

Computation by (not about) chemistry

Workshop on mathematical trends in reaction network theory
University of Copenhagen, July 2015

David Doty

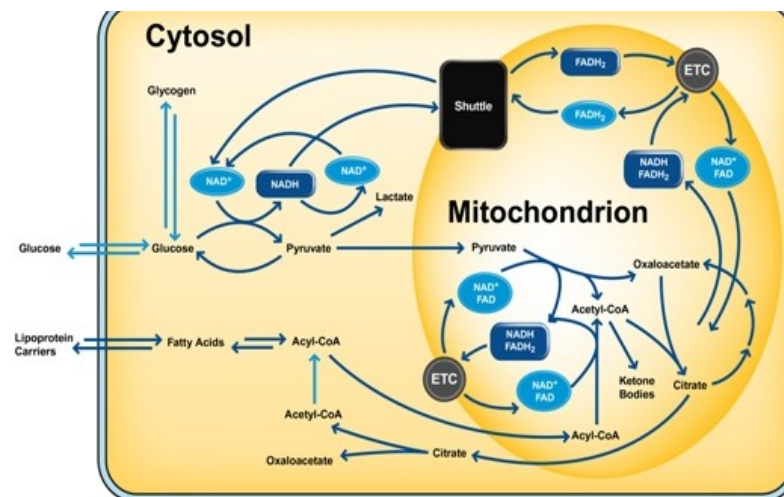


The software of life

How does the cell
compute?

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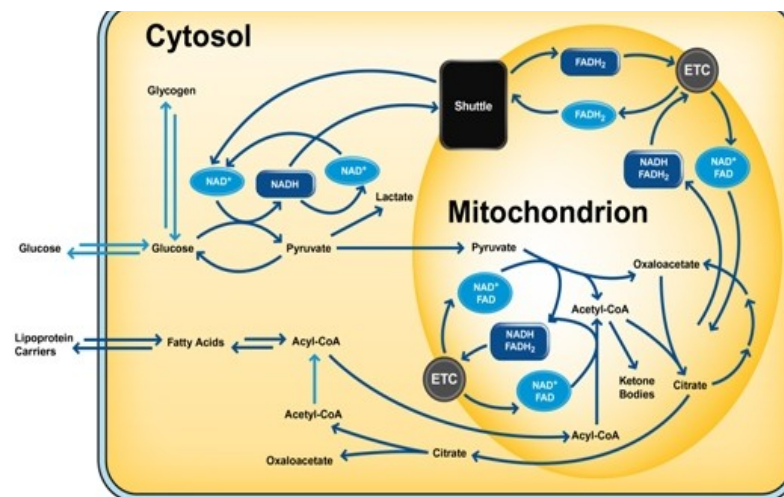


chemistry /
geometry

The software of life

~~How does the cell compute?~~

What is possible to compute with chemistry?
~~geometry~~



Chemical reaction networks (CRN)

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Chemical reaction networks (CRN)



(anonymous
waste product)

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(anonymous
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(anonymous
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What behavior is possible
for chemistry in principle?

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found in biology

inspiration



What behavior is possible for chemistry in principle?

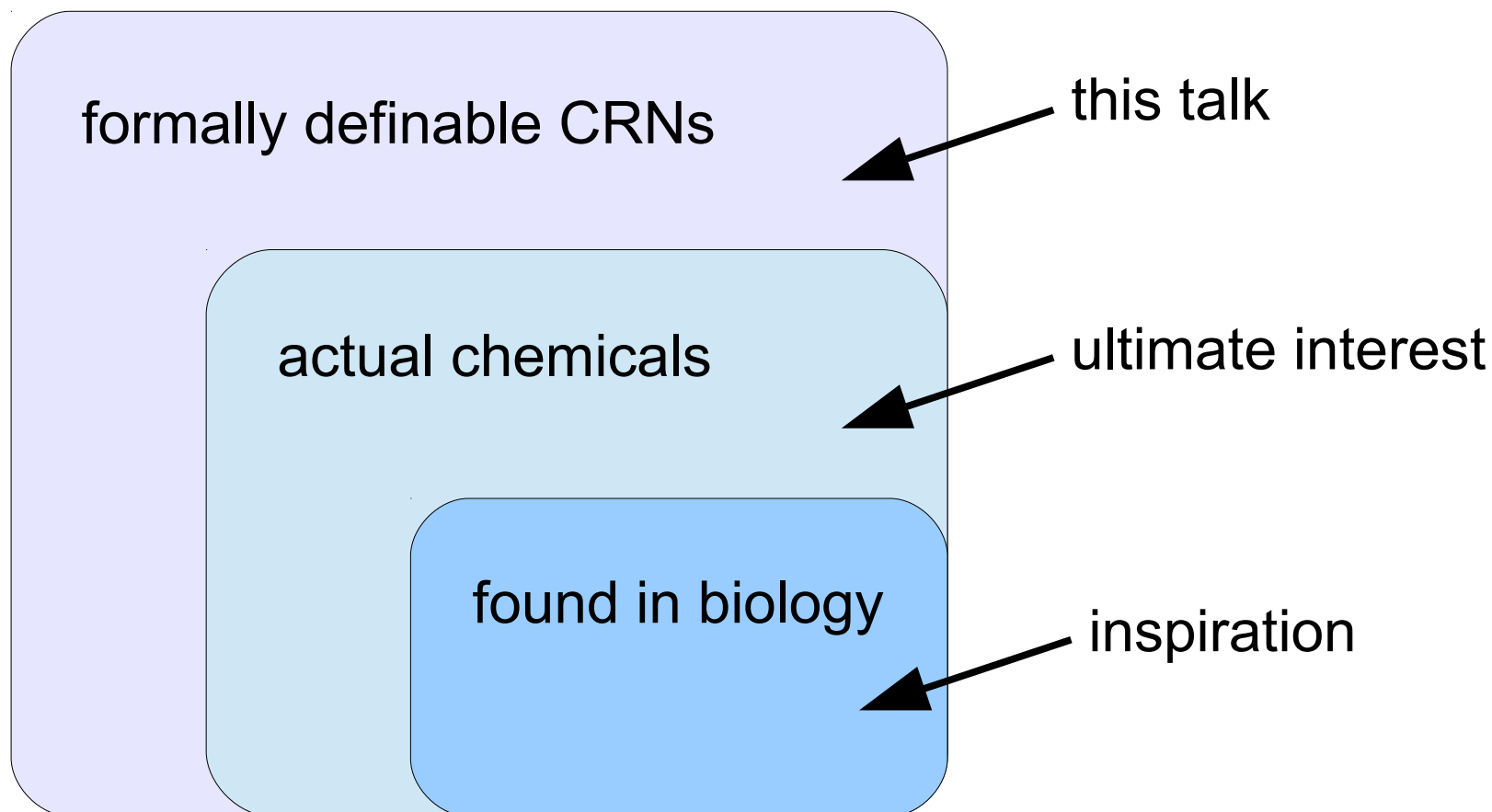
formally definable CRNs

this talk

found in biology

inspiration

What behavior is possible for chemistry in principle?



Can we compute with chemistry?

“Not every crazy CRN you scribble on paper describes actual chemicals!”

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Response to objection: Soloveichik et al. [*PNAS* 2010] showed a physical implementation of every CRN, using *DNA strand displacement*



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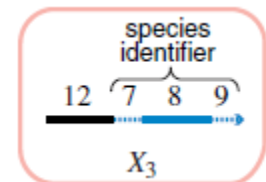
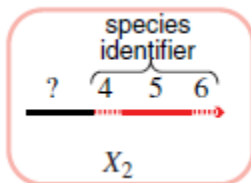
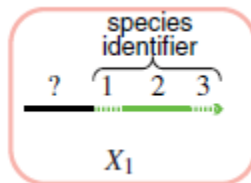
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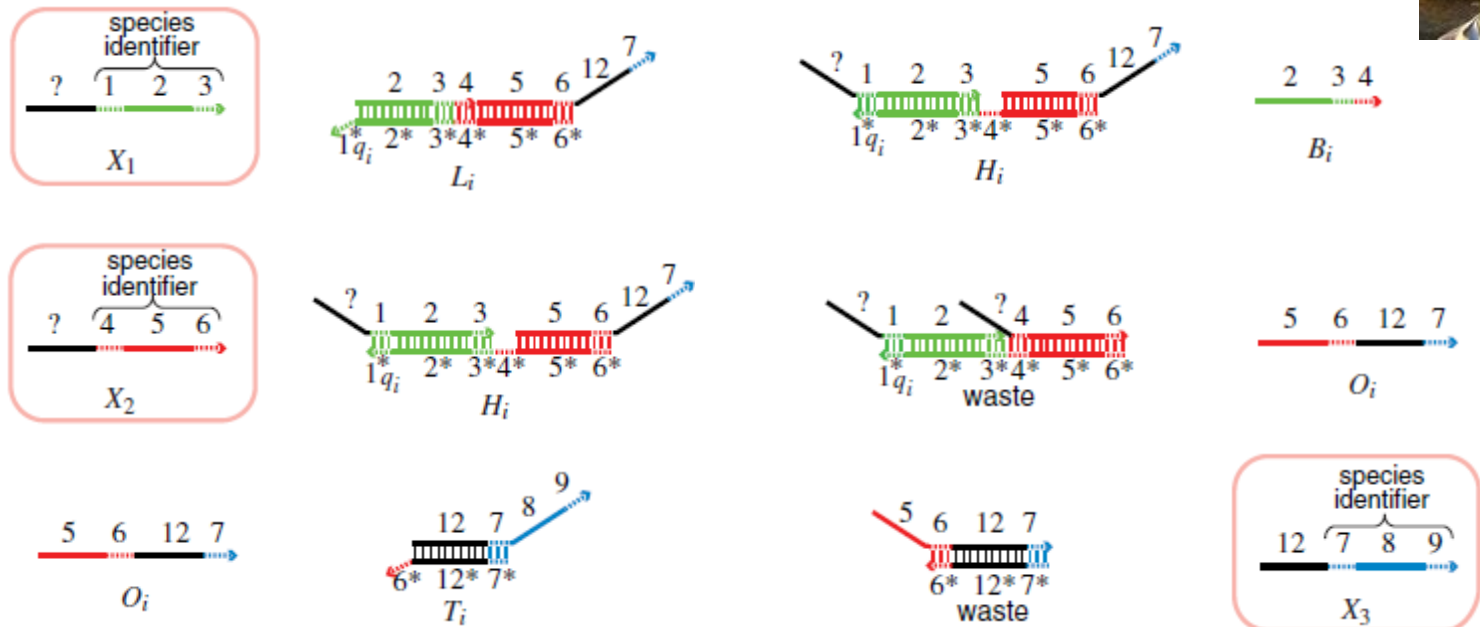
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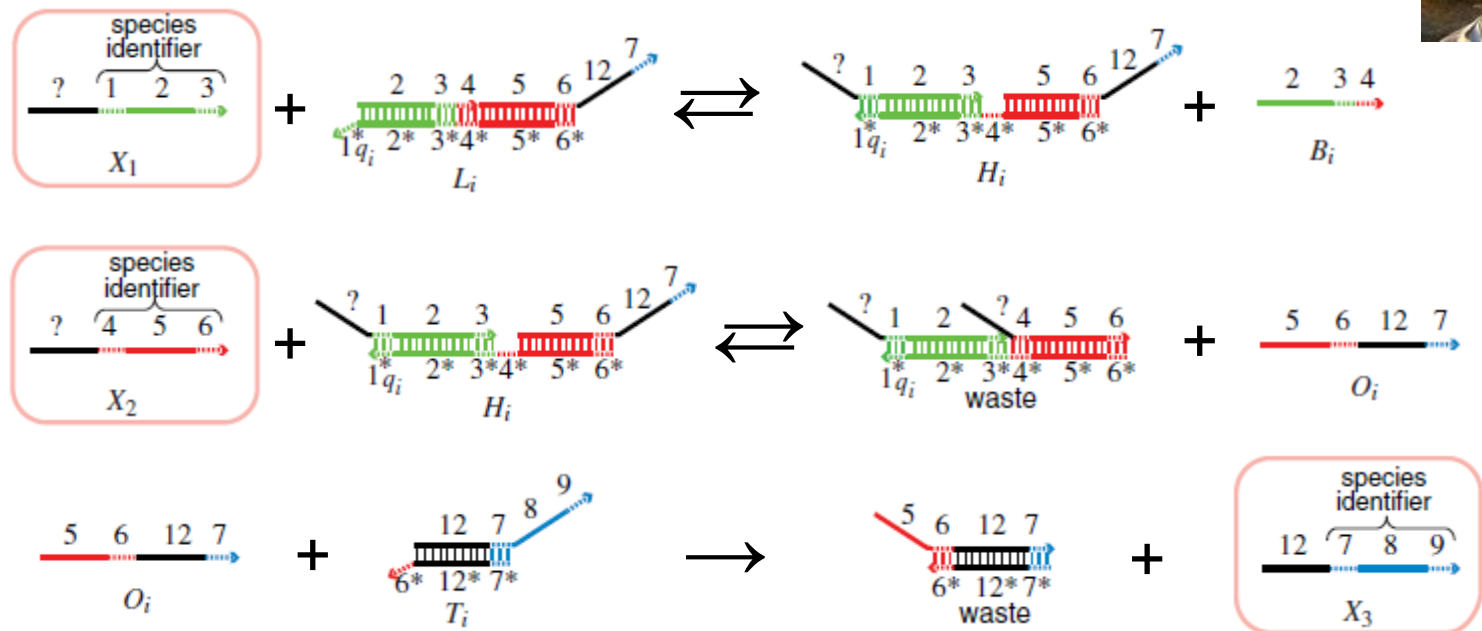
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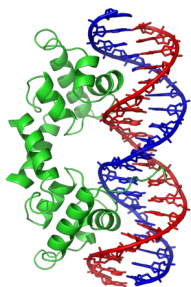
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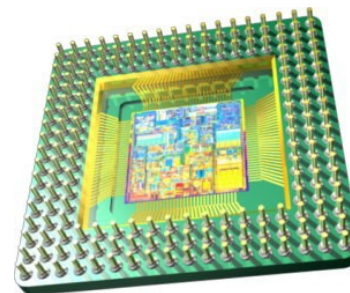
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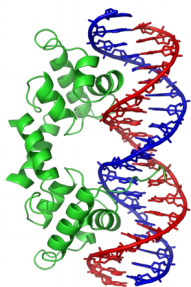
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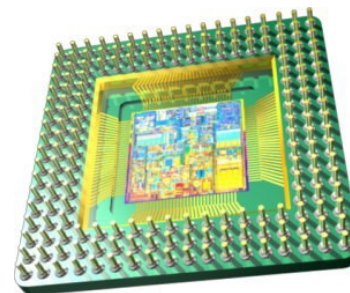
versus



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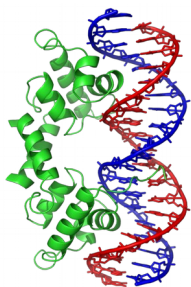


versus



speed?

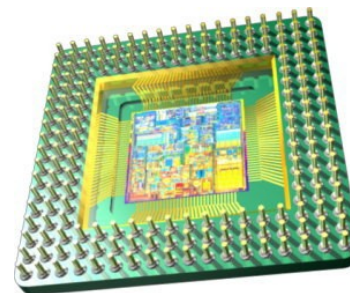
Why compute with chemistry?



slower

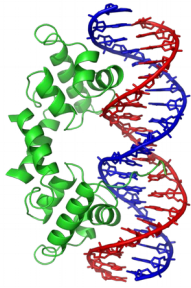
versus

speed?



faster

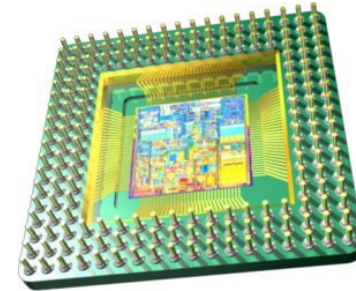
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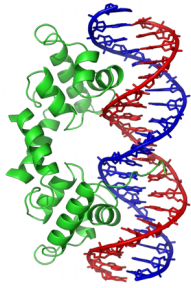
versus

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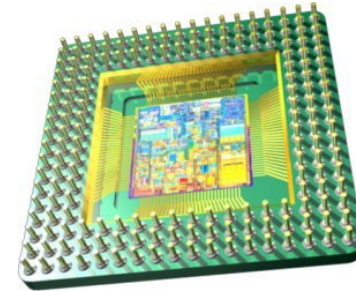
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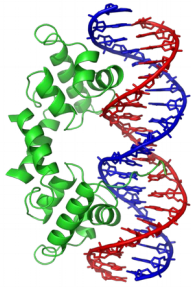


faster

~~speed?~~

component size?

Why compute with chemistry?



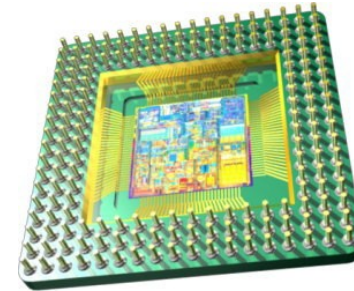
slower

≈ 10-100 nm

versus

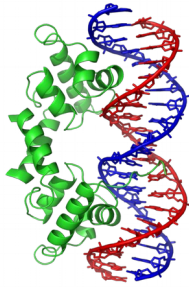
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component size?



faster

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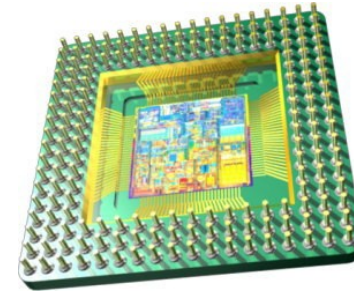
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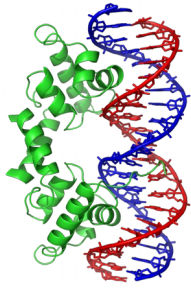
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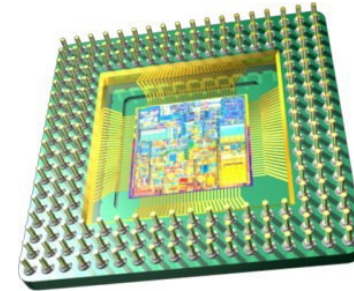
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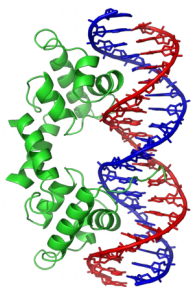
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Why compute with chemistry?



