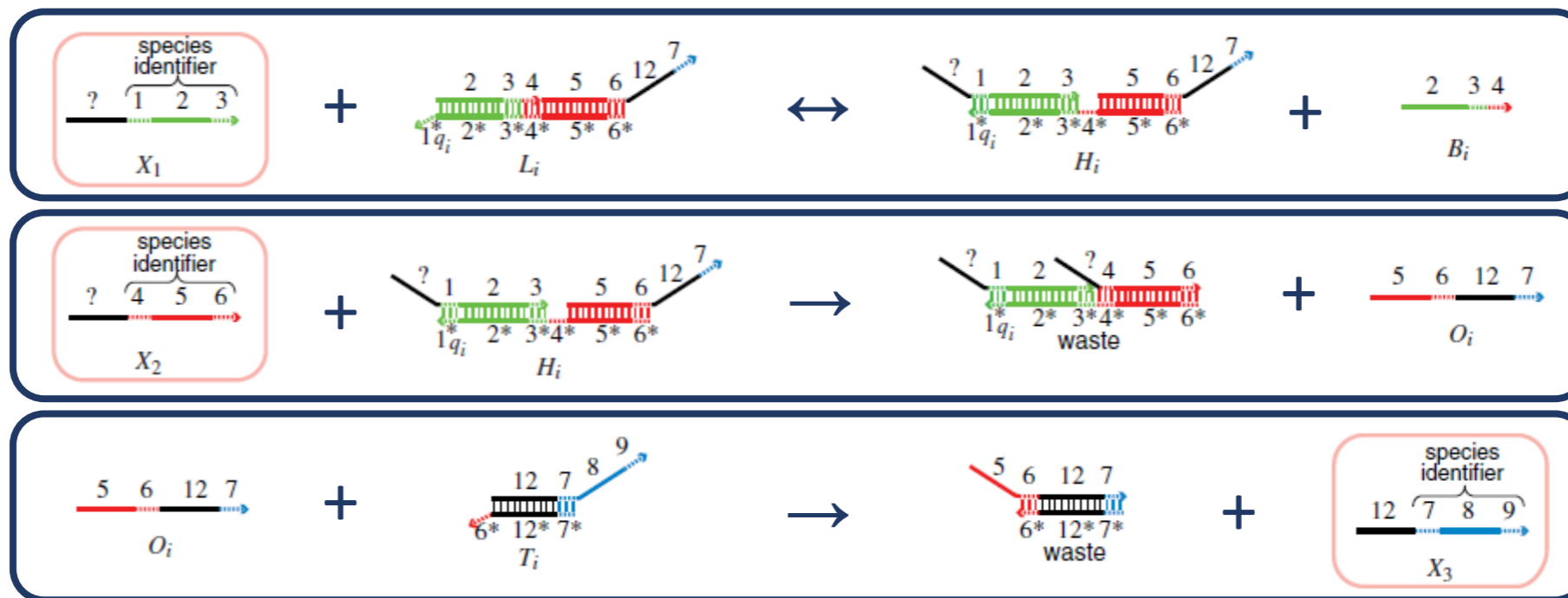
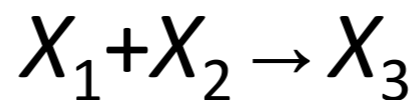
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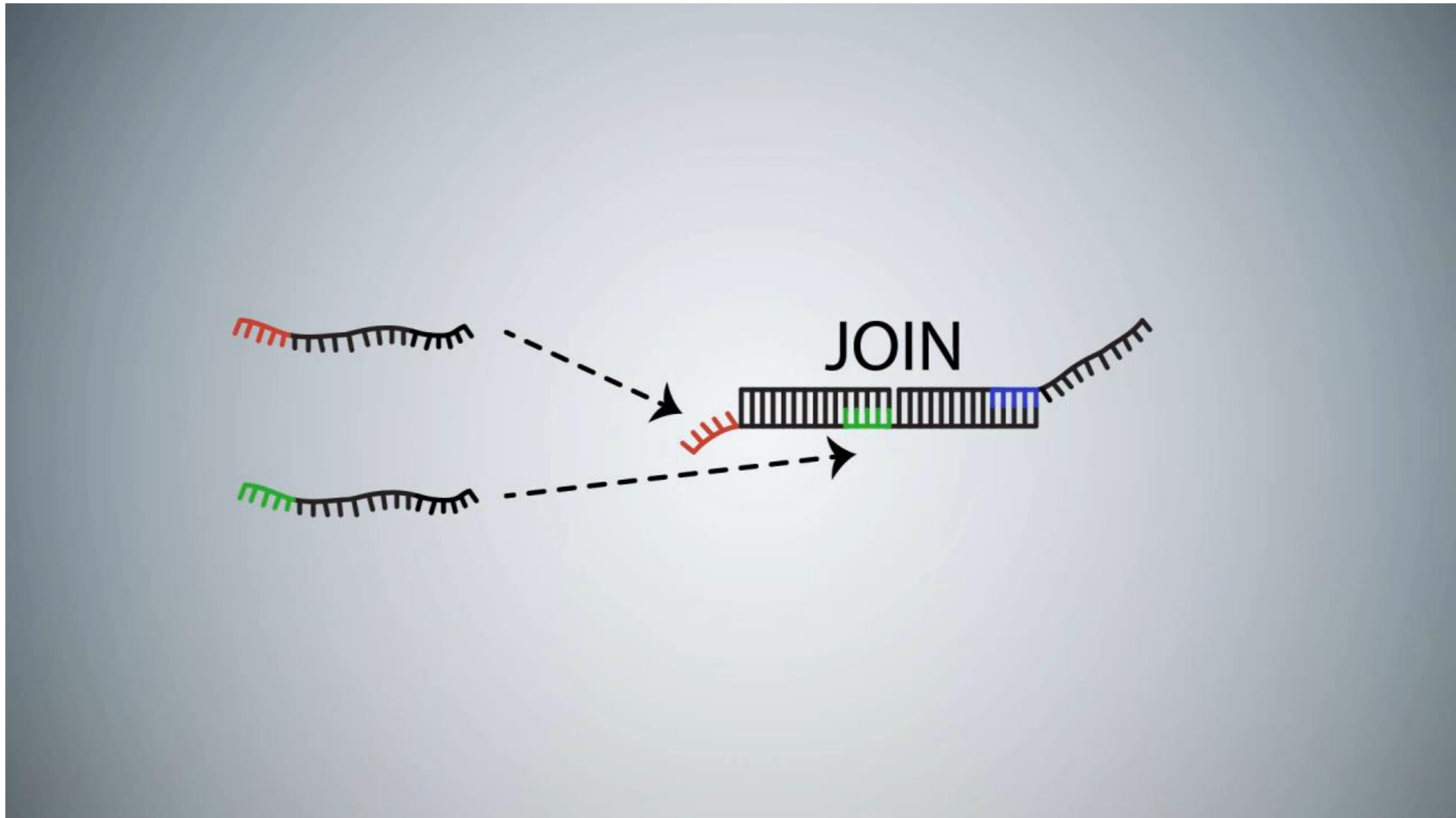
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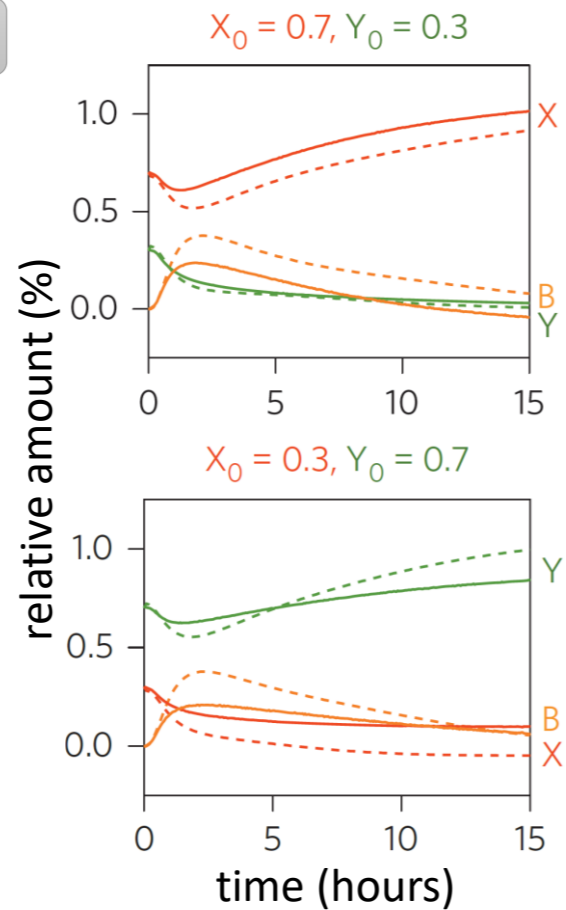
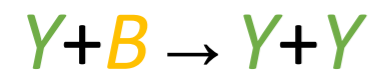
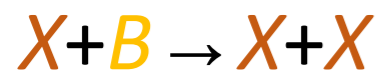
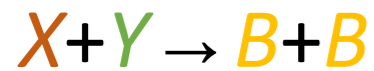
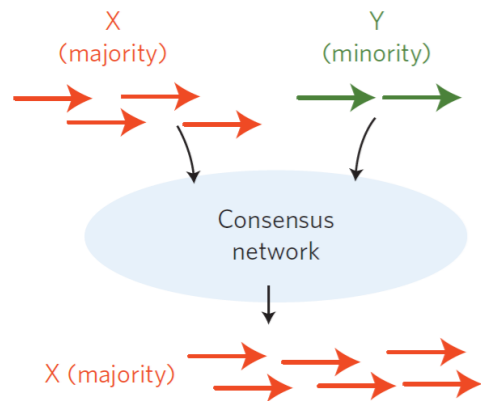


