Molecular Programming

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INRIA Scientific Board, Paris, 2011-11-18 http://lucacardelli.name

Smaller and Smaller

First working transistor

John Bardeen and Walter Brattain, Dec. 23, 1947.

First integrated circuit Jack Kilby, Sep. 1958.

50 years later

25nm NAND flash

Intel&Micron, Jan. 2010. ~50atoms.

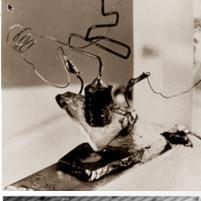
Single molecule transistor

Observation of molecular orbital gating. *Nature*, 2009; 462 (7276): 1039

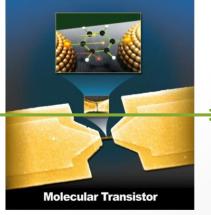
Molecules on a chip

~10 Moore's Law cycles left!







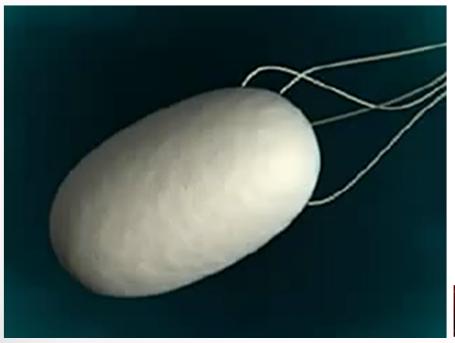




Placement and orientation of individual DNA shapes on lithographically patterned surfaces. Nature Nanotechnology 4, 557 - 561 (2009).

Building The Smallest Things

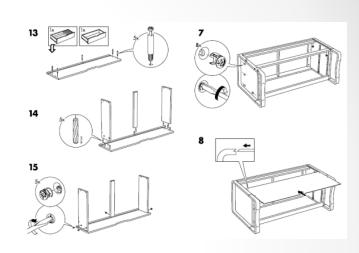
- How do we build structures that are by definition smaller than your tools?
- Basic answer: you can't. Structures (and tools) should build themselves!
- By programmed self-assembly.





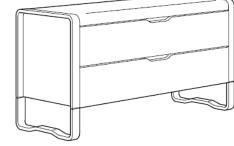
Molecular IKEA

- Nature can self-assemble. Can we?
- "Dear IKEA, please send me a chest of drawers that assembles itself."
- · We need a magical material where the pieces are pre-programmed to fit into to each other.
- At the molecular scale many such materials exist...





Add water



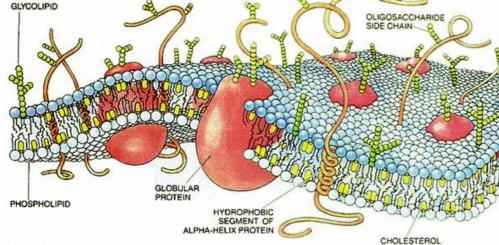
http://www.ikea.com/ms/en_US/custome r_service/assembly_instructions.html

Programmed Self-Assembly

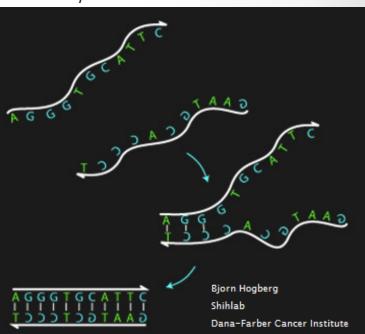
Proteins

Wikimedia





DNA/RNA



Membranes

Molecular Languages - modeling languages -

Chemistry

Chemical reactions

$$\circ$$
 A + B \rightarrow _r C + D

(a program)