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yes

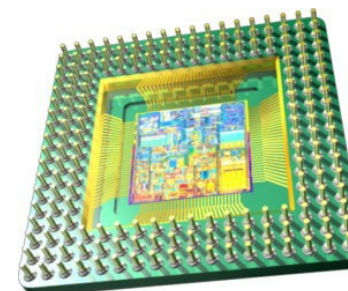
versus

~~speed?~~

~~component size?~~



Compatible with biological or other “wet environments”?

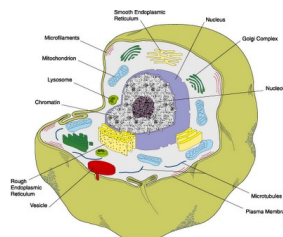


faster

≈ 10-100 nm

not easily

cells



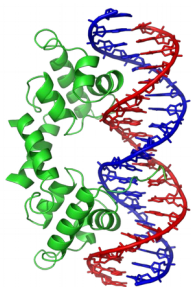
“smart drug” released only in certain cellular conditions

bioreactors



“chemical controller” to optimize yield of metabolically produced biofuels/drugs/etc.

Why compute with chemistry?



slower

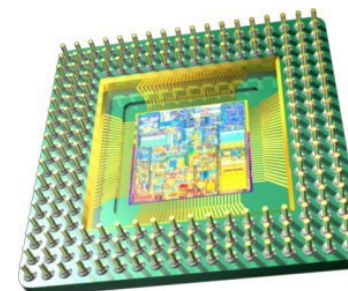
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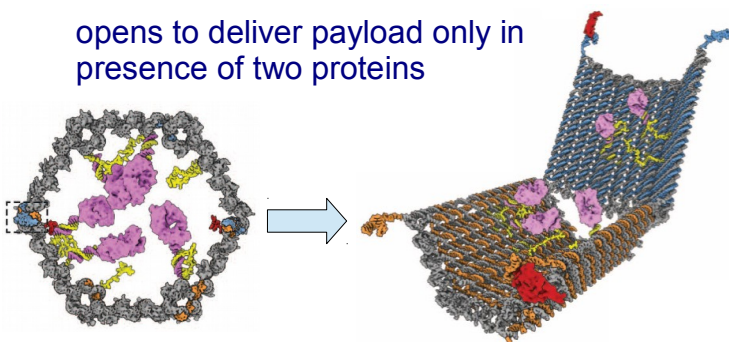
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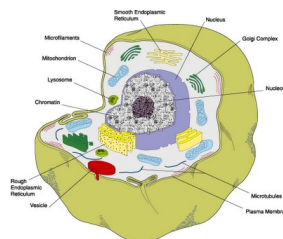
Compatible with biological or other “wet environments”?

opens to deliver payload only in presence of two proteins



Douglas et al, *Science* 2012

cells



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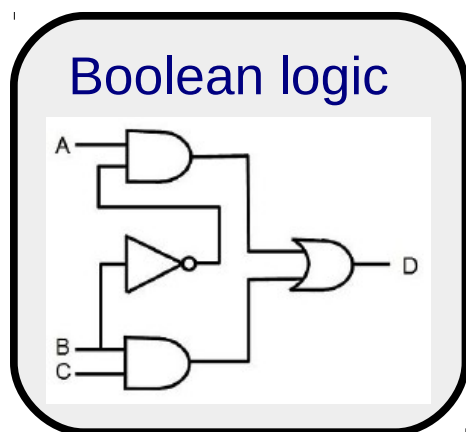
“chemical controller” to optimize yield of metabolically produced biofuels/drugs/etc.

What does it mean to compute with chemistry?

CRNs have a wide range of behaviors:

What does it mean to compute with chemistry?

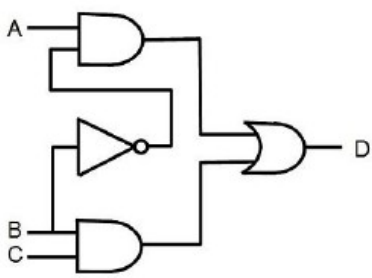
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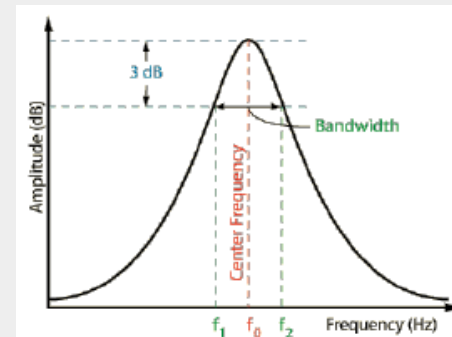
What does it mean to compute with chemistry?

CRNs have a wide range of behaviors:

Boolean logic



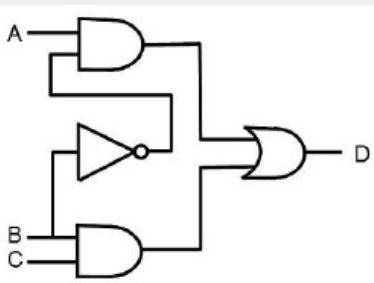
signal processing



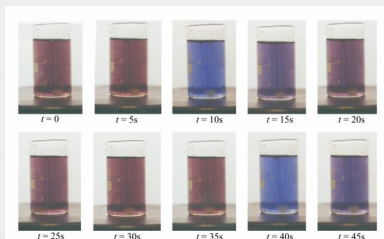
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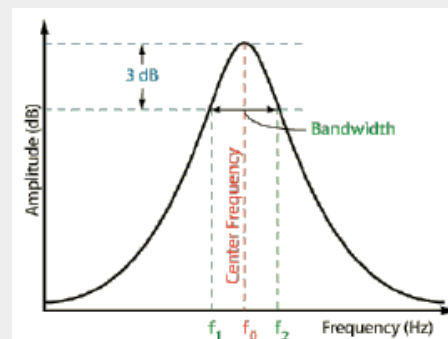
Boolean logic



oscillation



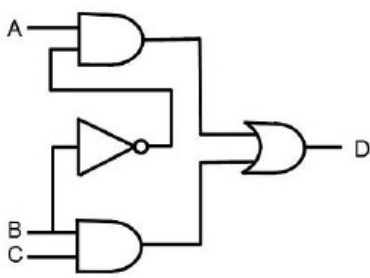
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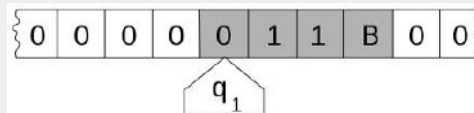
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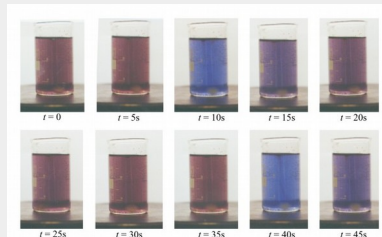
discrete algorithms



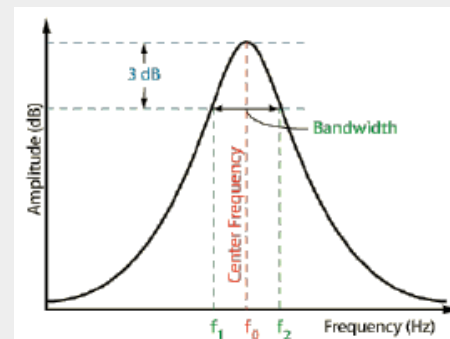
analog computing



oscillation



signal processing



Integer-valued kinetic CRN model

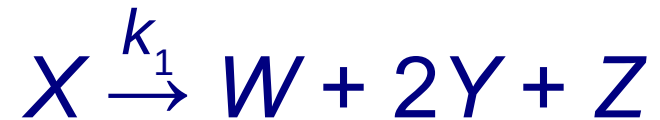
Integer-valued kinetic CRN model

- **species:** $\{X, Y, \dots\}$

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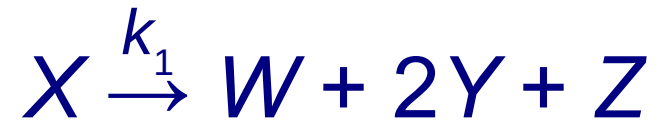
- **species:** $\{X, Y, \dots\}$

- **reactions:**



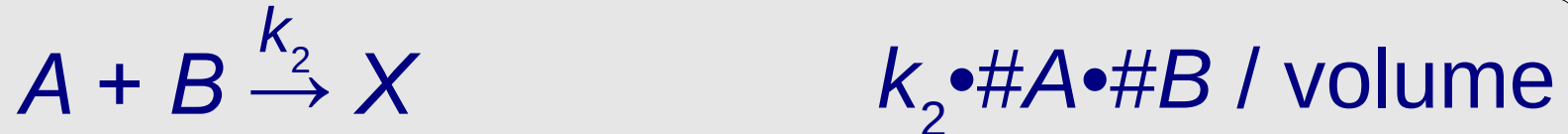
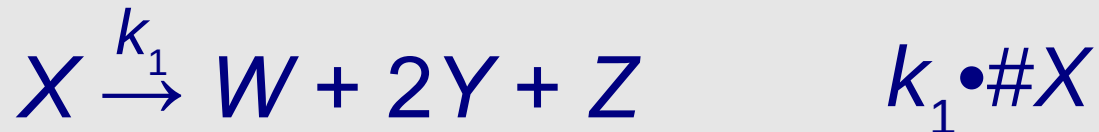
Integer-valued kinetic CRN model

- **species:** $\{X, Y, \dots\}$
- **state:** integer vector of *counts*
 $\mathbf{s} = (\#X, \#Y, \dots)$
- **reactions:**



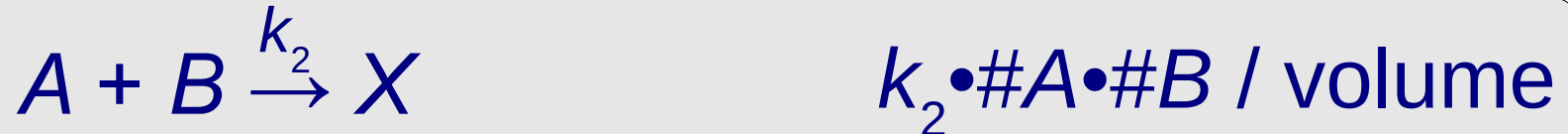
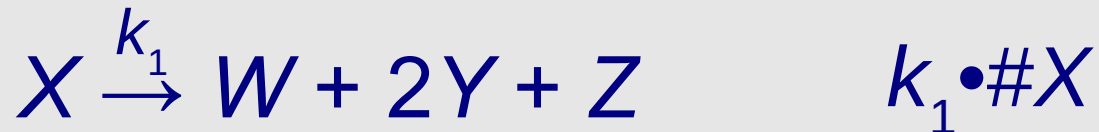
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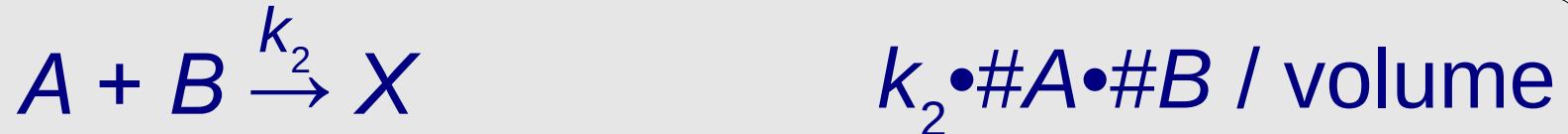
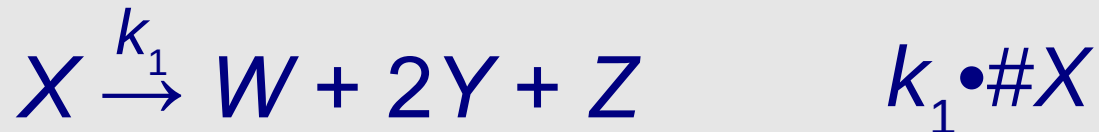
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$$\text{Prob}[\text{some reaction}] = \frac{\text{rate of that reaction}}{\text{sum of all reaction rates}}$$

Integer-valued kinetic CRN model

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$$\text{Prob}[\text{some reaction}] = \frac{\text{rate of that reaction}}{\text{sum of all reaction rates}}$$

$$E[\text{time until next reaction}] = 1 / \text{rate}$$

CRN function computation (example)

function: $f(x) = x/2$

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$\#Y = x/2$ expected at equilibrium

reactions: $X \xrightarrow{1} Y$
 $X \xrightarrow{1}$

$\#Y$ stabilizes, with
expected value $x/2$

CRN function computation (example)

function: $f(x) = x/2$



input species: X

output species: Y

initial state: $\{x X, 0 Y\}$

$\#Y = \frac{x/3}{x/2}$ expected at equilibrium



$\#Y$ stabilizes, with
expected value $\frac{x/2}{x/3}$

Rate-independent CRN computation

What can CRNs compute when we
don't know/can't control the rates?

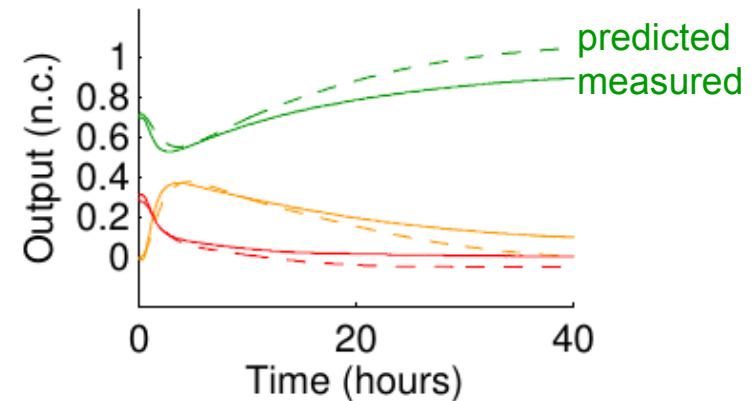
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Rate-independent CRN computation
(a.k.a. “stable”, “deterministic”)

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not the mass-action model!!

CRN function computation (example)

function: $f(x) = 2x$

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CRN function computation (example)

function: $f(x) = 2x$

input species: X

output species: Y

reactions: ??



CRN function computation (example)

function: $f(x) = 2x$

input species: X

output species: Y

reactions: $X \rightarrow 2Y$



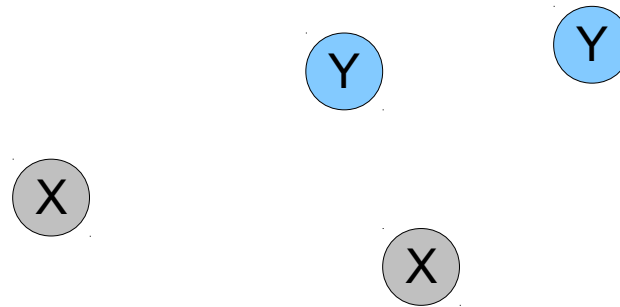
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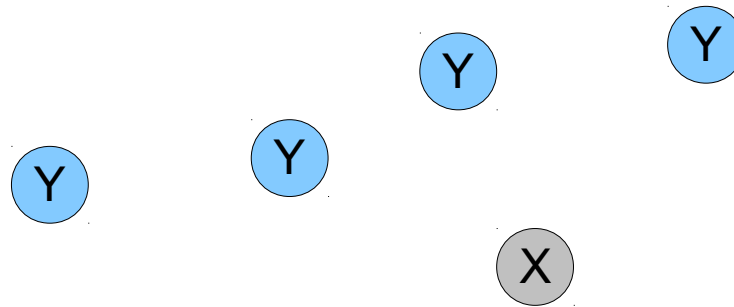
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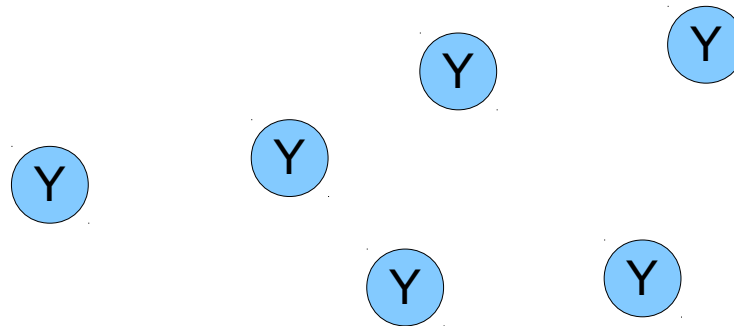
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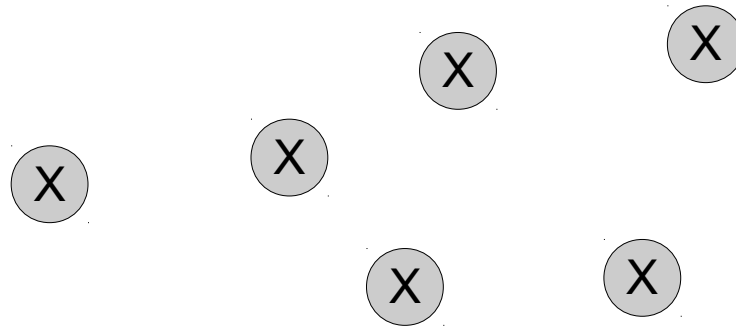
output species: Y

reactions: $X \rightarrow 2Y$



CRN function computation (example)

