

Outline

- The Language of Science
- The Language of Functions
- The Language of Objects
- The Language of Molecules

The Language of Science

The Language of Science

- Never much good at natural languages.
 Latin: bad. English: worse.
- Middle-school remedial lessons
 - English is the Language of Shakespeare Science!
 - Learned by Science Fiction, from Asimov to Zelazny.
- In Edinburgh, only two English speakers
 - Robin (textbook, slow English)
 - Lectures on Operational Semantics
 - Dave (American English has 5 vowels instead of 1)
 - Lectures on Hope
 - Corollary
 - Famous episode involving a Scottish milkman and 3 months worth of empty milk bottles stacked all over Gordon Plotkin's flat.

The Language of Functions

The Language of Functions

- Ironically attracted to artificial languages
 - ο Pisa: λ -calculus (+ Fortran, Lisp, Algol68, Simula)
 - Edinburgh: ML & Hope (+ Pascal, VAX assembly)
 - Murray Hill: C and Unix (during a snow storm at Dave's)

And their semantics

- Pisa: Scott–Strachey semantics.
- Edinburgh: CPOs.
- Murray Hill: the MacQueen-Sethi-Plotkin ideal model
 - More on this in a moment...

The Language of Objects

The Language of Objects

- The Practice of Objects Record
 - Pisa: Simula was my favorite language.: (t_{ei}: A_{ei}) ⇔ A_i: A_i
 - o Edinburgh: Added records and variants to ML. Tried to add record subtyping by type inference, but gave up.
 - Pisa again: Galileo, an ML-insipired database programming language with records and variants and Simula-inspired Vaziant subtyping.

[[t=a]]: [[t::A:] 👄 a:A A BK.A=AK

Where was the Theory of Objects VIIt:: A.J V It: Ail

- Logic languages: --> Predicate logic
- $\mathbf{I} t \mathbf{I} = \mathbf{I} t = \mathbf{O} \mathbf{I}$ Database languages: --> Relational calculus $[1, ..., t_i, ...] = [1, ..., t_{i-1}, ...]$ Ο
- Functional languages: $--> \lambda$ -calculus 0
- Imperative languages: --> Hoare Logic / Weakest Preconditions 0
- Modular languages: --> Algebraic semantics 0
- Object-oriented languages: --> ??? 0

Inspiration!

Type Operators

The Ideal Model

- In early Scott denotational models, types were "retracts" of the universal value set, which did not support subtyping.
- The Ideal Model was designed as a semantics for polymorphism, which was modeled as a "big ... o (a, b,) intersection" of domains.
- So, it accidentally provided a subset-based
 denotational semantics of subtyping (via non-empty intersections between domains)
 tuest, subset, su
- Therefore enabling:
 - records as functions (well-known lisp=hack)= folk, crew order. 3
 - record types as domains (label-dependent function types)
 - record subtyping as set inclusion of function spaces

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3 Hotave, VHOLO.ce. }

Result

• Paper:

Luca Cardelli, A Semantics of Multiple Inheritance, in Semantics of Data Types, Lecture Notes in Computer Science 173, 1984. Also to appear in Information and Control.

A Semantics of Multiple Inheritance

Luca Cardelli

AT&T Bell Laboratories Murray Hill, New Jersey 07974

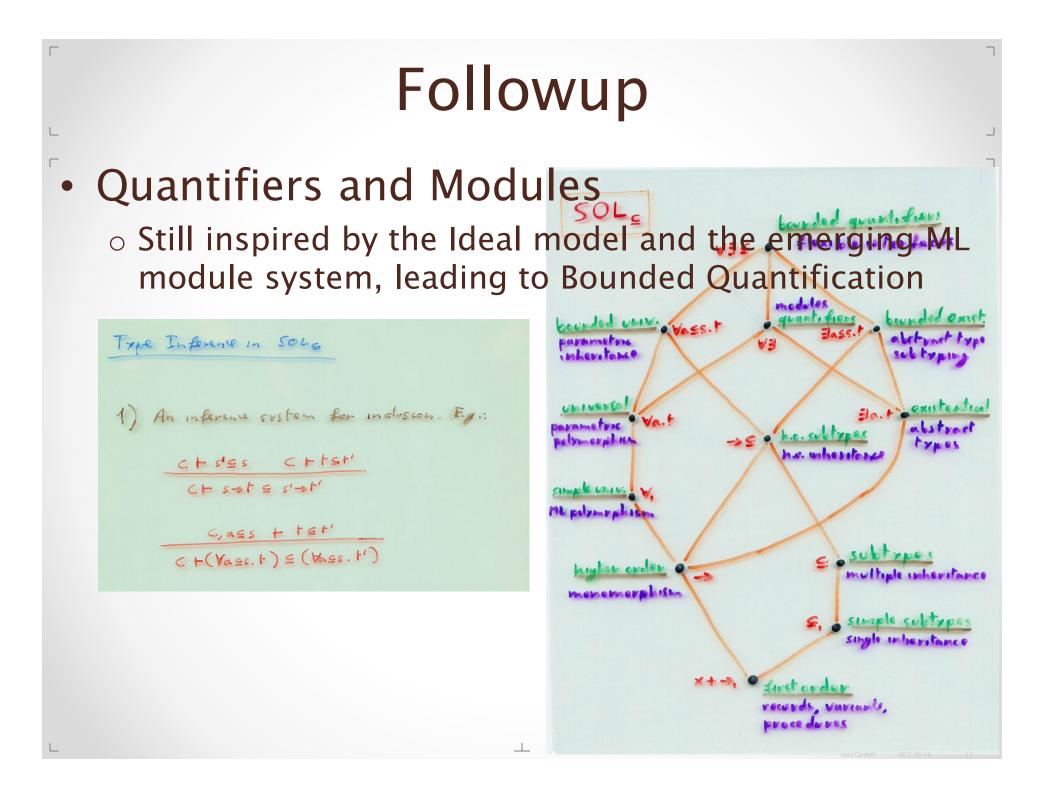
1. Introduction

There are two major ways of structuring data in programming languages. The first and common one, used for example in Pascal, can be said to derive from standard branches of mathematics. Data is organized as cartesian products (i.e. record types), disjoint sums (i.e. unions or variant types) and function spaces (i.e. functions and procedures).

The second method can be said to derive from biology and taxonomy. Data is organized in a hierarchy of classes and subclasses, and data at any level of the hierarchy inherits all the attributes of data higher up in the hierarchy. The top level of this hierarchy is usually called the class of all "objects"; every datum is an object and every datum inherits the basic properties of objects, like the ability to tell whether two objects are the same or not. Functions and procedures are also considered as local actions of objects, as opposed to global operations. 1984 Semantics of Data Types Talk:
 ... delivered by Dave

of Multiple Inheritance

Luca Cardelli



One Joint Paper

Persistence and Type Abstraction

Luca Cardelli David MacQueen

AT&T Bell Laboratories Murray Hill, NJ 07974

Introduction

Abstract types are a well known and effective way of structuring programs. The basic idea of information hiding can however conflict with the need to store data for long periods of time, ar make it accessible to different activities. In particular a typechecker must be able to recognize the occurrence of the *same* abstract type during different activations, and must enforce the privacy data representations.

To achieve this, the persistent storage of data must preserve type information, and mu respect type abstraction. The use of type abstractions in the presence of persistent storage require that abstract types be made persistent as well. Under these conditions, we can preserve typ security across distinct activations of the typechecker.

The following is a brief account of how various models of abstraction and persistence interact. We start by sketching a simple polymorphic language and its types and showing various ways of modeling type abstraction in such a language. We then discuss some basic notion underlying persistent storage of typed objects, such as the intern and extern primitives and the special type dynamic, and describe three persistence strategies. Finally we discuss the particular problem of persistent abstract types.

Values and Types

We will base our discussion on a simple polymorphic language in the tradition of ML [Miln 84] and Amber [Cardelli 84]. The simplified language we have in mind is closely related to th language SOL [Mitchell and Plotkin 85], variants of which are described in [Reynolds 85] an [Cardelli and Wegner 85].

The basis of this language is a slightly sugared applied lambda calculus that is adequate for expressing certain kinds of values. Type annotations are added to the basic expressions in such way that one calp statically determine a type for each expression. This type is a structure

[PDF] Persistence and type abstraction

L Cardelli... - Proceedings of the **Persistence** and ..., 1985 - lucacardelli.name Page 1. 231 **Persistence** and Type **Abstraction** Luca Cardelli David MacQueen AT&T Bell Laboratories Murray Hill, NJ ... The basic ideas of information hiding can however conflict with the Coll to Carde **data** for long periods of time, and make it accessible to different activities. ... <u>Cited by 34 - kelated articles - All 8 versions</u>

Archive

1982-03 "Basic Definitions and Facts" (?) 48 pages

| 1 | 3/26/22-1 |
|----|---|
| 12 | Davie Definitions and Facts |
| 4 | $\begin{array}{llllllllllllllllllllllllllllllllllll$ |
| 4 | Definition . Decore I is directed closed if for any directed $X \in I$, $\sqcup_D X \circ I$. |
| | Jemma 1. Da epo, IED directed cloud => < I, EI> in |
| 17 | a cpo, where EI is Ep reducted to I. |
| | Proo : Let X & I be a Ex- directed set. Then |
| | X is also Ep- directed (notion of directed set |
| | relatinger). Thus UNX exists and UNX & I mice |
| | I is directed-cloud. |
| | Claim: Wox is the Ez-least upper bound of X. |
| | x E D x for any xex => |
| | x = Up X for any XEX because x, Up XEI. |
| | ". Wox us an I upper bound . Furthermore, |
| | if us I & XEIU, then XEDU, to WoXEDU |
| | and \therefore $\square_{D} X \equiv_{I} \omega$, \therefore $\square_{D} X = \square_{I} X$. |
| | |
| | Corollary 2. D a cho, IED directed cloud => |
| | for any directed XEI, $\Box_{D}X = \Box_{I}X$. |
| | Mote : Boing directed complete impleir the excitence |
| | of a deast element \bot , since $X = \varphi$ is a directed set |
| | and is must be a least element. more portiendorly |
| | LI= LD since LD= WD\$ CI. |

