Speaking the Language of Molecules

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CISBIO Open Day London 2011-07-13 http://lucacardelli.name

Outline

- Molecular Structures
 - In Science and Engineering
 - Self-assembly
- Molecular Languages
 - Natural languages: proteins, genes, membranes
 - Modeling languages (systems biology)
 - Executable languages (nano-engineering)
- Molecular Computation
 - Molecular Programming
 - Molecular Compilation

Molecular Structures

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

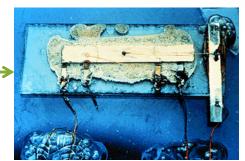
Intel&Micron, Jan. 2010. ~50atoms.

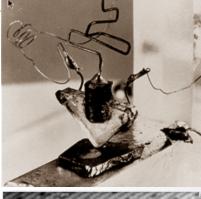
Single molecule transistor

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

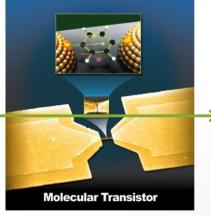
Molecules on a chip

~10 Moore's Law cycles left!





Scanning tunneling microscope image of a silicon surface showing 10nm is ~20 atoms across

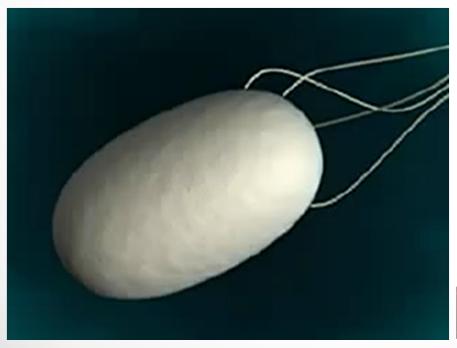




Placement and orientation of individual DNA shapes on lithographically patterned surfaces. Nature Nanotechnology 4, 557 - 561 (2009).

Building The Smallest Things

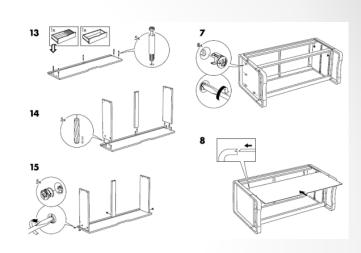
- How do we build structures that are by definition smaller than your tools?
- Basic answer: you can't. Structures (and tools) should build themselves!
- By programmed self-assembly.

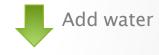


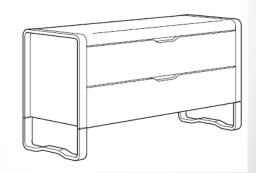


Molecular IKEA

- 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...