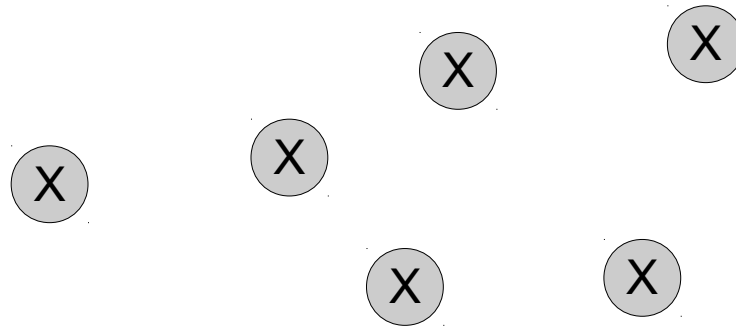
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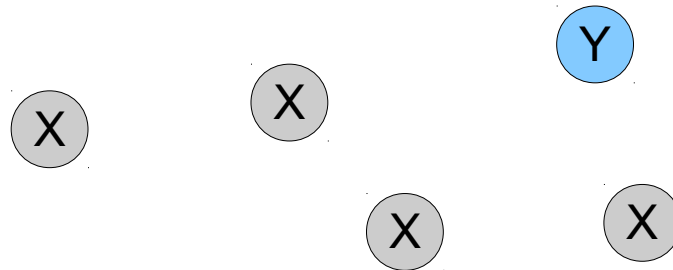
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CRN function computation (example)

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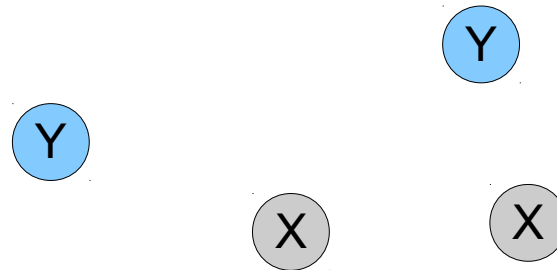
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CRN function computation (example)

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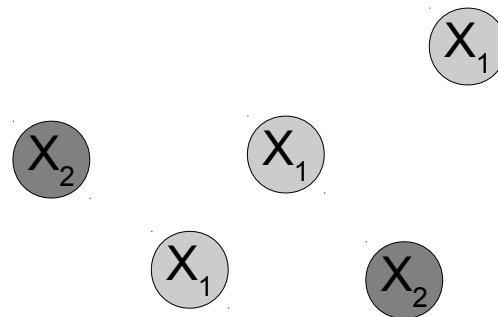
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CRN function computation (example)

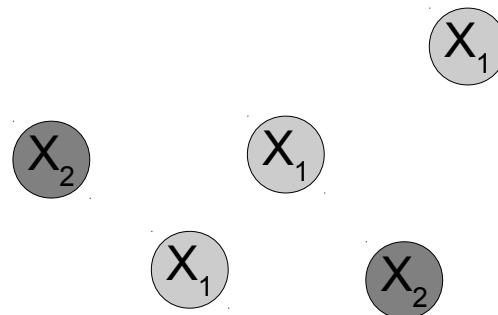
function: $f(x_1, x_2) = x_1 + x_2$



CRN function computation (example)

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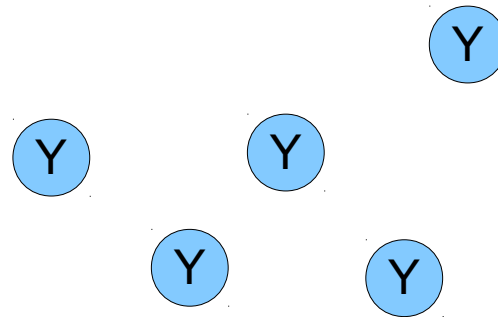
reactions: $X_1 \rightarrow Y$
 $X_2 \rightarrow Y$



CRN function computation (example)

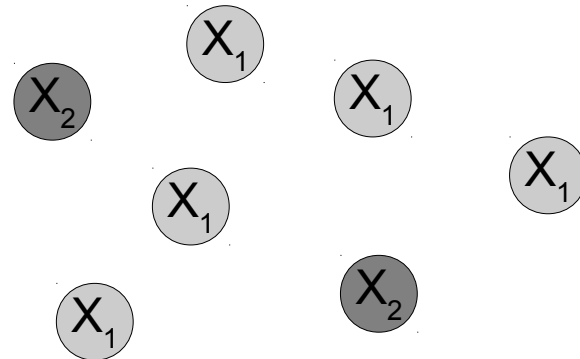
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CRN function computation (example)

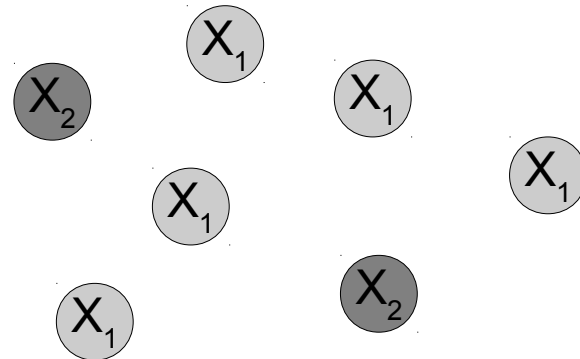
function: $f(x_1, x_2) = x_1 - x_2$



CRN function computation (example)

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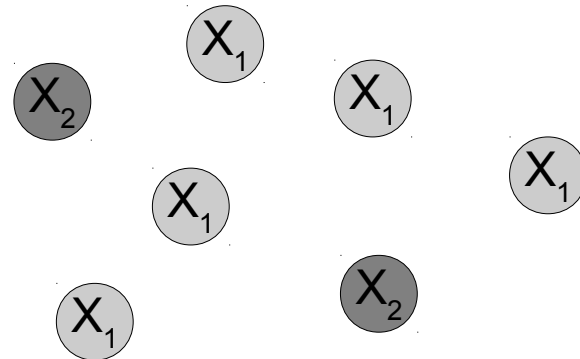
reactions: $X_1 \rightarrow Y$
 $X_2 + Y \rightarrow$



CRN function computation (example)

function: $f(x_1, x_2) = x_1 - x_2$

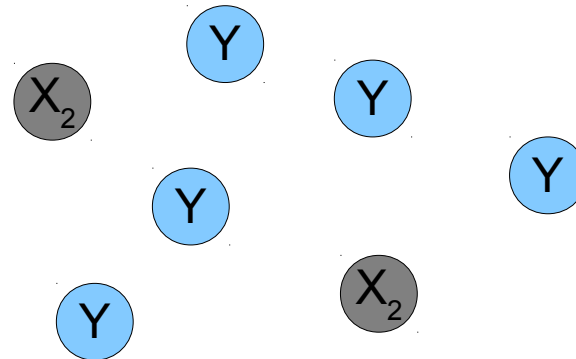
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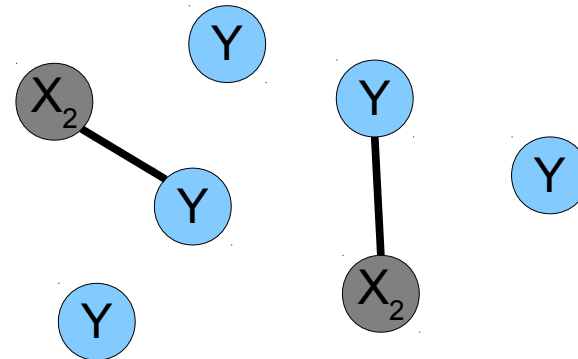
reactions: $X_1 \rightarrow Y$
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CRN function computation (example)

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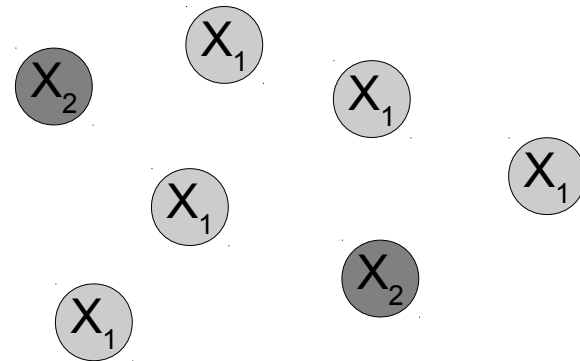
Y

Y

Y

CRN function computation (example)

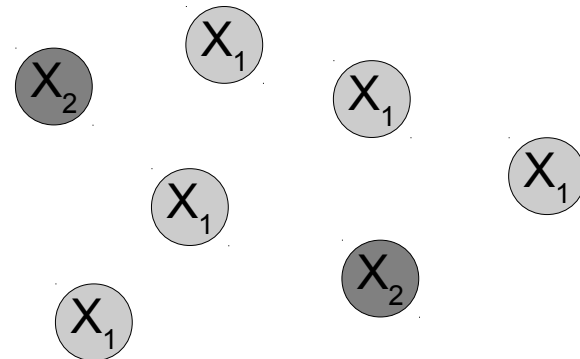
function: $f(x_1, x_2) = \min\{x_1, x_2\}$



CRN function computation (example)

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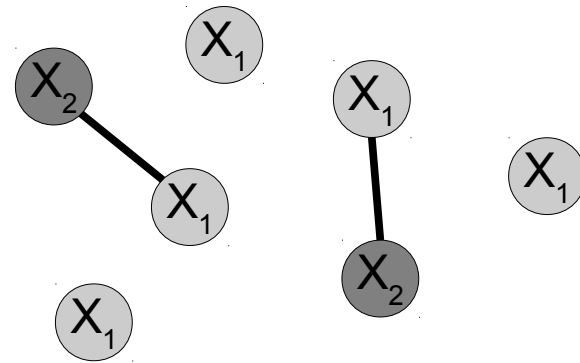
reactions: $X_1 + X_2 \rightarrow Y$



CRN function computation (example)

function: $f(x_1, x_2) = \min\{x_1, x_2\}$

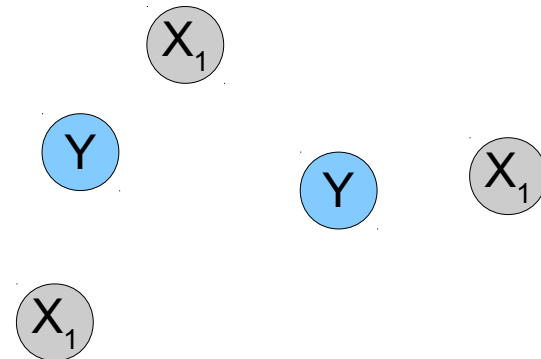
reactions: $X_1 + X_2 \rightarrow Y$



CRN function computation (example)

function: $f(x_1, x_2) = \min\{x_1, x_2\}$

reactions: $X_1 + X_2 \rightarrow Y$



CRN function computation (example)

function: $f(x_1, x_2) = \max\{x_1, x_2\}$

CRN function computation (example)

function: $f(x_1, x_2) = \max\{x_1, x_2\} = x_1 + x_2 - \min\{x_1, x_2\}$

x_1

x_1

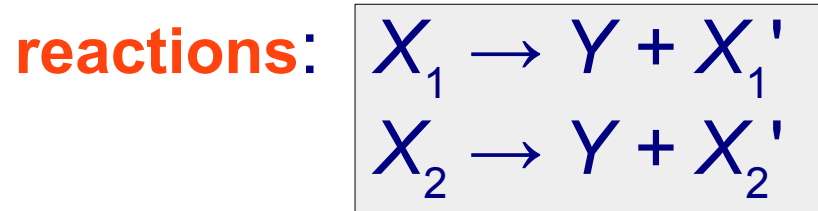
x_2

x_1

x_2

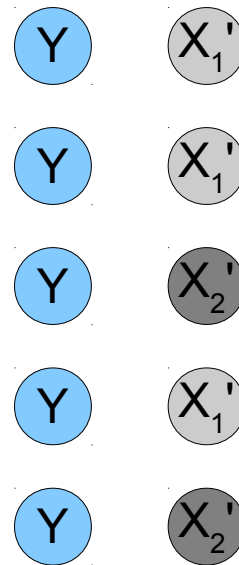
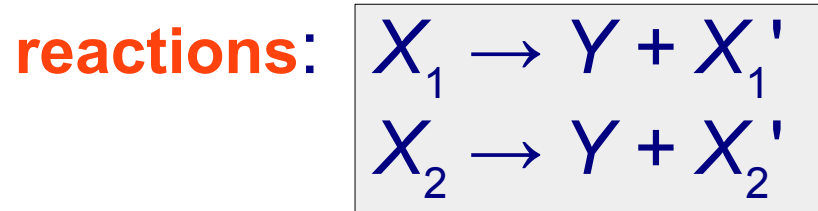
CRN function computation (example)

function: $f(x_1, x_2) = \max\{x_1, x_2\} = \boxed{x_1 + x_2} - \min\{x_1, x_2\}$



CRN function computation (example)

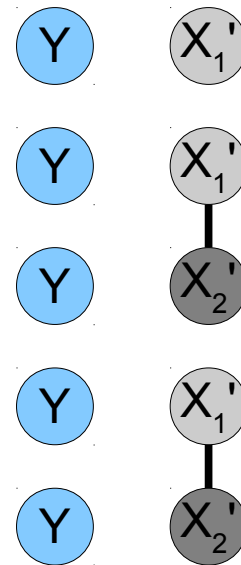
function: $f(x_1, x_2) = \max\{x_1, x_2\} = \boxed{x_1 + x_2} - \min\{x_1, x_2\}$



CRN function computation (example)

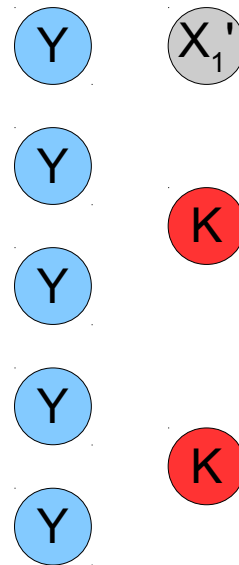
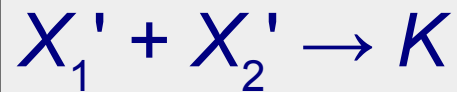
function: $f(x_1, x_2) = \max\{x_1, x_2\} = x_1 + x_2 - \min\{x_1, x_2\}$

reactions: $X_1 \rightarrow Y + X_1'$
 $X_2 \rightarrow Y + X_2'$
 $X_1' + X_2' \rightarrow K$



CRN function computation (example)

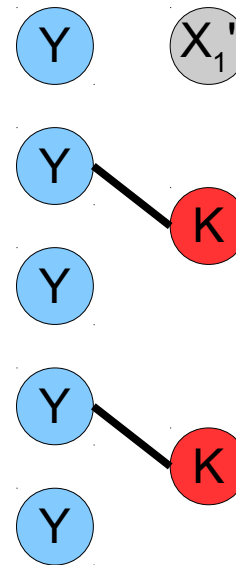
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 $X_2 \rightarrow Y + X_2'$
 $X_1' + X_2' \rightarrow K$
 $K + Y \rightarrow$



CRN function computation (example)

function: $f(x_1, x_2) = \max\{x_1, x_2\} = x_1 + x_2 - \min\{x_1, x_2\}$

reactions: $X_1 \rightarrow Y + X_1'$
 $X_2 \rightarrow Y + X_2'$
 $X_1' + X_2' \rightarrow K$
 $K + Y \rightarrow$



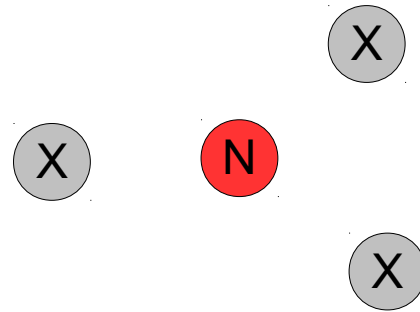
CRN predicate computation (example)

predicate: $p(x)$: parity of x (“yes” $\Leftrightarrow x$ is odd)

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initial state: $\{x X, 1 N, 0 Y\}$

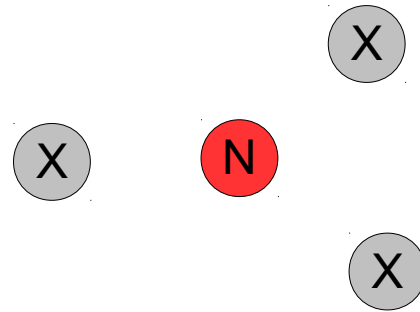


CRN predicate computation (example)

predicate: $p(x)$: parity of x (“yes” $\Leftrightarrow x$ is odd)

initial state: $\{x X, 1 N, 0 Y\}$

reactions: $N + X \rightarrow Y$
 $Y + X \rightarrow N$

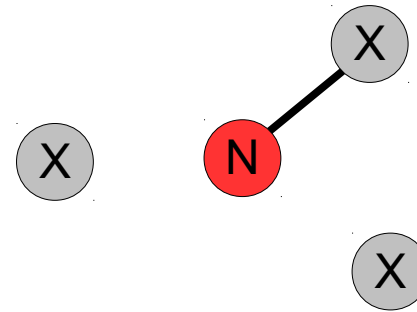


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X

Y

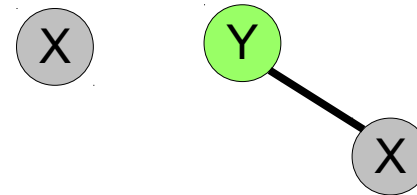
X

CRN predicate computation (example)

