Programming Cells

ATGCTTACCGGTACGTTTACGACTACGT AGCTAGCATGCTTACCGGTACGTTTACG

Andrew Phillips Biological Computation Group Microsoft Research

Image courtesy of James Brown, Haseloff Lab, University of Cambridge

Potential

Health

Program cells to control tumours



Environmentally Controlled Invasion of Cancer Cells by Engineered Bacteria

J. Christopher Anderson^{1,3}, Elizabeth J. Clarke³, Adam P. Arkin^{1,2*} and Christopher A. Voigt^{2,3}

ELSEVIER



Energy

Convert CO₂ into fuel



Produce vaccines



Production of the antimalarial drug precursor artemisinic acid in engineered yeast

Dae-Kyun Ro¹*, Eric M. Paradise²*, Mario Ouellet¹, Karl J. Fisher⁶, Karyn L. Newman¹, John M. Ndungu³, Kimberly A. Ho¹, Rachel A. Eachus¹, Timothy S. Ham¹, James Kirby², Michelle C. Y. Chang¹, Sydnor T. Withers², Yolchiro Shiba², Richmond Sarpong³ & Jay D. Keasling^{1,2,4,5}

Convert sunlight into electricity



Challenges

- Programming cells is hugely difficult
 - Issues of reliability, toxicity, strain on the host cell

- Systems are highly complex
 - Cannot be designed by trial and error
 - Requires use of computer software

 Could software for programming cells one day rival software for programming silicon?

DNA Structure (A-T,G-C)



www.wehi.edu.au/wehi-tv

DNA as a Computing Substrate

- Molecular Scale
 - Overcome limits to miniaturisation of silicon chips
 - 1GB of information in a millionth of a mm³
 - Can self-assemble
 - Clean and cheap to manufacture



• Programmable

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

- Interactions are directly controlled by the sequence
- Massively parallel

DNA Computers Inside Cells

DNA Drugs (Mullis)





Programmable Drugs (Shapiro)

An autonomous molecular computer for logical control of gene expression

Yaakov Benenson $^{1,2},$ Binyamin Gil 2, Uri Ben-Dor 1, Rivka Adar 2 & Ehud Shapiro 1,2



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DNA Strand Displacement

Computation solely by complementary base pairs sticking together (T-A and G-C)



Bernard Yurke

DNA Strand Displacement

Dynamic DNA nanotechnology using strand displacement reactions

David Yu Zhang¹ and Georg Seelig²



NATURE CHEMISTRY | VOL 3 | FEBRUARY 2011

Enzyme-Free Nucleic Acid Logic Circuits

Georg Seelig,¹ David Soloveichik,² David Yu Zhang,² Erik Winfree^{2,3}*



SCIENCE VOL 314 8 DECEMBER 2006

Engineering Entropy-Driven Reactions and Networks Catalyzed by DNA

David Yu Zhang, ¹† Andrew J. Turberfield, ² Bernard Yurke, ³* Erik Winfree¹†



SCIENCE VOL 318 16 NOVEMBER 2007

Selective cell death mediated by small conditional RNAs

Suvir Venkataraman^a, Robert M. Dirks^{a,b}, Christine T. Ueda^b, and Niles A. Pierce^{a,c,1}



PNAS | September 28, 2010

DNA Strand Displacement Language

DNA Strand Displacement (DSD) Language

Step 1: Program circuit design



Step 2: Compile circuit behaviour



Step 3: Simulate circuit



Step 4: Compile circuit to DNA or RNA



Step 5: Insert circuit into cells



Phillips, Cardelli. Royal Society Interface, 2009 Lakin, Youssef, Cardelli, Phillips. Royal Society Interface, 2011 Lakin, Youssef, Polo, Emmott, Phillips. Bioinformatics, 2011