Ordinary Differential Equations

```
\circ d[A]/dt = -r[A][B] ... (a semantics)
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Rich analytical techniques based on Calculus

- But prone to combinatorial explosion
 - E.g., due to the peculiarities of protein interactions

High(er)-Level Languages

Gene Networks

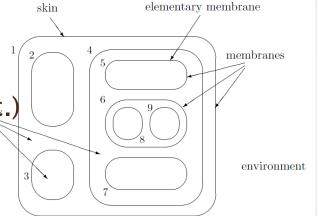
- Synchronous Boolean networks
 - · Stewart Kauffman, etc.
- Asynchronous Boolean networks
 - · René Thomas, etc.

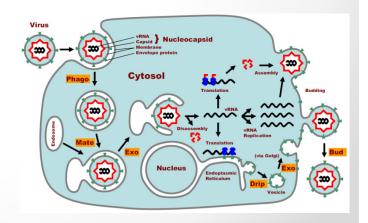
Protein Networks

- \circ Process Algebra (stochastic π -calculus etc.)
 - · Priami, Regev-Shapiro, etc.
- o Graph Rewriting (kappa, BioNetGen etc.)
 - Danos-Laneve, Fontana & al., etc.

Membrane Networks

- Membrane Computing
 - Gheorghe Păun, etc.
- o Brane Calculi
 - · Luca Cardelli, etc.

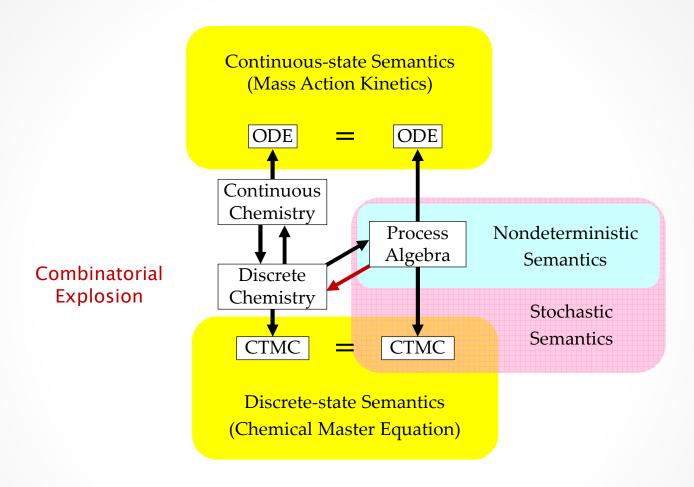




Molecular Languages

- Reaction-Based $(A + B \rightarrow C + D)$ (Chemistry)
 - Limited to finite set of species (no polymerization)
 - Practically limited to small number of species (no run-away complexation)
- Interaction–Based (A = !r; C) (Process Algebra)
 - Reduces combinatorial complexity of models by combining independent submodels connected by interactions.
- Rule-Based $(A{-}:B{p} \rightarrow A{p}:B{-})$ (Logic, Graph Rewriting)
 - Further reduces model complexity by describing molecular state, and by allowing one to 'ignore the context': a *rule* is a reaction in an unspecified (complexation/phosphorylatio) context.
 - Similar to informal descriptions of biochemical events ("narratives").
- Syntactic connections
 - The latter two can be translated (to each other and) to the first, but doing so may introduce an infinite, or anyway extremely large, number of species.

Semantic Connections



These diagrams commute via appropriate maps.

L. Cardelli: "On Process Rate Semantics" (TCS)

L. Cardelli: "A Process Algebra Master Equation" (QEST'07)

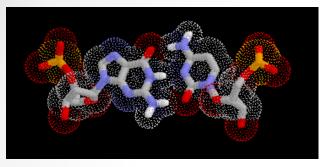
But what about Execution?

- Chemistry is not easily executable
 - Please Mr Chemist, execute me these reactions that I just made up.
- Similarly, the molecular languages seen so fare are descriptive (modeling) languages
- How can we actually execute molecular languages? With real molecules?

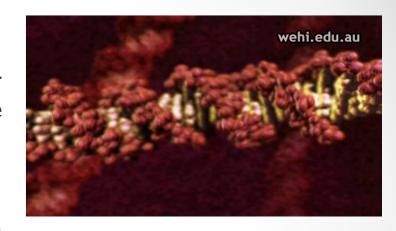
Molecular Languages

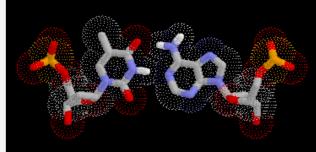
- executable languages -

DNA



GC Base Pair Guanine-Cytosine

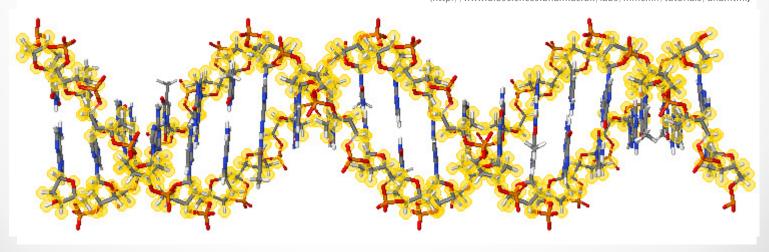




TA Base Pair Thymine-Adenine

Interactive DNA Tutorial

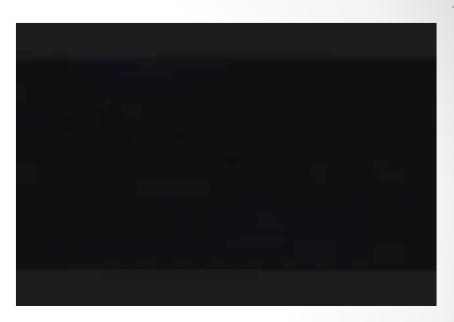
(http://www.biosciences.bham.ac.uk/labs/minchin/tutorials/dna.html)



Sequence of Base Pairs (GACT alphabet)

Robust, and Long

- DNA in each human cell:
 - 3 billion base pairs
 - 2 meters long, 2nm thick
 - o folded into a 6μm ball