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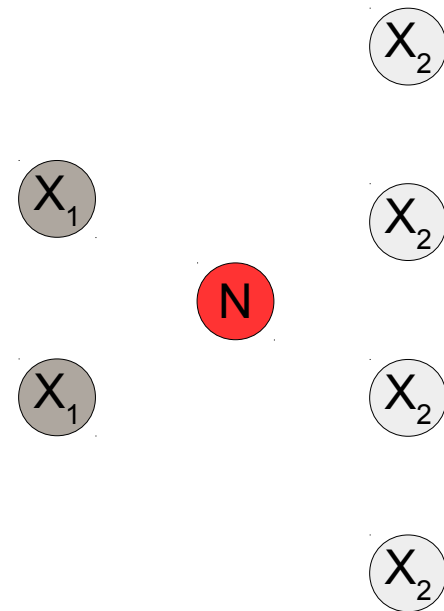
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CRN predicate computation (example)

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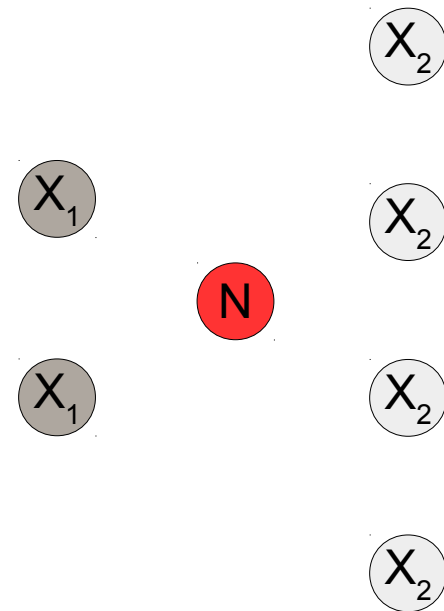


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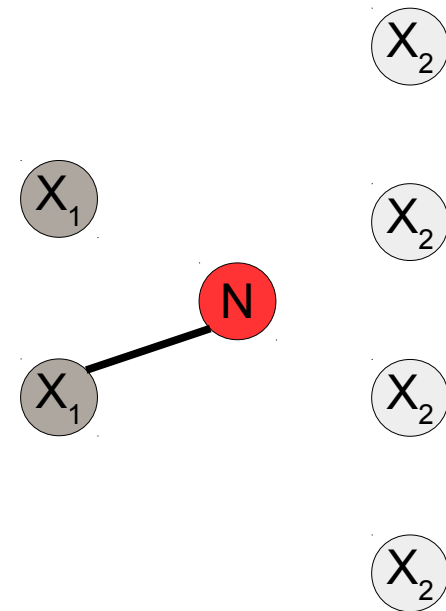


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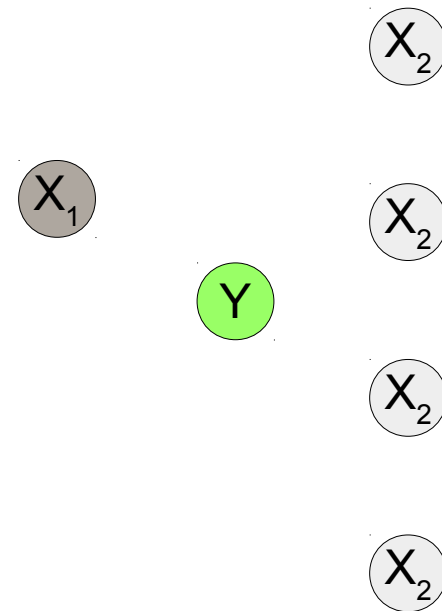


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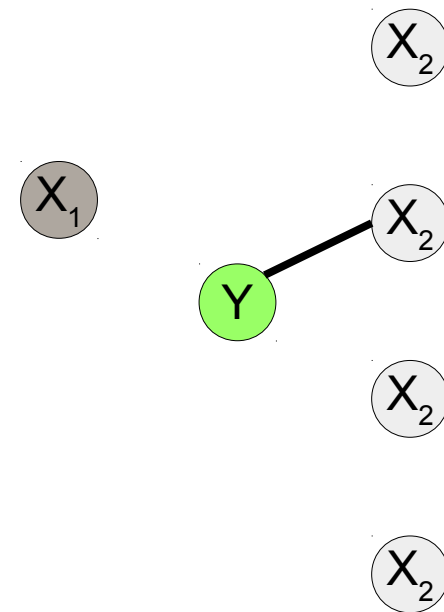


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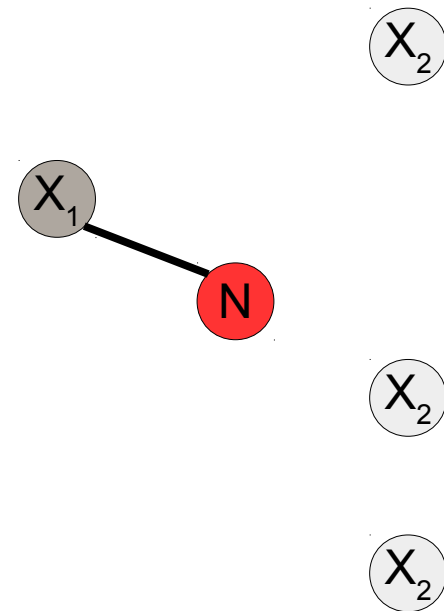


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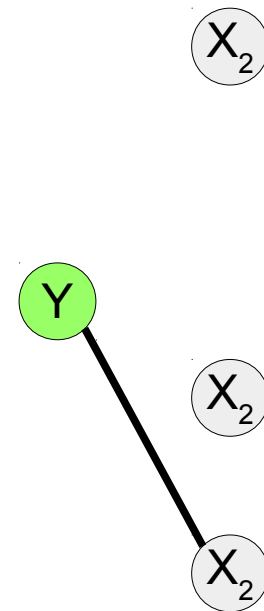


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predicate: $p(x_1, x_2): "x_1 = x_2"?$

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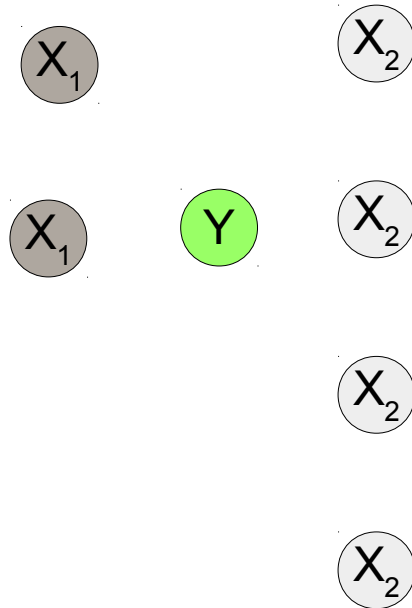
initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

Y

CRN predicate computation (example)

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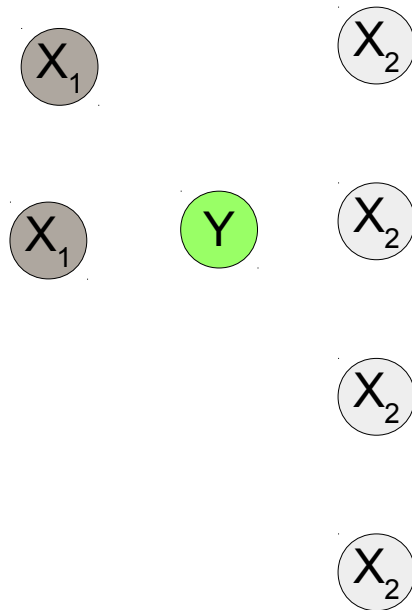


CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
 $X_1 + Y \rightarrow X_1 + N$
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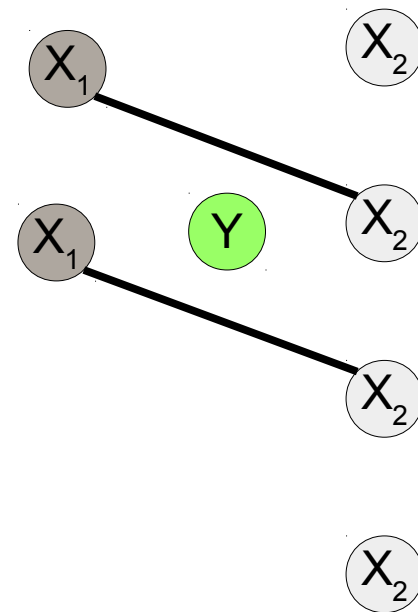


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reactions:

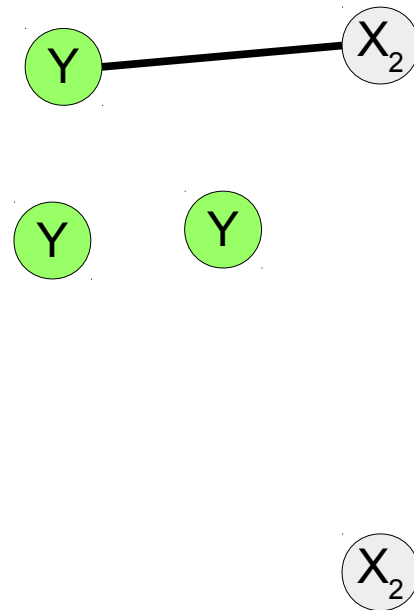


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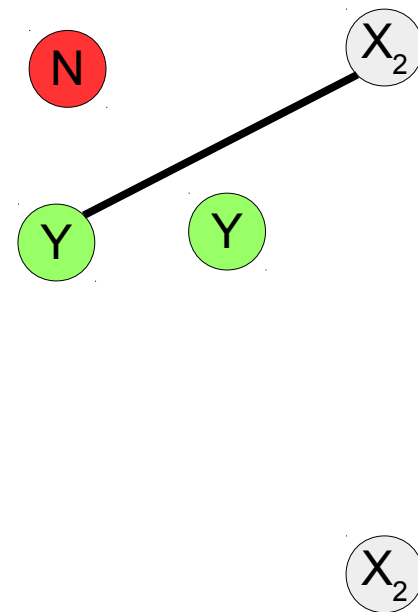


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reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
 $X_1 + Y \rightarrow X_1 + N$
 $X_2 + Y \rightarrow X_2 + N$

N

X_2

N

Y

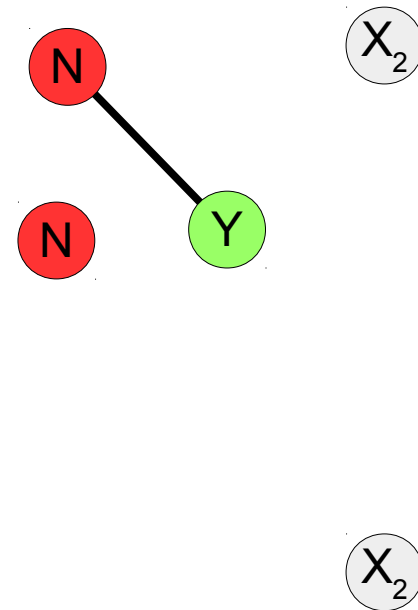
X_2

CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
 $X_1 + Y \rightarrow X_1 + N$
 $X_2 + Y \rightarrow X_2 + N$



CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
 $X_1 + Y \rightarrow X_1 + N$
 $X_2 + Y \rightarrow X_2 + N$

N

Y

X_2

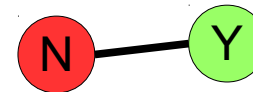
X_2

CRN predicate computation (example)

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initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
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X_2

X_2

CRN predicate computation (example)

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reactions: $X_1 + X_2 \rightarrow Y$
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X_2

Y

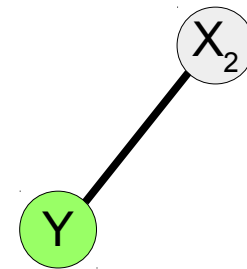
X_2

CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions:

$$X_1 + X_2 \rightarrow Y$$
$$Y + N \rightarrow Y$$
$$X_1 + Y \rightarrow X_1 + N$$
$$X_2 + Y \rightarrow X_2 + N$$


CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

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reactions:

$$X_1 + X_2 \rightarrow Y$$
$$Y + N \rightarrow Y$$
$$X_1 + Y \rightarrow X_1 + N$$
$$X_2 + Y \rightarrow X_2 + N$$

X_2

N

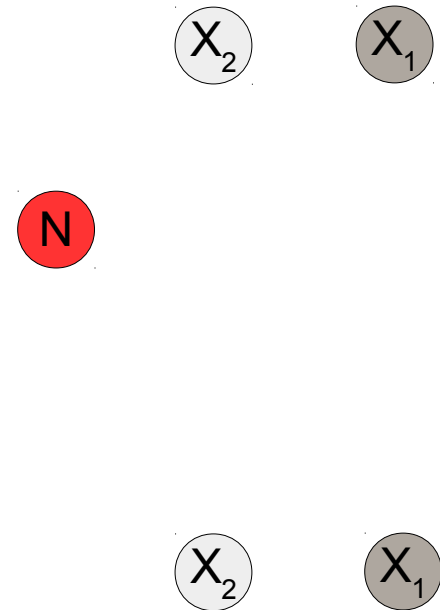
X_2

CRN predicate computation (example)

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initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
 $X_1 + Y \rightarrow X_1 + N$
 $X_2 + Y \rightarrow X_2 + N$



CRN predicate computation (example)

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reactions:



CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions:

$$X_1 + X_2 \rightarrow Y$$
$$Y + N \rightarrow Y$$
$$X_1 + Y \rightarrow X_1 + N$$
$$X_2 + Y \rightarrow X_2 + N$$

Y

N

Y

CRN predicate computation (example)

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reactions: $X_1 + X_2 \rightarrow Y$
 $Y + N \rightarrow Y$
 ~~$X_1 + Y \rightarrow X_1 + N$~~
 ~~$X_2 + Y \rightarrow X_2 + N$~~

Y

N

Y

CRN predicate computation (example)

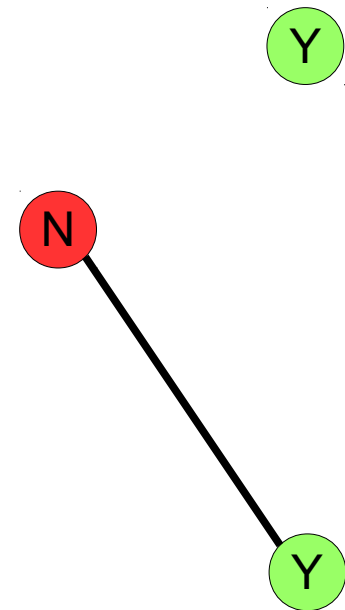
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initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$

$Y + N \rightarrow Y$

~~$X_1 + Y \rightarrow X_1 + N$~~
 ~~$X_2 + Y \rightarrow X_2 + N$~~



CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $x_1 = x_2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 Y\}$

reactions: $X_1 + X_2 \rightarrow Y$

$Y + N \rightarrow Y$

~~$X_1 + Y \rightarrow X_1 + N$~~
 ~~$X_2 + Y \rightarrow X_2 + N$~~

Y

Y

CRN predicate computation (example)

predicate: $p(x_1, x_2): "3x_1 > x_2/2"?$

initial state: $\{ x_1 X_1, x_2 X_2, 1 N \}$

CRN predicate computation (example)

predicate: $p(x_1, x_2):$ “ $3x_1 > x_2/2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 N\}$

reactions: $X_1 \rightarrow 3Z_1$

CRN predicate computation (example)

predicate: $p(x_1, x_2): "3x_1 > x_2/2"?$

initial state: $\{ x_1 X_1, x_2 X_2, 1 N \}$

reactions: $X_1 \rightarrow 3Z_1$
 $2X_2 \rightarrow Z_2$

CRN predicate computation (example)

predicate: $p(x_1, x_2)$: “ $3x_1 \geq x_2/2$ ”?

initial state: $\{x_1 X_1, x_2 X_2, 1 N\}$

reactions: $X_1 \rightarrow 3Z_1$

$2X_2 \rightarrow Z_2$

$N + Z_1 \rightarrow Y$

$Y + Z_2 \rightarrow N$

Stable predicate computation (definition)

task: compute $p(x_1, \dots, x_k) \in \{\text{yes}, \text{no}\}$, $x_1, \dots, x_k \in \mathbb{N}$