Output = Input1 AND Input2

Input 1

Input 2

TATTCC CCCAAAACAAAACAAAACAA

CCCTTTTCTAAACTAAACAA GCTA

Output

Output = Input1 AND Input2



Output = Input1 AND Input2



Output = Input1 AND Input2



Output = Input1 AND Input2

Output







🔄 Visual DSD - lepton.research.microsoft.com	
Examples: Compile Simulate Analyse	Pause Compilation: Default
Simulation: Stochastic 🔹 View: 💌	
Code DNA Input	Compilation Simulation Analysis
	Species Reactions Graph Text Domains SBML
	Pan Zoom Layout Zoom S2 % Fit Layout
CCCAAAACAAAACAAAACAAAACAA ATAAGG GGGTTTTGTTTGTTTGTTTGTTTGTTTGTTTGT	Horizontal Aspect Ratio Group Initial Nodes
TATTCC CCCAAAACAAAACAAAACAA	
CCCTTTTCTAAACTAAACAA GCTA	
	· · ·

Large-Scale Logic Circuits

Scaling Up Digital Circuit Computation with DNA Strand Displacement Cascades

Lulu Qian¹ and Erik Winfree^{1,2,3}*





Scaling Up DNA Computation

John H. Reif

"In addition to biochemistry laboratory techniques, computer science techniques were essential."

"Computer simulations of seesaw gate circuitry optimized the design and correlated experimental data."

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Turing-Powerful Circuits

Encoding a Stack

Encoding state transitions



Model-Checking a DNA Ripple Carry Adder



1	nput	A	Input B			Output X		Output C	Result
MSB	LSB	Value	MSB	LSB	Value	LSB	MSB	Value	Value
0	0	0	0	0	0	0	0	0	0
0	0	0	0	1	1	1	0	0	1
0	0	0	1	0	2	0	1	0	2
0	0	0	1	1	3	1	1	0	3
0	1	1	0	0	0	1	0	0	1
0	1	1	0	1	1	0	1	0	2
0	1	1	1	0	2	1	1	0	3
0	1	1	1	1	3	0	0	1	4
1	0	2	0	0	0	0	1	0	2
1	0	2	0	1	1	1	1	0	3
1	0	2	1	0	2	0	0	1	4
1	0	2	1	1	3	1	0	1	5
1	1	3	0	0	0	1	1	0	3
1	1	3	0	1	1	0	0	1	4
1	1	3	1	0	2	1	0	1	5
1	1	3	1	1	3	0	1	1	6

Lakin, Phillips. DNA Computing, 2011

Localised Circuits

Organization of Intracellular Reactions with Rationally Designed RNA Assemblies

Camille J. Delebecque, 1,2,3,4 Ariel B. Lindner, 3,4* Pamela A. Silver, 1,2* Faisal A. Aldaye1,2



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Hairpins tethered to origami

- Increased speed
- Reduced interference



²⁰ Chandran, Gopalkrishnan, Phillips, Reif. DNA Computing, 2011

Programming Living Cells

The "software" of a cell: DNA \rightarrow RNA \rightarrow Protein



DNA transcription (real time)

Messenger RNA translation

Information storage and processing within the cell is more efficient by many orders of magnitude than electronic digital computation, with respect to both information density and energy consumption.

DNA can function across species

A "multi-platform" code



Glowing jellyfish



Glowing bacteria

Moving DNA from one organism to another

DNA can completely reprogram a cell



Creation of a Bacterial Cell Controlled by a Chemically Synthesized Genome

Sciencexpress / www.sciencexpress.org / 20 May 2010

M. mycoides Extraction Sequencing ...GTTTCTCCATACCCGTTTTTTTGGGCTAGC... M. capricolum **Synthesis** Insertion Transformation

Low-Level DNA Language

A simplified view of DNA instructions



High-Level DNA Language

Given a design, automatically determine the DNA



A language for programming cells

Genetic Engineering of Cells (GEC)