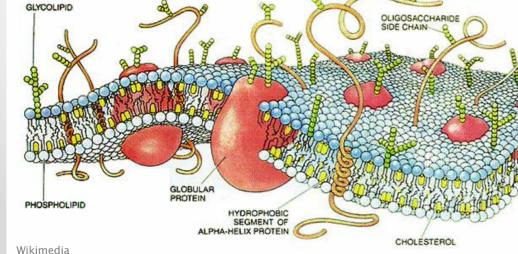


http://www.ikea.com/ms/en_US/custome r_service/assembly_instructions.html

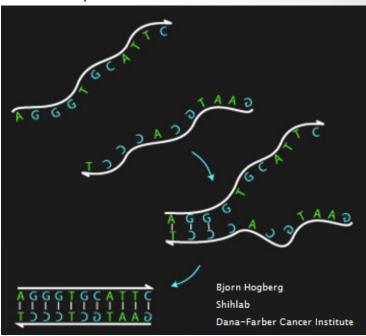
Programmed Self-Assembly

Proteins





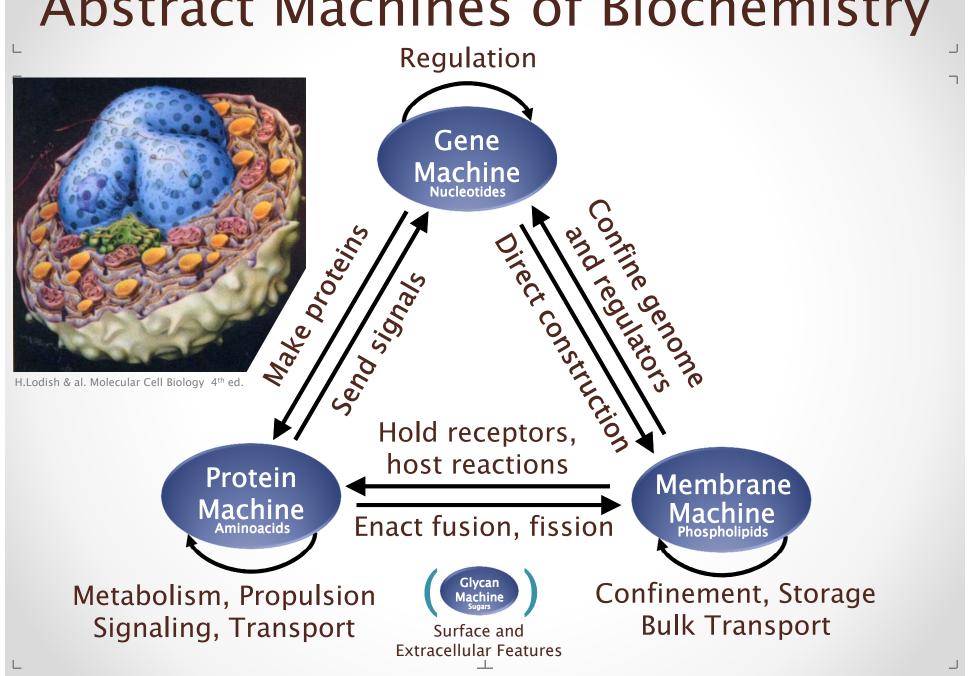
DNA/RNA



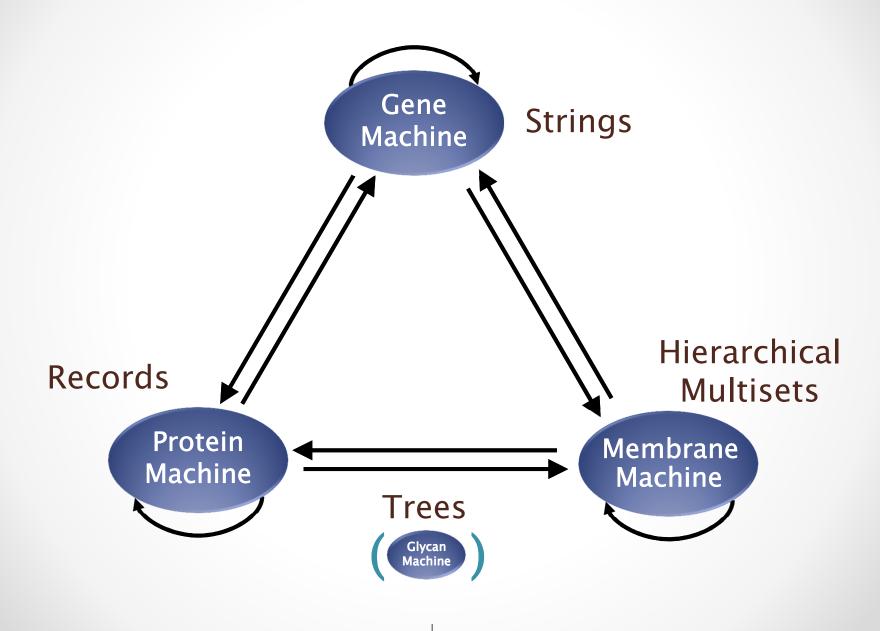
Membranes

Molecular Languages - natural languages -

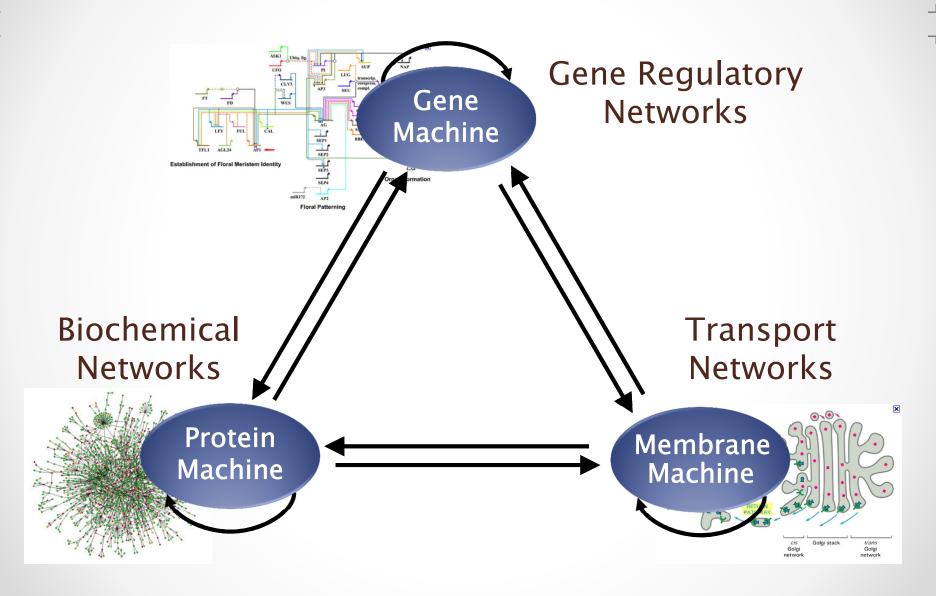
Abstract Machines of Biochemistry



Bioinformatics (Data Structures)



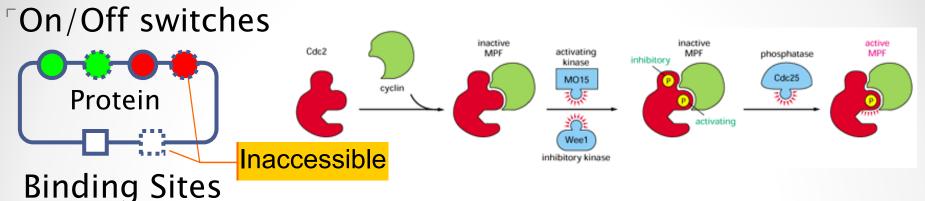
Systems Biology (Networks)



Computation (Languages) Gene Machine Protein Membrane Machine Machine

The Protein Machine

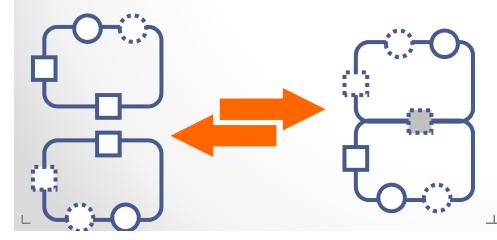
cf. BioCalculus [Kitano&Nagasaki], k-calculus [Danos&Laneve]





Switching accessible switches

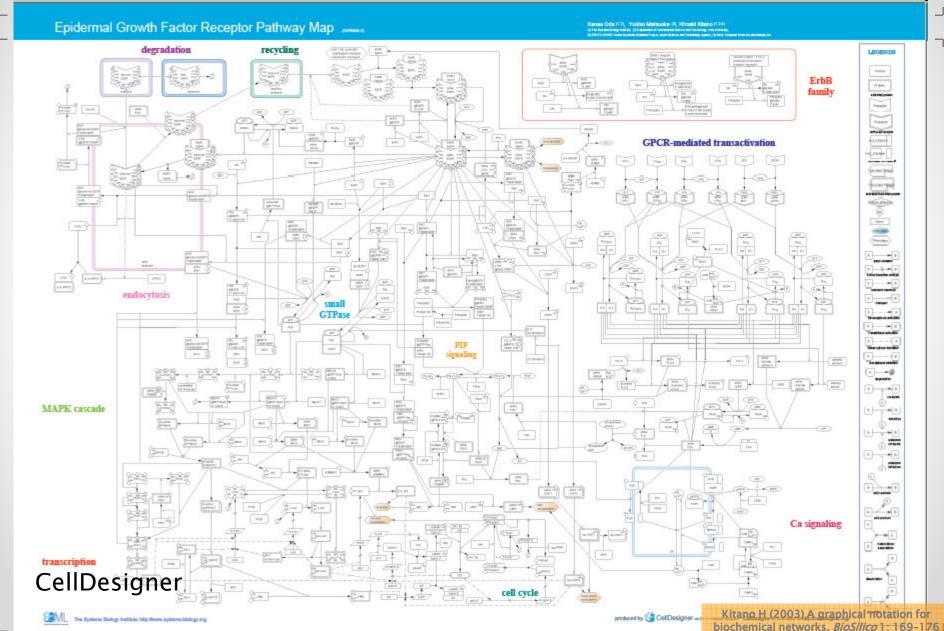
- May cause other switches and binding sites to become (in)accessible.



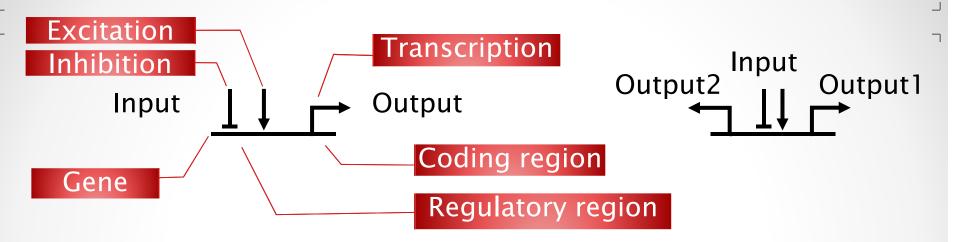
Binding accessible sites

 May cause other switches and binding sites to become (in)accessible.