DNA strand displacement implementing $A+B \rightarrow C$



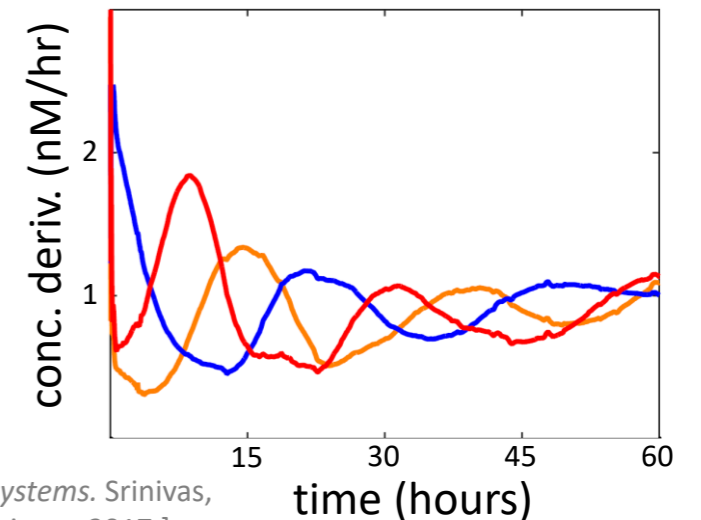
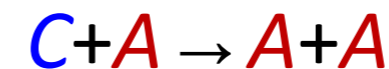
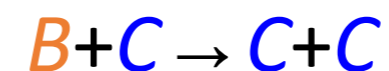
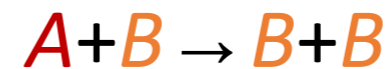
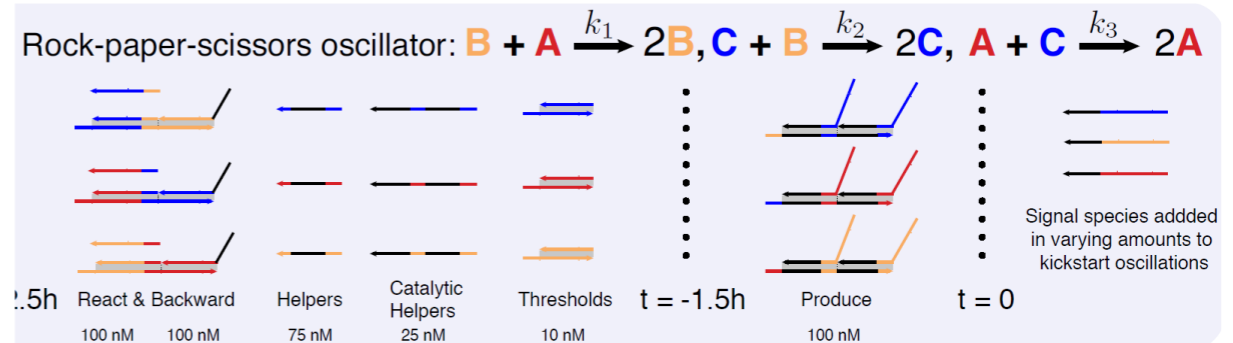
Experimental implementations of synthetic chemical reaction networks with DNA

Analog majority computation



[Programmable chemical controllers made from DNA. Chen, Dalchau, Srinivas, Phillips, Cardelli, Soloveichik, Seelig, *Nature Nanotechnology* 2013.]

Chemical oscillator

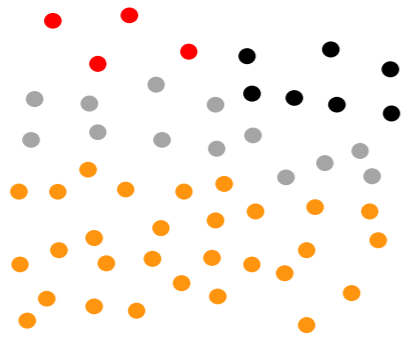


[Enzyme-free nucleic acid dynamical systems. Srinivas, Parkin, Seelig, Winfree, Soloveichik, *Science* 2017.]

