# Computation with reaction networks 

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Mathematical Modeling of Chemical Reaction Networks Monday, July 24, 2023

## Autonomous computation in a cell

- Goal: Construct synthetic reaction networks that perform pre-determined task.
- Field is called "computation with chemical reaction networks".
- Use sub-cellular substrate (eg. proteins) to devise a "chemical" computer instead of a silicon-based one.
- Tasks can be simple, eg. arithmetic or complex, eg. machine learning.


## Elementary operations

- Implementing identity function.
- Suppose we want to assign a pre-determined value $a$ to the species $X$.
- Think of species $A$ as input species whose concentration $a$ is input value.
- Think of species $X$ as output species whose concentration $x$ will be the output of the computation.
- Want: Concentration of input species never changes. Output species converges to the correct value.


## Elementary operations: identity

Reference: Buisman, Eikelder, Hilbers, Liekens: Computing algebraic functions with biochemical reaction networks

$$
\begin{aligned}
& A \xrightarrow{k_{1}} A+X \\
& X \xrightarrow{k_{2}} 0
\end{aligned}
$$

- $X$ degrades autonomously with rate constant $k_{2}$.
- A catalyzes production of $X$ with rate constant $k_{1}$.
- We do not include the resource (raw material from which $X$ is assembled) and waste (that $X$ breaks down into).
- Set rate constants $k_{1}$ and $k_{2}$ to 1 .

$$
\begin{aligned}
& \dot{a}=0 \\
& \dot{x}=a-x
\end{aligned}
$$

Clearly at steady state $x=a=a(0)$.

## Elementary operations: identity

$$
\begin{aligned}
& A \xrightarrow{k_{1}} A+X \\
& X \xrightarrow{k_{2}} 0 \\
& \dot{a}=0 \\
& \dot{x}=k_{1} a-k_{2} x
\end{aligned}
$$

Explicit solution:

$$
\begin{aligned}
& a(t)=a(0) \\
& x(t)=\frac{1}{k_{2}}\left(k_{1} a_{0}-\left(k_{1} a_{0}-k_{2} x_{0}\right) e^{-k_{2} t}\right) \\
& x(t) \xrightarrow{t \rightarrow \infty} \frac{k_{1}}{k_{2}} a_{0}
\end{aligned}
$$

- Note: The convergence happens at an exponential rate.
- We implemented "multiplying by a fixed constant".


## Digital vs. Analog

- Usual silicon computers are digital.
- Boolean arithmetic and discretization are used to approximate continuous processes.
- A "chemical computer" is nonlinear and analog.
- When performing arithmetic, we are implementing discrete operations which are approximated with continuous processes.


## Elementary operations: multiplication

$$
\begin{gathered}
A+B \xrightarrow{k_{1}} A+B+X \\
X \xrightarrow{k_{2}} 0 \\
\dot{a}=0, \quad \dot{b}=0 \\
\dot{x}=k_{1} a b-k_{2} x
\end{gathered}
$$

Explicit solution:

$$
\begin{aligned}
& a(t)=a(0) \\
& x(t)=\frac{1}{k_{2}}\left(k_{1} a_{0} b_{0}-\left(k_{1} a_{0} b_{0}-k_{2} x_{0}\right) e^{-k_{2} t}\right) \\
& x(t) \xrightarrow{t \rightarrow \infty} \frac{k_{1}}{k_{2}} a_{0} b_{0}
\end{aligned}
$$

## Elementary operations: division

Exercise: How would you implement division?

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$$
\left.\begin{array}{rl} 
& A \xrightarrow{k_{1}} A+X \\
B+X & \xrightarrow{k_{2}} B
\end{array}\right]=k_{1} a-k_{2} b x .
$$

Explicit solution:

$$
\begin{gathered}
x(t)=\frac{1}{k_{2} b_{0}}\left(k_{1} a_{0}-\left(k_{1} a_{0}-k_{2} b_{0} x_{0}\right) e^{-k_{2} b_{0} t}\right) \\
x(t) \xrightarrow{t \rightarrow \infty} \frac{k_{1} a_{0}}{k_{2} b_{0}}
\end{gathered}
$$

## Elementary operations: addition

Exercise: How would you implement addition?

## Elementary operations: addition

Exercise: How would you implement addition?

$$
\begin{aligned}
& A \xrightarrow{1} A+X \\
& B \xrightarrow{1} B+X \\
& X \xrightarrow{1} 0 \\
& \dot{x}=a+b-x
\end{aligned}
$$

## Elementary operations: subtraction

Exercise: How would you implement subtraction? Think first about why subtraction is trickier than all previous operations?

## Elementary operations: subtraction

Exercise: How would you implement subtraction?
Think first about why subtraction is trickier than all previous operations?

- For the previous operations, the input was 1 or 2 positive reals and the output was a positive real. Positive reals can be coded as concentrations of species.
- Applying subtraction to two positive reals can result in a negative real. How to handle this?
- Define proper subtraction (also called monus)

$$
a \dot{-} b= \begin{cases}0 & \text { if } a<b \\ a-b & \text { if } a \geq b\end{cases}
$$

- Many possibilities. Sometimes dual rail construction. One variable tracks positive part while another tracks negative part. $x=x^{+}-x^{-}$with $x^{+}, x^{-}>0$.


## Elementary operations: subtraction

- Non-dual rail construction.
- We need multiple dynamic (non-input) variables.

$$
\begin{aligned}
A & \xrightarrow{k_{1}} A+X \\
B & \xrightarrow{k_{1}} B+Y \\
X & \xrightarrow{k_{1}} 0 \\
X+Y & \xrightarrow{k_{2}} 0
\end{aligned}
$$

$$
\begin{aligned}
\dot{x} & =k_{1} a-k_{1} x-k_{2} x y \\
\dot{y} & =k_{1} b-k_{2} x y
\end{aligned}
$$

- Explicit solution seems difficult.
- Need nonlinear analysis (note the nonlinear term $x y$ ).