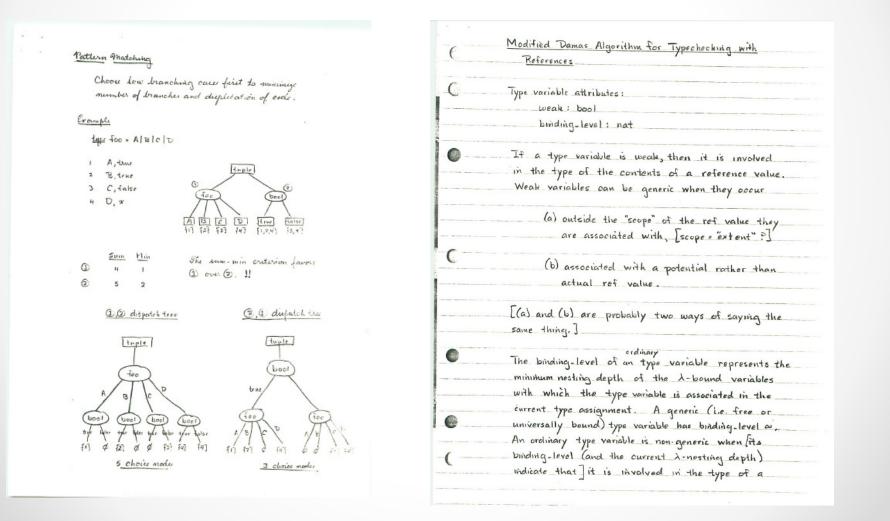
1982-04 "Semantics of Data Types"

Semantics of Data Types 4/13/82-1 φοy≡ id_{D→D} yoq∃id_p DĴD⇒D Deln: CED is a cloure iff CEC= c and idgec. C = {ceD | c is a cloure}. Dely: Let close : (D->D) -> (D->D) be quien by close (f) = ∐ (Fuid)". and IED by $\Gamma = \underline{\lambda} x. \mathcal{V}(close(g(x))).$ Lomma O. I is (4 of) a continuous function. Lemma 1. $\forall f \in D$. $\Gamma(f) \equiv f$ and $\Gamma(f) \equiv id_p$. Lumma 2, ce C => r(c)= c. Lamma 3. VfeD. I(+) eC. Prop 4. C = fiepts (q(I)) (where finiple (+) is the set of all fixed points of +). Con 5. C is a domain (i.e. a complete lattice). [Farshi]. Brop G. I is a cloure. (IEC). Con 7, C = ran(I).

Archive

1983-?? "Pattern Matching"

1983-03 "Modified Damas Algorithm for Typechecking with References"



Archive

1984-05 "Algorithm for Compiling Pattern Matching"

5/23/84 -1

Algorithm for compiling pattern matching, (a match) We show with a list of ruly, each of which is a pair considing of a briding pattern and a body expression.

rule := pat : pattern , body : expr.

The goal is to construct a decision or decisionination true that imities the matching of a pattern against an argument wake to select a particular rule body to evaluate (in an environment given by briding the transition in the corresponding pattern to the appropriate argument components).

There are two functions to be performed.

(1) discrimination of call

(2) brinding of pattern variables

The will deal with ally with (1) ,

The basic data structure is a block, which is a sectangular array of pattern elements on items arranged in now and columns. Each now is associated with a particular hule, and there is at most one now for a given rule in a block. Each column contains pattern subterms occurring at the same portion in a common pattern scheme

(The one I failed to implement in my ML compiler.)

End of Story

→ California

The Language of Distributed Objects

• Leading to ...

• \rightarrow England

The Language of Mobile Processes

• Leading to ...

The Language of Biological Processes

• Leading to ...

The Language of Molecules

Molecular Programming 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/phosphorylation) context.
 - Similar to informal descriptions of biochemical events ("narratives").
- Different levels of representation efficiency
 - 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.

But what about Execution?

- Chemistry is not easily executable
 - Please Mr Chemist, execute me these reactions that I just made up.

Description

 Molecular languages used in systems biology are descriptive (modeling) languages

Compilation

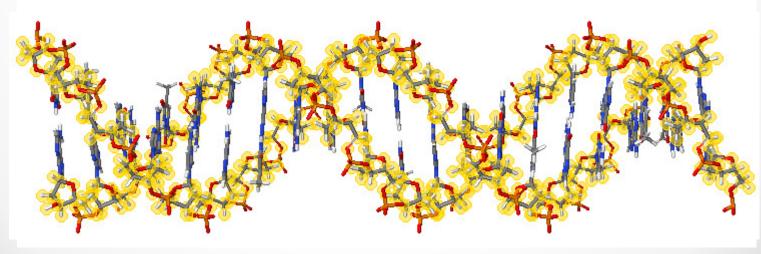
• How can we compile *arbitrary* molecular programs?

Execution

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

DNA as an Engineering Material

- This is why DNA/RNA is important: it is programmable matter.
- Not the only one, in principle, but the only one for which we have a well-developed manufacturing technology.



Sequence of Base Pairs (GACT alphabet)

Molecular Control Systems

Sensing

Reacting to forcesBinding to molecules

Actuating

Releasing moleculesProducing forces

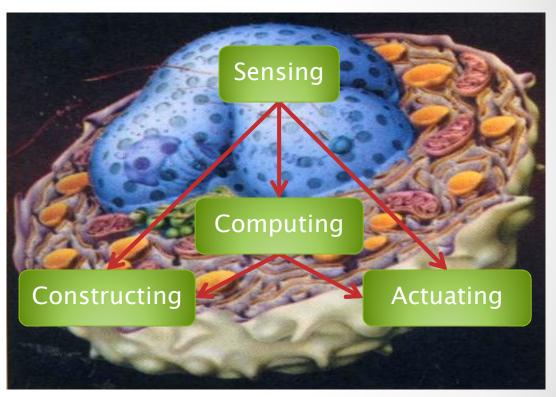
Constructing

- o Chassis
- o Growth

Computing

- Signal Processing
- Decision Making

Control Systems



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

"Embedded" DNA Computing

(Synthetic Biology)

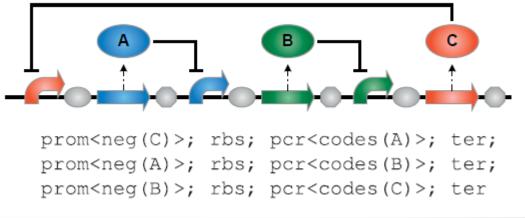
- Using bacterial machinery (e.g.) as the hardware. Using embedded gene networks as the software.
- MIT Registry of Standard Biological Parts
- GenoCAD

• Meaningful sequences [Cai et al.]



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

- GEC
 - [Pedersen & Phillips]



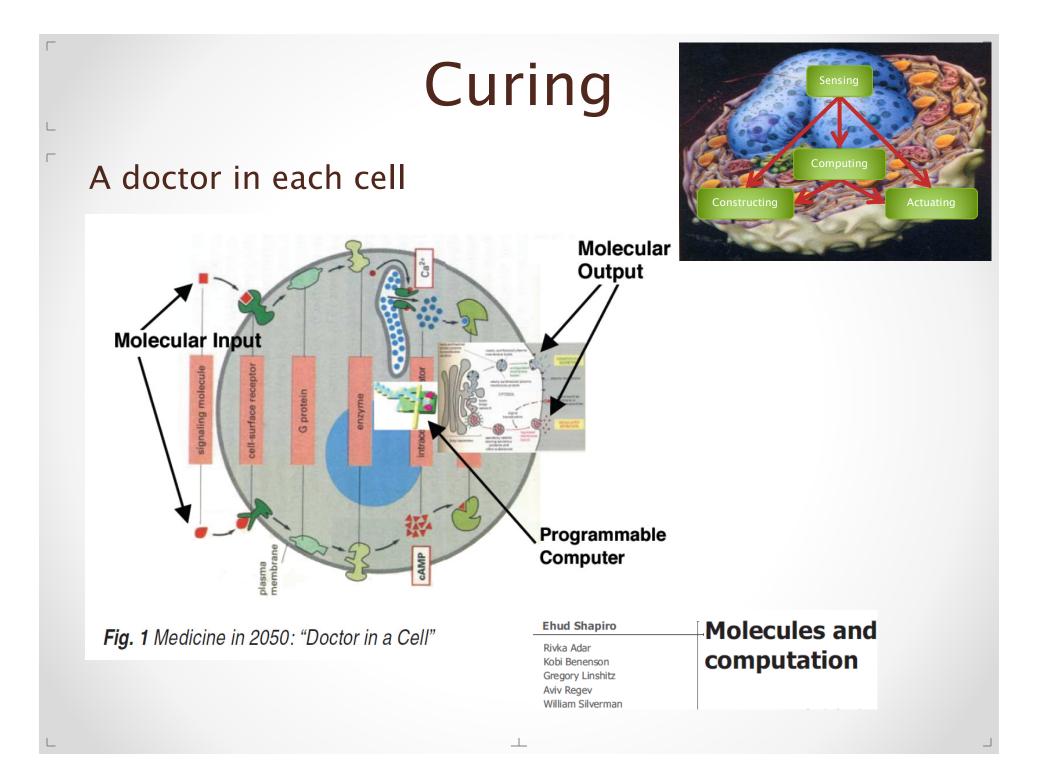
"Autonomous" DNA Computing

(Nano-engineering with biological materials)

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 (in the future)