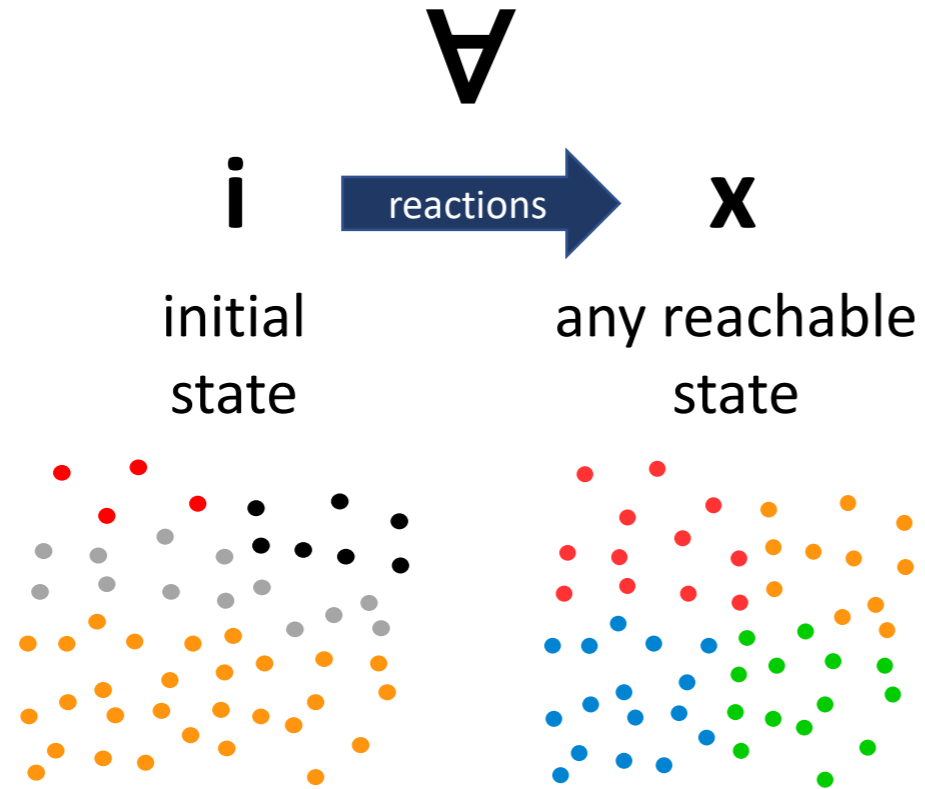
Defining **stable** computation

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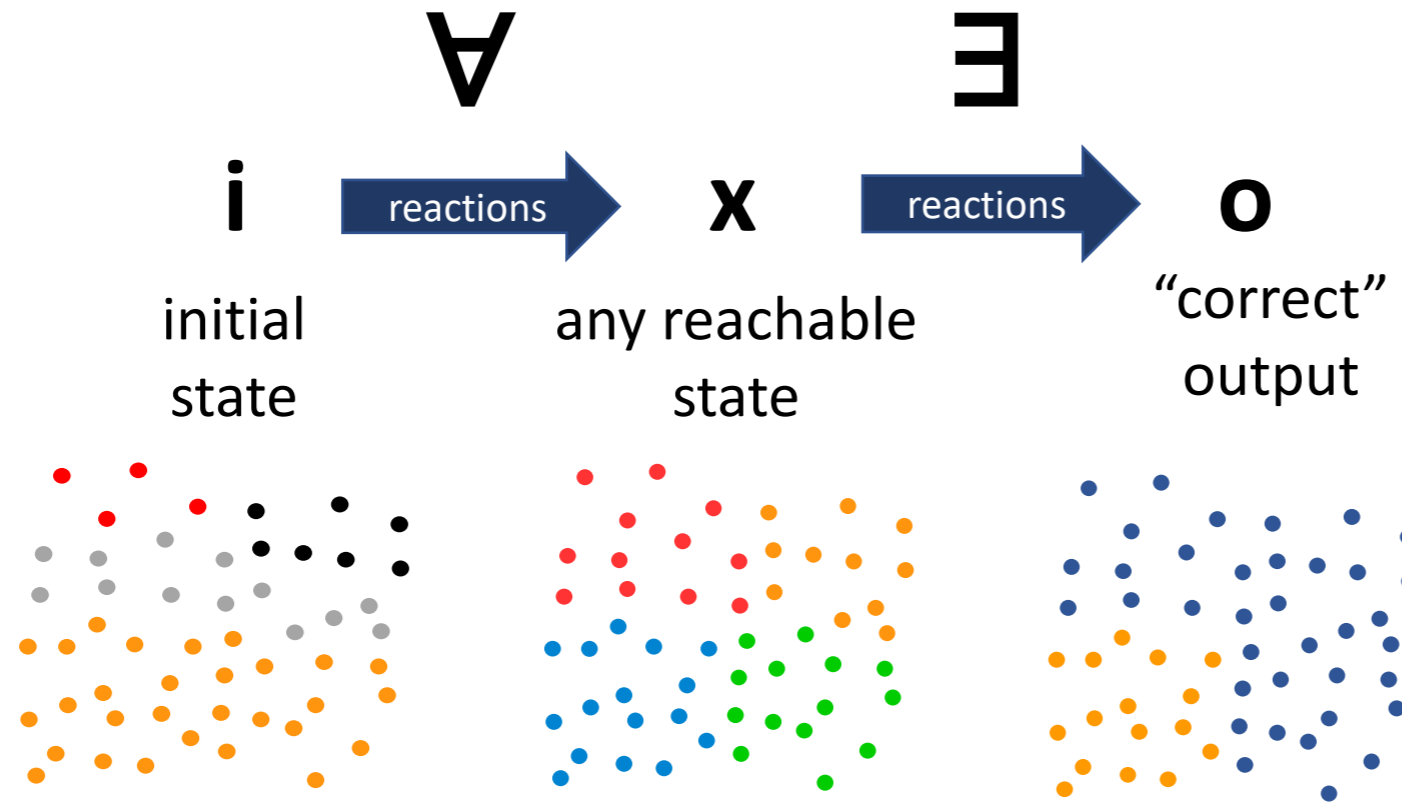
initial
state



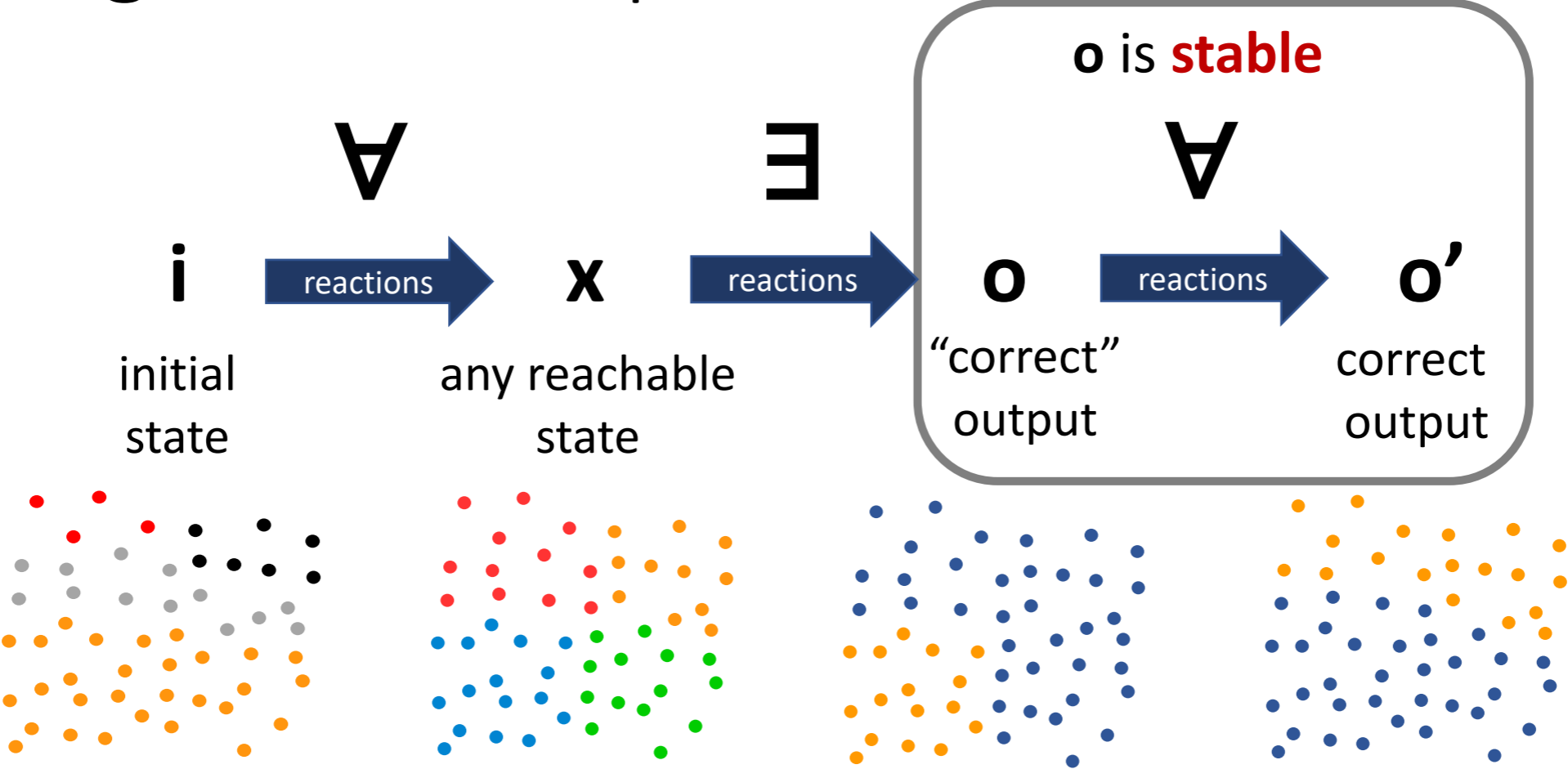
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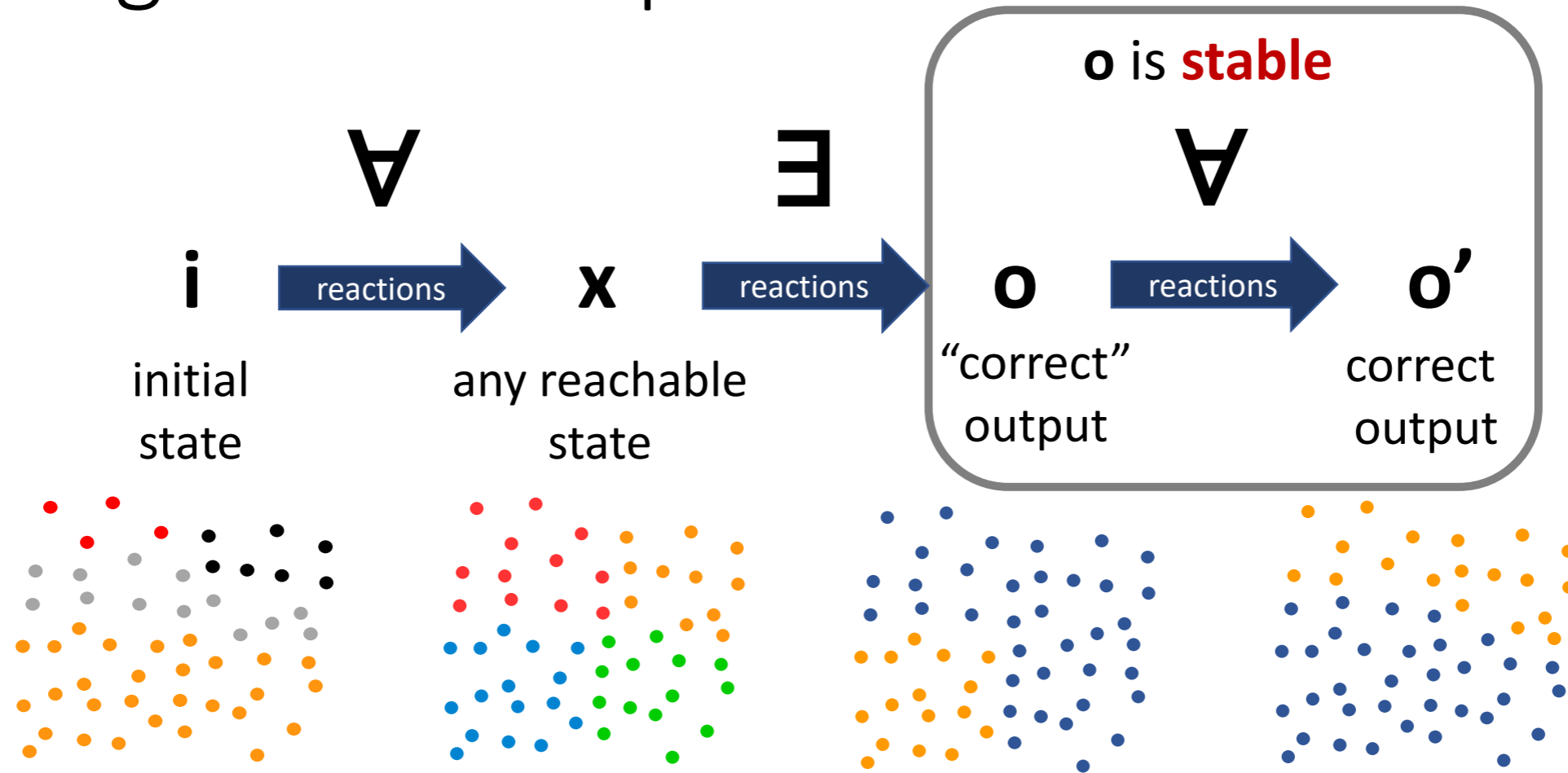
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(assuming finite set of reachable states) equivalent to:
The system will reach the correct output with probability 1.