 - 750 MegaBytes
- A huge amount for a cell
 - Every time a cell replicates it has to copy 2 meters of DNA reliably.
 - To get a feeling for the scale disparity, compute:
- DNA in human body
 - 10 trillion cells
 - 133 Astronomical Units long
 - 7.5 OctaBytes
- DNA in human population
 - 20 million light years long



DNA wrapping into chromosomes

wehi.edu.au



Andromeda Galaxy
2.5 million light years away

Natural DNA Operation

DNA can support structural and computational complexity.



DNA replication in *real time*

In Humans: 50 nucleotides/second Whole genome in a few hours (with parallel processing)

In Bacteria: 1000 nucleotides/second (higher error rate)



DNA transcription in real time

RNA polymerase II: 15-30 bases/second

Drew Berry http://www.wehi.edu.au/wehi-tv

Unnatural DNA Operation

Sensing

- Reacting to forces
- Binding to molecules

Actuating

- Releasing molecules
- Producing forces

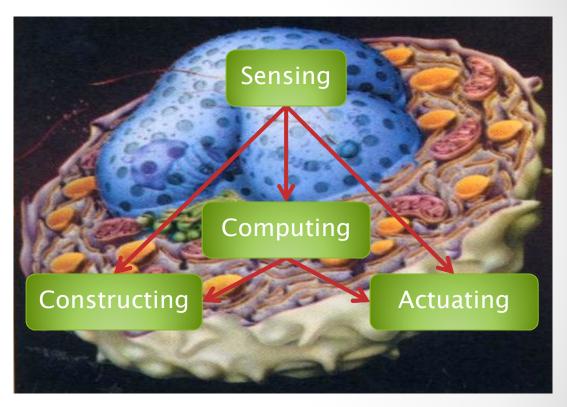
Constructing

- o Chassis
- Growth

Computing

- Signal Processing
- Decision Making

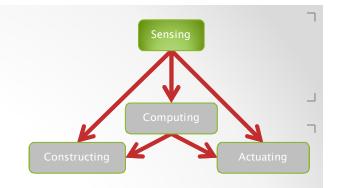
Nanoscale Control Systems

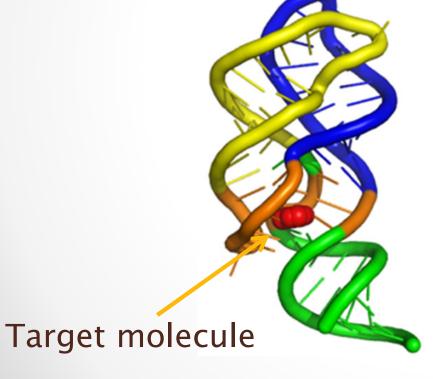


Nucleic Acids can do all this. And interface to biology.

Sensing

Aptamers: natural or artificially evolved DNA molecules that stick to other molecules (highly selectively).





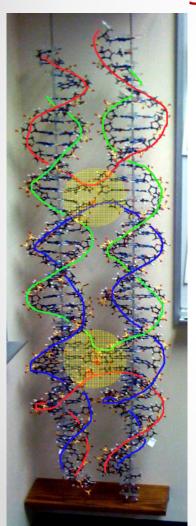
Adenine riboswitch aptamer

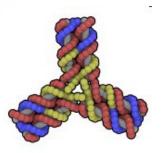
Structural basis for discriminative regulation of gene expression by adenine- and guanine-sensing mRNAs. Chem Biol. 2004 Dec;11(12):1729-41.

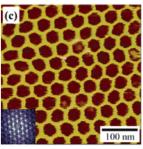
Constructing

Computing Constructing Actuating

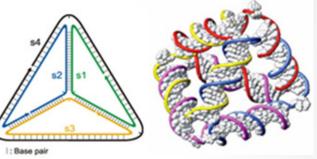
Crosslinking











Chengde Mao, Purdue

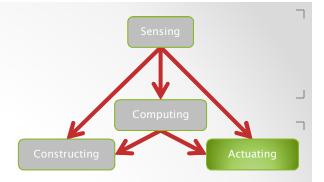
Andrew Turberfield, Oxford

Folding DNA into Twisted and Curved Nanoscale Shapes

Hendrik Dietz, Shawn M. Douglas, & William M. Shih Science, 325:725-730, 7 August 2009.



Actuating

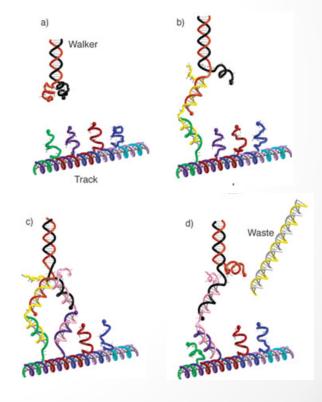


DNA tweezers

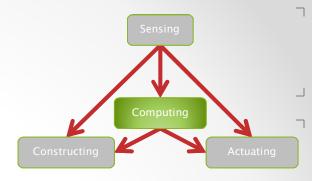


Bernard Yurke, Boise State

DNA walkers



Computing

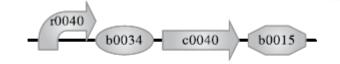


- Sensors and Actuators at the 'edge' of the system
 - They can use disparate technologies and phenomena
- Computation in the 'kernel' of the system
- Compositionality in the kernel
 - The components should use uniform inputs and outputs
 - The components should be 'computationally complete'

"Embedded" Computing

(Synthetic Biology)

- Using bacterial machinery (e.g.) as the hardware.
 Using embedded gene networks as the software.
- MIT Registry of Standard Biological Parts
- GenoCAD
 - Meaningful sequences [Cai et al.]



r0040:prom; b0034:rbs; c0040:pcr; b0015:ter

- GEC
 - [Pedersen & Phillips]