Molecular Interaction Maps (Kohn/Kitano)

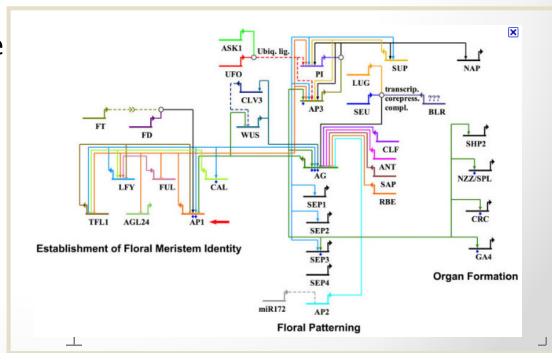


The Gene Machine

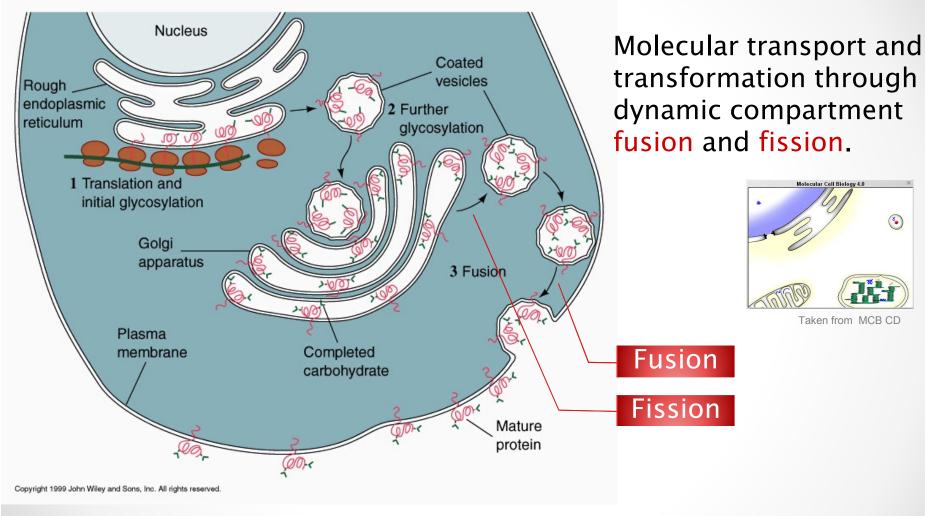


Regulation of a gene influences transcription. The regulatory region has precise DNA sequences meant for binding regulators.

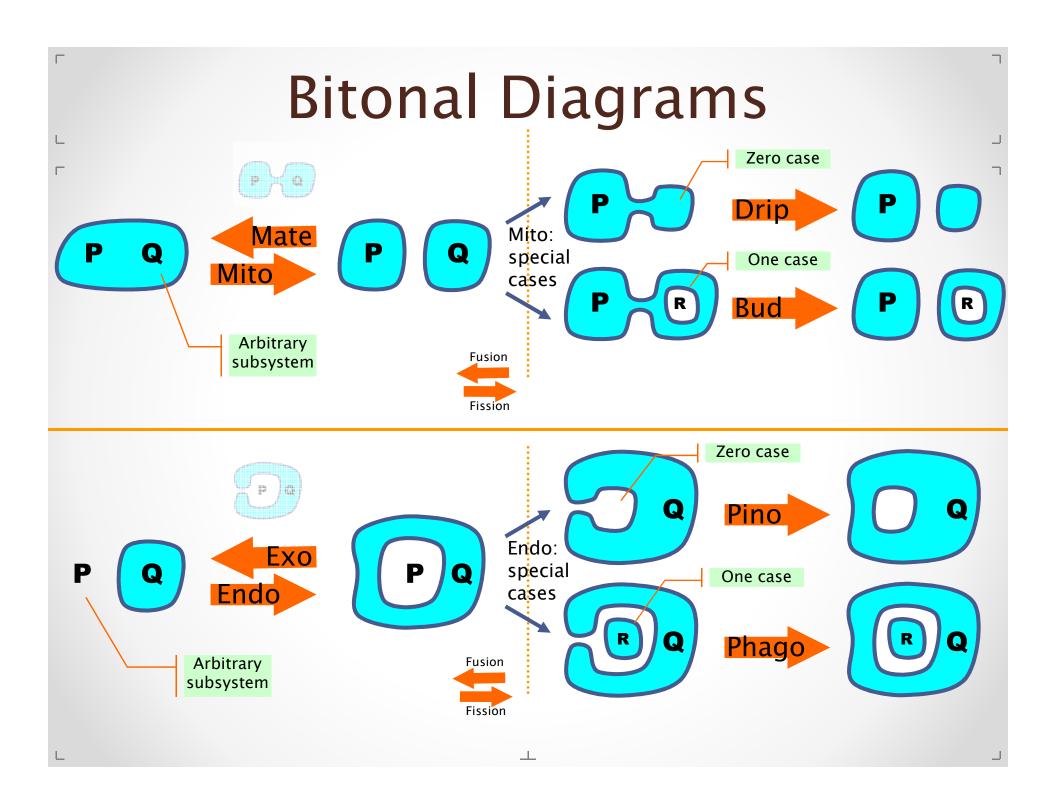
Transcription produces molecules (RNA or, through RNA, proteins) that bind to regulatory region of other genes (or that are endproducts).



The Membrane Machine



Voet, Voet & Pratt Fundamentals of Biochemistry Wiley 1999. Ch10 Fig 10-22.



Molecular Languages - modeling languages -

From Instructions to Programs

- We have seen the instruction sets:
 - Proteins complexation, phosphorilation
 - Genes activation, inhibition
 - Membranes fusion, fission
- How do we combine them into programs?
 - I.e., into models (quantitative programs)
- How do we study their semantics?
 - I.e., their kinetics (quantitative semantics)

Chemistry

Chemical reactions

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

(a program)

Ordinary Differential Equations

```
\circ d[A]/dt = -r[A][B] ... (a semantics)
```

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

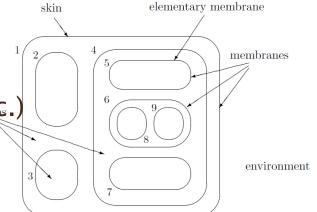
- Synchronous Boolean networks
 - · Stewart Kauffman, etc.
- Asynchronous Boolean networks
 - · René Thomas, etc.

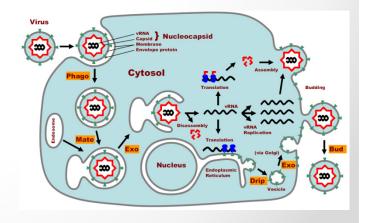
Protein Networks

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

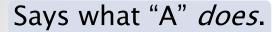
Membrane Networks

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





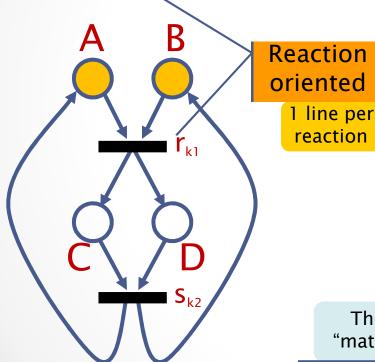
Reactions vs. Reagents



$$r: A + B \rightarrow_{k1} C + D$$

s:
$$C + D \rightarrow_{k2} A + B$$

Does A C or D?



become

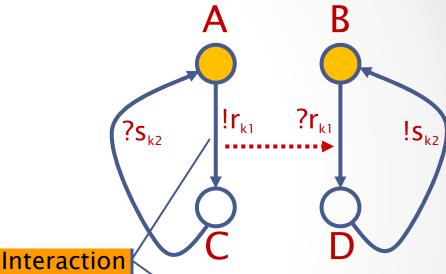
oriented

1 line per agent

The same "math model"

CTMC

Says what "A" is.