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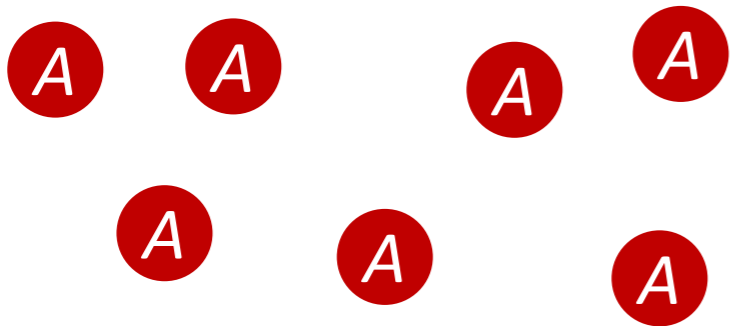
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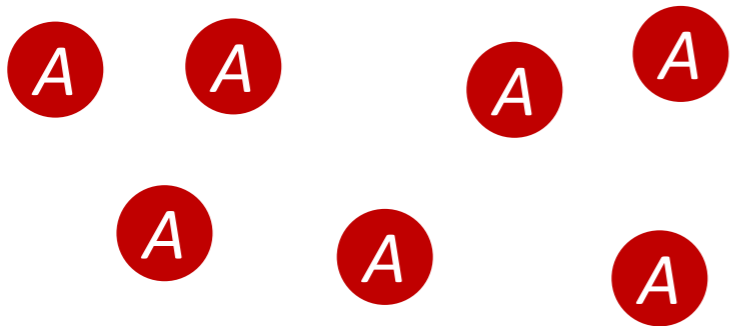


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$$B+A \rightarrow B+B$$

A votes no; B votes yes

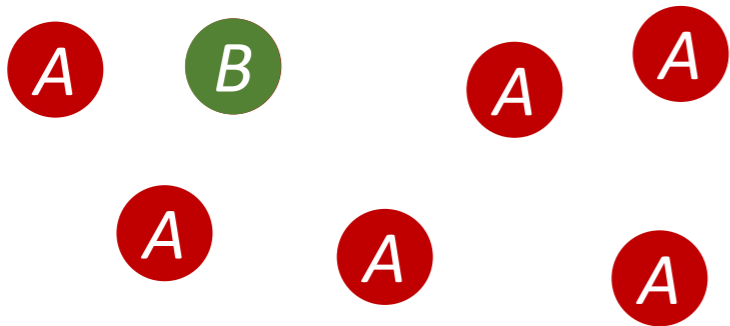


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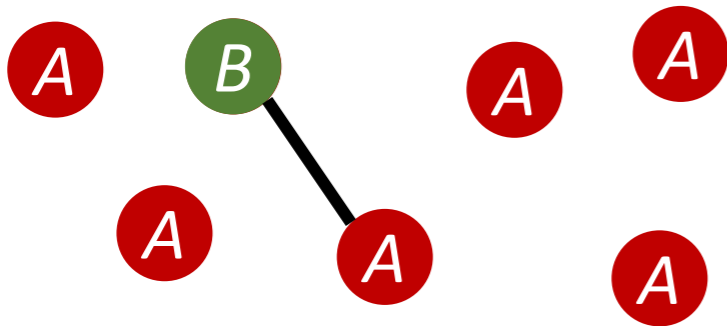


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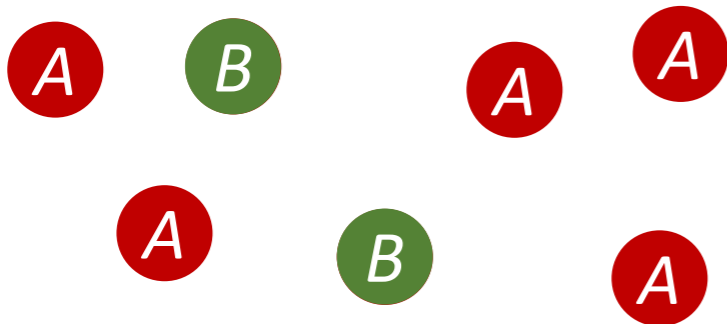


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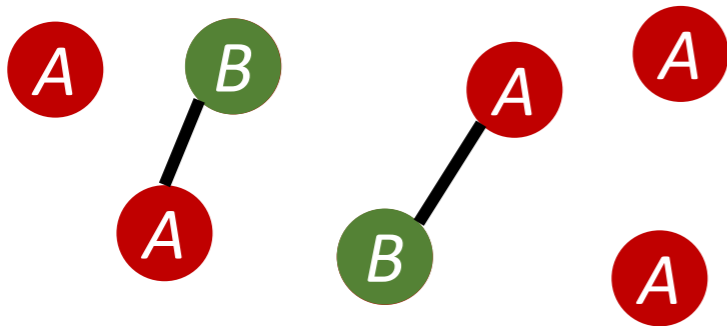


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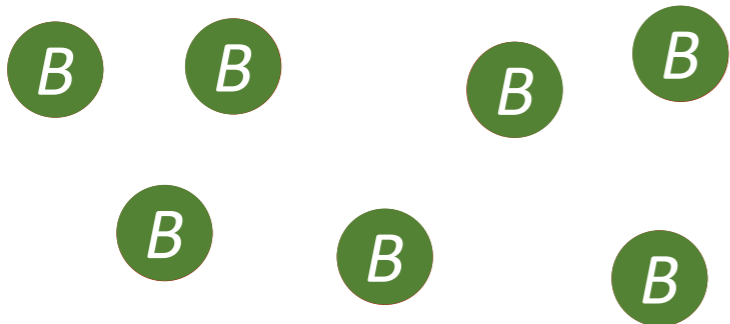


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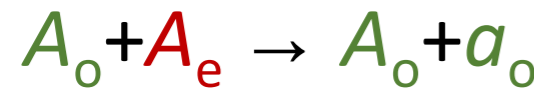
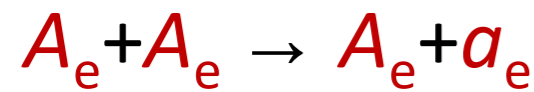
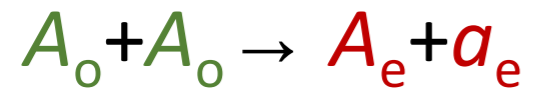
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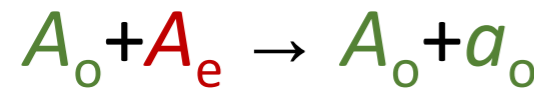
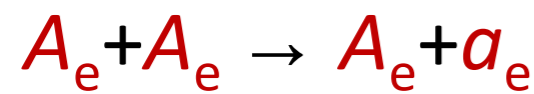
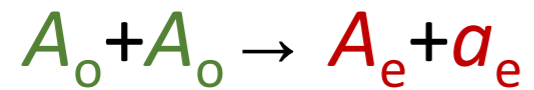