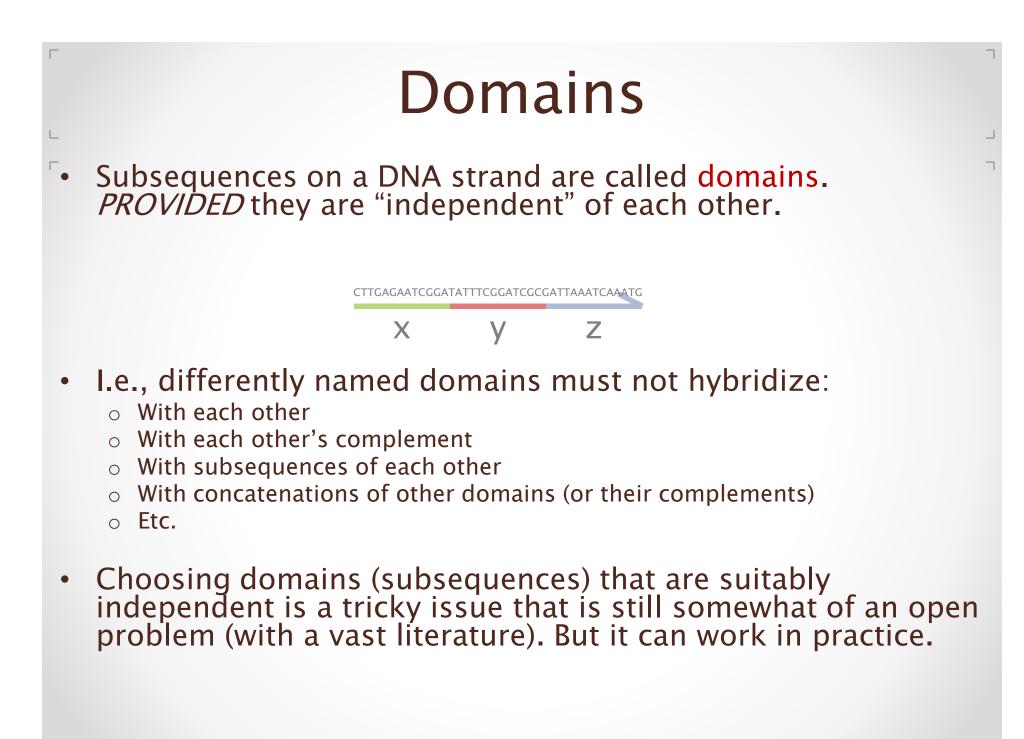


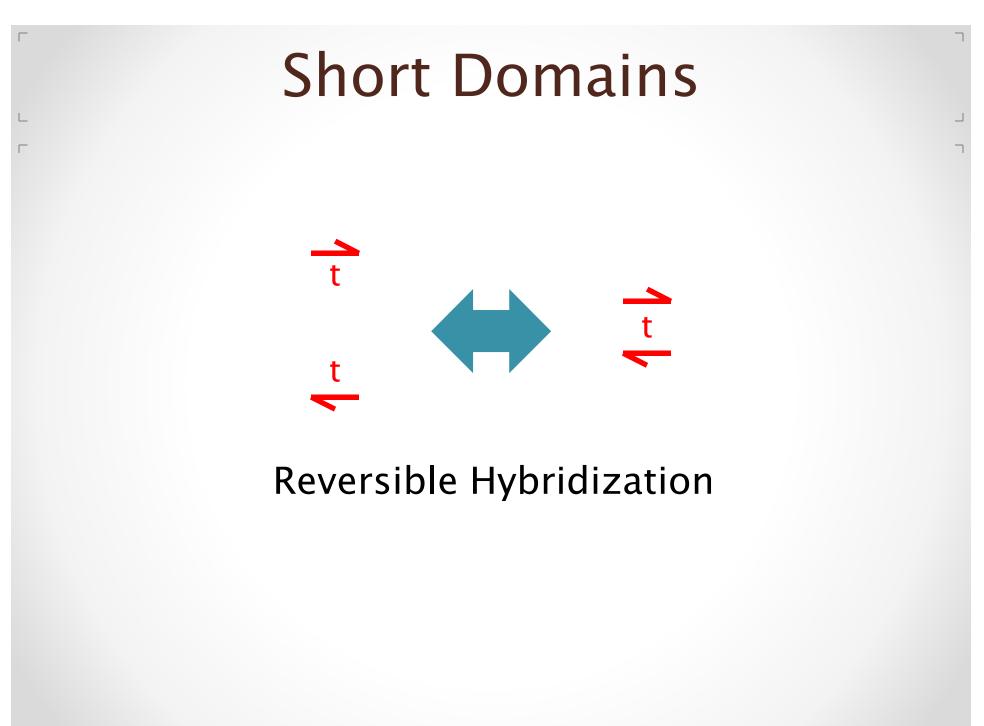
Modern DNA Computing

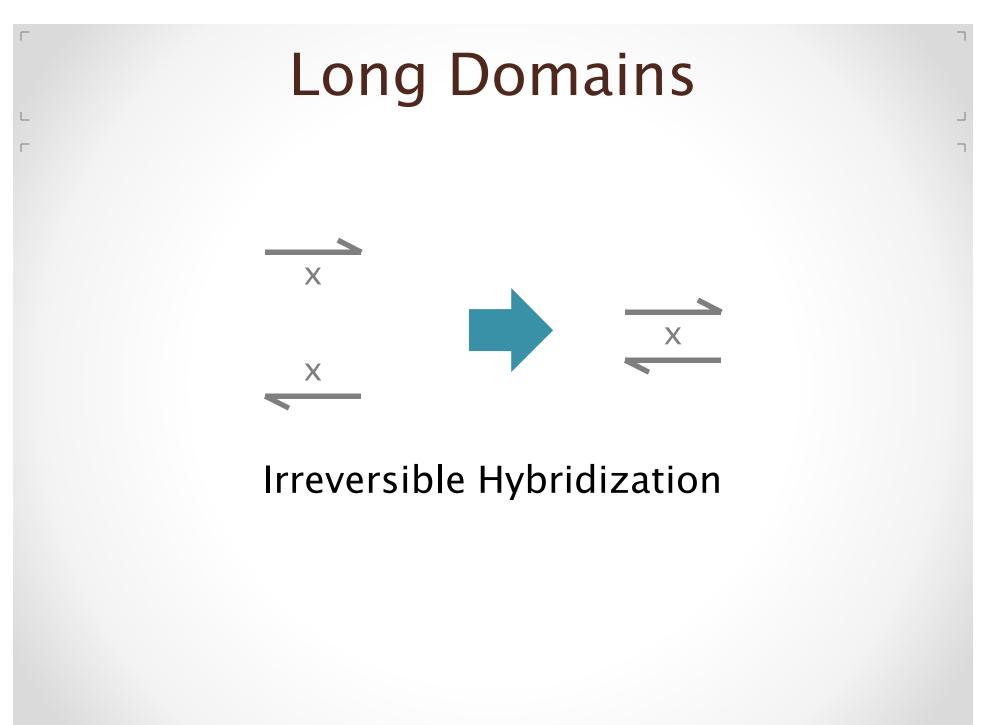
Non-goals

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

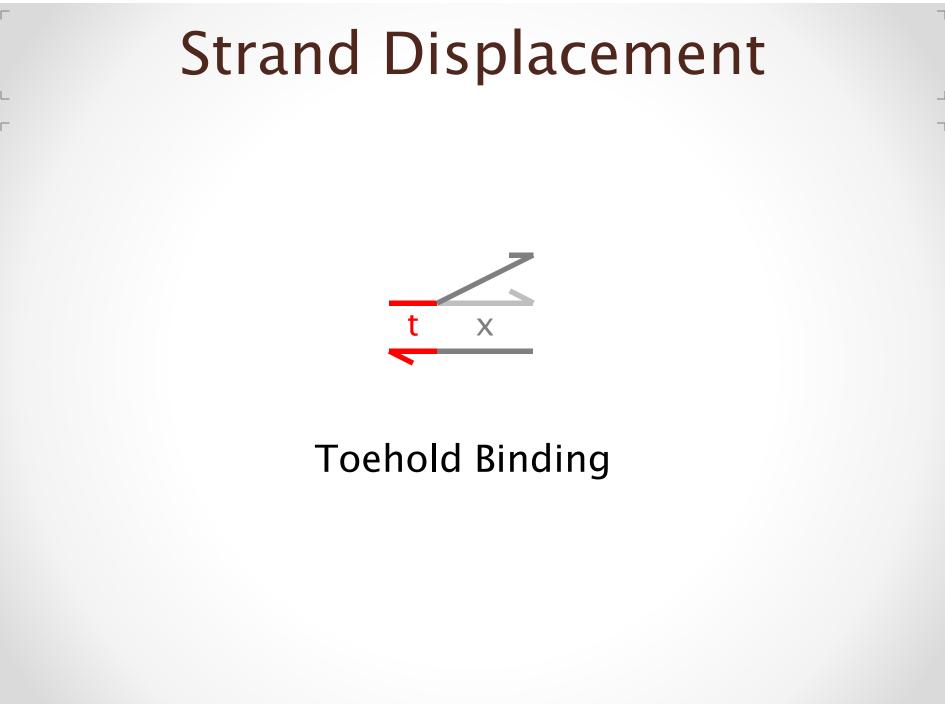
For general 'molecular programming' To precisely control the organization and dynamics of matter and information at the molecular level. To interact algorithmically with biological entities.