 $A = !r_{k1}; C$

 $C = ?s_{k2}; A$

 $B = ?r_{k1}; D$

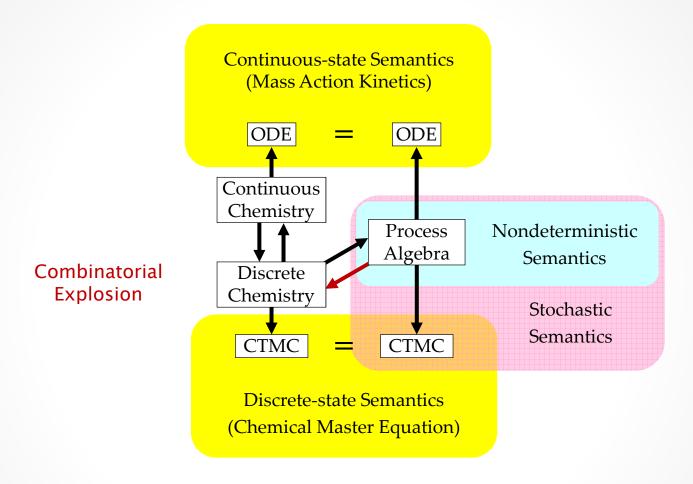
 $D = !s_{k2}; B$

becomes C not D!

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?

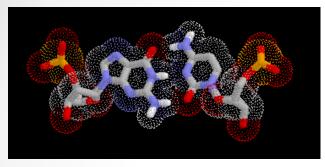
- Chemistry is not easily executable
 - Please Mr Chemist, execute me these reactions that I just made up
- Similarly, the molecular languages seen so fare are descriptive (modeling) languages

 How can we actually execute molecular languages? With real molecules?

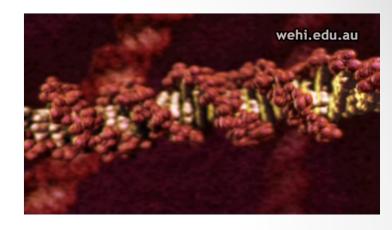
Molecular Languages

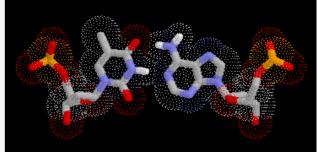
- executable languages -

DNA



GC Base Pair Guanine-Cytosine

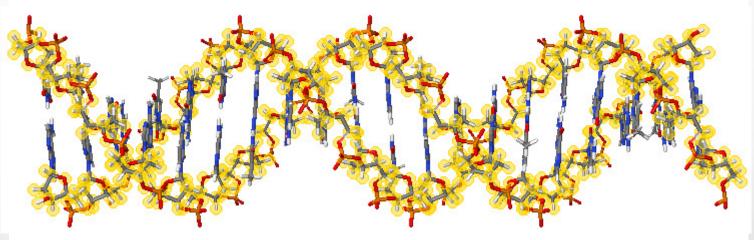




TA Base Pair Thymine-Adenine

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

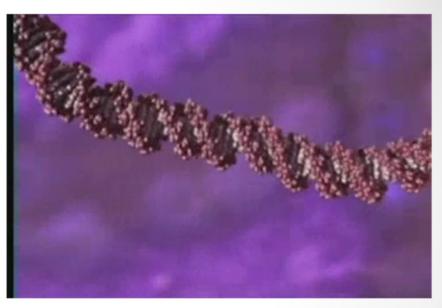
- Every time a cell replicates it has to copy 2 meters of DNA reliably.
- To get a feeling for the scale disparity, compute:

DNA in human body

- o 10 trillion cells
- 133 Astronomical Units long
- 7.5 OctaBytes

DNA in human population

20 million light years long



DNA wrapping into chromosomes

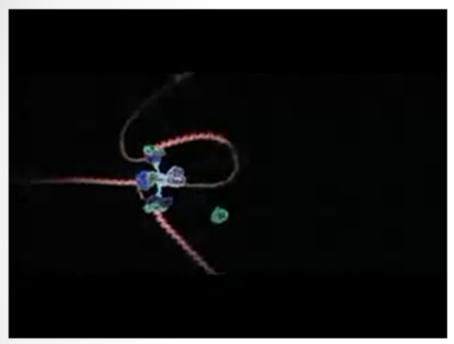
wehi.edu.au



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

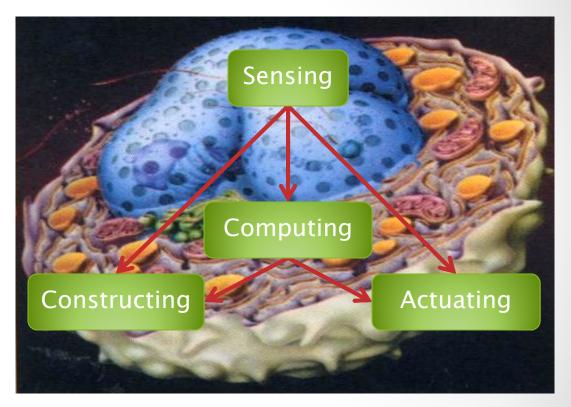
Constructing

- Chassis
- Growth

Computing

- Signal Processing
- Decision Making

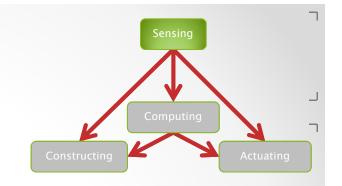
Nanoscale Control Systems

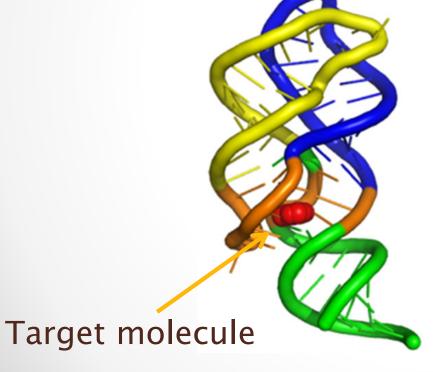


Nucleic Acids can do all this. And interface to biology.

Sensing

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



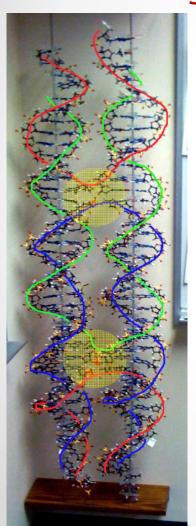


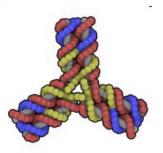
Adenine riboswitch aptamer

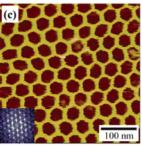
Structural basis for discriminative regulation of gene expression by adenine- and guanine-sensing mRNAs. Chem Biol. 2004 Dec;11(12):1729-41.