two leaders XOR their parity,
and one becomes follower

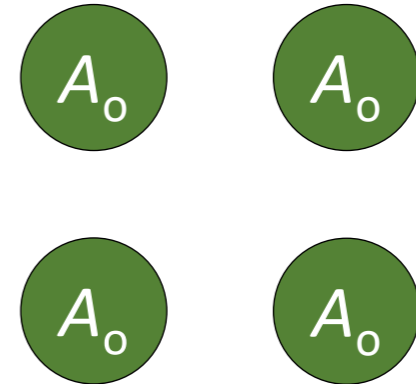
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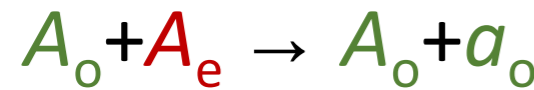
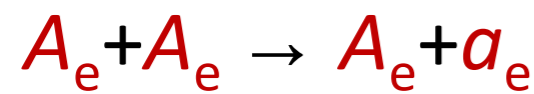
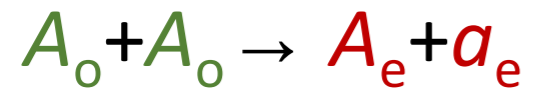
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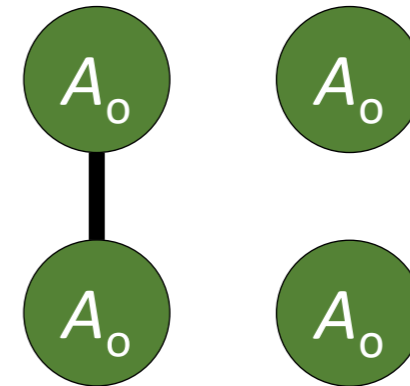
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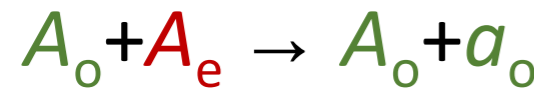
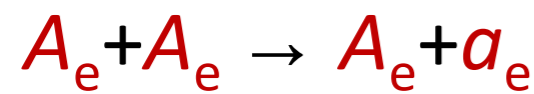
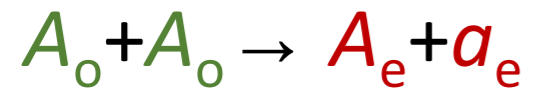
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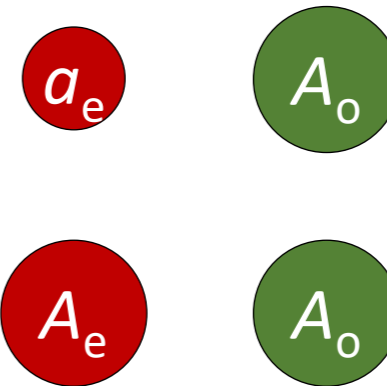
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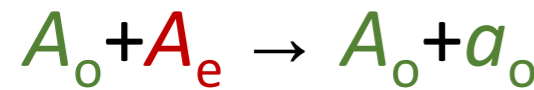
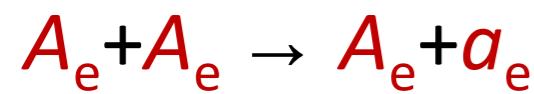
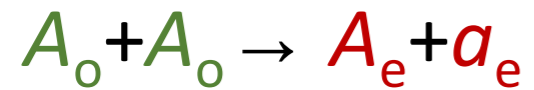
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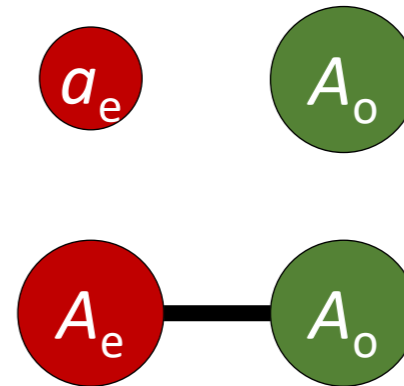
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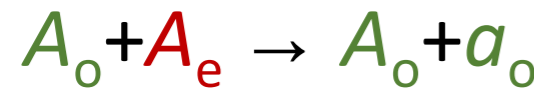
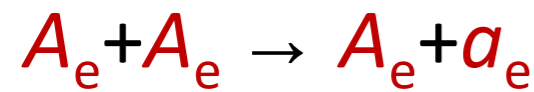
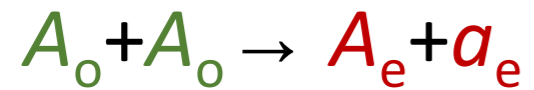
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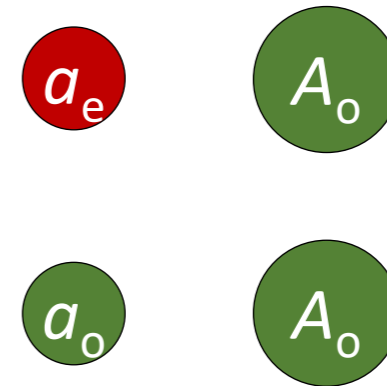
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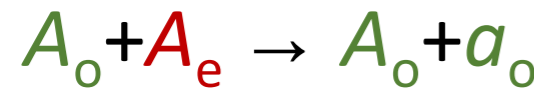
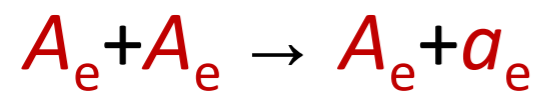
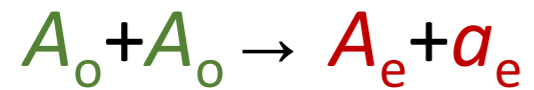
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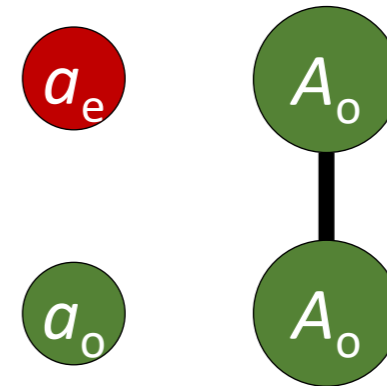
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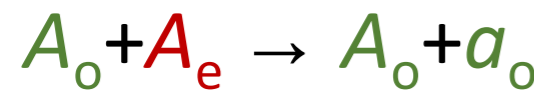
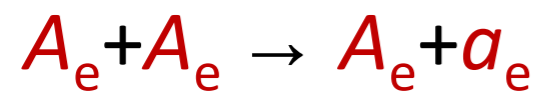
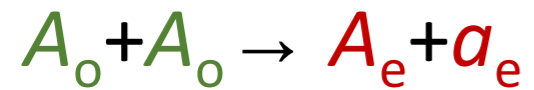
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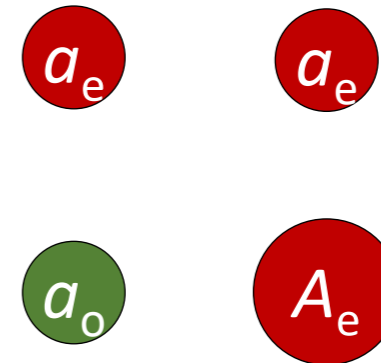
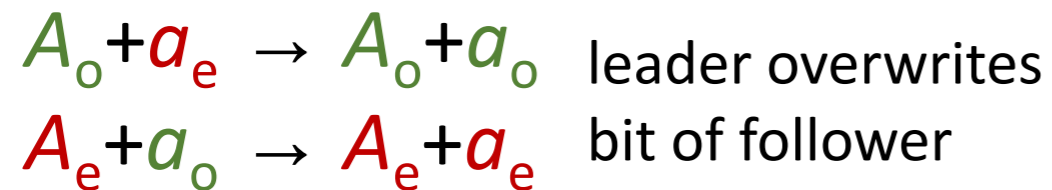
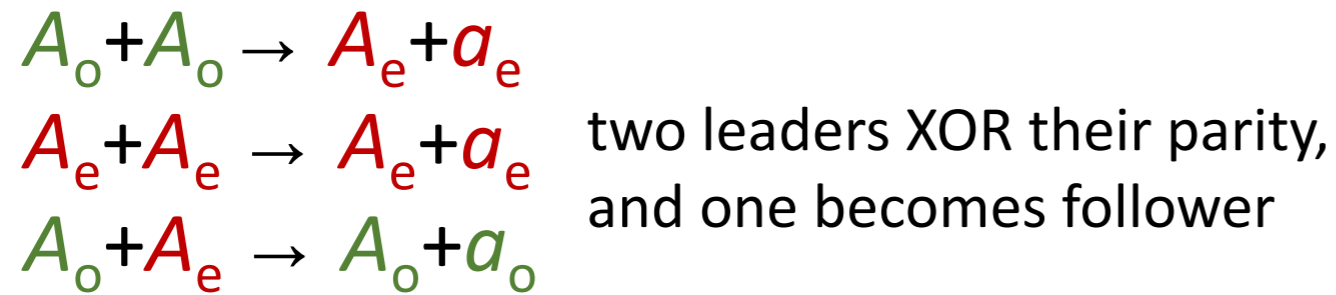
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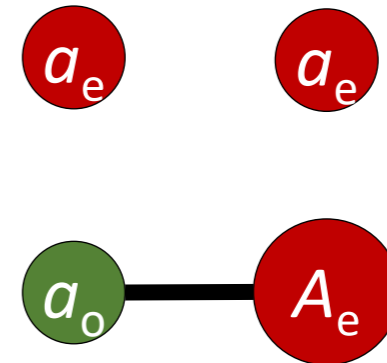
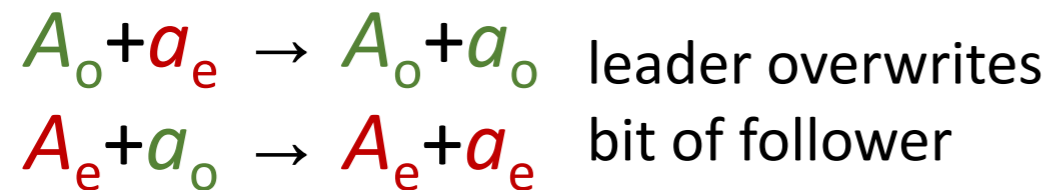
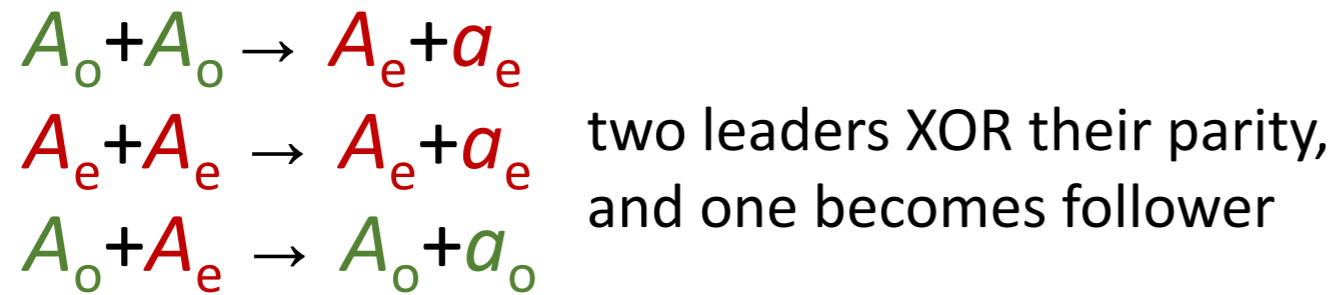
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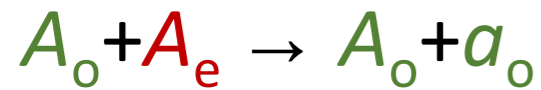
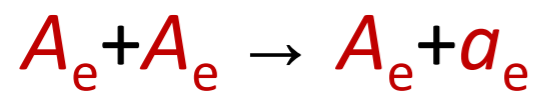
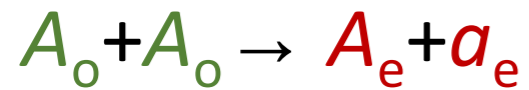
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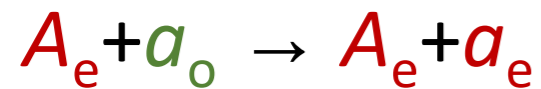
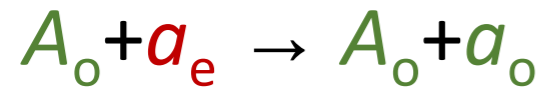
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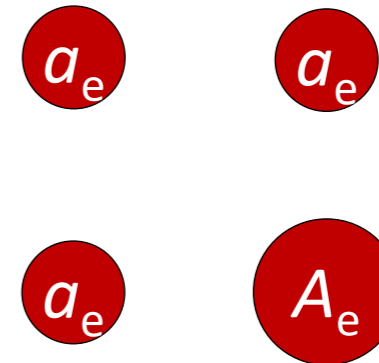
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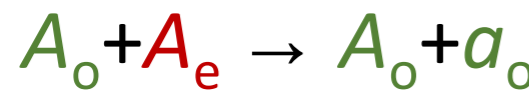
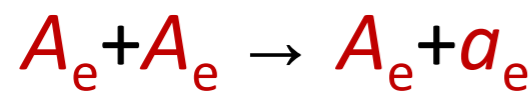
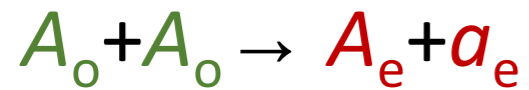
leader overwrites
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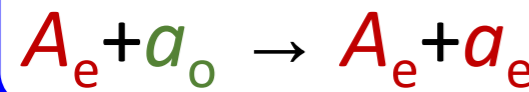
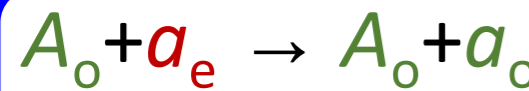
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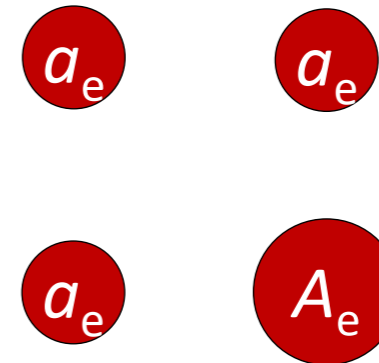
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Not execution bounded!