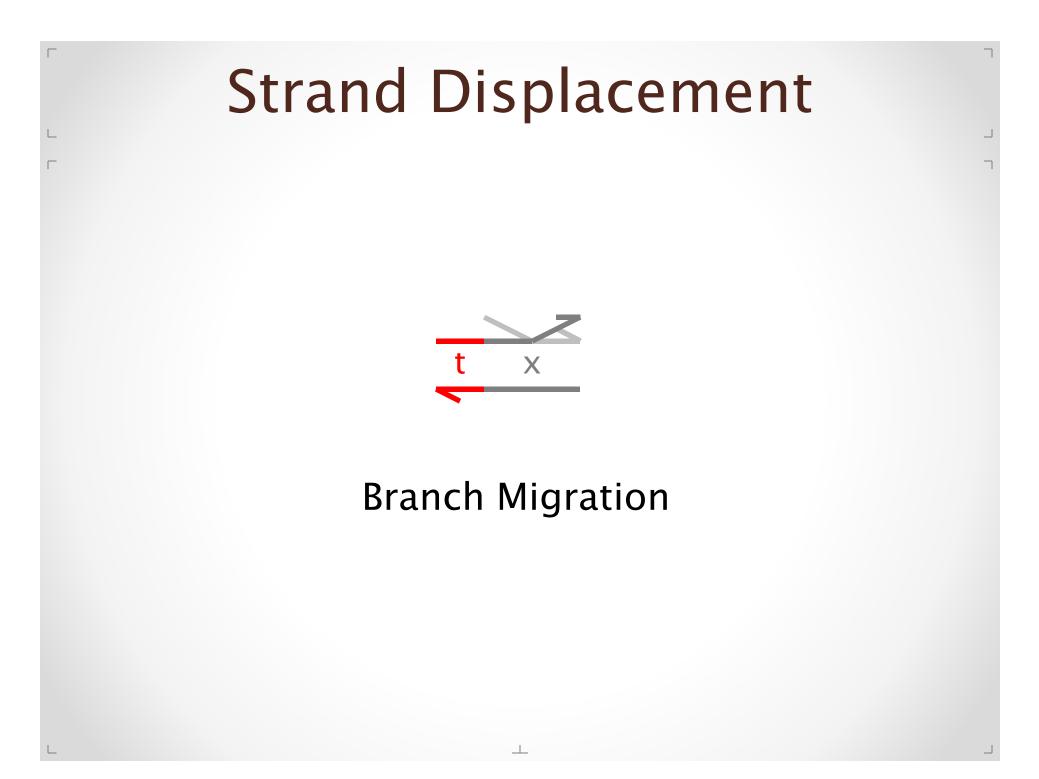


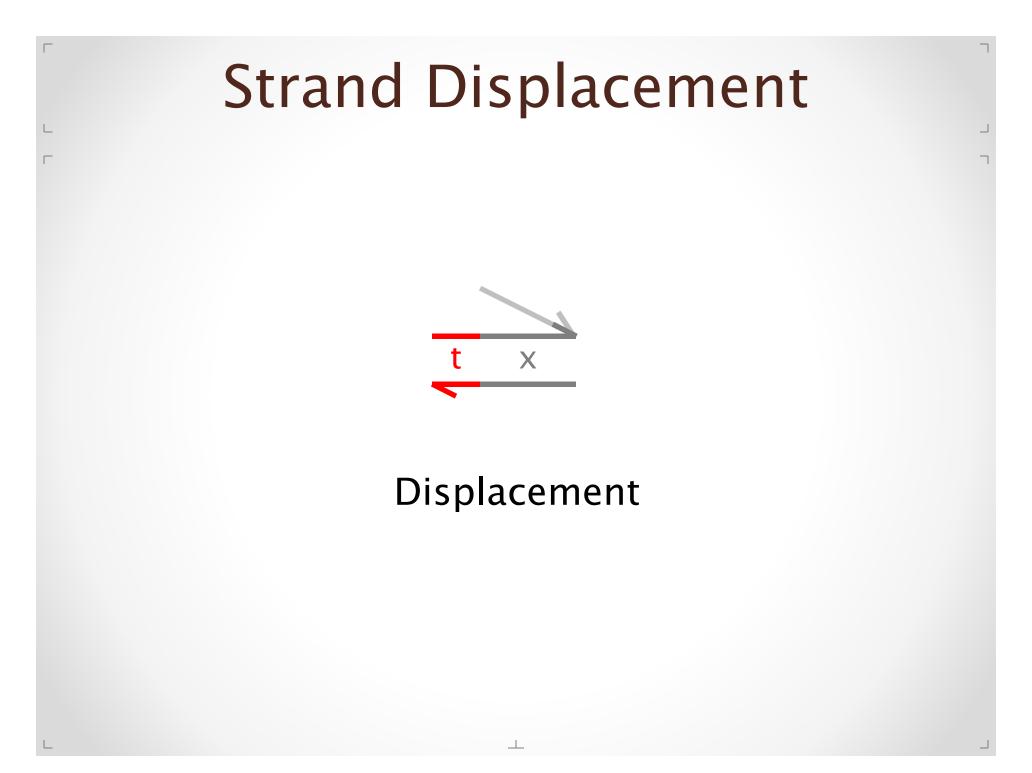




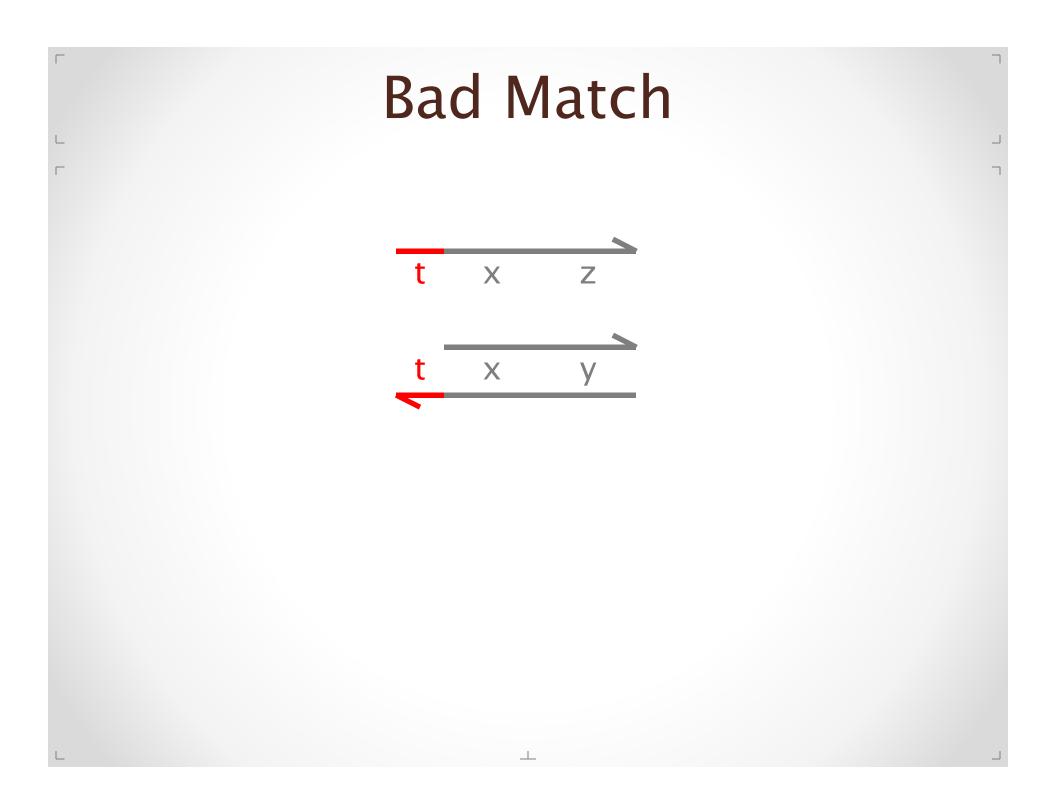
Strand Displacement t Х Х "Toehold Mediated"

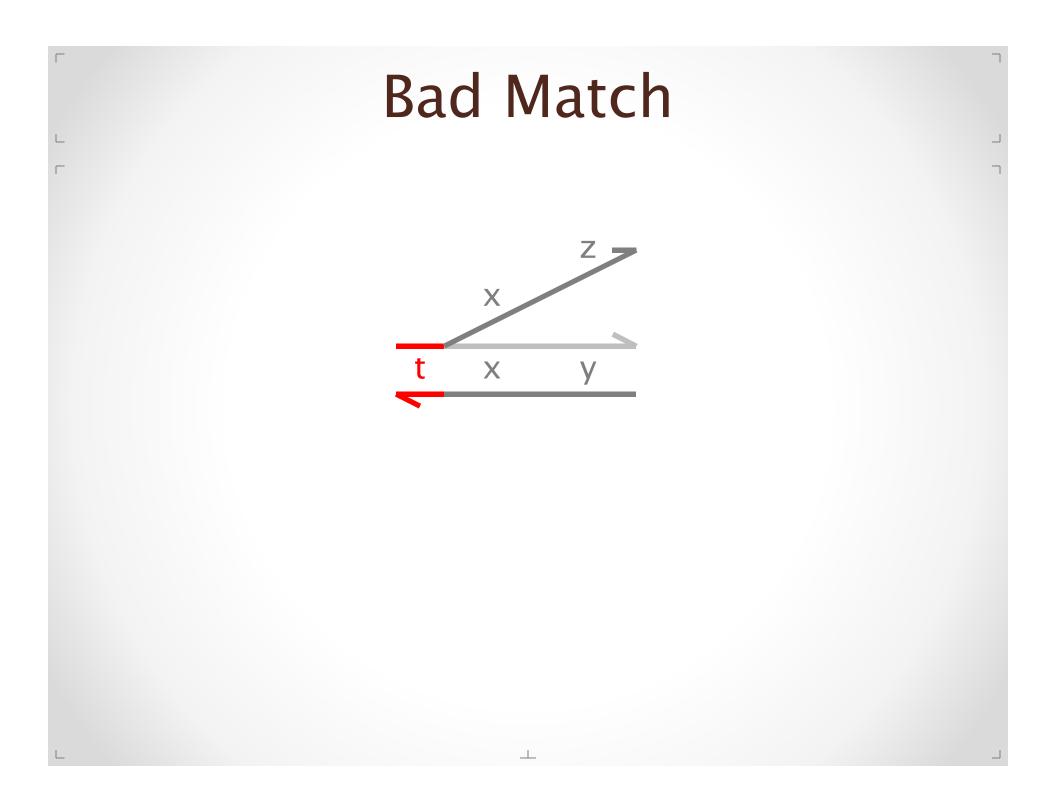


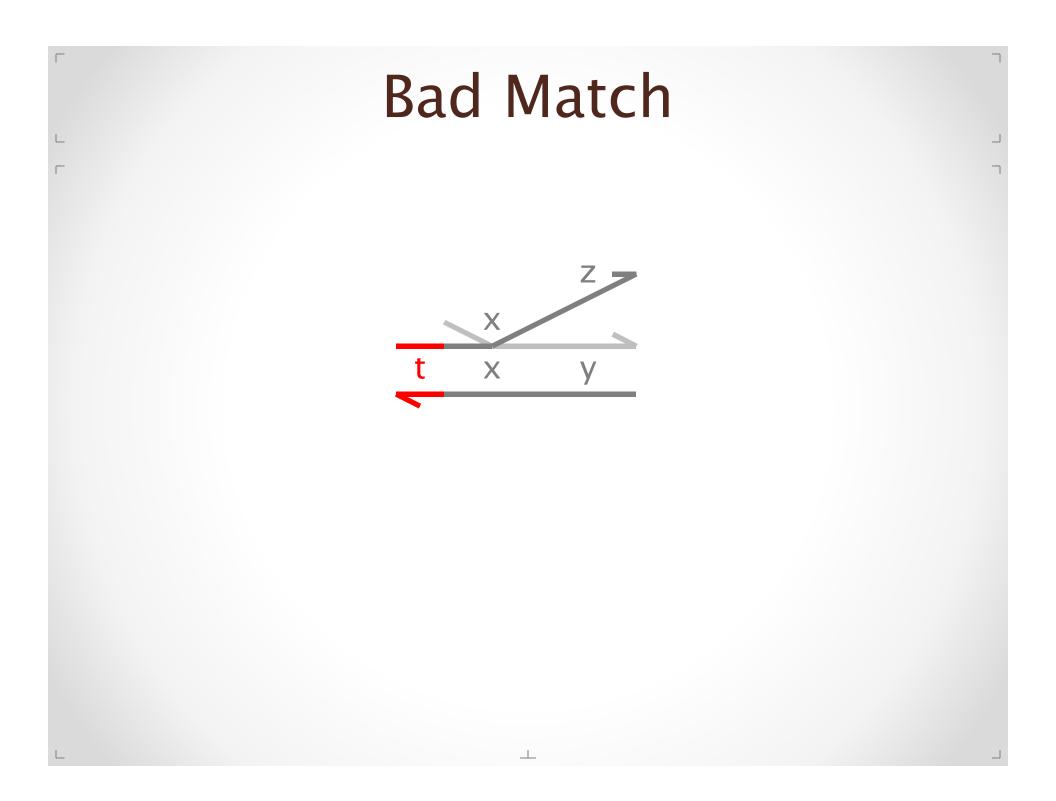




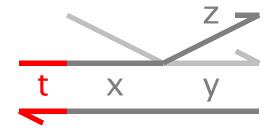
Strand Displacement Х Х Irreversible release