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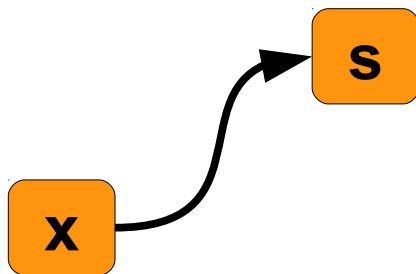
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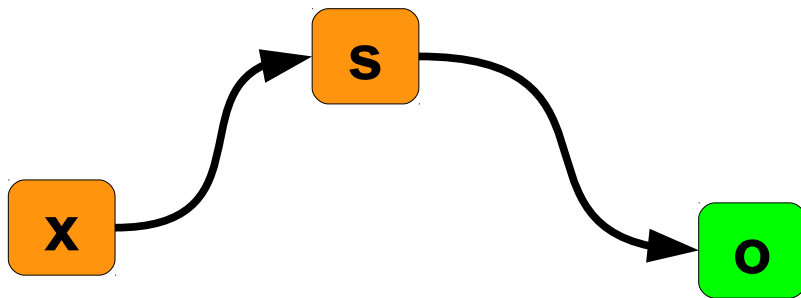
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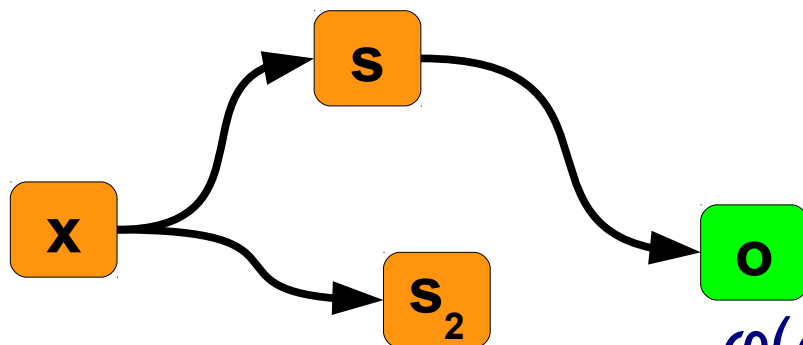
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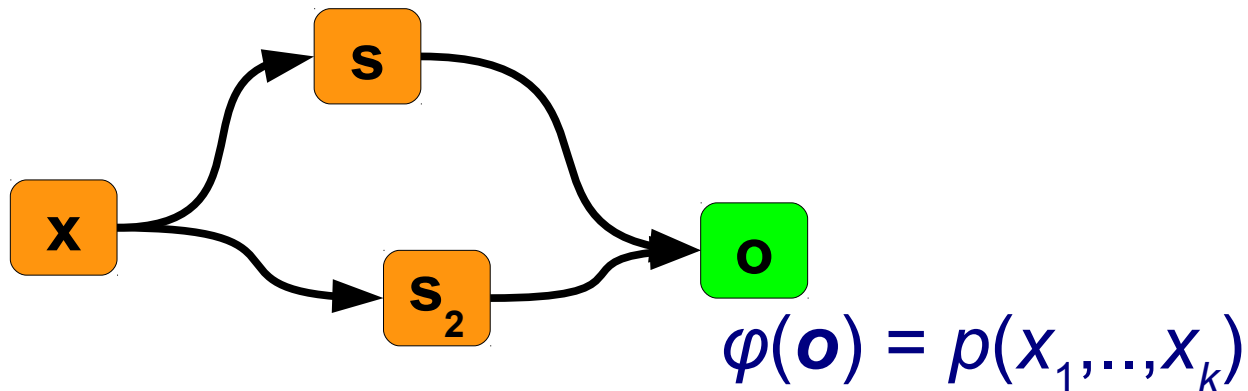
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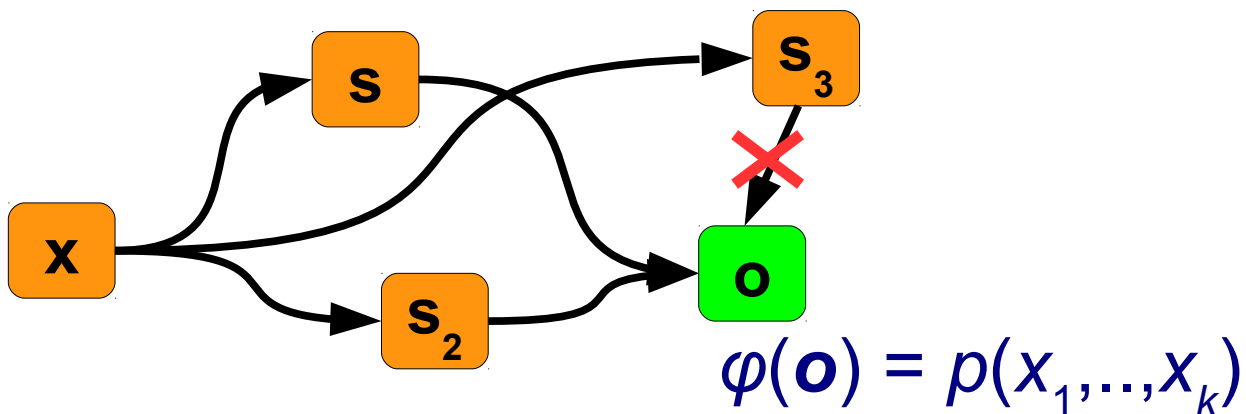
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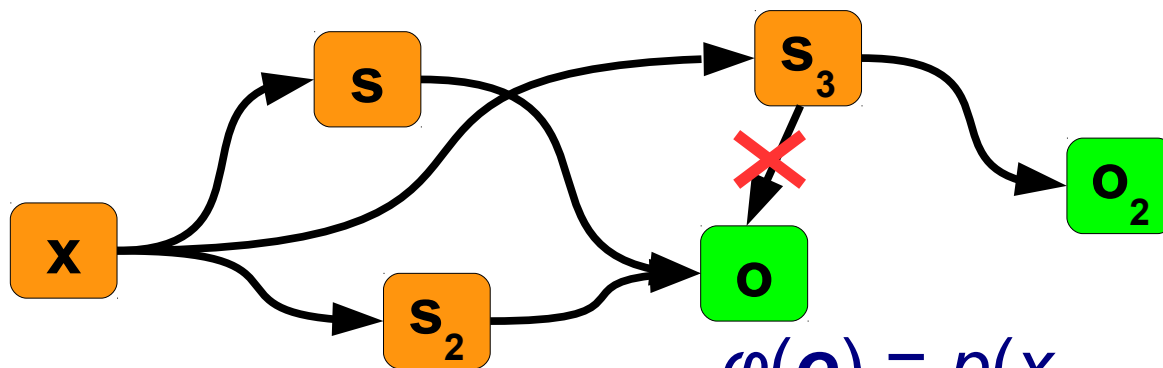
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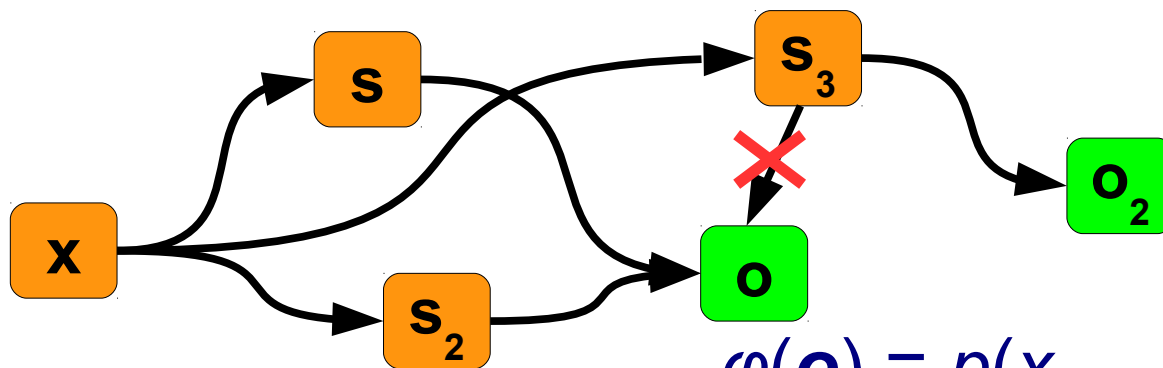
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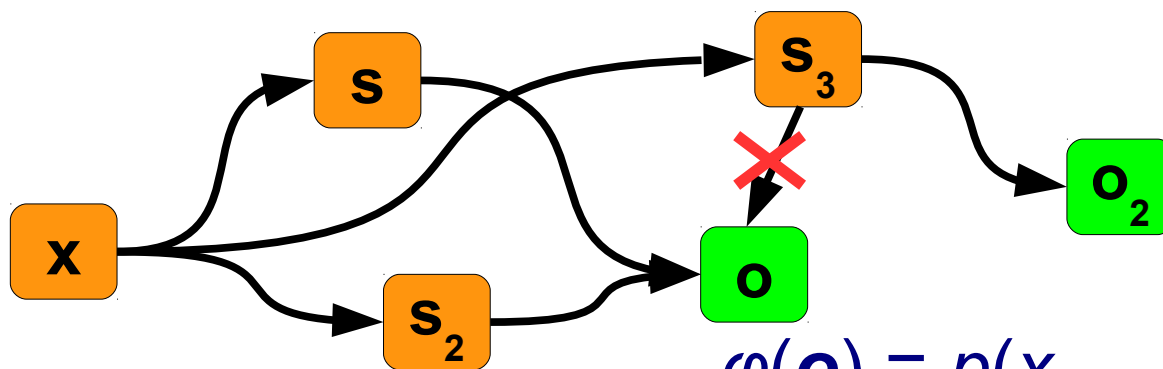
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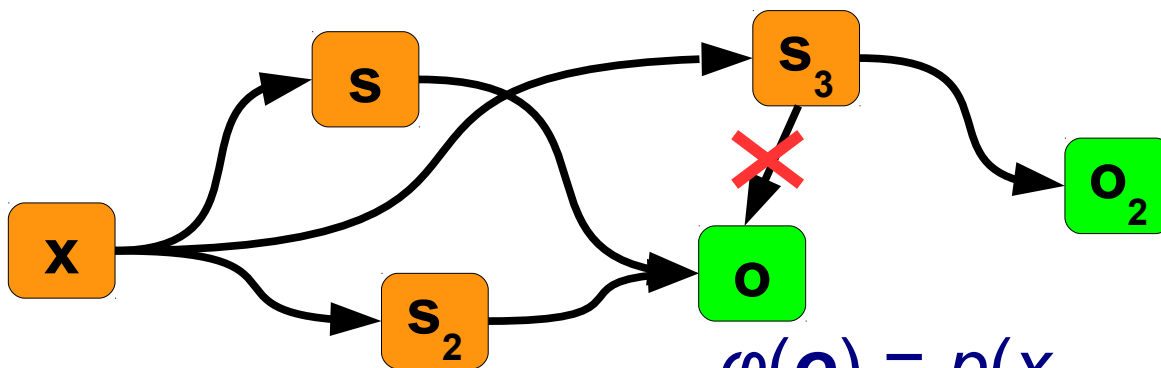
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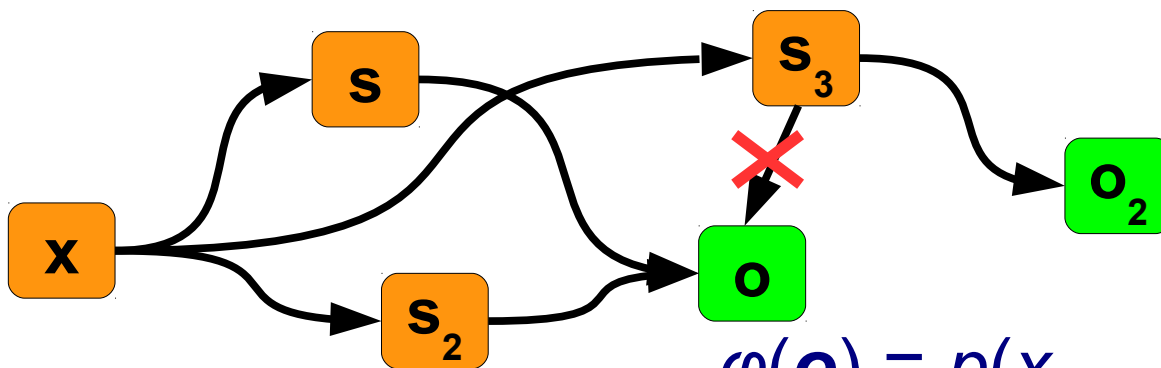
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Stable computation characterization

Theorem: A predicate/function is stably computed by a CRN if and only if it is *semilinear*.

[Angluin, Aspnes, Diamadi, Fisher, Peralta, [Principles of Distributed Computing 2004](#)]

[Angluin, Aspnes, Eisenstat, [Principles of Distributed Computing 2006](#)]

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predicates

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semilinear function \approx piecewise linear

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Real-valued CRNs

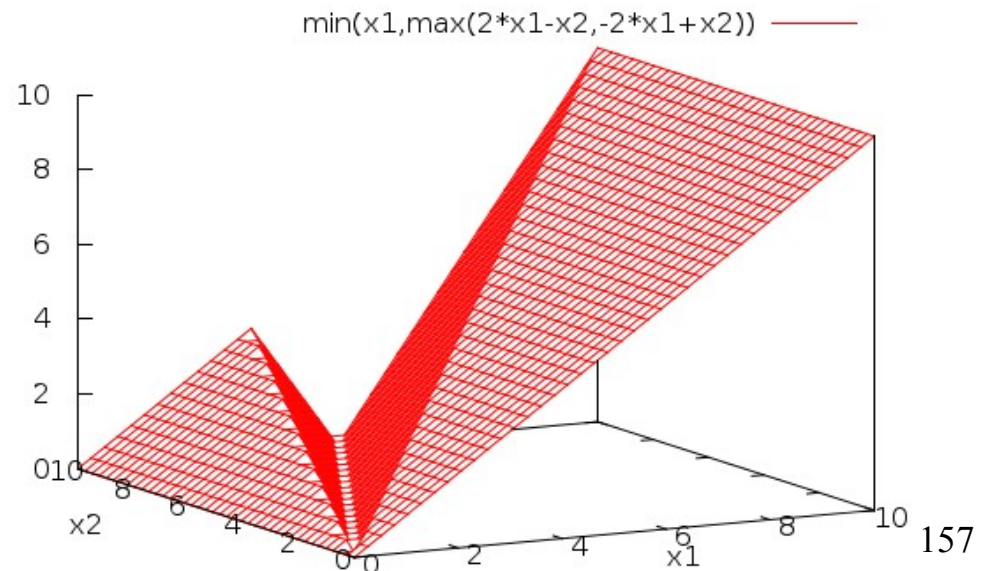
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Real-valued version: A function is stably computed by a **real-valued** CRN if and only if it is *continuous* and *piecewise linear*.

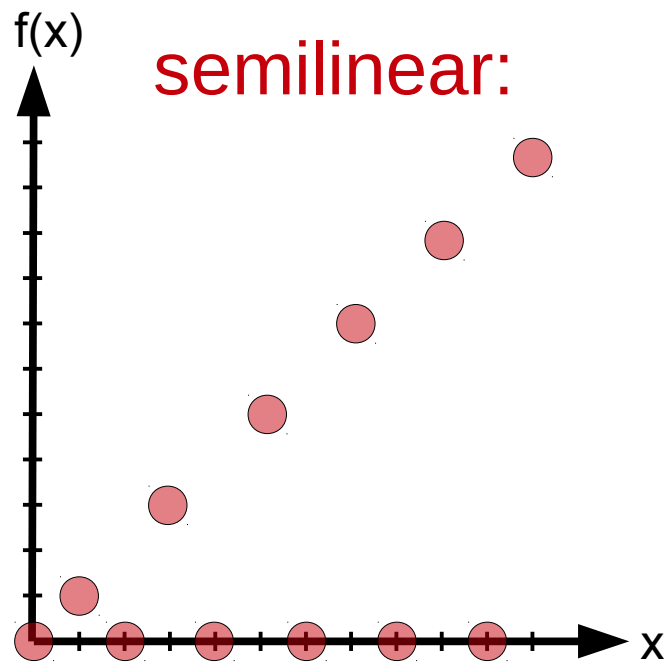
[Chen, D, Soloveichik, [Innovations in Theoretical Computer Science 2014](#)]



Real-valued CRNs

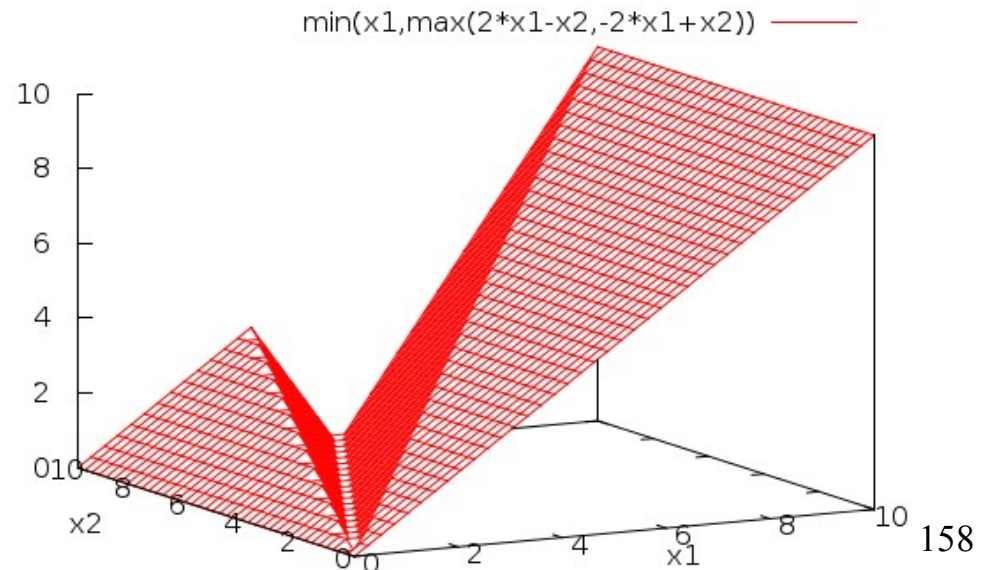
Theorem from previous slide: A function is stably computed by an **integer-valued** CRN if and only if it is *semilinear*.

≈ piecewise linear functions with “discontinuous” pieces



Real-valued version: A function is stably computed by a **real-valued** CRN if and only if it is *continuous* and *piecewise linear*.