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[Draief, Vojnovic. Convergence speed of binary interval consensus. *SIAM Journal on Control and Optimization*, 50(3):1087–1109, 2012]

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Limits of stable computation

Theorem: $\varphi: \mathbb{N}^k \rightarrow \{Y, N\}$ is stably computable by a CRN if and only if φ is *semilinear*.

semilinear = Boolean combination of threshold and mod predicates:

take weighted sum $s = w_1 \cdot x_1 + \dots + w_k \cdot x_k$ of inputs $x_1 \dots x_k$ and ask if

$s \leq \text{constant } c$?

$s \equiv c \pmod{m}$ for constants c, m ?

$a > b$? $a = b$? a is odd? $a > 1$? $a > 1$ and b is odd?

NOT $a = b^2$? a is a power of 2? a is prime?

[Angluin, Aspnes, Diamadi, Fischer, Peralta, Computation in networks of passively mobile finite-state sensors, *PODC* 2004]

[Angluin, Aspnes, Eisenstat, Stably computable predicates are semilinear, *PODC* 2006]

Outline

- Formal definition of chemical reaction networks
- Execution bounded chemical reaction networks and linear potential functions
- What is “computation” with chemical reactions?
- **Limitations of computation with execution bounded chemical reaction networks**
- Possibilities of computation with execution bounded chemical reaction networks

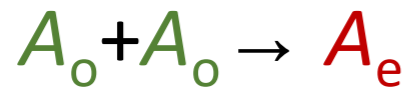
Noncollapsing CRNs

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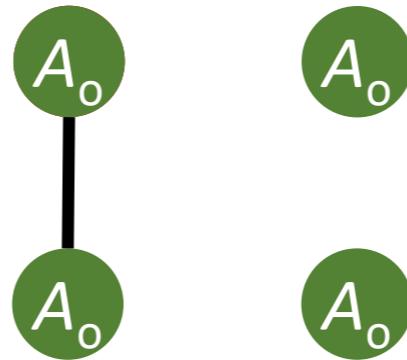
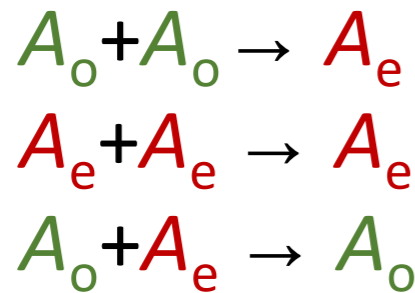


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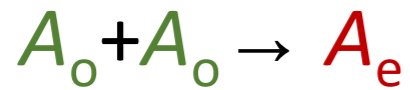


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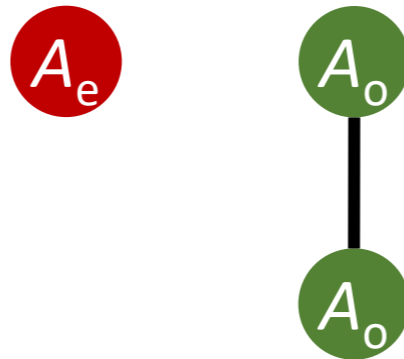
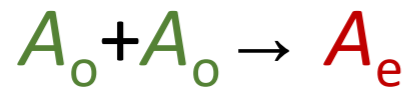


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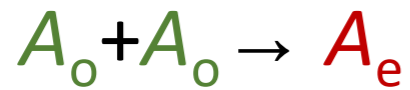


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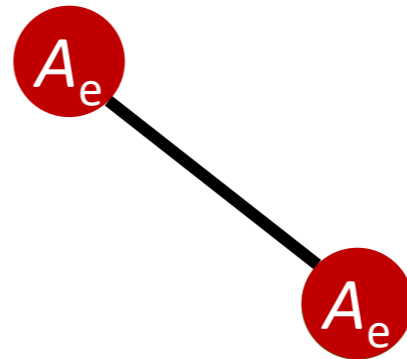
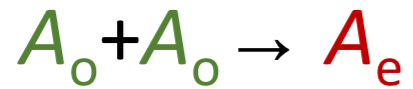


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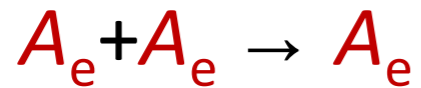
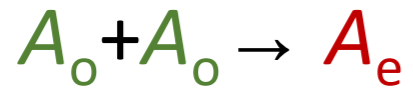


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Eventually constant predicates

Definition: A predicate $\varphi: \mathbb{N}^k \rightarrow \{Y, N\}$ is **eventually constant** if, for some $c \in \mathbb{N}$, $\varphi(\mathbf{x})$ is constant on all inputs $\mathbf{x} \geq (c, c, \dots, c)$.