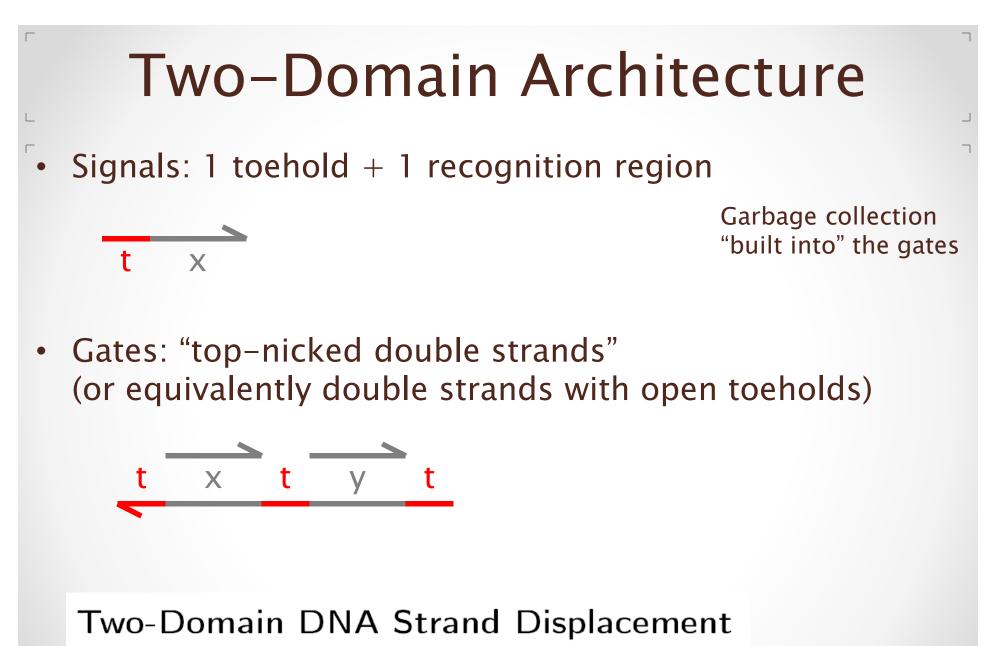




Bad Match

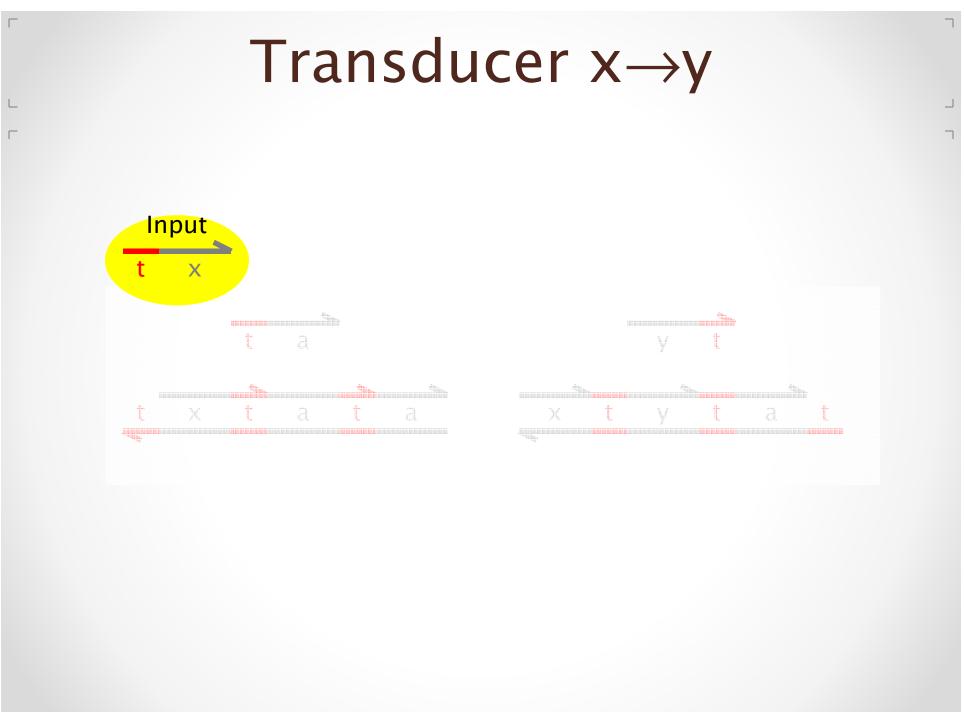


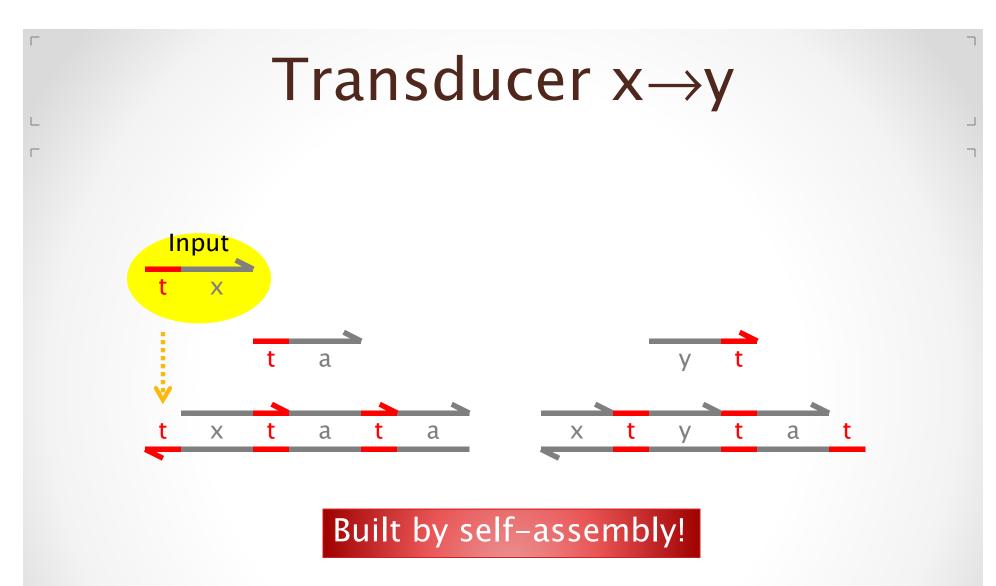
Cannot proceed Hence will undo



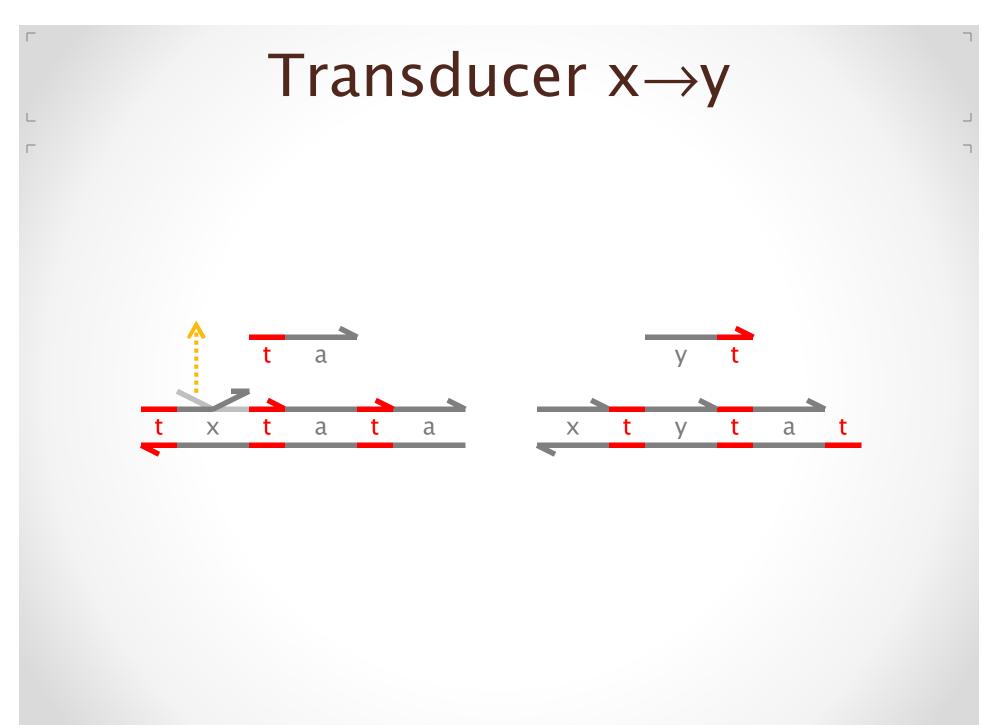
In S. B. Cooper, E. Kashefi, P. Panangaden (Eds.): Developments in Computational Models (DCM 2010). EPTCS 25, 2010, pp. 33-47. May 2010.

Luca Cardelli



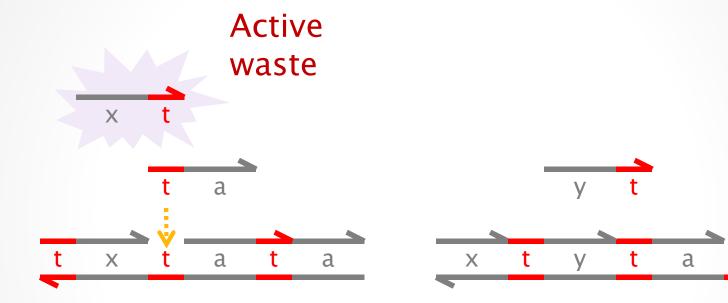


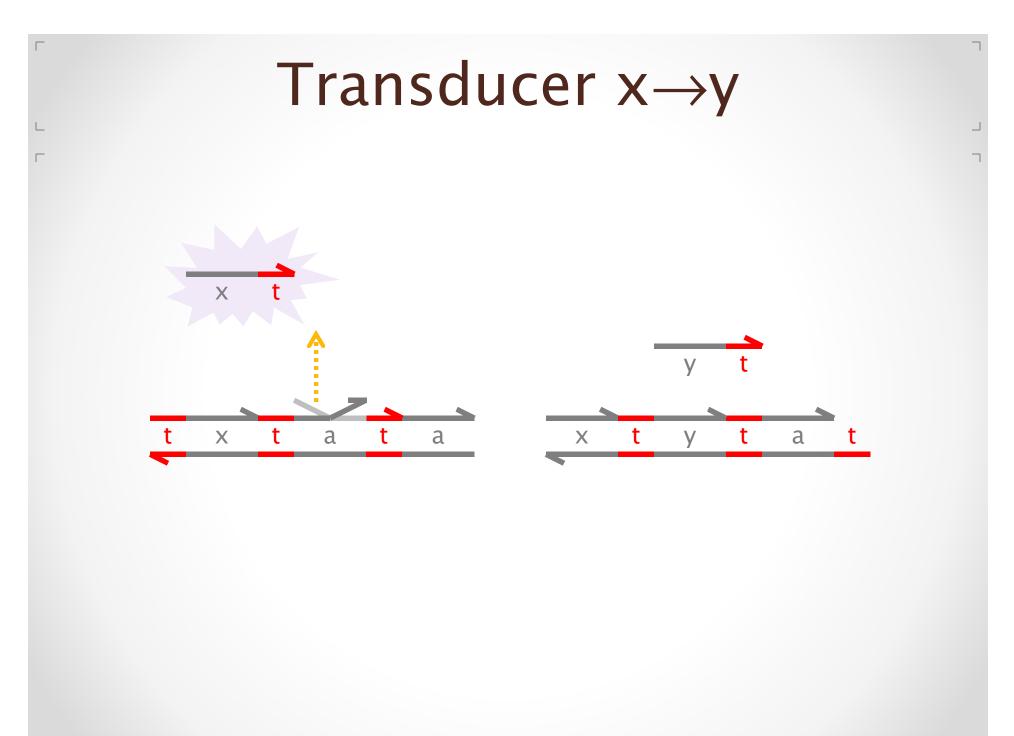
ta is a *private* signal (a different 'a' for each xy pair)