```
prom<neg(C)>; rbs; pcr<codes(A)>; ter;
prom<neg(A)>; rbs; pcr<codes(B)>; ter;
prom<neg(B)>; rbs; pcr<codes(C)>; ter
```

"Autonomous" Computing

(Nano-engineering)

- Mix & go
 - All (or most) parts are synthesized
 - No manual cycling (cf. early DNA computing)
 - In some cases, all parts are made of DNA (no enzyme/proteins)
- Self-assembled and self-powered
 - Can run on its own (e.g. environmental sensing)
 - Or be embedded into organisms, but running 'separately'

Curing

A doctor in each cell

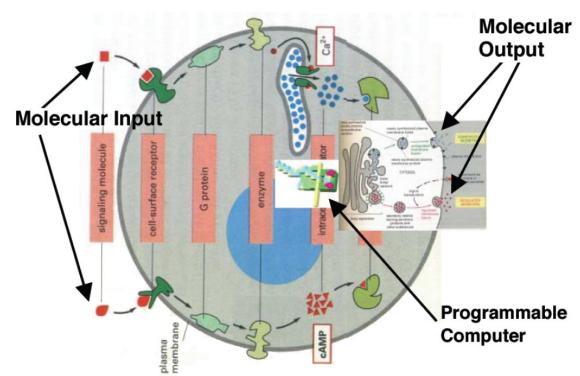
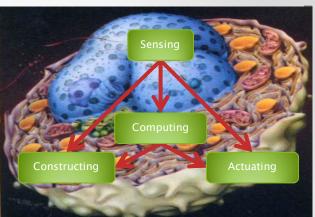


Fig. 1 Medicine in 2050: "Doctor in a Cell"

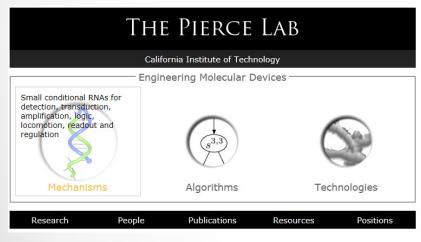
Ehud Shapiro

Rivka Adar Kobi Benenson Gregory Linshitz Aviv Regev William Silverman Molecules and computation



RNA operation in (dead) cells

- Using RNA Hybridization Chain Reaction for imaging of mRNA expression.
 - The programmability of orthogonal RNA reactions enables spatial imaging with 5 simultaneous targets.





Molecular Computation

DNA Computing

Non-goals

- Not to solve NP-complete problems.
- Not to replace electronics.
- Not necessarily using genes or producing proteins.

For general 'molecular programming'

- To precisely control the organization and dynamics of matter and information at the molecular level.
- To interact algorithmically with biological entities.
- The use of DNA is "accidental": no genes involved.
- o In fact, no material of biological origin.

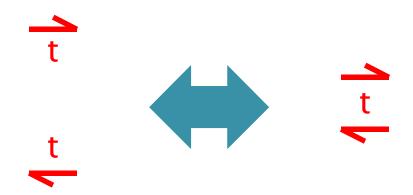