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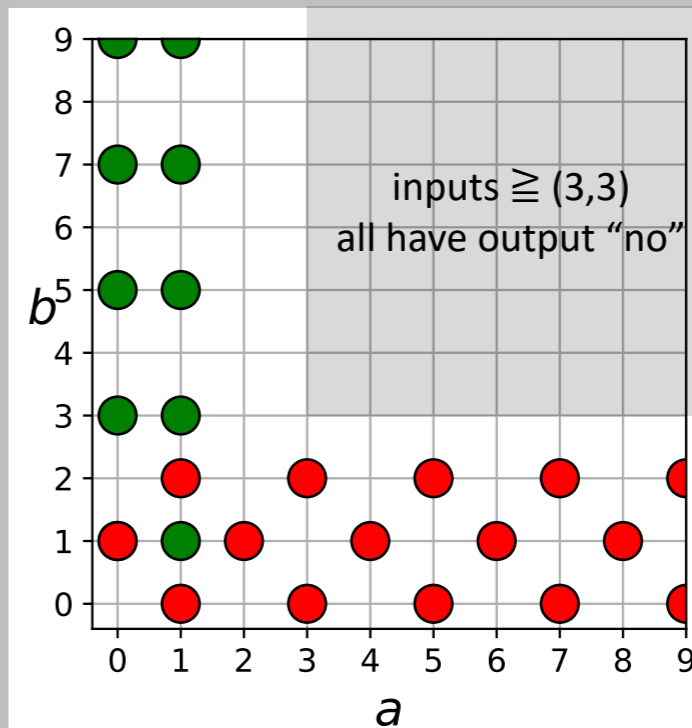
majority ($a \geq b$?)

parity (a is odd?)

equality ($a=b$?)

and most anything interesting.

Example of eventually constant predicate:
 $a < 2$ and b is odd, or $b < 3$ and $a+b$ is odd



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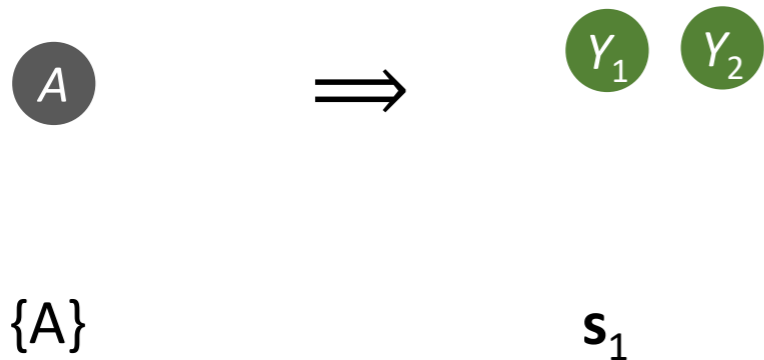
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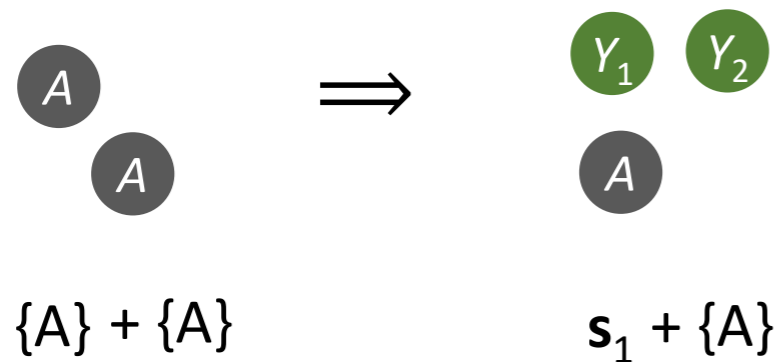
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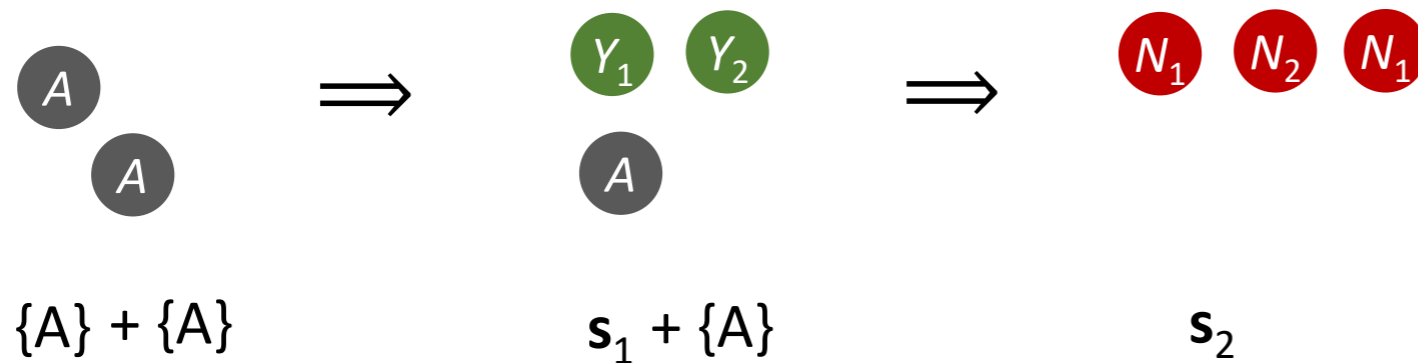
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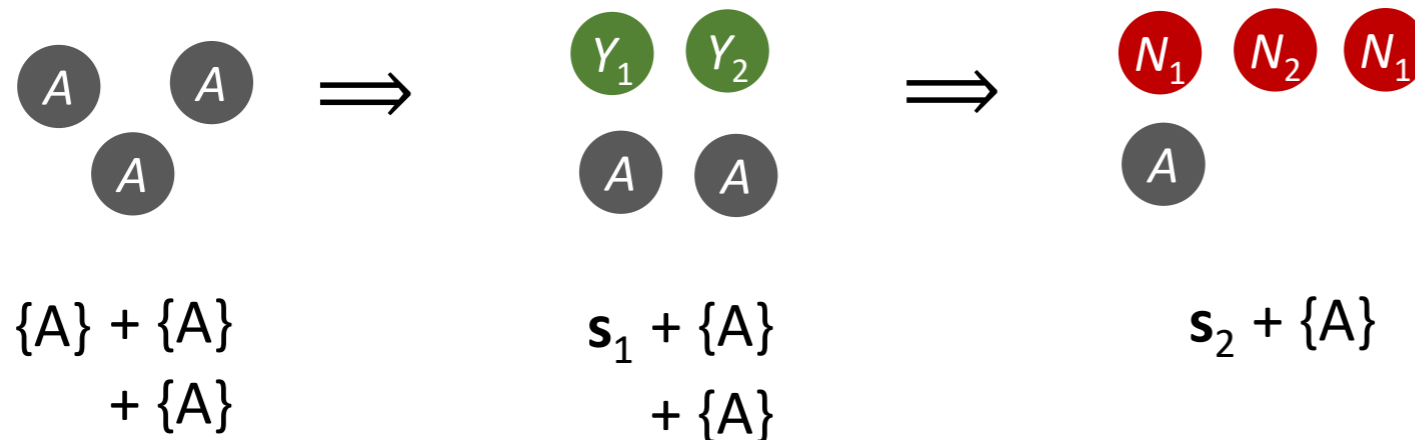
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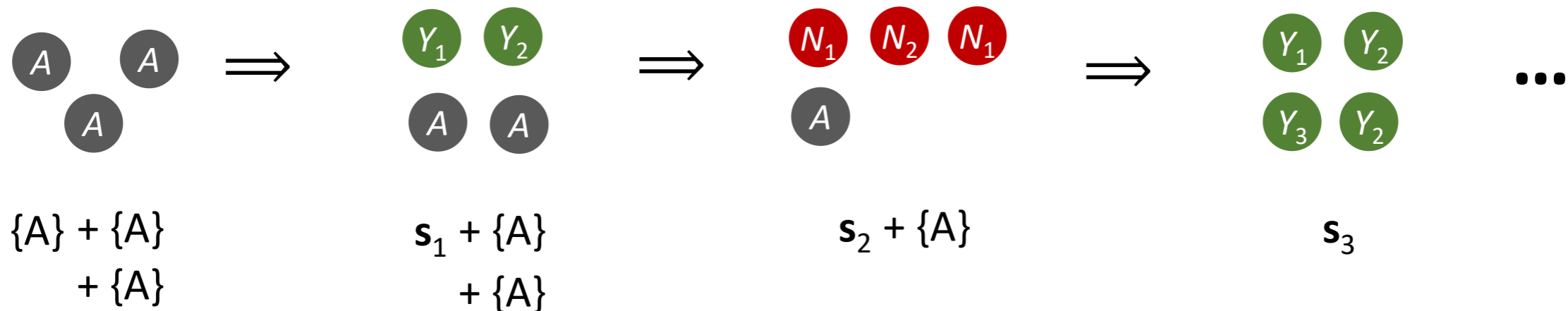
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- Since Φ is nonnegative, at some point we cannot continue. QED

Outline

- Formal definition of chemical reaction networks
- Execution bounded chemical reaction networks and linear potential functions
- What is “computation” with chemical reactions?
- Limitations of computation with execution bounded chemical reaction networks
- **Possibilities of computation with execution bounded chemical reaction networks**

Are execution bounded CRNs good for any computation?