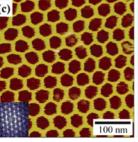
Constructing

Crosslinking









Chengde Mao, Purdue

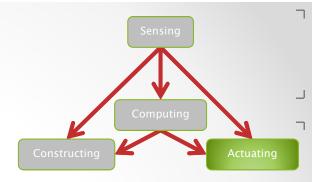
Andrew Turberfield, Oxford

Folding DNA into Twisted and Curved Nanoscale Shapes

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



Actuating

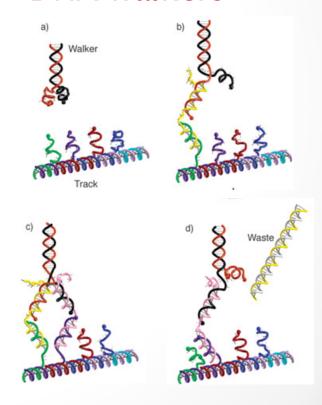


DNA tweezers

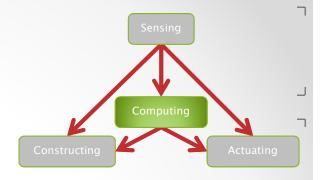


Bernard Yurke, Boise State

DNA walkers



Computing

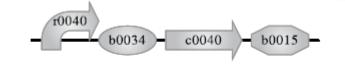


- Sensors and Actuators at the 'edge' of the system
 - They can use disparate technologies and phenomena
- Computation in the 'kernel' of the system
- Compositionality in the kernel
 - The components should use uniform inputs and outputs
 - The components should be 'computationally complete'

"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

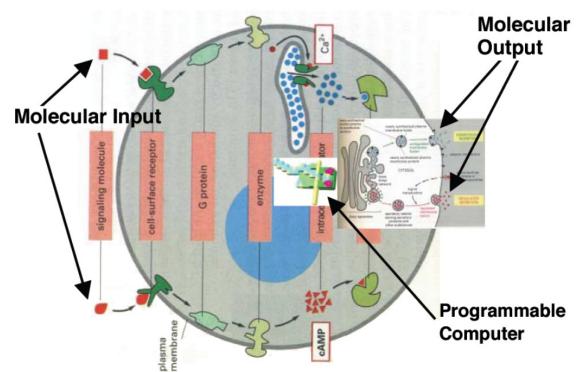
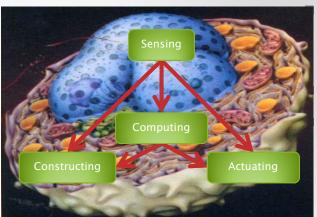


Fig. 1 Medicine in 2050: "Doctor in a Cell"

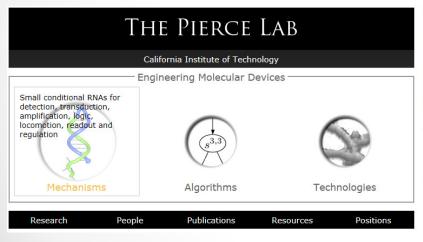
Ehud Shapiro

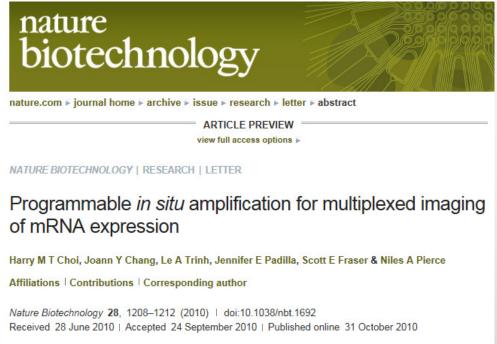
Rivka Adar Kobi Benenson Gregory Linshitz Aviv Regev William Silverman Molecules and computation



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.
- In fact, no material of biological origin.

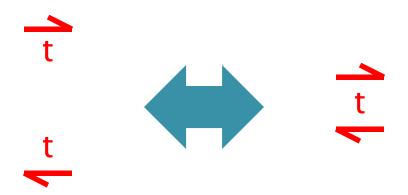
Domains

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



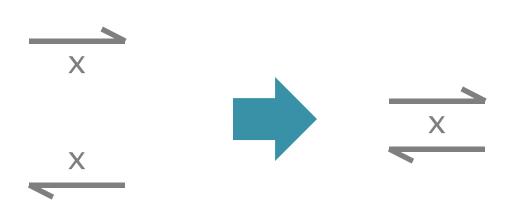
- I.e., differently named domains must not hybridize:
 - With each other
 - With each other's complement
 - With subsequences of each other
 - With concatenations of other domains (or their complements)
 - o Etc.
- Choosing domains (subsequences) that are suitably independent is a tricky issue that is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

Short Domains

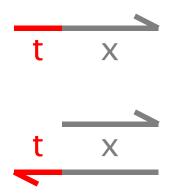


Reversible Hybridization

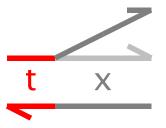
Long Domains



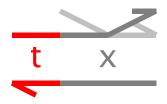
Irreversible Hybridization



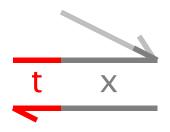
"Toehold Mediated"



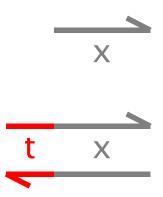
Toehold Binding



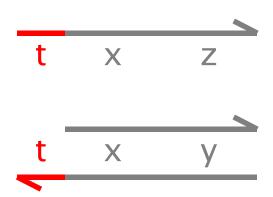
Branch Migration

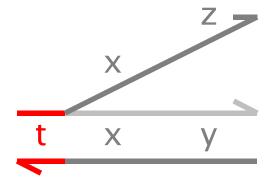


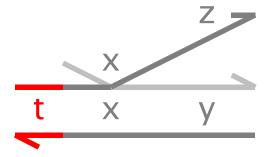
Displacement

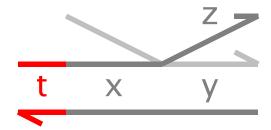


Irreversible release









Cannot proceed Hence will undo

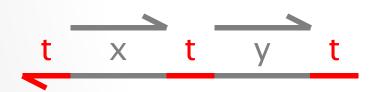
Two-Domain Architecture

Signals: 1 toehold + 1 recognition region



Garbage collection "built into" the gates

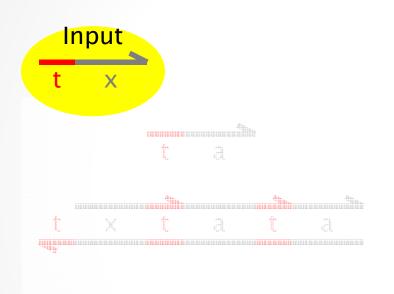
 Gates: "top-nicked double strands" (or equivalently double strands with open toeholds)

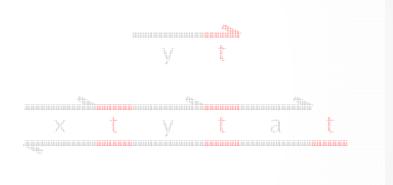


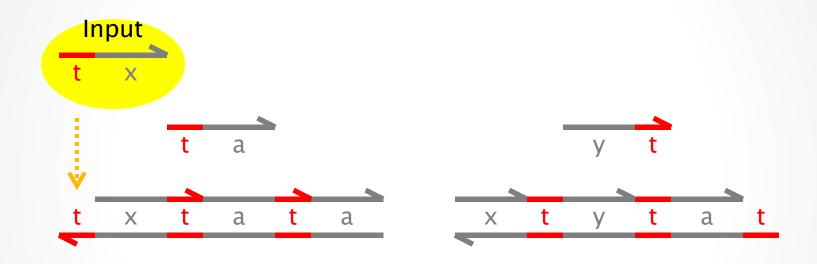
Two-Domain DNA Strand Displacement

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.