## Elementary operations: subtraction

Exercise: Find the positive steady states.

## Elementary operations: subtraction

- Steady states are

$$
\begin{aligned}
x^{*} & =a_{0}-b_{0}, \\
y^{*} & =\frac{k_{1} b_{0}}{k_{2}\left(a_{0}-b_{0}\right)}
\end{aligned}
$$

Exercise: Calculate the Jacobian matrix at the steady state and determine its linear stability.

## Elementary (unary) operation: $n$th root

$$
\begin{aligned}
& A \xrightarrow{k_{1}} A+X \\
n X & \xrightarrow{k_{2}} 0
\end{aligned}
$$

$$
\dot{x}=k_{1} a-k_{2} x^{n}
$$

## Exercises

(1) Show that the steady state for subtraction is exponentially stable. Does this prove that for all initial values we will have convergence to the steady state?
(2) Some of the operations (eg. multiplication, $n$th root) required higher-than-bimolecular complexes. Can you modify the construction such that only bimolecular complexes are involved?

## Structure of a Neural Network

Input layer $L^{0}$

Hidden
layer $L^{1}$

Hidden layer $L^{m-1}$

Output
layer $L^{m}$


Source: David F. Anderson, Badal Joshi, and Abhishek Deshpande. On reaction network implementations of neural networks, Journal of Royal Society Interface, Vol. 18, Issue 177, (April 2021), arXiv.

## Hardwired neural network

A pair of consecutive layers $L^{\ell-1}$ and $L^{\ell}$ along with all edges between the two layers, encode a function $\psi^{\ell}: \mathbb{R}^{c_{\ell-1}} \rightarrow \mathbb{R}^{c_{\ell}}$ which is defined via

$$
\psi^{\ell}(y)=\varphi\left(W^{\ell} y+\beta^{\ell}\right) .
$$

Taking compositions, a hardwired neural network is then simply a visual representation for the function $\Psi_{(G, \mathcal{P}, \varphi)}: \mathbb{R}^{c_{0}} \rightarrow \mathbb{R}_{\geq 0}^{c_{m}}$ defined via

$$
\Psi_{(G, \mathcal{P}, \varphi)}=\psi^{m} \circ \psi^{m-1} \circ \cdots \circ \psi^{1}
$$

The function $\varphi$ is called activation function.

## Hardwired neural network

Commonly used activation functions:
(1) ReLU (Rectified Linear Units)

$$
\varphi_{0}(y)=\max (0, y)
$$

(2) For $h \geq 0$,

$$
\varphi_{h}(y)=\frac{1}{2}\left(y+\sqrt{y^{2}+4 h}\right) .
$$

Plugging $h=0$ in $\varphi_{h}$ gives ReLU. $\varphi_{h}$ is smoothed version of ReLU while being strictly monotone.


## Hardwired neural network

| Which aspect of neural network is implemented chemically? | Chemical implementation of a single directed edge ( $\tilde{X}^{\prime}, \widetilde{X}$ ) of the neural network | Which term results in the ODE for the species $X$ ? |
| :---: | :---: | :---: |
| Closeness to ReLU | $H \longrightarrow H+X$ | $h$ |
| Input $\tilde{X}^{\prime}$ and weight of the edge $\left(\tilde{X}^{\prime}, \widetilde{X}\right)$ | $\begin{gathered} X^{\prime}+W^{+}+X \longrightarrow X^{\prime}+W^{+}+2 X \\ X^{\prime}+W^{-}+X \longrightarrow X^{\prime}+W^{-} \end{gathered}$ | $\begin{gathered} \left(w^{+}-w^{-}\right) x^{\prime} x \\ \text { where } w:=w^{+}-w^{-} \end{gathered}$ <br> implements the edge weight |
| Additive node bias of $\tilde{X}$ | $\begin{gathered} B^{+}+X \longrightarrow B^{+}+2 X \\ B^{-}+X \longrightarrow B^{-} \end{gathered}$ | $\begin{gathered} \left(b^{+}-b^{-}\right) x \\ \text { where } b:=b^{+}-b^{-} \end{gathered}$ <br> implements the node bias |
| $q$-polynomial decay <br> Stability/convergence from $\infty$ | $\begin{gathered} q X \longrightarrow X \\ (q>1) \end{gathered}$ | $(q-1) x^{q}$ |

Chemical implementation of a single directed edge ( $\widetilde{X}^{\prime}, \widetilde{X}$ ) along with nodes $\widetilde{X}^{\prime}$ and $\widetilde{X}$ of the neural network.

$$
\frac{d}{d t} x(t)=h+\left(\left(b^{+}-b^{-}\right)+\left(w^{+}-w^{-}\right) x^{\prime}\right) x(t)-(q-1) x(t)^{q}
$$

## Hardwired neural network

$$
\begin{aligned}
\frac{d}{d t} x(t) & =h+\left(\left(b^{+}-b^{-}\right)+\sum_{i=1}^{c}\left(w_{x_{i}^{\prime}, x}^{+}-w_{x_{i}^{\prime}, x}^{-}\right) x_{i}^{\prime}\right) x(t)-x(t)^{2} \\
& =h+\rho_{x} x(t)-x(t)^{2}
\end{aligned}
$$



Step two: node $\tilde{X}$ along with all its $c$ inputs.

Thank you!