Yes! Execution bounded CRNs *can* stably compute all semilinear predicates if the CRN is *leader-driven*: it starts with an “initial leader”, e.g., to compute majority ($a \geq b$?), start in initial state $\{1 L, a A, b B\}$... these are execution bounded from such states, but not from states with multiple leaders.

We also relax the voting requirement and allow only the leader to vote.
(though this requirement can be relaxed; not shown in slides)

Single-voting CRNs

Definition: A CRN computing a predicate $\varphi: \mathbb{N}^k \rightarrow \{Y, N\}$ is **single-voting** if all states reachable from the input have a single voter.

Such CRNs are leader-driven: valid initial configurations have a single leader/voter molecule, and only the leader votes.

Semilinear predicates are Boolean combinations of threshold and mod predicates

Recall: Theorem: $\varphi: \mathbb{N}^k \rightarrow \{Y,N\}$ is stably computable by a CRN if and only if φ is *semilinear*. (= Boolean combination of threshold and mod predicates)

To show execution bounded CRNs can compute all semilinear predicates, it suffices to show:

- They can compute all threshold predicates.
- They can compute all mod predicates.
- They can be composed to compute AND, OR, and NOT of other CRNs.

Execution bounded CRNs can compute threshold predicates

Theorem: Every threshold predicate $[w_1x_1 + \dots + w_kx_k \leq c?]$ can be stably computed by a single-voting execution bounded CRN.

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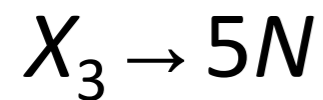
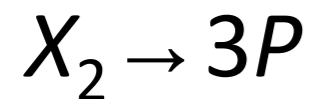
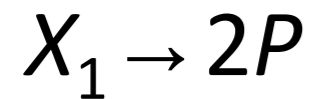
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Proof by example: To compute $[2x_1 + 3x_2 - 5x_3 \leq 4?]$, in addition to inputs X_1, X_2, X_3 , start with 1 L_Y (yes voter/leader) and 4 N, and have reactions:

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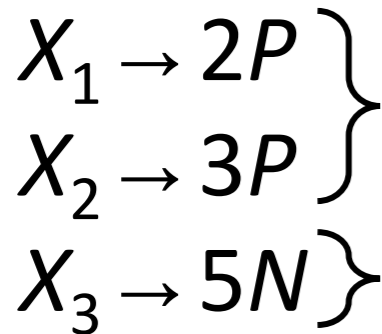
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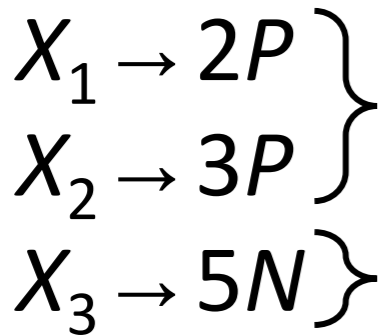
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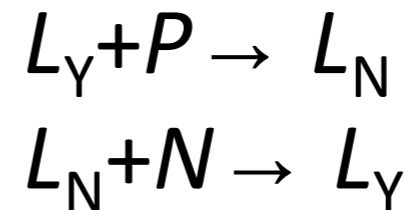
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Now we compute majority $[P \leq N?]$



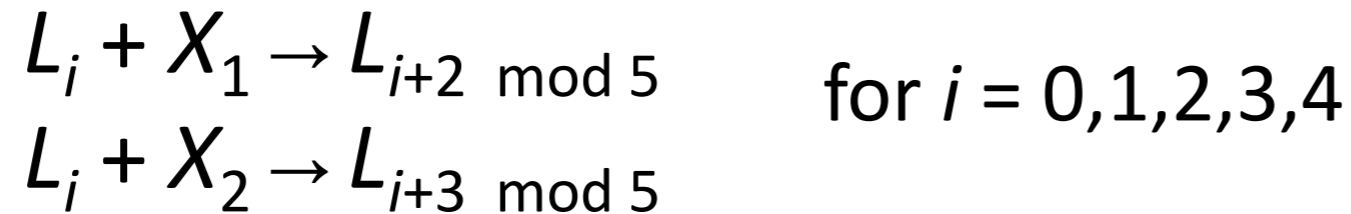
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L_4 votes **yes**, L_0, L_1, L_2, L_3 vote **no**

Composing CRNs to compute Boolean combinations of predicates

Theorem: If single-voting, execution bounded CRNs C_1 and C_2 stably compute predicates $\varphi_1: \mathbb{N}^k \rightarrow \{Y,N\}$ and $\varphi_2: \mathbb{N}^k \rightarrow \{Y,N\}$, then there are single-voting, execution bounded CRNs stably computing $[\varphi_1 \text{ and } \varphi_2]$, $[\varphi_1 \text{ or } \varphi_2]$, and $[\text{not } \varphi_1]$.

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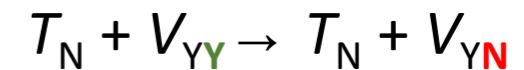
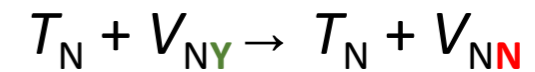
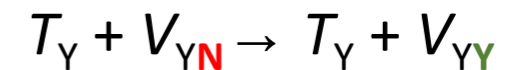
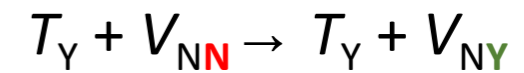
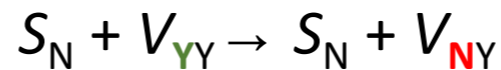
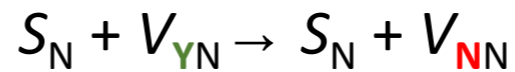
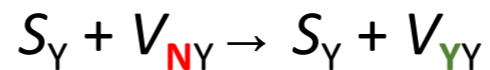
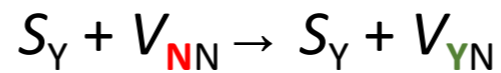
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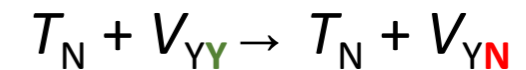
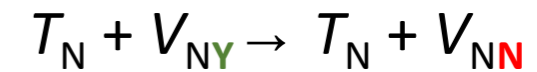
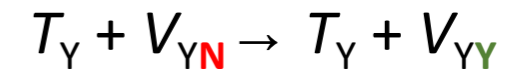
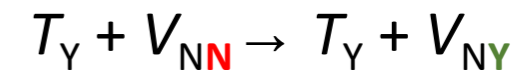
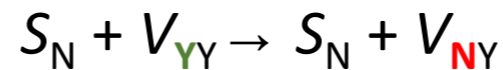
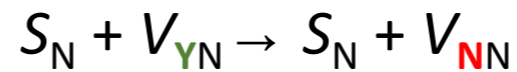
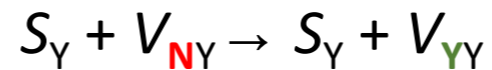
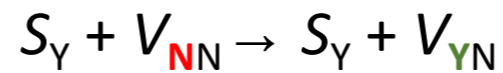
$V_{NN}, V_{NY}, V_{YN}, V_{YY}$; start with 1 V_{NN}

Let S_Y, S_N be yes and no voters of C_1

Let T_Y, T_N be yes and no voters of C_2

C_1 is execution bounded, so can only switch between S_Y and S_N a finite number of times, limiting how many times we can flip between $V_{N?}$ and $V_{Y?}$, so full CRN is execution bounded.

Voters of C_1 and C_2 influence global voters:



Open question

- Using standard stochastic model of chemical kinetics (not shown), the execution bounded CRNs stably computing semilinear predicates can be shown to take expected time $O(n \log n)$ to converge.

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- Conjecture: Execution bounded CRNs require $\Omega(n)$ time to stably compute any non-eventually constant predicate (e.g., majority or parity).

Thank you!

Questions?