Domains

• Subsequences on a DNA strand are called domains. *PROVIDED* they are "independent" of each other.



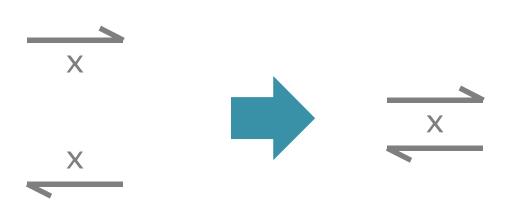
- I.e., differently named domains must not hybridize:
 - With each other
 - With each other's complement
 - With subsequences of each other
 - With concatenations of other domains (or their complements)
 - o Etc.
- Choosing domains (subsequences) that are suitably independent is a tricky issue that is still somewhat of an open problem (with a vast literature). But it can work in practice.

Short Domains

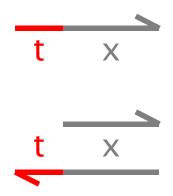


Reversible Hybridization

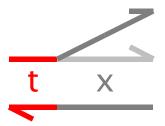
Long Domains



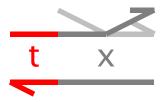
Irreversible Hybridization



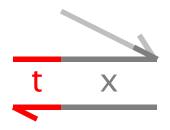
"Toehold Mediated"



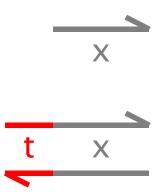
Toehold Binding



Branch Migration

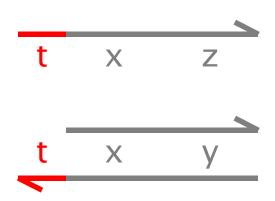


Displacement

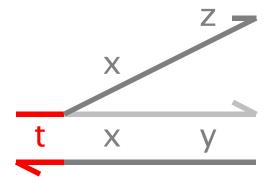


Irreversible release

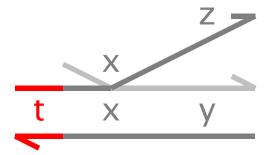
Bad Match



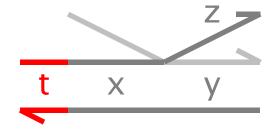
Bad Match



Bad Match



Bad Match



Cannot proceed Hence will undo

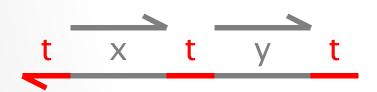
Two-Domain Architecture

Signals: 1 toehold + 1 recognition region



Garbage collection "built into" the gates

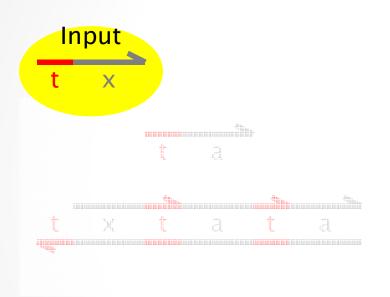
 Gates: "top-nicked double strands" (or equivalently double strands with open toeholds)

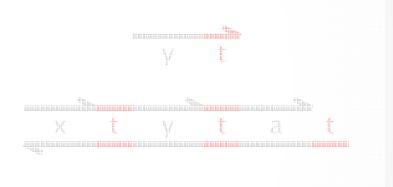