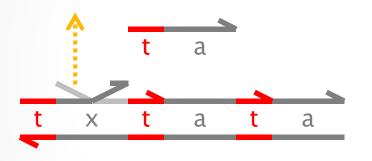


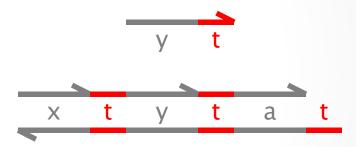


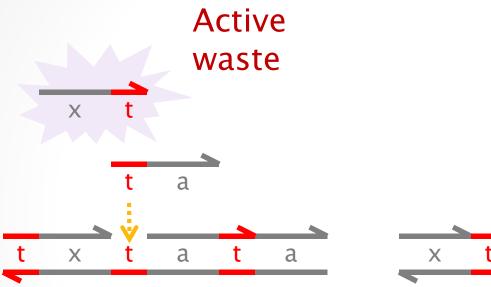


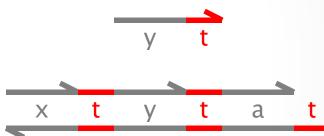
Built by self-assembly!

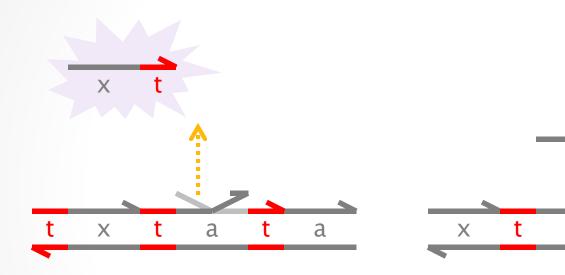
ta is a *private* signal (a different 'a' for each xy pair)

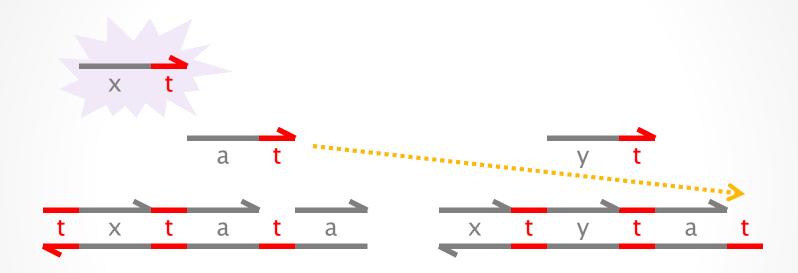




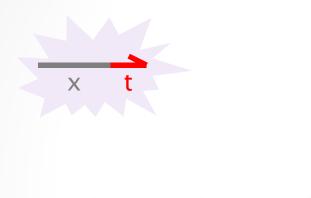


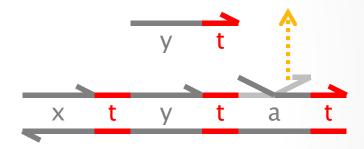


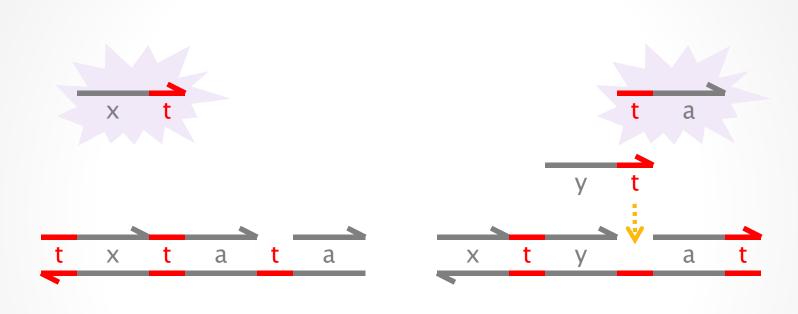


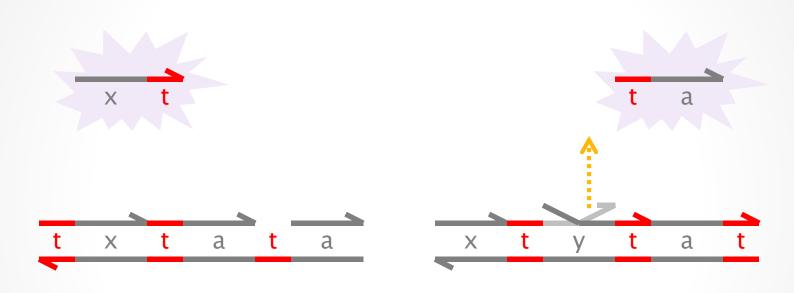


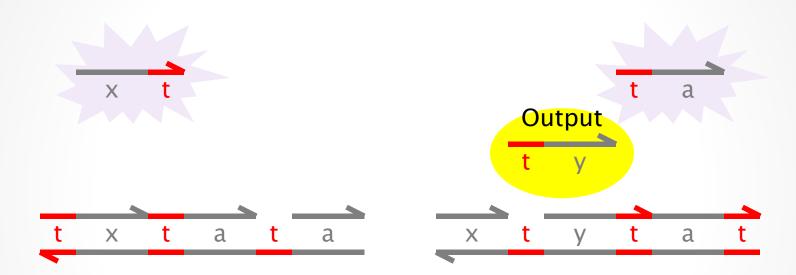
So far, a tx *signal* has produced an at *cosignal*. But we want signals as output, not cosignals.









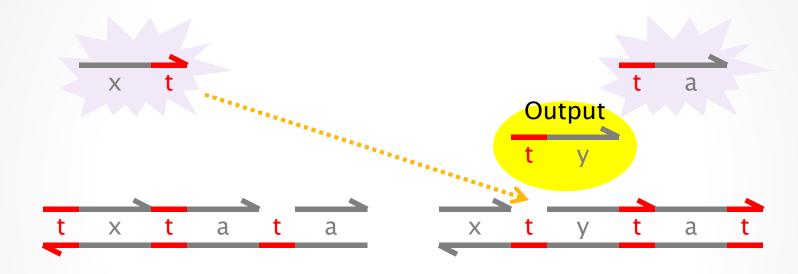


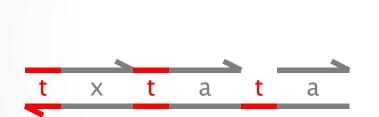
Here is our output ty signal.