[Chen, D, Soloveichik, [Innovations in Theoretical Computer Science 2014](#)]



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$n = \#$ molecules in initial state \approx volume

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polylogarithmic time = “fast” = polynomial in binary expansion of n

linear time = “slow” = exponential in binary expansion of n

Time complexity in CRNs

time until next reaction = exponential random variable

reaction



expected time

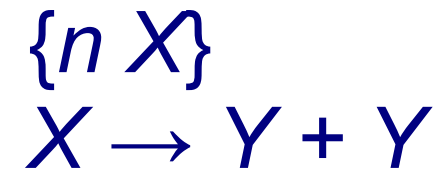
$$1 / \#X$$

$$\text{volume} / (\#A \cdot \#B)$$

Time complexity (example)

$$\{n\ X\}$$
$$X \rightarrow Y + Y$$

Time complexity (example)



E[time to consume all X] =

Time complexity (example)

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Time complexity (example)

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$$\begin{aligned} E[\text{time to consume all } X] = & E[\text{time to consume 1}^{\text{st}} X] && 1/n \\ & + E[\text{time to consume 2}^{\text{nd}} X] && 1/(n-1) \\ & + E[\text{time to consume 3}^{\text{rd}} X] && 1/(n-2) \\ & + \dots \\ & + E[\text{time to consume final } X] && 1/1 \end{aligned}$$

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Time complexity (example)

$\{n X\}$, volume n
 $X + X \rightarrow Y$

Time complexity (example)

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“finite density
constraint”



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volume

$\#X^2$

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$$E[\text{time to consume all } X] = n/n^2 + n/(n-2)^2 + n/(n-4)^2 + \dots + n$$

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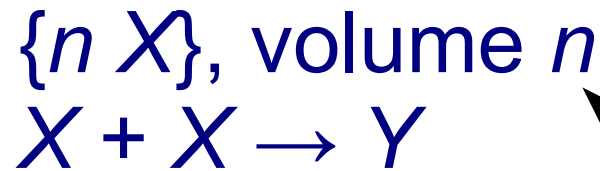
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Is there a faster CRN for dividing
by 2 from initial state $\{n X\}$?

(Open problem)

Time complexity (leader election)

$\{n L\}$, volume n

$L + L \rightarrow L$

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 - if we require 0 probability of error, **no** [D and Soloveichik, in submission]

What if we allow a small probability of error?
(rate-dependent CRN computation)

CRNs with small probability of error are Turing universal

[Angluin, Aspnes, Eisenstat, [Symposium on Distributed Computing 2006](#)]

[Soloveichik, Cook, Winfree, Bruck, [Natural Computing 2008](#)]

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(Informally) A CRN can simulate any algorithm, with a small chance of error.

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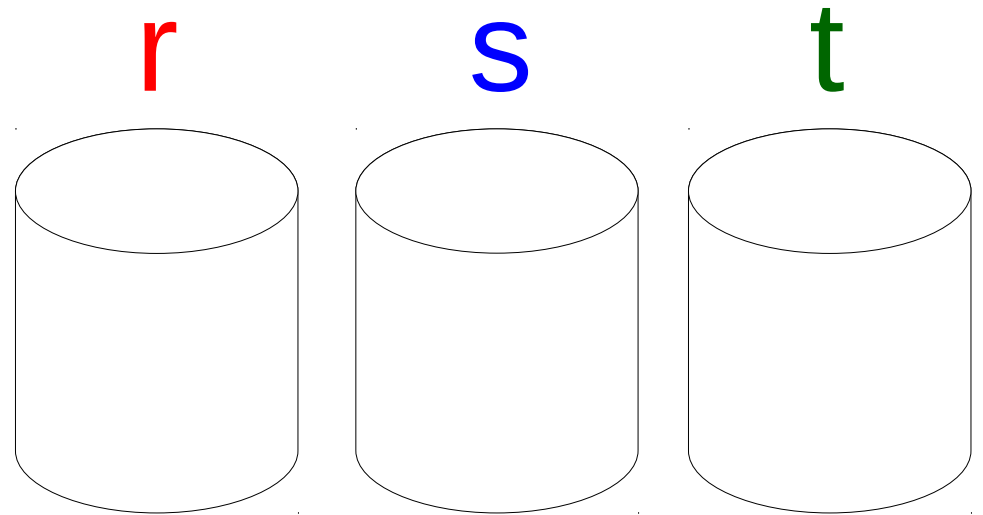
Implication: *General CRN long-term behavior cannot be predicted faster than by simulating.*

[Angluin, Aspnes, Eisenstat, [Symposium on Distributed Computing 2006](#)]

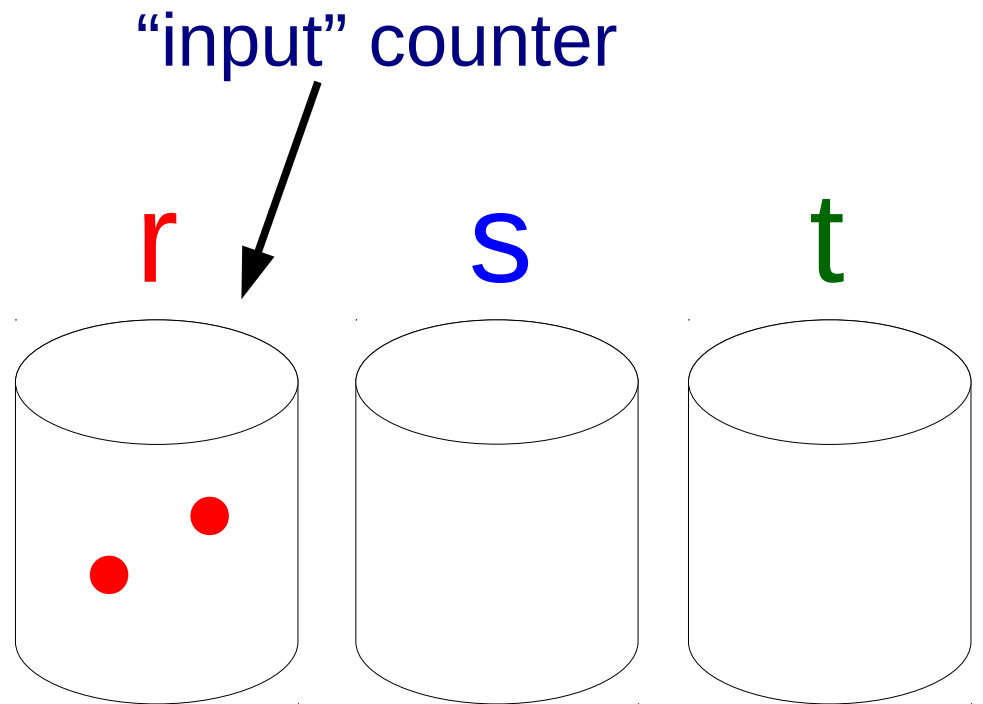
[Soloveichik, Cook, Winfree, Bruck, [Natural Computing 2008](#)]

Counter (register) machine

Counter (register) machine

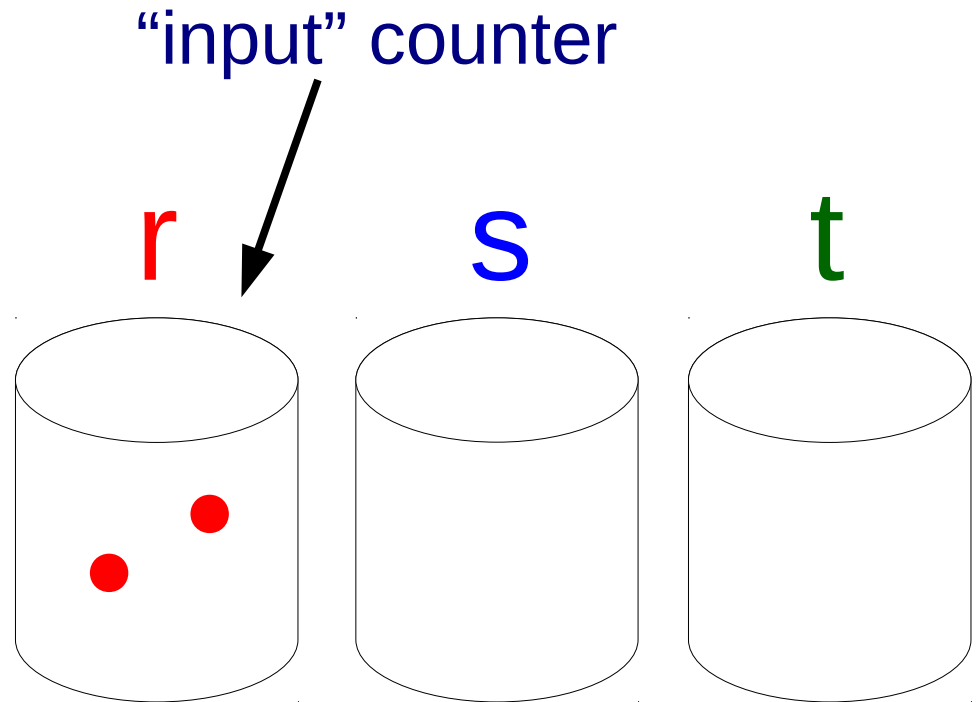


Counter (register) machine



Counter (register) machine

- 1) $dec(r)$
- 2) $inc(s)$
- 3) $inc(s)$
- 4) $inc(s)$
- 5) $dec(t)$
- 6) $inc(s)$



Counter (register) machine

1) *dec(r)*

2) *inc(s)*

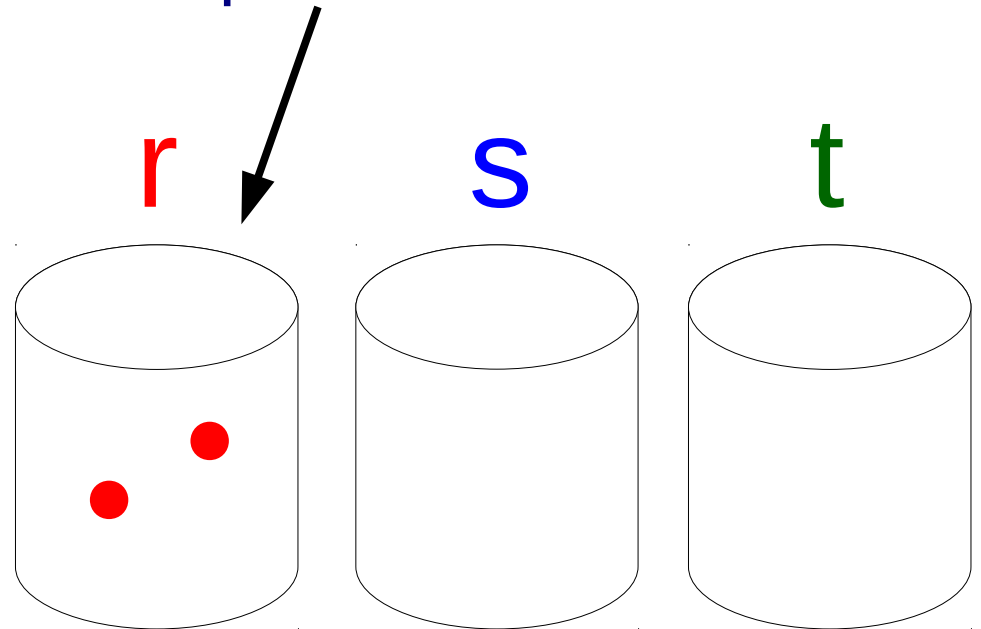
3) *inc(s)*

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5) *dec(t)*

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“input” counter



Counter (register) machine

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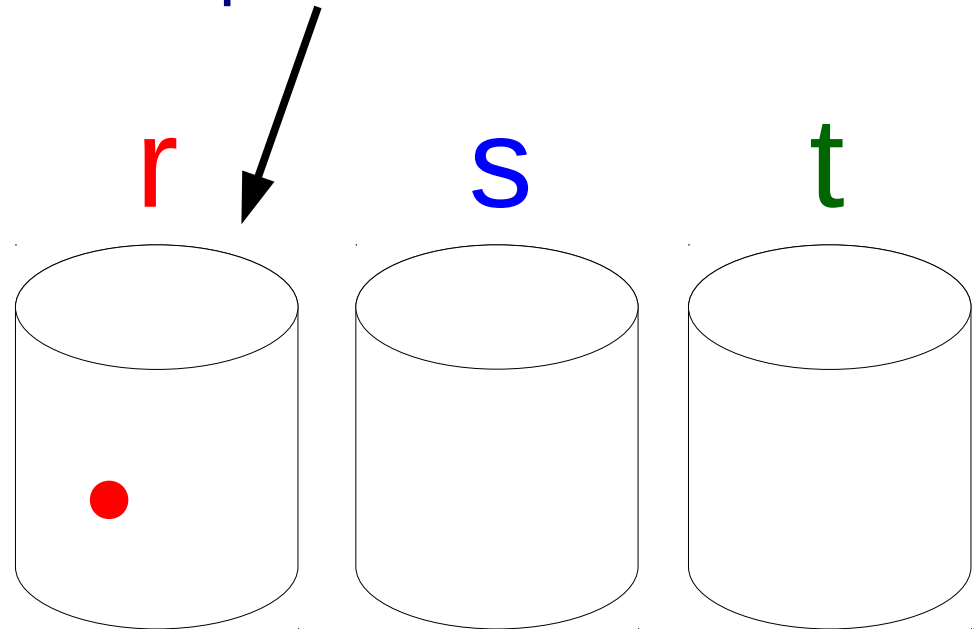
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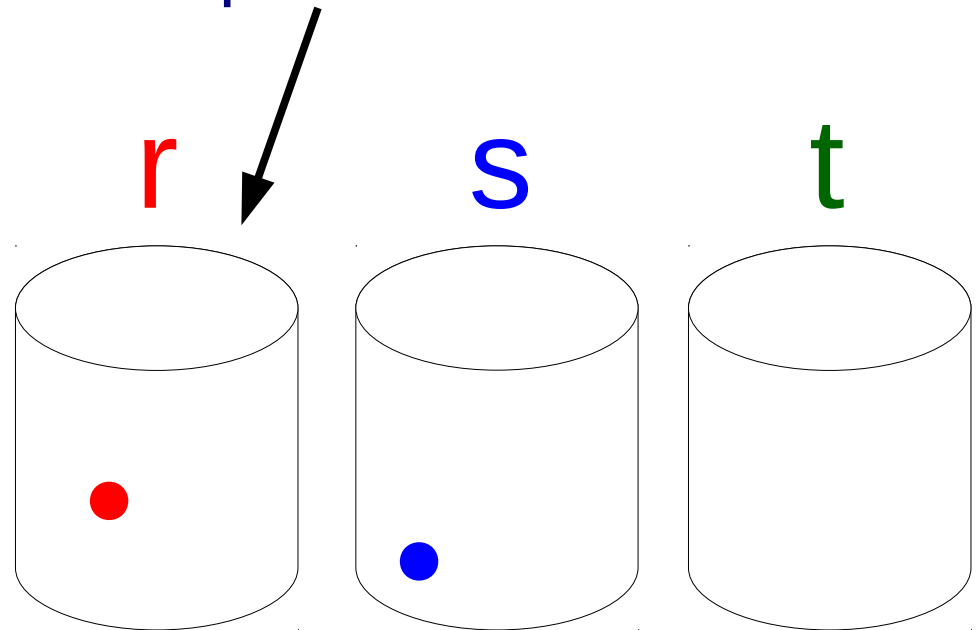
3) $inc(s)$