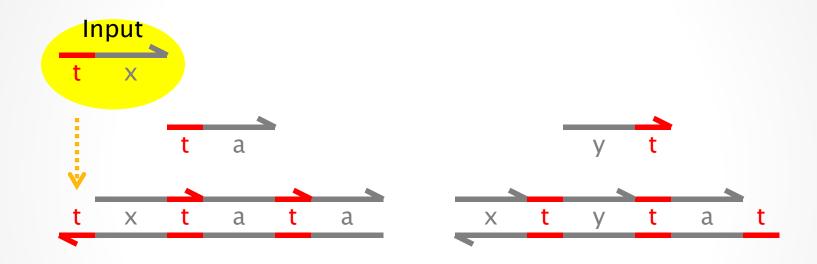
Two-Domain DNA Strand Displacement

Luca Cardelli

In S. B. Cooper, E. Kashefi, P. Panangaden (Eds.): Developments in Computational Models (DCM 2010). EPTCS 25, 2010, pp. 33-47. May 2010.

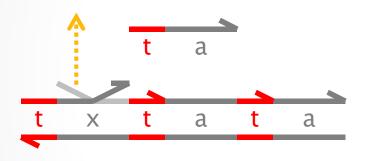


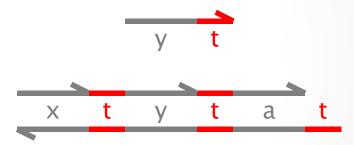


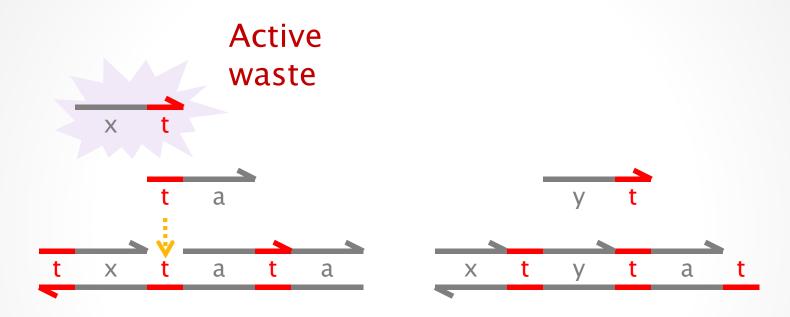


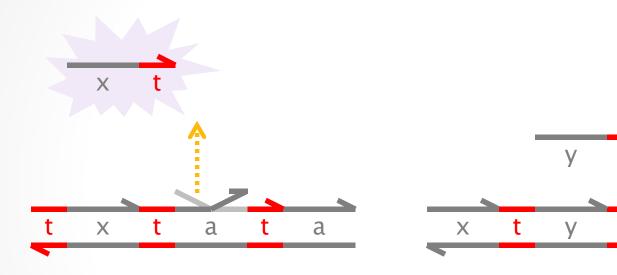
Built by self-assembly!

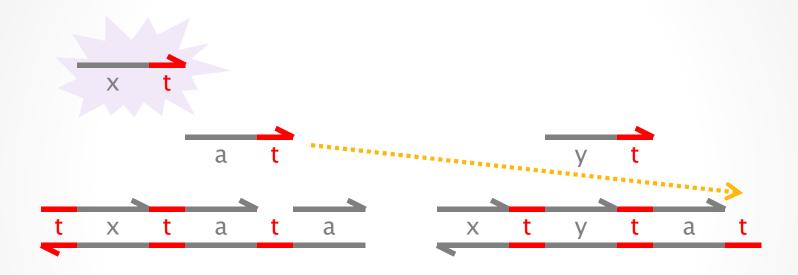
ta is a *private* signal (a different 'a' for each xy pair)



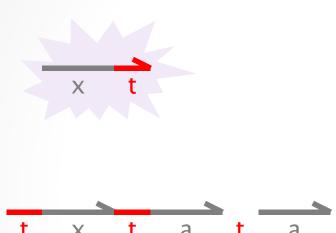


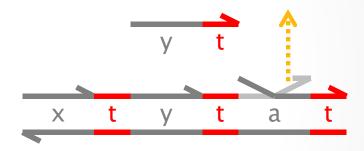


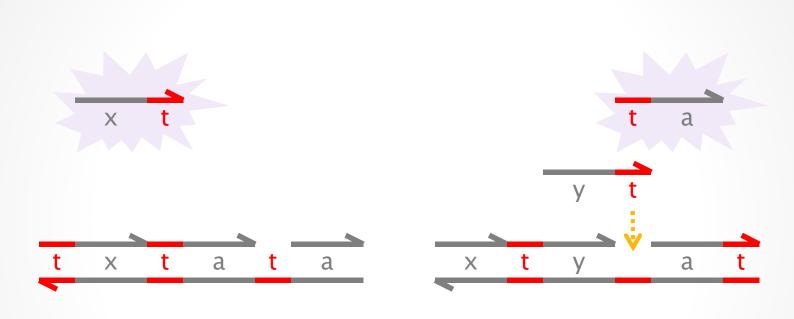


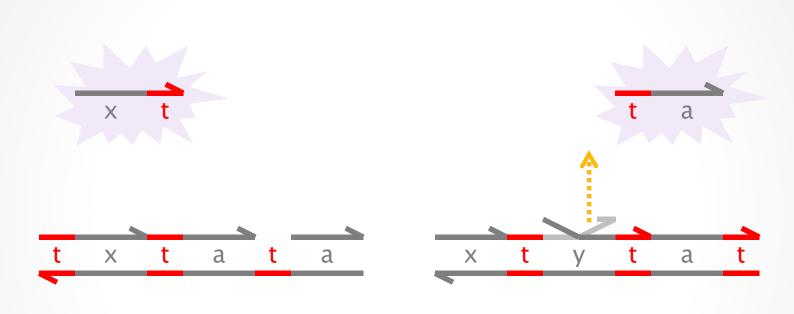


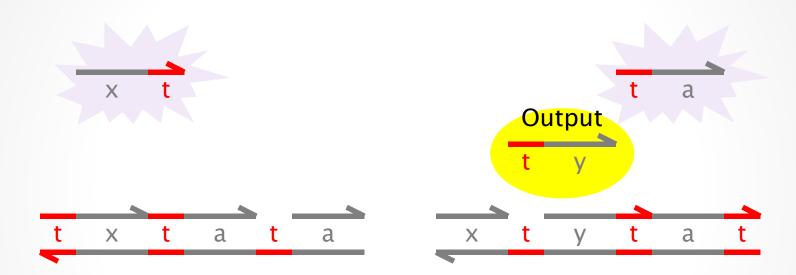
So far, a tx *signal* has produced an at *cosignal*. But we want signals as output, not cosignals.









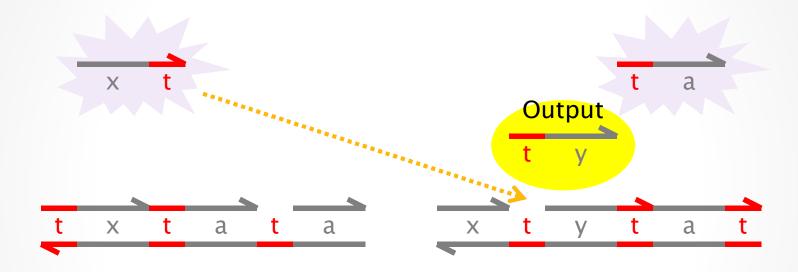


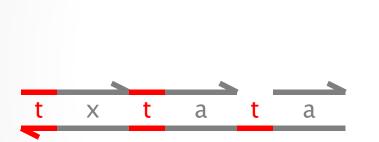
Here is our output ty signal.

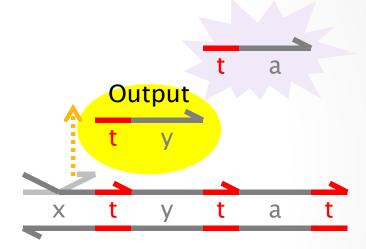
But we are not done yet:

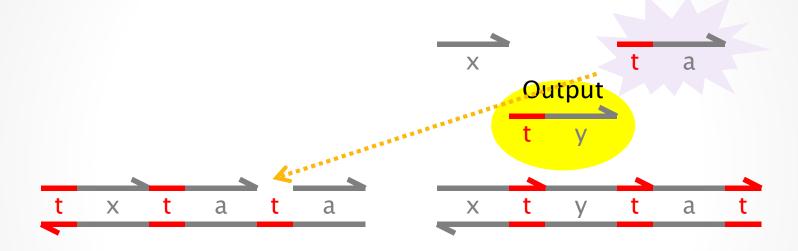
- 1) We need to make the output irreversible.
- 2) We need to remove the garbage.

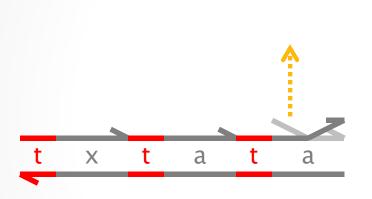
We can use (2) to achieve (1).

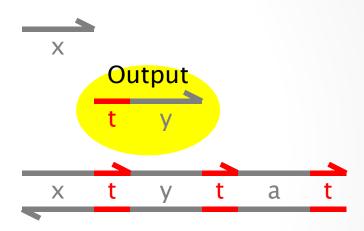


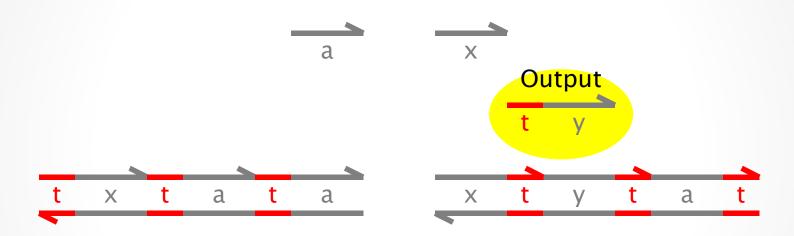


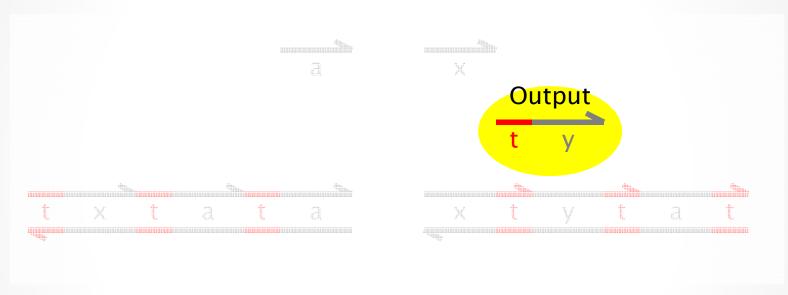






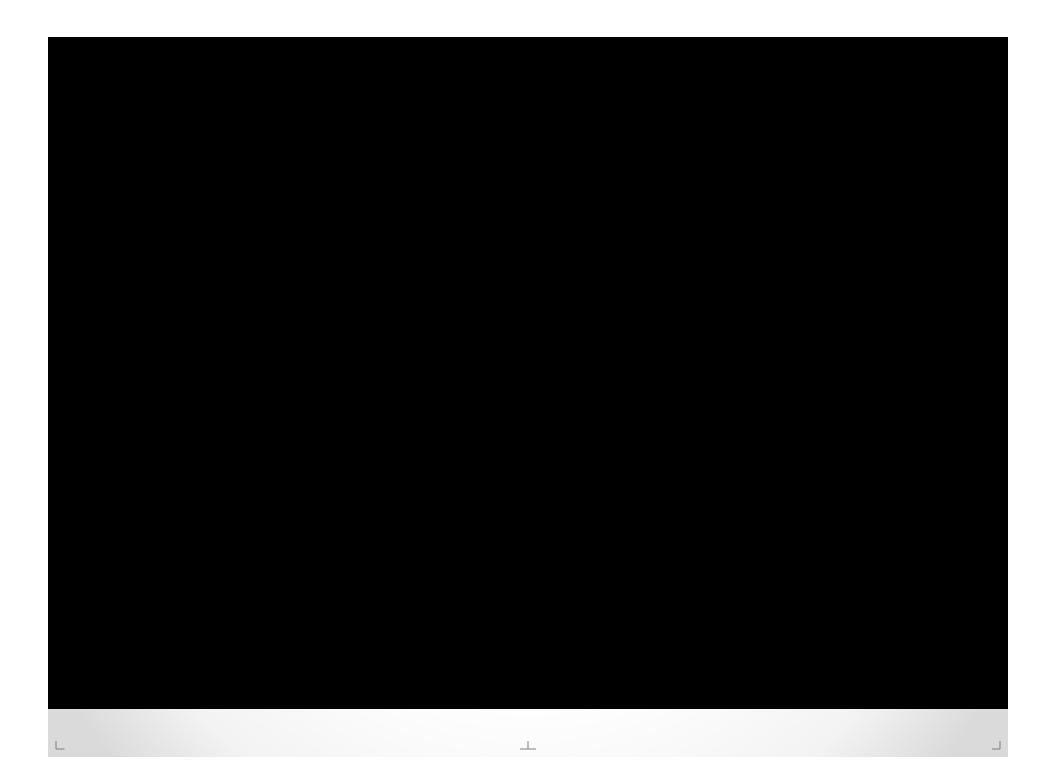


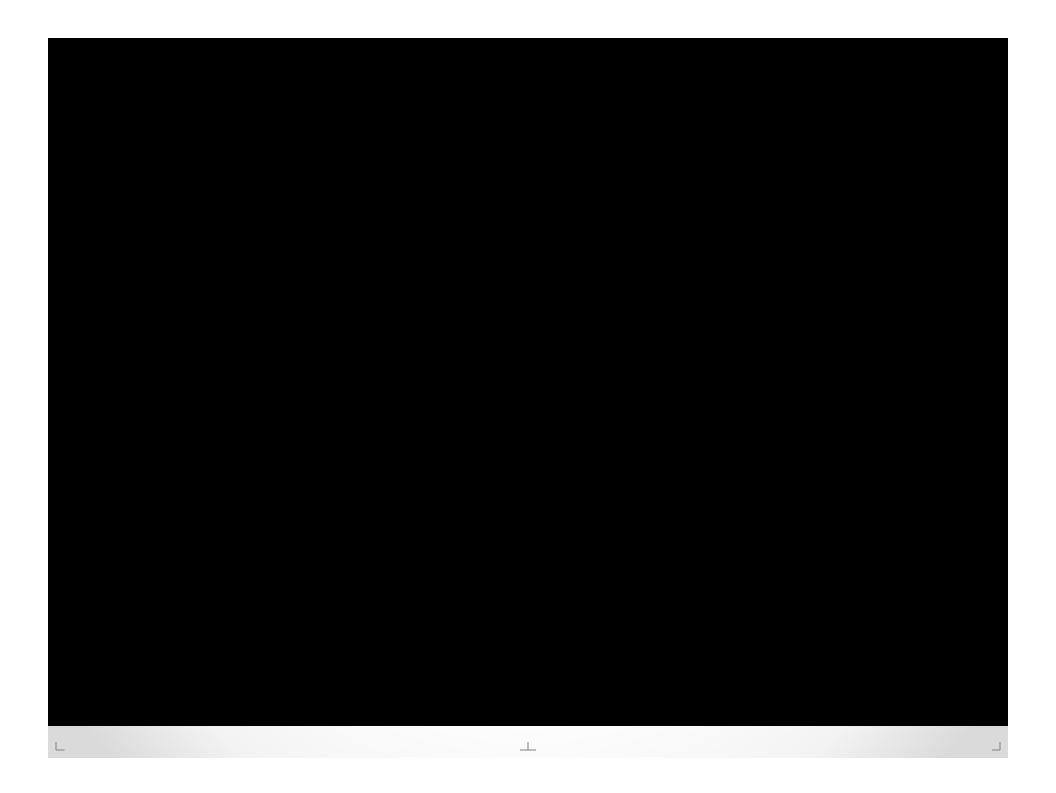




Done.