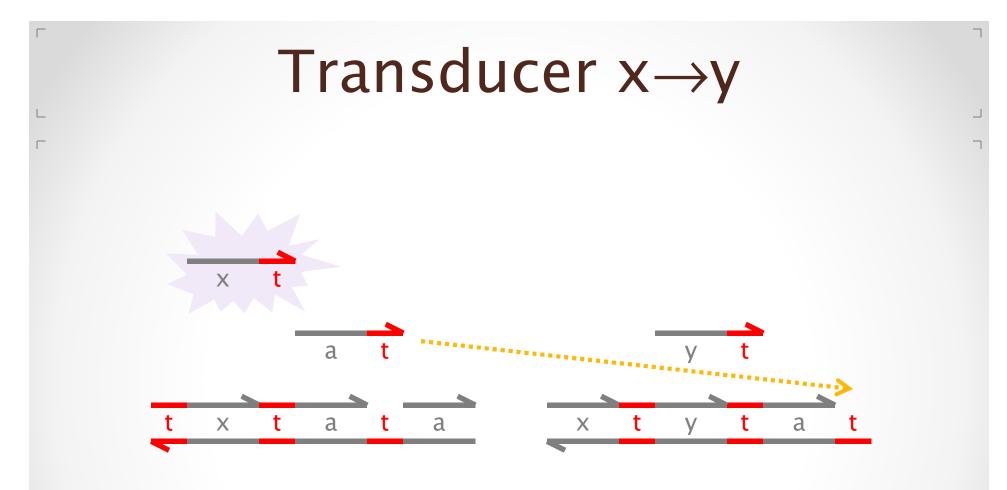


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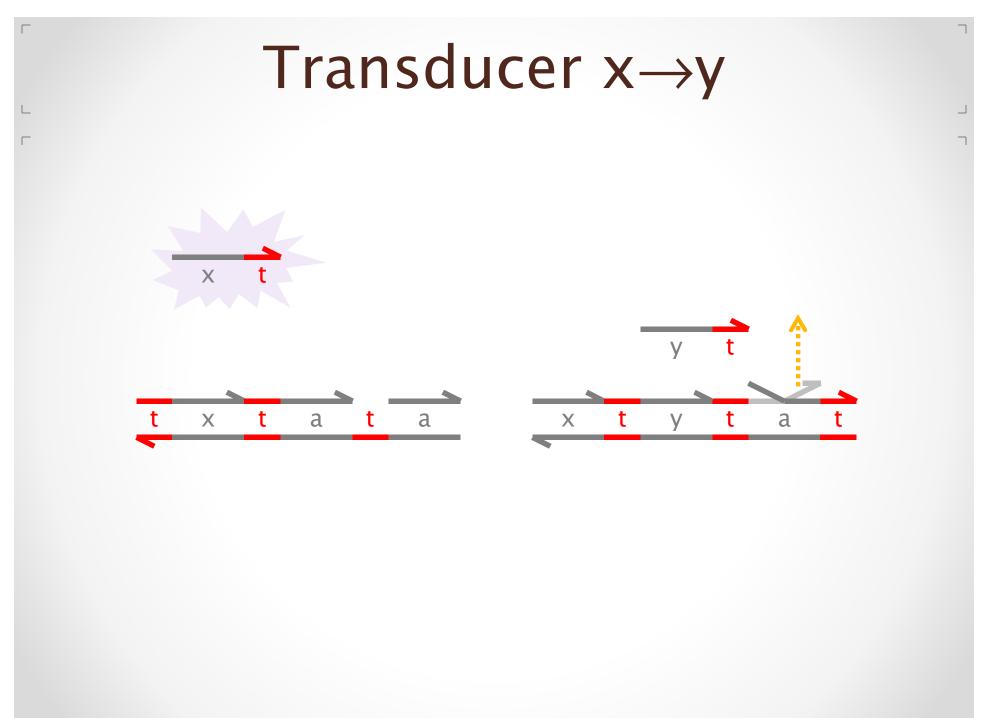
Transducer x→y

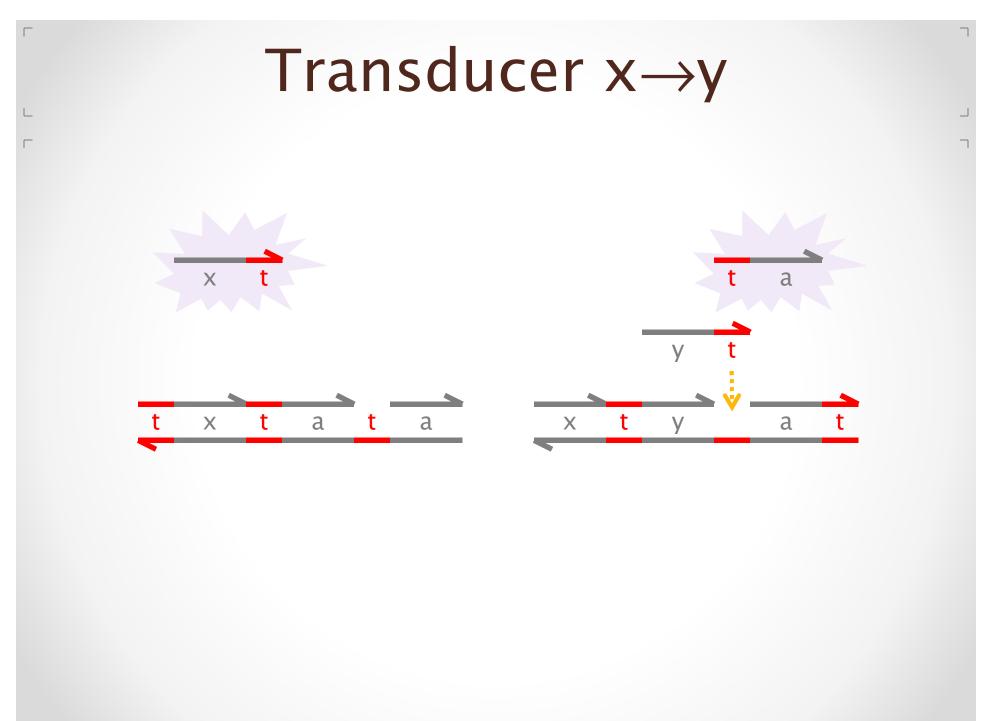


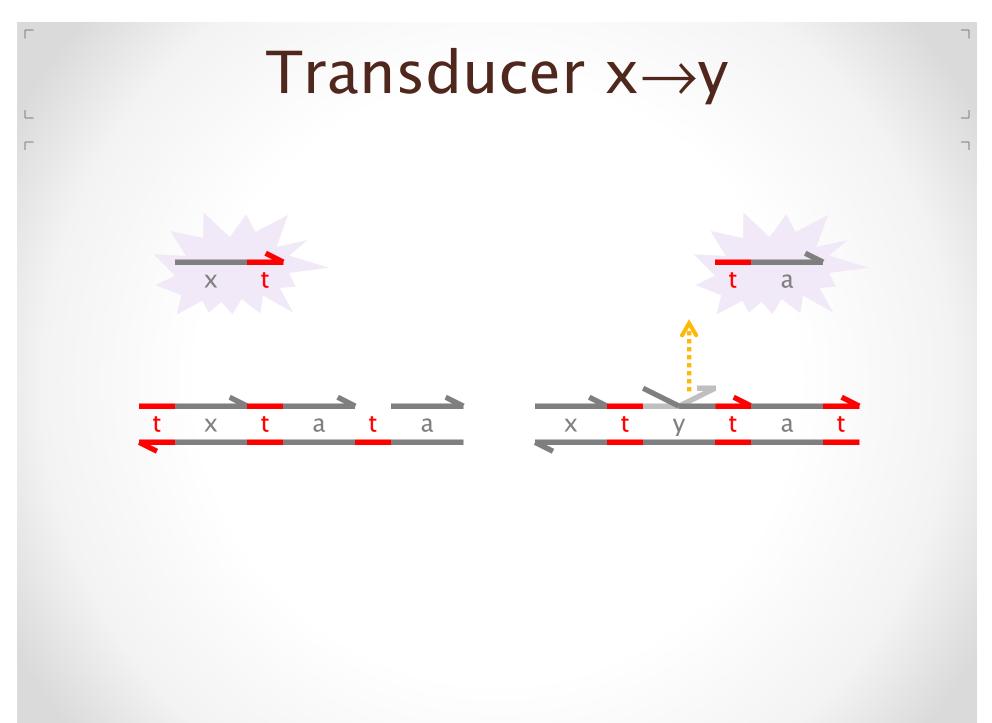


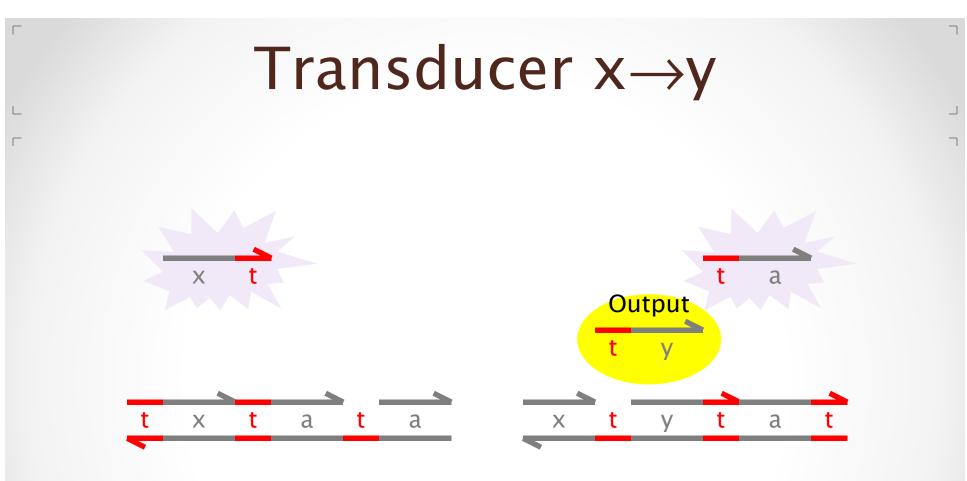


So far, a tx *signal* has produced an at *cosignal*. But we want signals as output, not cosignals.