Step 1: Program device design



Step 2: Compile device behaviour



Step 3: Simulate device



Step 4: Compile device to DNA



Step 5: Insert DNA into cells



With Michael Pedersen, Matthew Lakin

Pedersen & Phillips. Royal Society Interface, 2009

Programming a receiver device

Signal crosses cell wall and binds to Receiver



Programming a receiver device



Programming a receiver device



Characterising a receiver device

Model Simulation



Experimental Data



Spatial Receiver Device

Model Simulation

Experimental Data





Programming Turing Patterns

Cells that communicate to perform complex functions









Modelling Biophysics: Thresholding



With Tim Rudge, James Brown, Jim Haseloff

Modelling Biophysics: Thresholding



MHC class I Antigen Presentation



© Diego Accorsi 2011. A Master's Research Project submitted in conformity with the requirements for the degree of Master of Science in Biomedical Communications (MScBMC), Faculty of Medicine, University of Toronto.

Peptide Optimisation

- Peptide-MHC complexes generally need to be stable for many hours or days at the cell surface
- *Peptide optimisation:* high affinity peptides are preferentially selected for presentation
- Optimisation in the ER is typically limited to tens of minutes
 - How is such high optimisation is achieved in so little time?
 - What are the main mechanisms of optimisation?

MHC Class I Model



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Automatic generation of reactions



Automatic generation of ODEs

 $\varnothing \quad \rightleftharpoons \frac{g_M}{d_M}$ M $\varnothing \qquad \rightleftharpoons \stackrel{g_T}{\rightleftharpoons} \quad T$ $M+T \quad \rightleftharpoons \overset{b_T}{\underset{u_T}{\Rightarrow}} TM$ $Me \rightarrow^{d_{Me}} \varnothing$ $\varnothing \qquad \rightleftharpoons \stackrel{g_i}{\underset{d_{\mathcal{D}}}{\rightleftharpoons}} \quad P_i$ MP_i $TM + P_i \quad \rightleftharpoons \stackrel{c}{\underset{u_i \cdot q}{\leftarrow}}$ TMP_i $TMP_i \rightarrow u_T \cdot v \quad T + MP_i$ $MP_i \rightarrow^e MeP_i$ $MeP_i \rightarrow^{u_i}$ Me

 $[M]' = \sum_{i} u_i [MP_i] + u_T [TM] + g_M$ $-(b\sum_{i}[P_i]+b_T[T]+d_M)[M]$ $[T]' = u_T[TM] + g_T + u_T v \sum_{i} [TMP_i] - (b_T[M] + d_T)[T]$ $[MP_i]' = b[M][P_i] + u_T v[TMP_i] - (u_i + e)[MP_i]$ $[TM]' = b_T[M][T] + q \sum_i u_i [TMP_i] - (u_T + c \sum_i [P_i])[TM]$ $[TMP_i]' = ba[TM][P_i] - (u_iq + u_Tv)[TMP_i]$ $[P_i]' = u_i[MP_i] + u_iq[TMP_i] + g_i$ $-(b[M]+c[TM]+d_P)[P_i]$ $[MeP_i]' = e[MP_i] - u_i[MeP_i]$ $[Me]' = \sum u_i [MeP_i] - d_{Me}[Me]$

Phillips, Dalchau et al. PLoS Computational Biology, 2011

Steady-state ODE analysis

$$x = u_T v/q \qquad C = c[TM]^*/d_P$$

Steady-state experiments

Tapasin enhances peptide presentation by $1/u_i$ Measured off-rates u_i for SIINFEKL peptides



Delayed egress vs. Enhanced off-rate

Measured transit time of MHC to cell surface



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Fast transit => fast optimisation => enhanced off-rate

Time-dependent experiments

Representative peptides P_{low}, P_{med}, P_{high}



Identify differences between alleles

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Allele-specific peptide binding





Phillips, Dalchau et al. PLoS Computational Biology, 2011

Optimisation of HIV peptides



Biological Computation Group



Biological Modelling Engine: A common language runtime for Biological Computation

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DNA Computing



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Stephen Emmott



Leonard Goldstein (Cambridge)



Mark Howarth (Oxford)





Tim Joern Elliott Werner (Southampton)