N.B. the gate is consumed: it is the energy source.





General n×m Join-Fork

- Easily generalized to 2+ inputs (with 1+ collectors).
- Easily generalized to 2+ outputs.

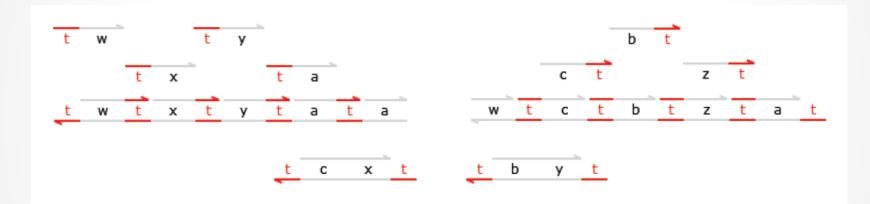
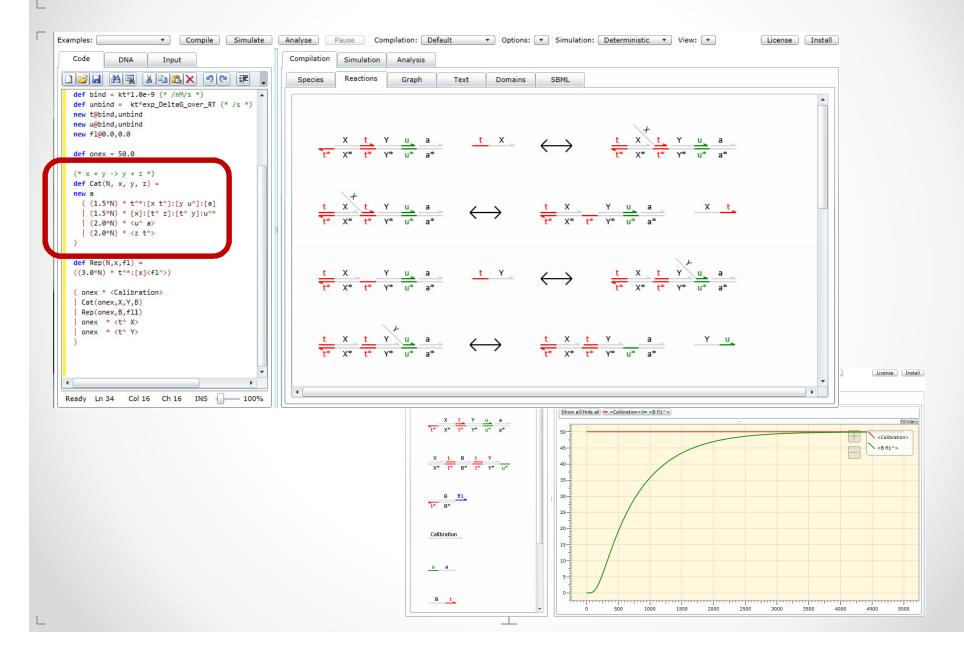
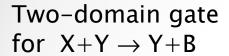


Figure 9: 3-Join $J_{wxyz} \mid tw \mid tx \mid ty \rightarrow tz$: initial state plus inputs tw, tx, ty.

DNA Programming



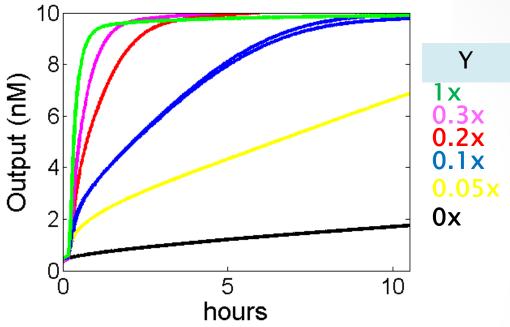
Experiments



$$X+Y \rightarrow Y+B$$

$$35C$$

$$1x = 50nM$$



Yuan-Jyue Chen and Georg Seelig U.Washingon.

	X+Y→Y+B	Concentration
LG1	X T Y U1 a	1.5x
LG2	X T B T Y X* T* B* T* Y* U1*	1.5x
input	X	1x
Catalyst		0x, 0.05x,0.1x,0.2x,0.3x,1x
~В	B T	2x
R1	U1 a	2x
B readout	B RQ ROX	3x

Verification

Verification Issues

Environment

- The nano-environment is messy (stochastic noise, failures, etc.)
- o But we should al least ensure our designs are logically correct

Verifying Components

- Reversible reactions (infinite traces)
- o Interferences (deadlocks etc.) between copies of the same gate
- o Interferences (deadlocks etc.) between copies of different gates
- Removal of active byproducts (garbage collection) is tricky

Verifying Populations

- Gates come in (large) populations
- Each population shares private domains (technologically unavoidable)
- o Correctness of populations means proofs with large state spaces

Correctness

The spec of a transducer:

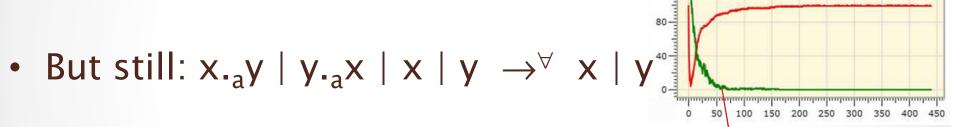
$$x.y \mid x \rightarrow y$$

- o Is it true at all?