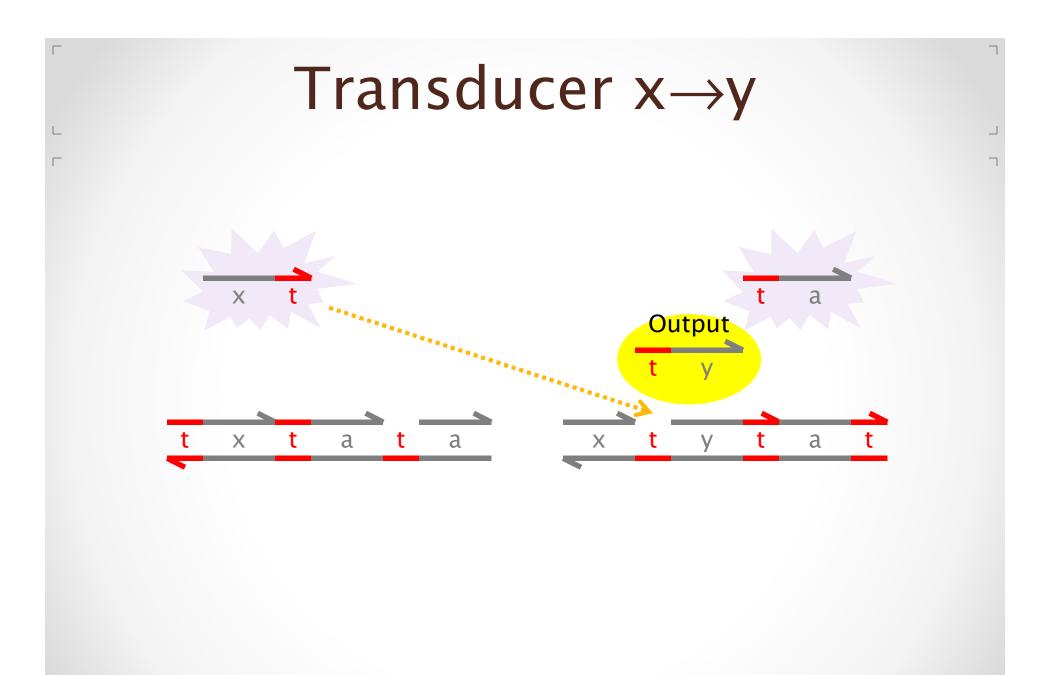




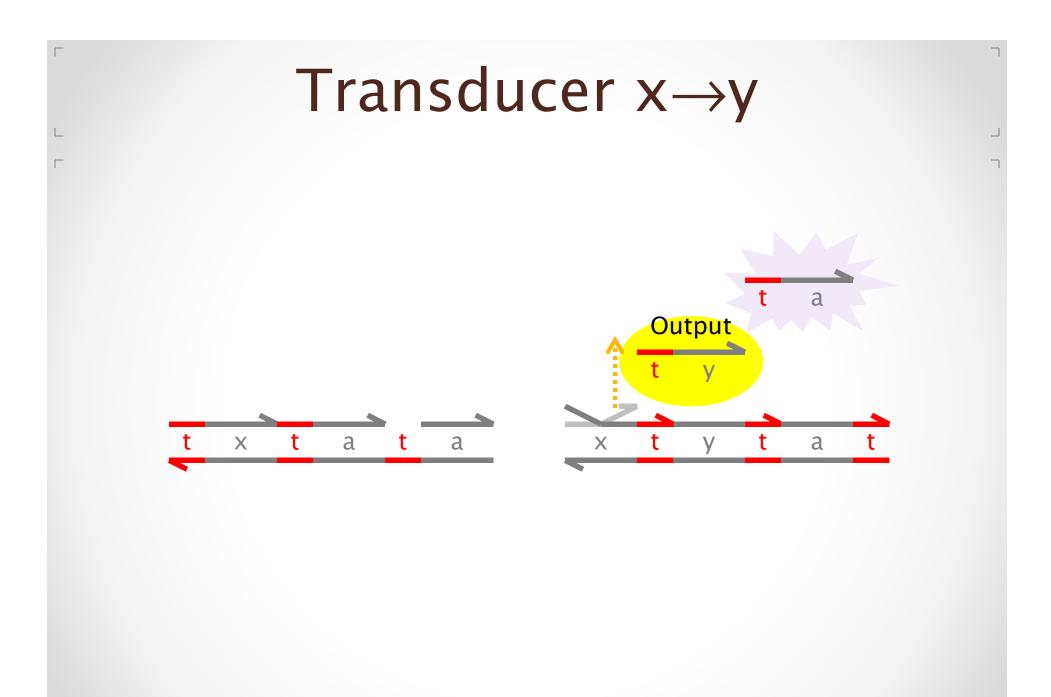


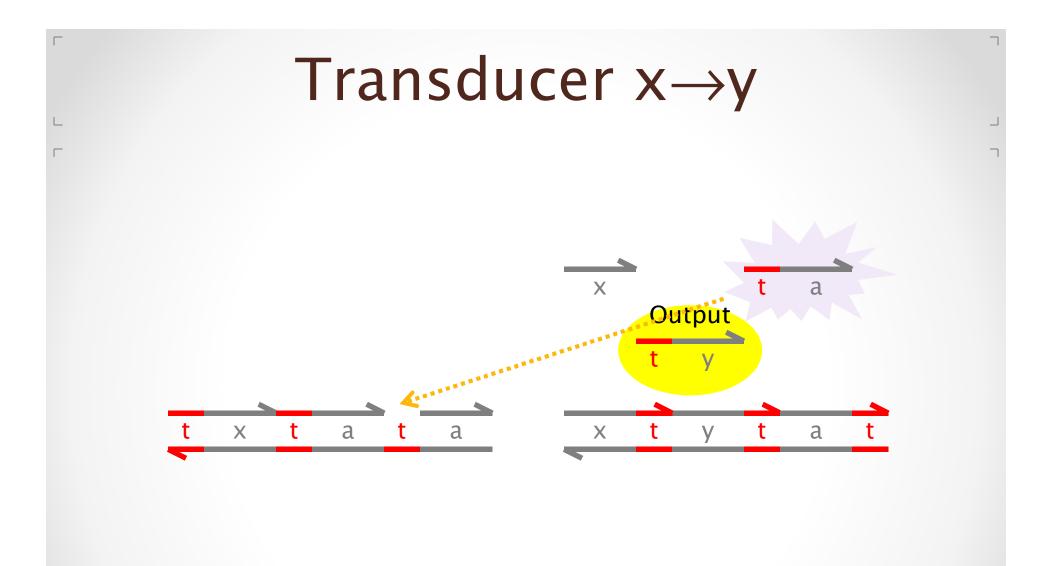
Here is our output ty signal.

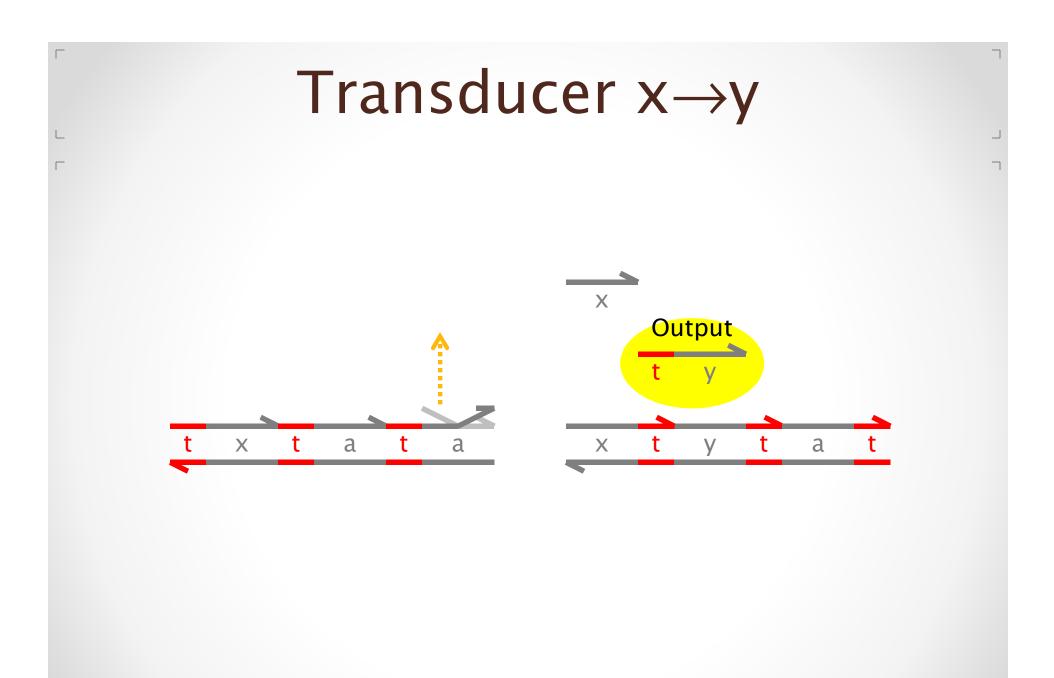
But we are not done yet: 1) We need to make the output irreversible. 2) We need to remove the garbage. We can use (2) to achieve (1).