4) $inc(s)$

5) $dec(t)$

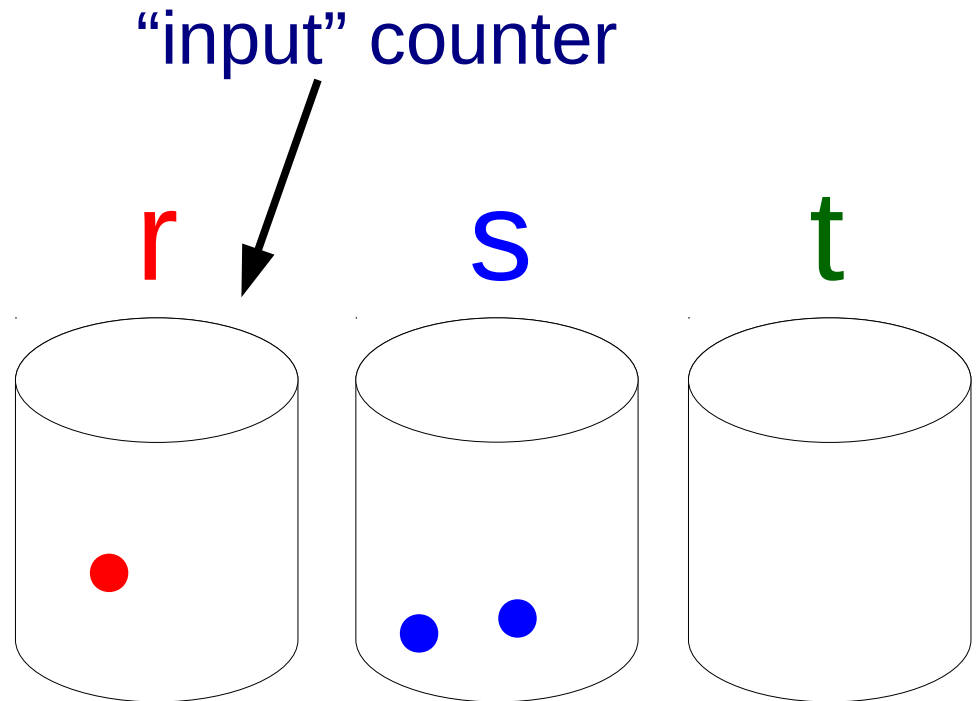
6) $inc(s)$

“input” counter



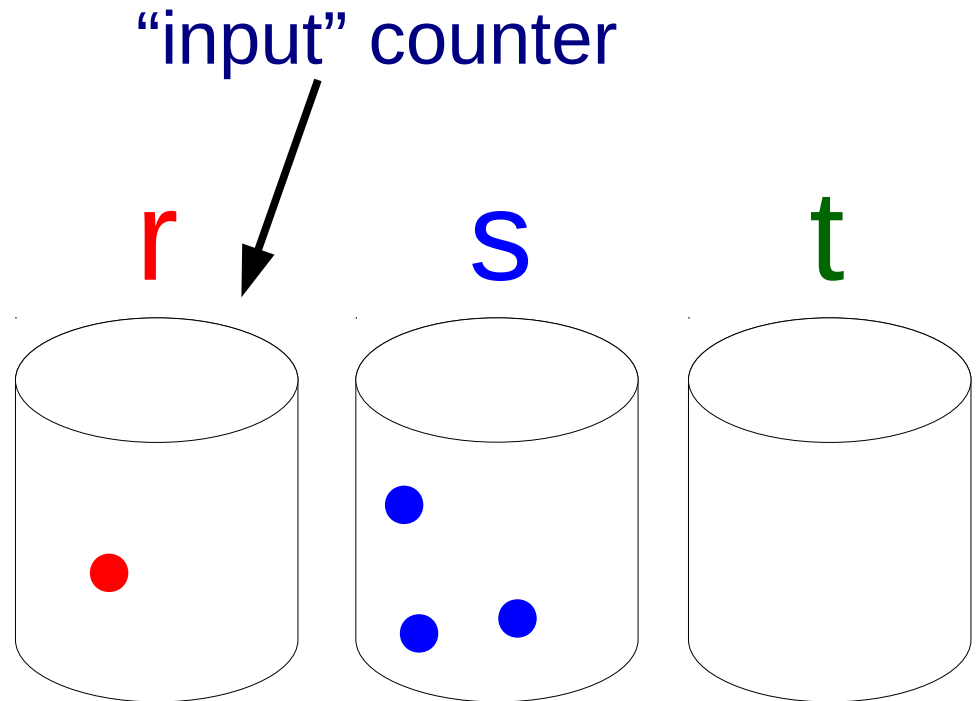
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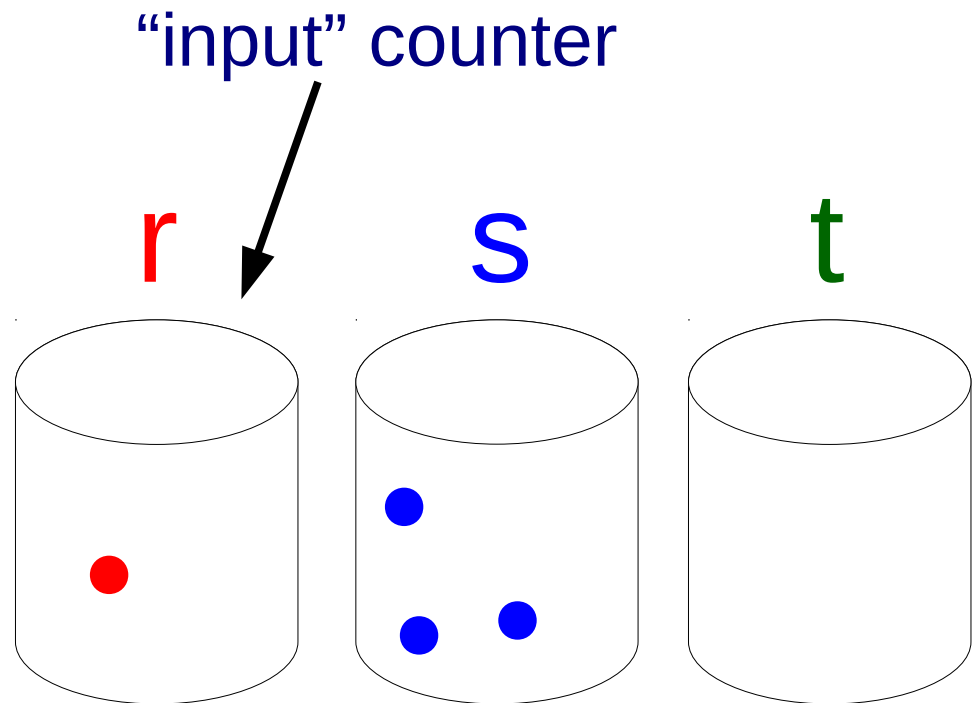
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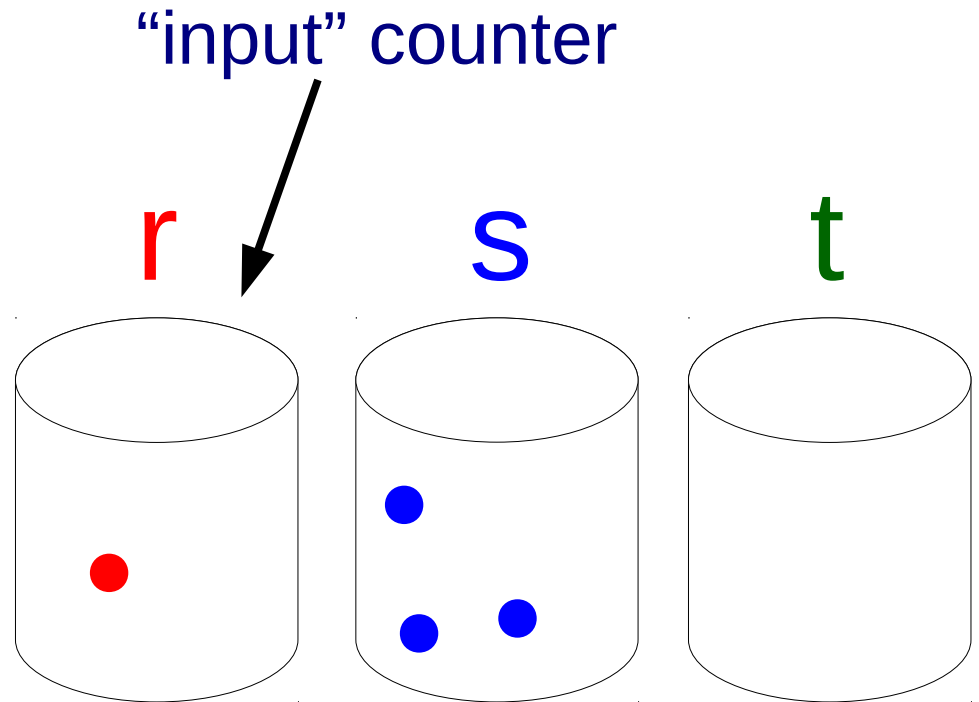
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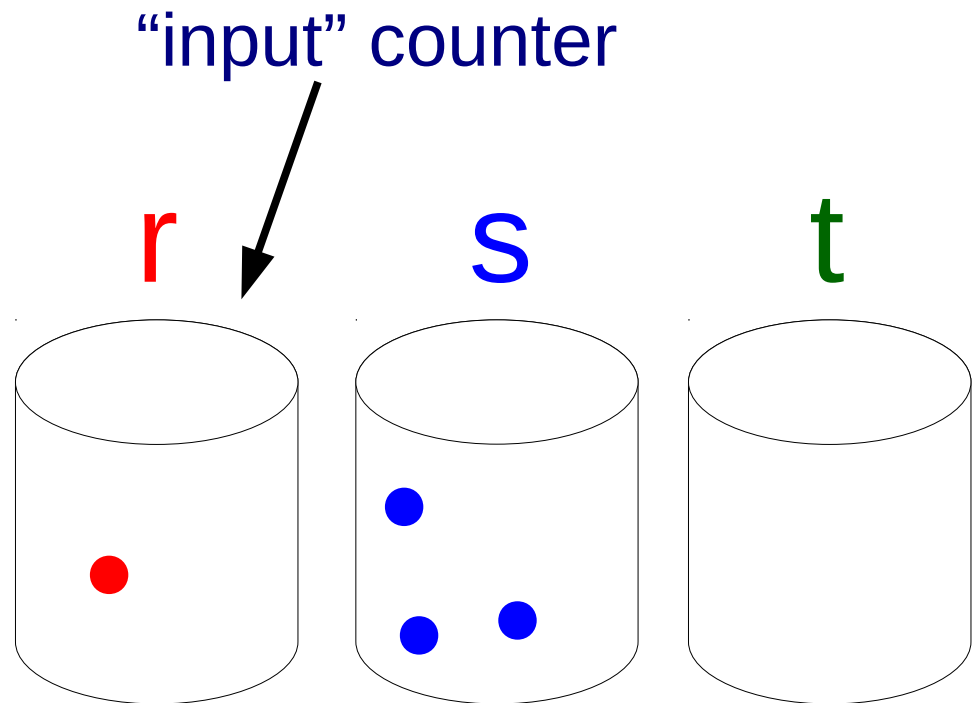
Counter (register) machine

- 1) *dec*(r) if empty goto 6
- 2) *inc*(s)
- 3) *inc*(s)
- 4) *inc*(s)
- 5) *dec*(t) if empty goto 1
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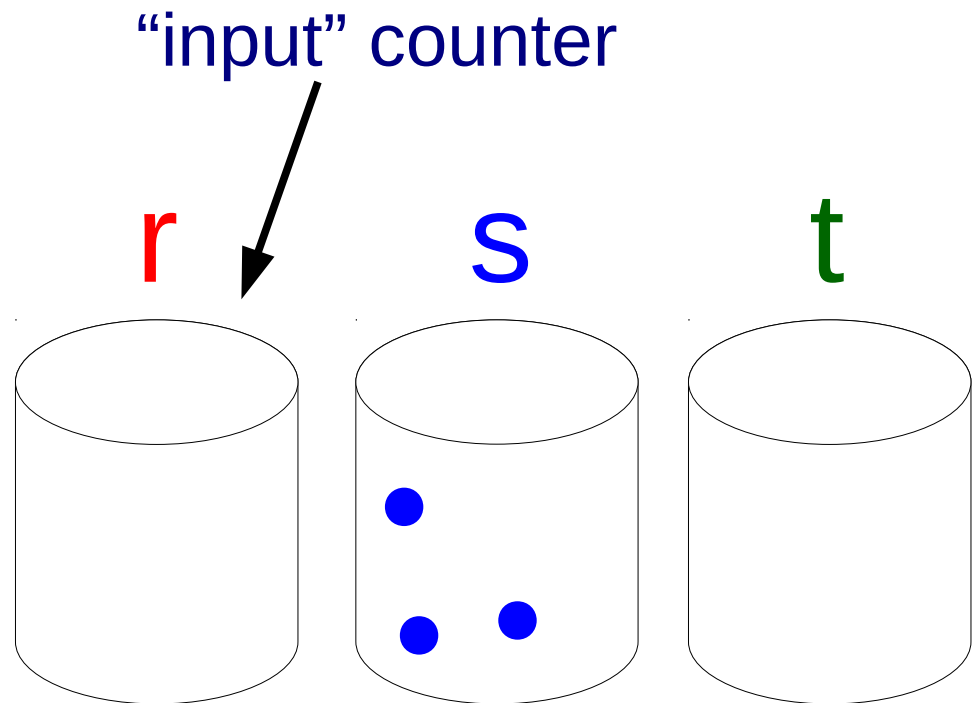
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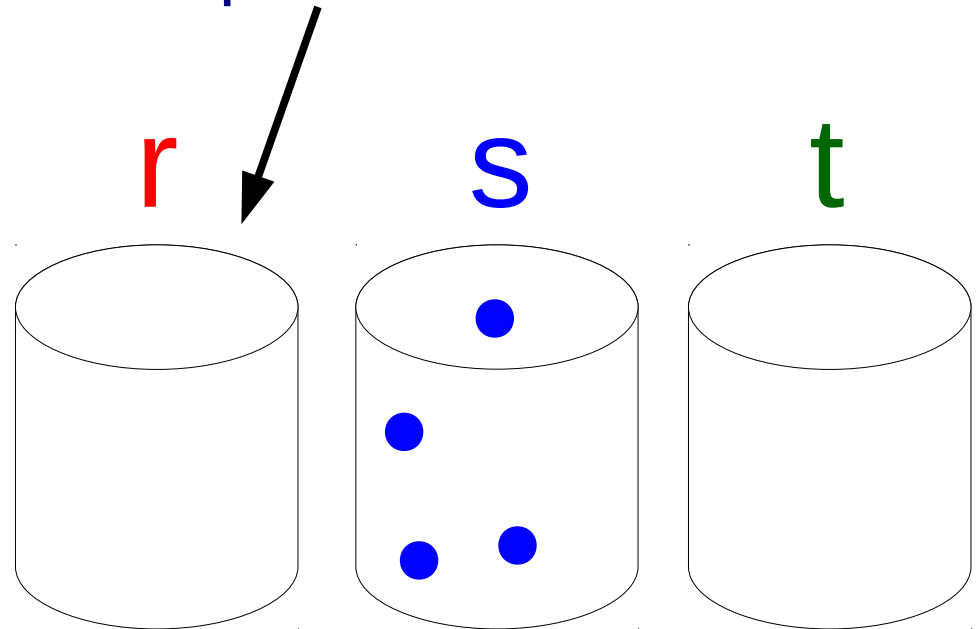
3) *inc*(s)

4) *inc*(s)

5) *dec*(t) if empty goto 1

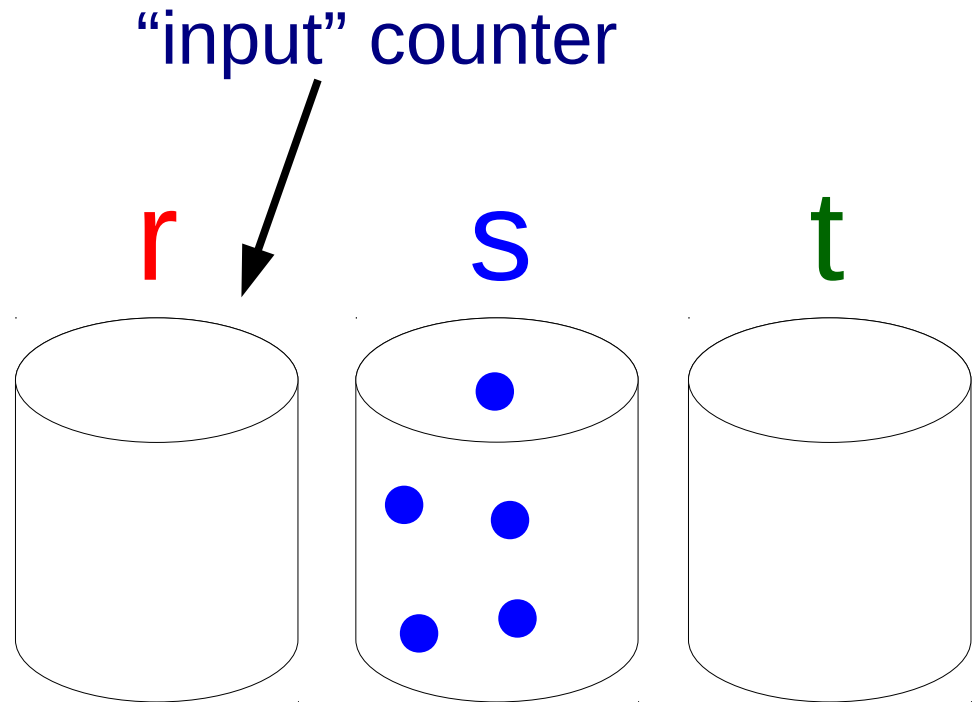
6) *inc*(s)