- o Is it true possibly, necessarily, or probabilistically?
- Is it true in the context of a population of identical transducers?
- o Is it true in all possible contexts?
- o If false, does it become true for infinite populations?

Interfering Transducers

 Let a be the private transducer domain, but let's share it between x.y and y.x

Interference: x_ay | y_ax | x → ∀ x

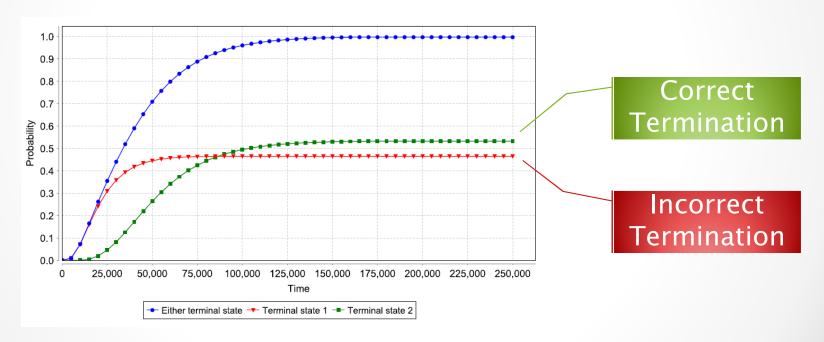


 A large population of such gates in practice does not deadlock easily. Stuck gates in a population of 200

 The wisdom of crowds: individuals can be wrong, but the population is all right.

Modelchecking DNA Systems

- Using the PRISM stochastic modelchecker
 - \circ Termination probability of interfering transducers $x \mid x_a y \mid y_a z$



L. Cardelli, M. Kwiatkowska, M. Lakin, D. Parker and A. Phillips.

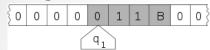
Design and Analysis of DNA Circuits using Probabilistic Model Checking.

http://gav.comlab.ox.ac.uk/papers/dna-pmc.pdf. September 2010

Conclusions

A Brief History of DNA

Turing Machine, 1936



Transistor, 1947



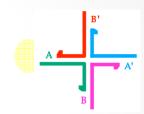
Digital Computers

Computer programming

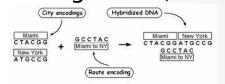
DNA, -3,800,000,000



Structural DNA, 1982



DNA Algorithm, 1994



Software

systematic manipulation of information 20th century Matterware??

systematic manipulation of matter 21th century **DNA Computers**

Molecular programming

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