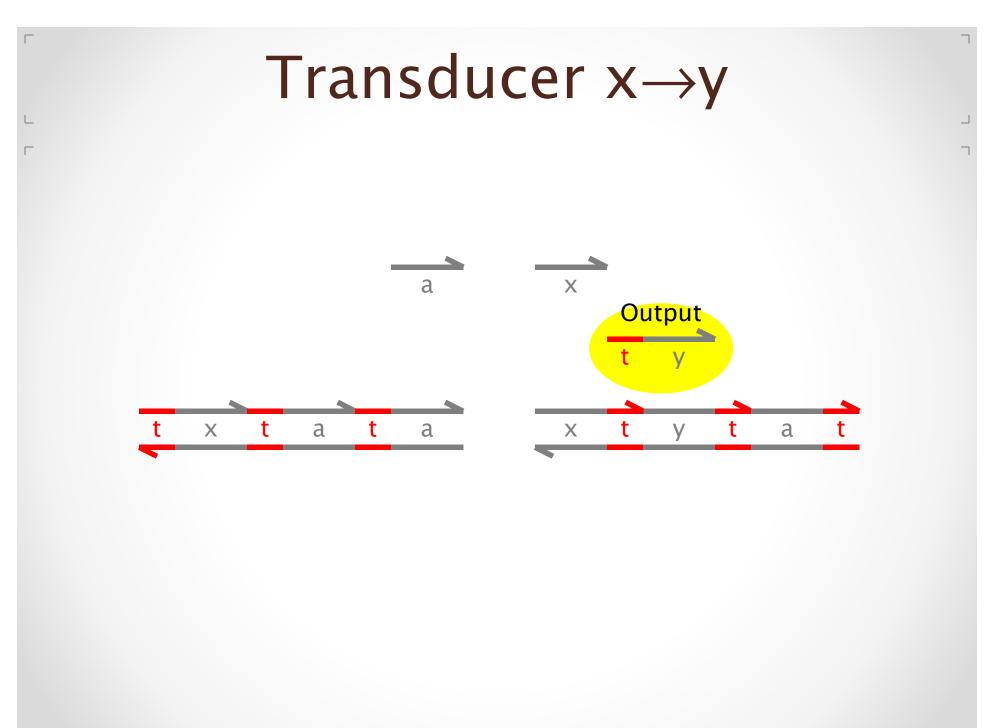
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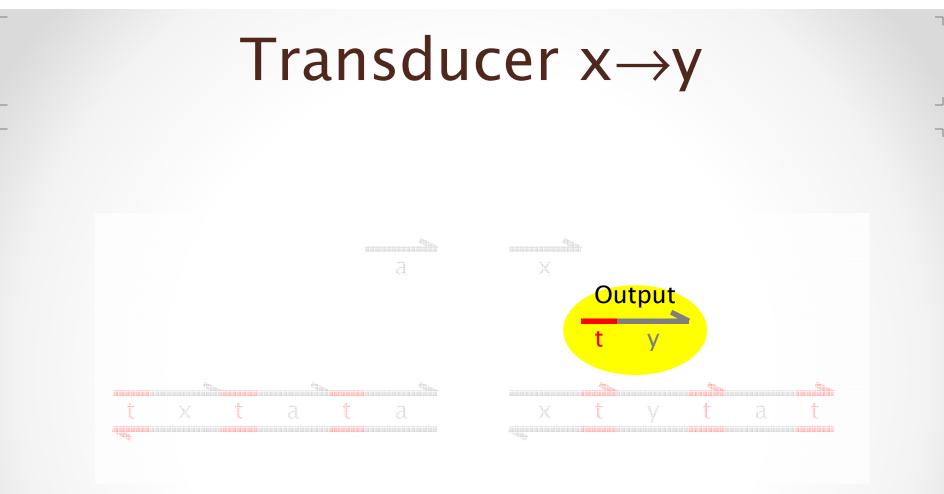






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Done.

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

 \perp

General n×m Join-Fork

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

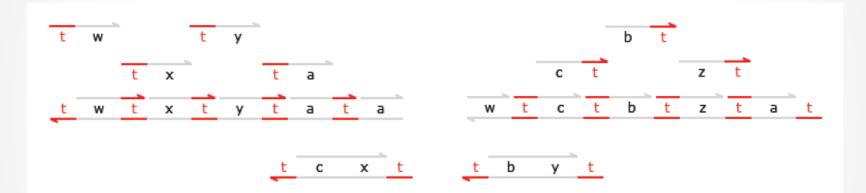
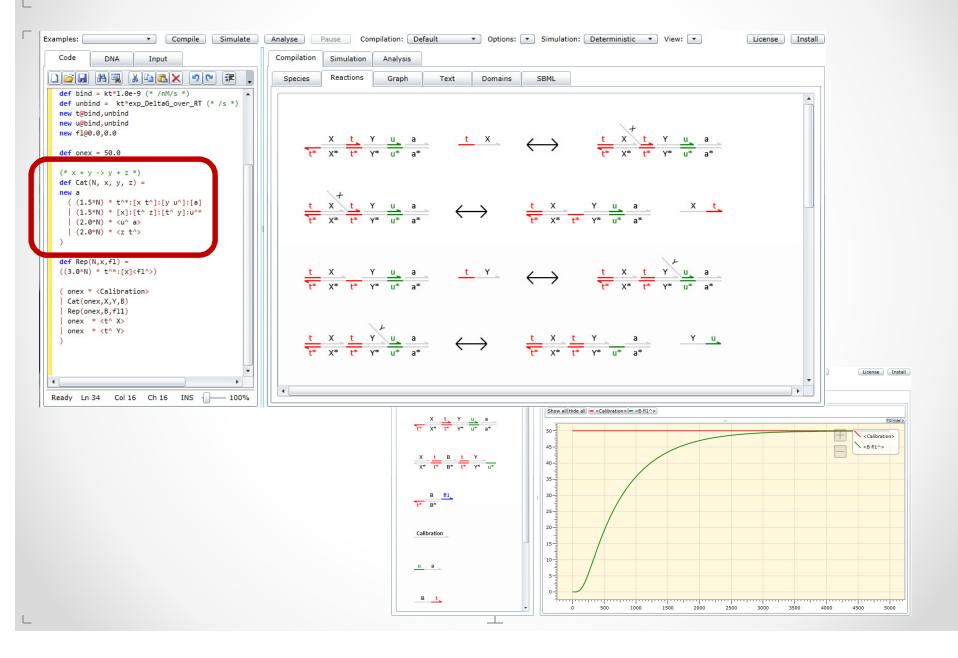


Figure 9: 3-Join $J_{wxyz} | tw | tx | ty \rightarrow tz$: initial state plus inputs tw, tx, ty.

DNA Programming



Debugging

Big Networks

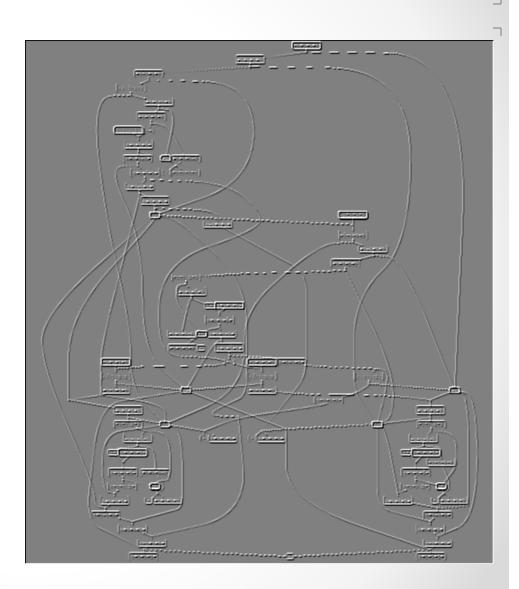
- Two-domain DNA gates for 1 Approximate Majority switch.
- Initial species: 17
- Total number of species: 85 (including run-time produced ones)
- Total number of reactions: 104

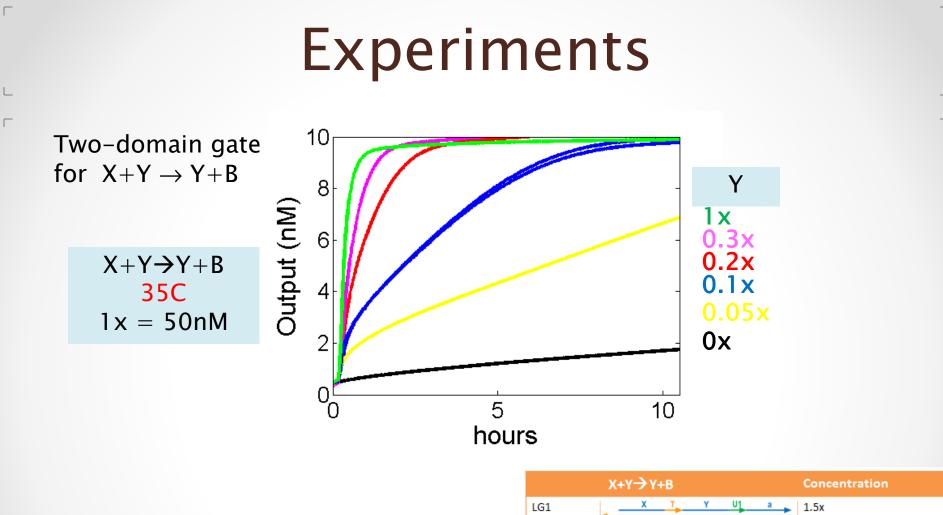
Analysis

- Gate correctness
- Circuit correctness
- Compiler correctness
- Currently, by simulation
- Increasingly, by modelchecking:

Design and Analysis of DNA Strand Displacement Devices using Probabilistic Model Checking

Matthew R. Lakin *† David Parker ^{‡†} Luca Cardelli* Marta Kwiatkowska [‡] Andrew Phillips*[§]





Yuan–Jyue Chen and Georg Seelig U.Washingon.

| | X+Y→Y+B | Concentration |
|--------------|---|---------------------------------|
| LG1 | X T Y U1 a T* X* T* Y* U1* a* | 1.5x |
| LG2 | X* T B T Y X* T* B* T* Y* U1* | 1.5x |
| input | <u>т х</u> | 1x |
| Catalyst | T Y | 0x, 0.05x, 0.1x, 0.2x, 0.3x, 1x |
| ~в | B T | 2x |
| R1 | U1 a | 2x |
| B readout | T* B* ROX | 3х |

Summary

Executable chemistry

 Given an arbitrary finite chemical network, compile it systematically and execute it.

[D. Soloveichik, G. Seelig, E. Winfree. DNA as a Universal Substrate for Chemical Kinetics. PNAS 107 no. 12, 5393-5398, 2010.]

 Finite chemical networks have the computing power of (stochastic) Petri Nets. Population protocols (such as AM) are also well-characterized. [D.Angluin, J.Aspnes, D.Eisenstat, E.Ruppert: The Computational Power of Population Protocols].

Executable bio-chemistry

- In addition, DNA supports polymerization, which gives the computing power of Turing Machines.
- Then the programming language cannot be just chemical reactions, but has to be something more like process algebra or term-rewriting systems.

Conclusions

Conclusions

• The Language of Functions

• The Language of Life