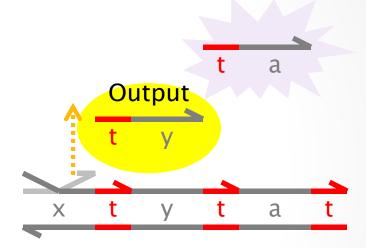
But we are not done yet:

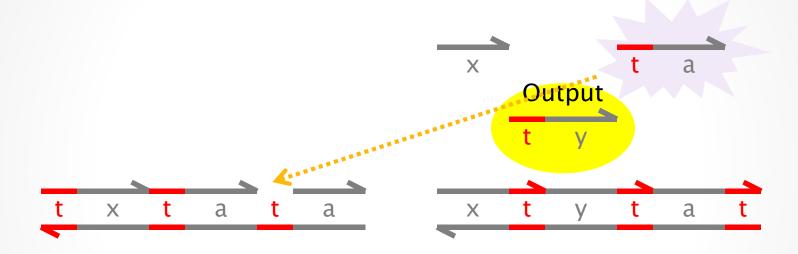
- 1) We need to make the output irreversible.
- 2) We need to remove the garbage.
- We can use (2) to achieve (1).

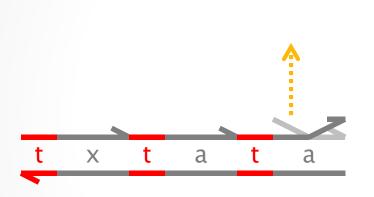
1

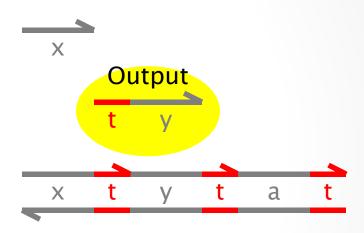


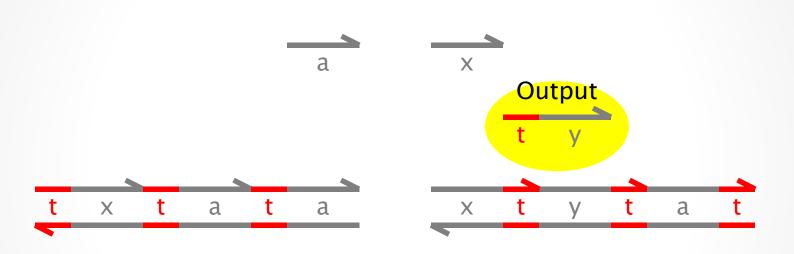


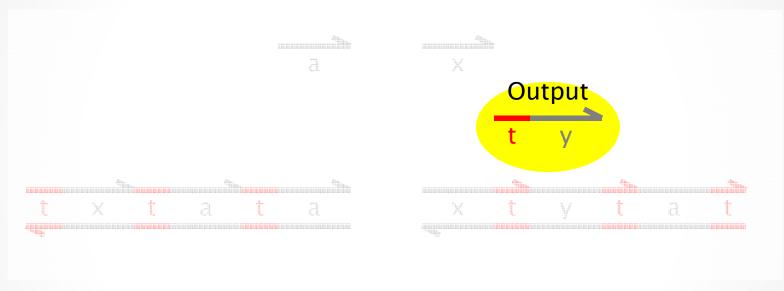






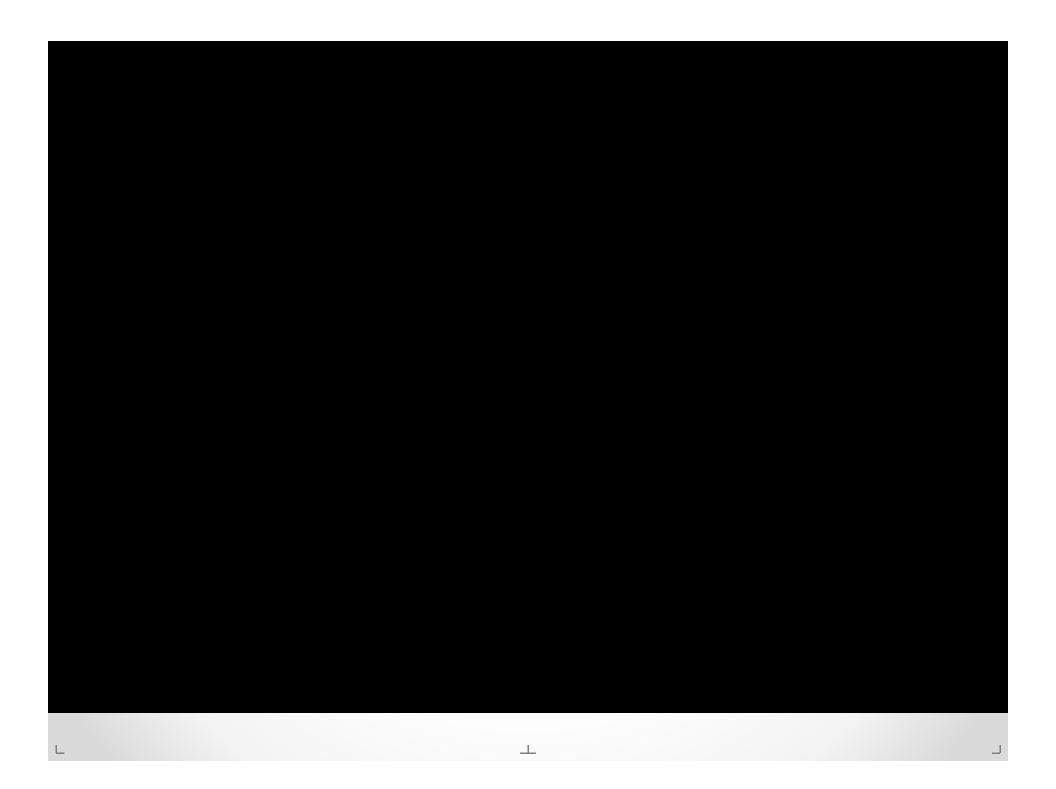






Done.

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.

