

Computation with reaction networks

Badal Joshi California State University San Marcos

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

$$A \xrightarrow{k_1} A + X$$
$$X \xrightarrow{k_2} 0$$

- 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.

$$\dot{a} = 0$$
$$\dot{x} = a - x$$

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

Elementary operations: identity

$$A \xrightarrow{k_1} A + X$$
$$X \xrightarrow{k_2} 0$$
$$\dot{a} = 0$$
$$\dot{x} = k_1 a - k_2 x$$

Explicit solution:

$$a(t) = a(0)$$
$$x(t) = \frac{1}{k_2} \left(k_1 a_0 - (k_1 a_0 - k_2 x_0) e^{-k_2 t} \right)$$
$$x(t) \xrightarrow{t \to \infty} \frac{k_1}{k_2} a_0$$

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

Digital vs. Analog

- Usual silicon computers are <u>digital</u>.
- 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

$$A + B \xrightarrow{k_1} A + B + X$$
$$X \xrightarrow{k_2} 0$$

$$\dot{a} = 0, \quad \dot{b} = 0$$

 $\dot{x} = k_1 a b - k_2 x$

Explicit solution:

$$a(t) = a(0)$$

$$x(t) = \frac{1}{k_2} \left(k_1 a_0 b_0 - (k_1 a_0 b_0 - k_2 x_0) e^{-k_2 t} \right)$$

$$x(t) \xrightarrow{t \to \infty} \frac{k_1}{k_2} a_0 b_0$$

Elementary operations: division

Exercise: How would you implement division?

Elementary operations: division

Exercise: How would you implement division?

$$\begin{array}{c} A \xrightarrow{k_1} A + X \\ B + X \xrightarrow{k_2} B \end{array}$$

$$\dot{x} = k_1 a - k_2 b x$$

Explicit solution:

$$x(t) = \frac{1}{k_2 b_0} \left(k_1 a_0 - (k_1 a_0 - k_2 b_0 x_0) e^{-k_2 b_0 t} \right)$$
$$x(t) \xrightarrow{t \to \infty} \frac{k_1 a_0}{k_2 b_0}$$

Elementary operations: addition

Exercise: How would you implement addition?

Elementary operations: addition

Exercise: How would you implement addition?

$$A \xrightarrow{1} A + X$$
$$B \xrightarrow{1} B + X$$
$$X \xrightarrow{1} 0$$

$$\dot{x} = a + b - x$$

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

Exercise: How would you implement subtraction? Think first about <u>why</u> 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 <u>monus</u>)

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

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

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

$$A \xrightarrow{k_1} A + X$$
$$B \xrightarrow{k_1} B + Y$$
$$X \xrightarrow{k_1} 0$$
$$X + Y \xrightarrow{k_2} 0$$
$$\dot{x} = k_1 a - k_1 x - k_2 x y$$

$$\dot{y} = k_1 b - k_2 x y$$

- Explicit solution seems difficult.
- Need nonlinear analysis (note the nonlinear term xy).

Exercise: Find the positive steady states.

• Steady states are

$$x^* = a_0 - b_0,$$

$$y^* = \frac{k_1 b_0}{k_2 (a_0 - b_0)}$$

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

Elementary (unary) operation: nth root

$$\begin{array}{c} A \xrightarrow{k_1} A + X \\ nX \xrightarrow{k_2} 0 \end{array}$$

$$\dot{x} = k_1 a - k_2 x^n$$



- 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?
- Some of the operations (eg. multiplication, nth root) required higher-than-bimolecular complexes. Can you modify the construction such that only bimolecular complexes are involved?

Structure of a Neural Network



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

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

Taking compositions, a hardwired neural network is then simply a visual representation for the function $\Psi_{(G,\mathcal{P},\varphi)}: \mathbb{R}^{c_0} \to \mathbb{R}^{c_m}_{\geq 0}$ defined via

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

The function φ is called activation function.

Hardwired neural network

Commonly used activation functions: ReLU (Rectified Linear Units)

$$\varphi_0(y) = \max(0, y)$$

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

Plugging h = 0 in φ_h gives ReLU. φ_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}', \tilde{X}) of the neural network | Which term results in the ODE for the species X ? |
|--|---|--|
| Closeness to ReLU | $H \longrightarrow H + X$ | h |
| Input \widetilde{X}' and weight of the edge $(\widetilde{X}', \widetilde{X})$ | $X' + W^+ + X \longrightarrow X' + W^+ + 2X$ $X' + W^- + X \longrightarrow X' + W^-$ | $(w^+ - w^-)x'x,$ where $w \coloneqq w^+ - w^-$ implements the edge weight |
| Additive node bias of \widetilde{X} | $B^+ + X \longrightarrow B^+ + 2X$ $B^- + X \longrightarrow B^-$ | $(b^+ - b^-)x,$ where $b := b^+ - b^-$ implements the node bias |
| q-polynomial decay Stability/convergence from ∞ | $qX \longrightarrow X$ $(q > 1)$ | $(q-1)x^q$ |

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

$$\frac{d}{dt}x(t) = h + \left(\left(b^+ - b^-\right) + \left(w^+ - w^-\right)x'\right)x(t) - (q-1)x(t)^q.$$

Hardwired neural network



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

Thank you!