“input” counter



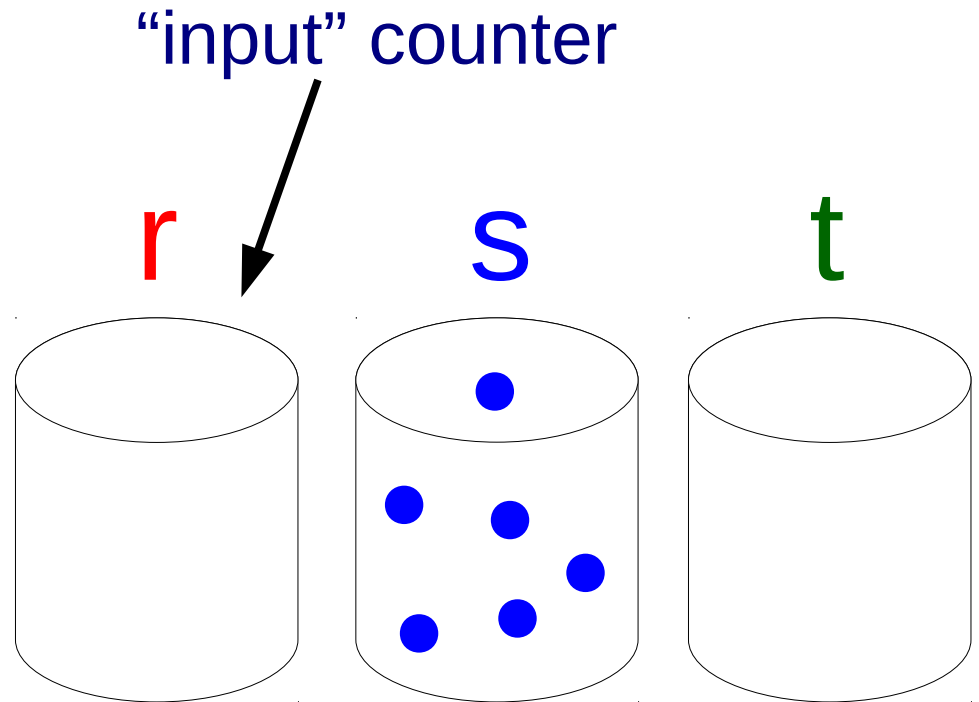
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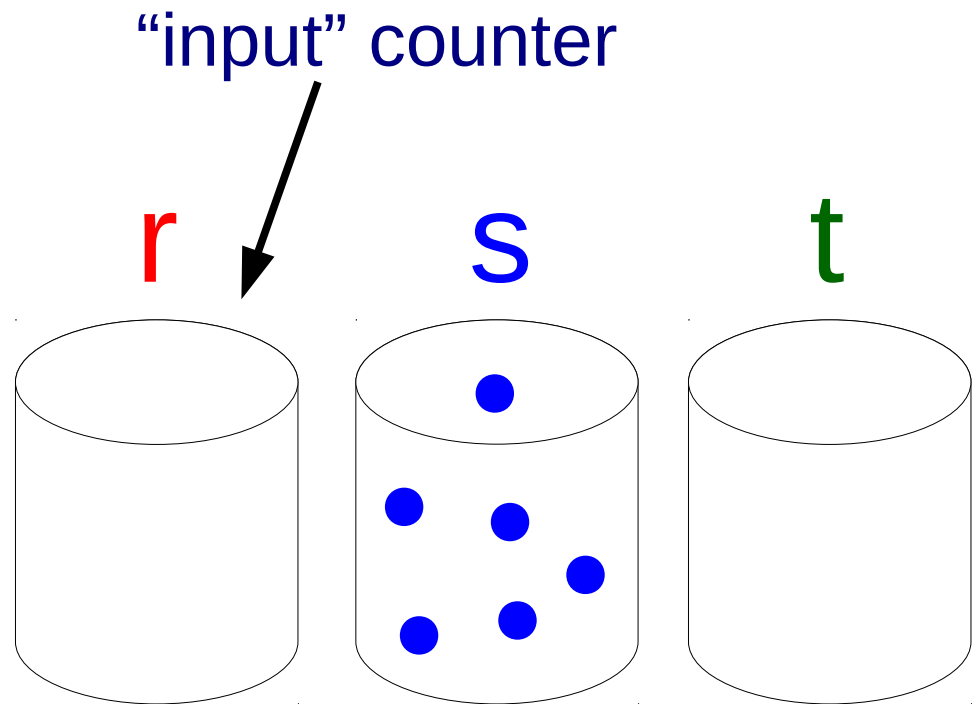
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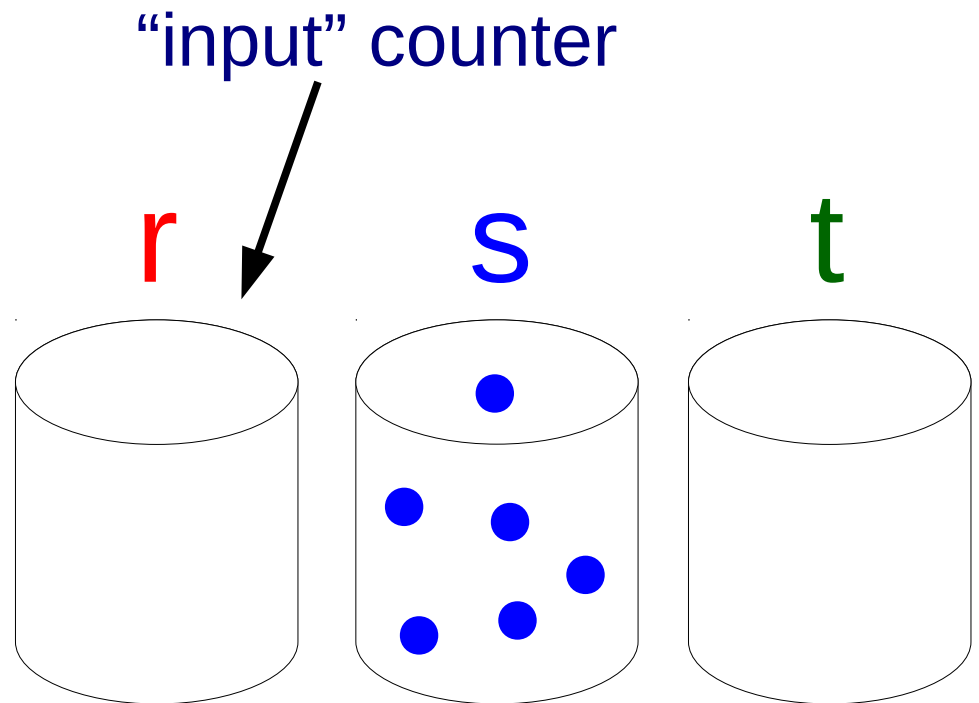
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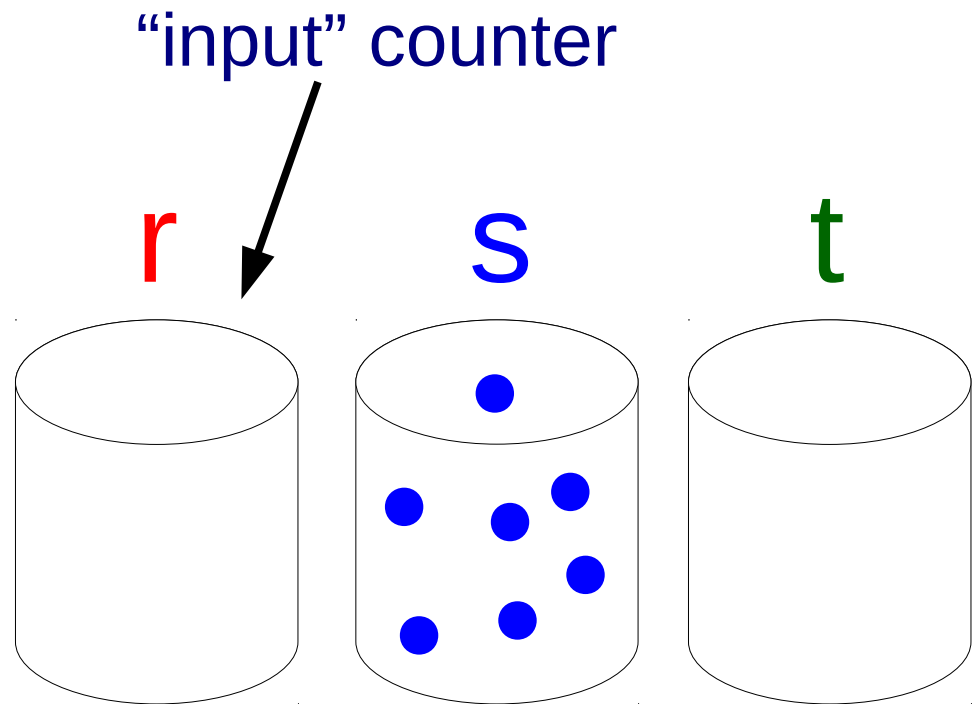
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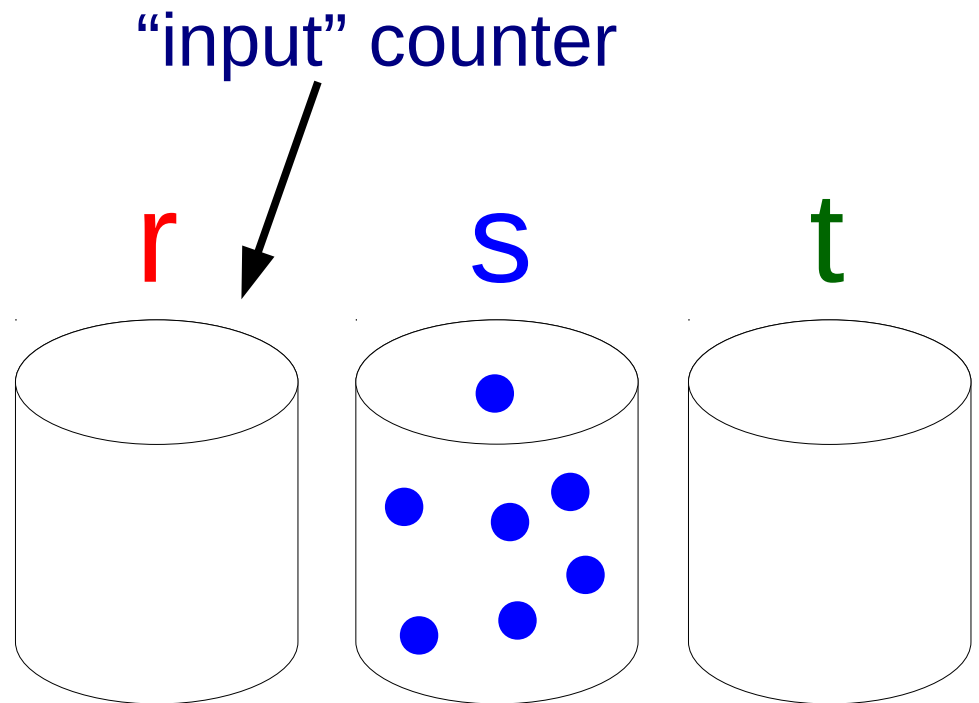
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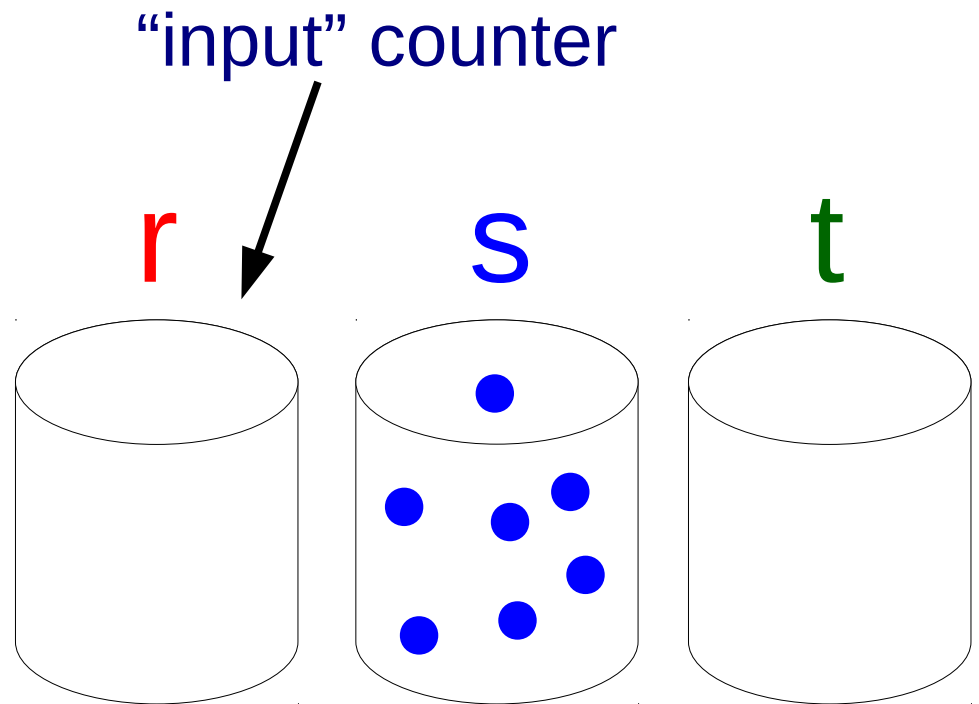
HALT



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HALT



computes $f(n) = 3n+1$

CRNs can simulate counter machines

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Counter machine:

$r = \text{input } n$, start line 1

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- 3) $\text{inc}(s)$
- 4) $\text{dec}(s)$ if zero goto 2

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$r = \text{input } n$, start line 1

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CRN:

initial state $\{n R, 1 L_1\}$

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CRNs can simulate counter machines

Counter machine:

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CRNs can simulate counter machines

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CRN:

initial state $\{n R, 1 L_1\}$



CRNs can simulate counter machines with probability < 1

Counter machine:

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- 1) $\text{inc}(r)$
- 2) $\text{dec}(r)$ if zero goto 1
- 3) $\text{inc}(s)$
- 4) $\text{dec}(s)$ if zero goto 2

CRN:

initial state $\{n R, 1 L_1\}$



Need to be
very slow!

How to slow down reaction $L_2 \rightarrow L_1$?

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Use a **clock**:

$1 C_1, 1 F$

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Use a **clock**:

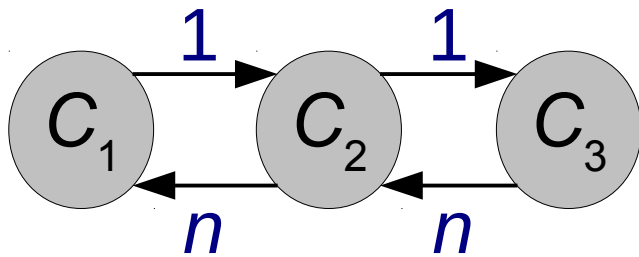
$1 C_1, 1 F$, $n B$



How to slow down reaction $L_2 \rightarrow L_1$?

Use a **clock**:

$1 C_1, 1 F$, $n B$

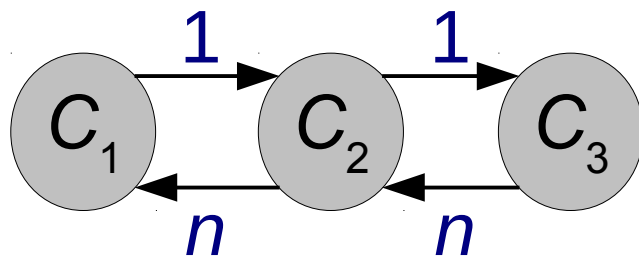


reverse-biased random walk

How to slow down reaction $L_2 \rightarrow L_1$?

Use a **clock**:

$1 C_1, 1 F$, $n B$



C_3 appears after
expected time $\approx n^2$

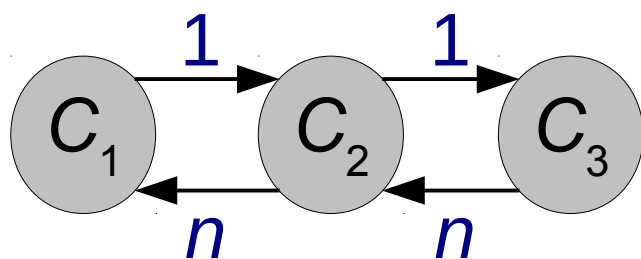
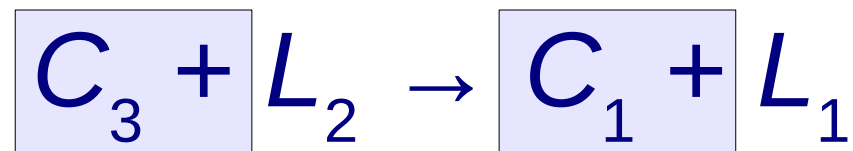
reverse-biased random walk

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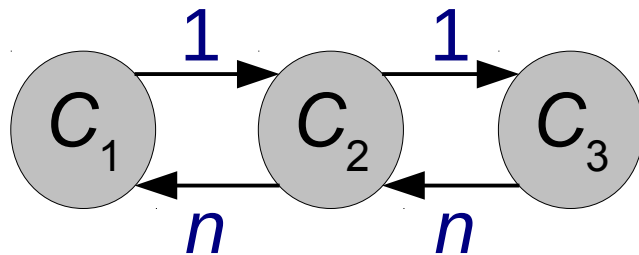
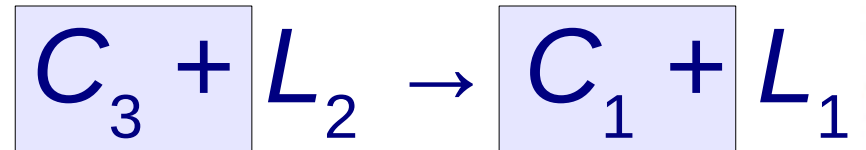
reverse-biased random walk

How to slow down reaction $L_2 \rightarrow L_1$?

Use a **clock**:

$1 C_1, 1 F$

, $n B$



reverse-biased random walk

C_3 appears after
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$$E[\text{time for } L_2 + R \rightarrow L_3] \leq n$$

Probability 1 computation

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- Errr... isn't that stable computation?

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Consider...

$$Y \xrightarrow{2} 2Y$$

$$Y \xrightarrow{1}$$

stably computes $\varphi = \text{no}$
(prob < 1)

initial state $\{1Y, 1N\}$

Probability 1 computation

- Errr... isn't that stable computation?
- With finite state space, yes.

Consider...

$$Y \xrightarrow{2} 2Y$$

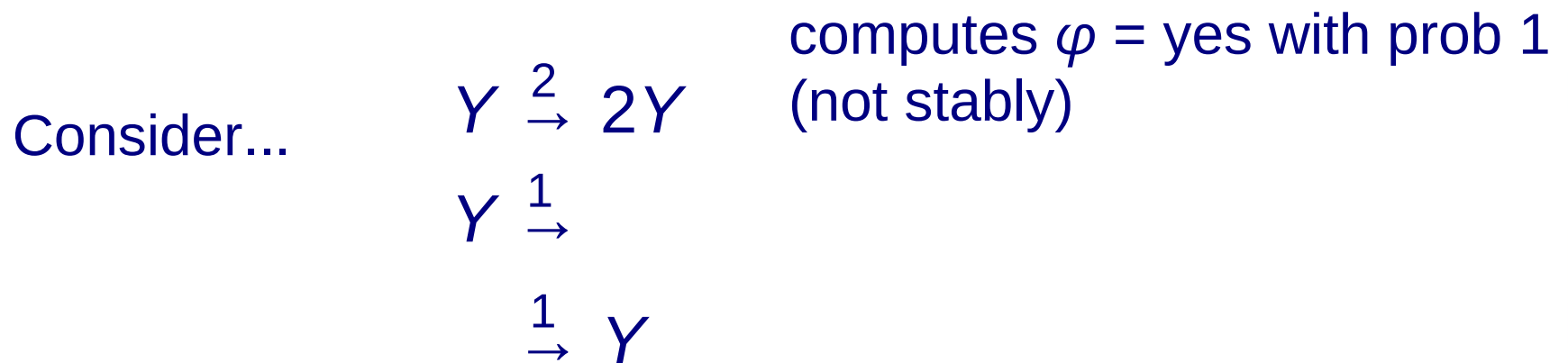
$$Y \xrightarrow{1}$$

$$\xrightarrow{1} Y$$

computes $\varphi = \text{yes}$ with prob 1
(not stably)

Probability 1 computation

- Errr... isn't that stable computation?
- With finite state space, yes.



Theorem: All (Turing) computable predicates/functions can be computed by a CRN with probability 1.

[Cummings, D, Soloveichik, [DNA Computing](#) 2014]

Acknowledgments

Ho-Lin Chen



Rachel Cummings



David Soloveichik