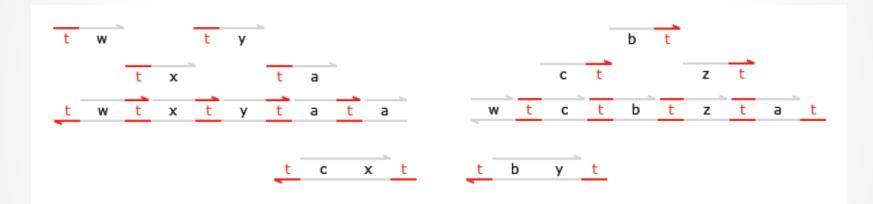
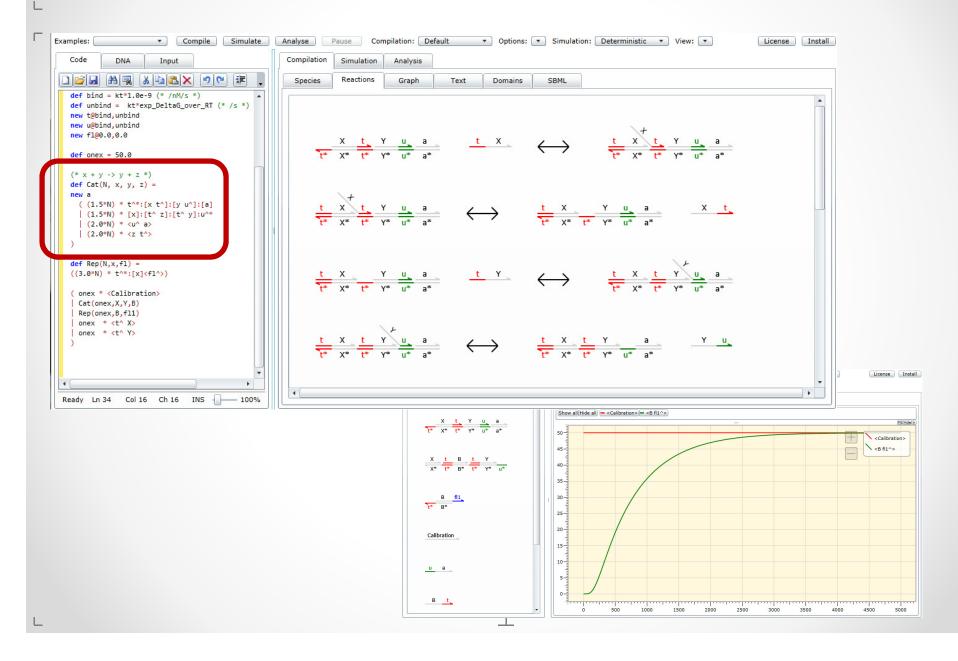
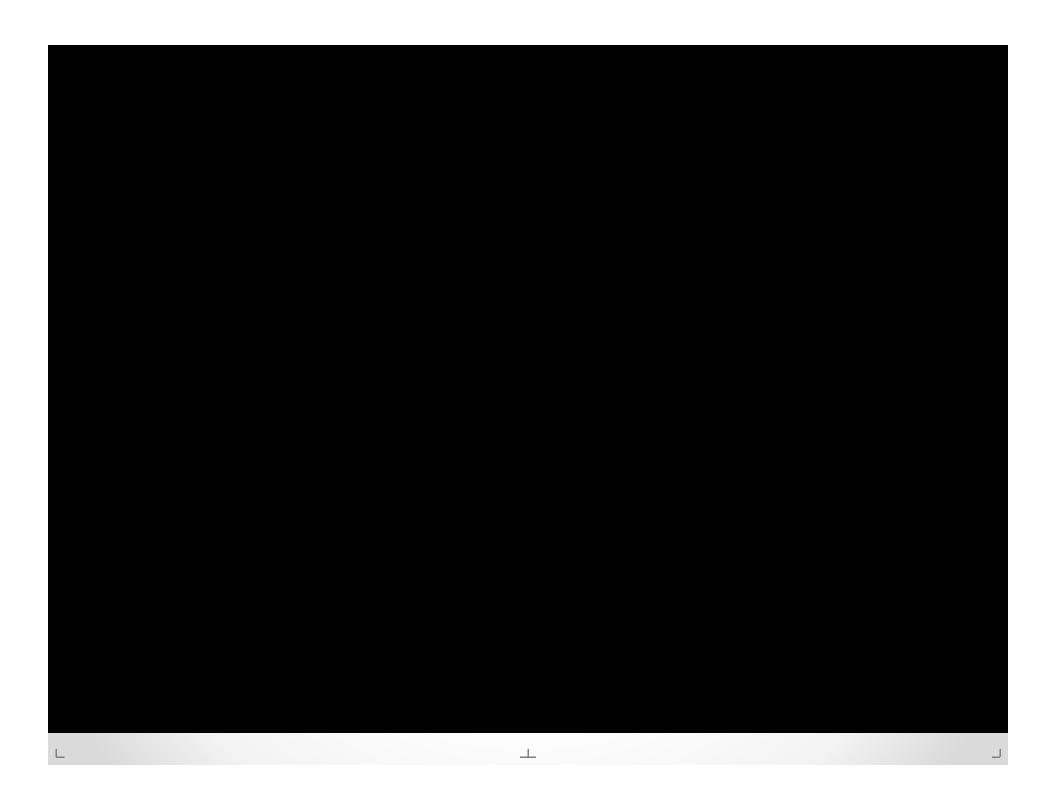


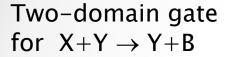
Figure 9: 3-Join $J_{wxyz} \mid tw \mid tx \mid ty \rightarrow tz$: initial state plus inputs tw, tx, ty.

DNA Programming



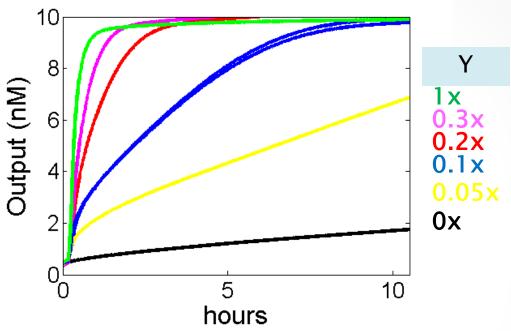


Experiments



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

 $35C$
 $1x = 50nM$

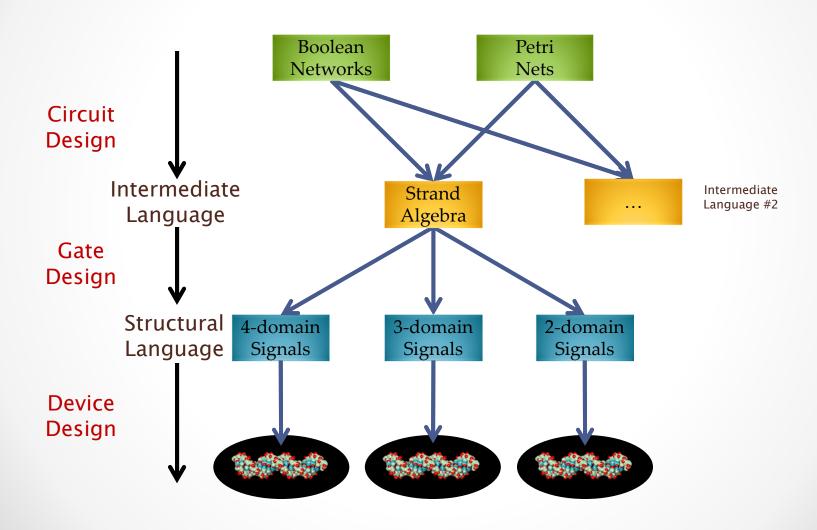


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	X+Y→Y+B	Concentration
LG1	X T Y U1 a	1.5x
LG2	X T B T Y X* T* B* T* Y* U1*	1.5x
input	X	1x
Catalyst		0x, 0.05x,0.1x,0.2x,0.3x,1x
~B	B T	2x
R1	U1 a	2x
B readout	B RQ ROX	3x

Molecular Programming

Molecular Compilation



Conclusions

Summary

Molecular Structures

 Hard to build... but they can build themselves under program control!

Molecular Languages

- Natural and unnatural
- Concurrent, quantitative

Molecular Compilation

Molecular architectures, verification, optimization

Molecular Programming

o In silico, in vitro, in vivo...

Acknowledgments

- Microsoft Research
 - Andrew Phillips
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- U.Washington
 - Seelig Lab