Design of Bio-molecular Feedback Systems

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Outline

 Part 1: Overview of synthetic biology and simple modules

Part 2: The challenge of composing modules together

Part 3: Fabrication technology

Part 1

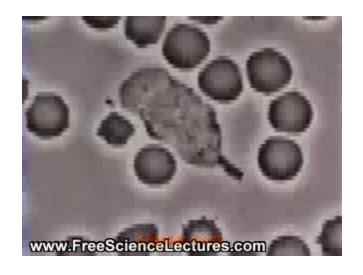
Overview of synthetic biology and simple modules

- History and basic technology
- Simple modules

Why designing bio-molecular feedback systems?

MEDICAL APPLICATIONS

(e.g. targeted drug delivery)



ALTERNATIVE ENERGY

(e.g. bio-fuels)

Making bacteria that...

- Produce hydrogen or ethanol
- Transform waste into energy



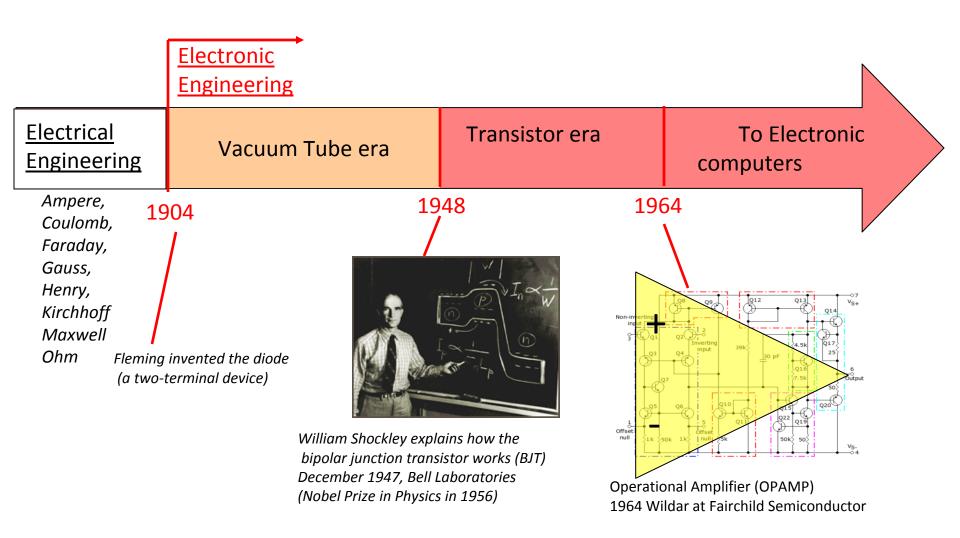
COMPUTING APPLICATIONS

(e.g. molecular computing)

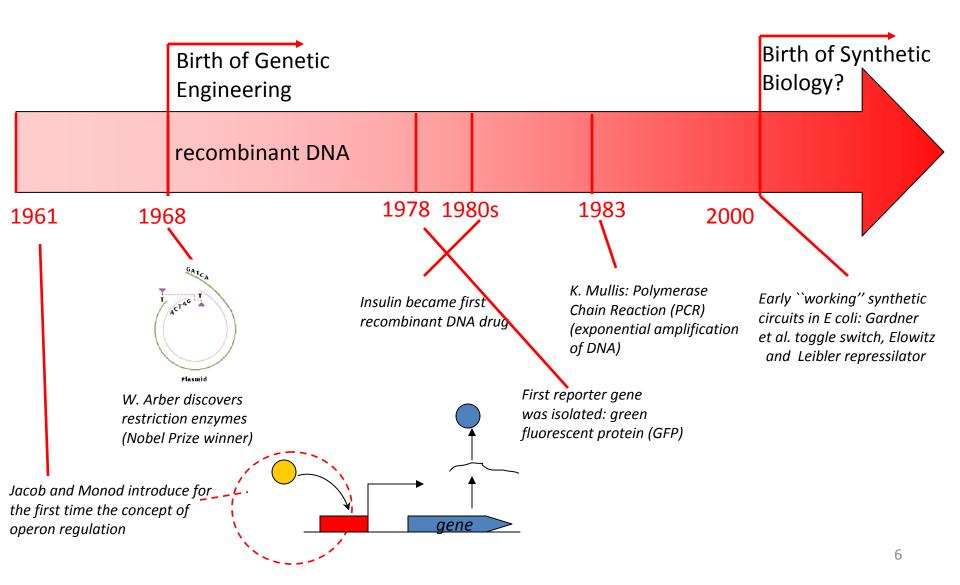
BIO-SENSING

(e.g. detecting pathogens or toxins)

Synthetic Biology: A Historical Perspective



Synthetic biology: A historical perspective



Key enabling technology

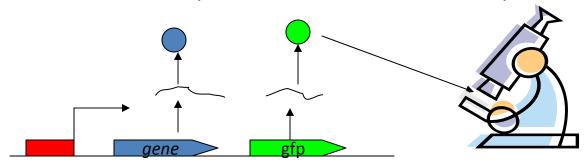
Recombinant DNA technology: allows to cut and paste pieces of DNA at desired locations cleaved by restriction enzymes

Chromosome

Plasmids

Bacterium

<u>Fluorescent Proteins</u>: allow through fluorescence microscopy to measure the concentration of a protein and thus the level of expression of the corresponding gene

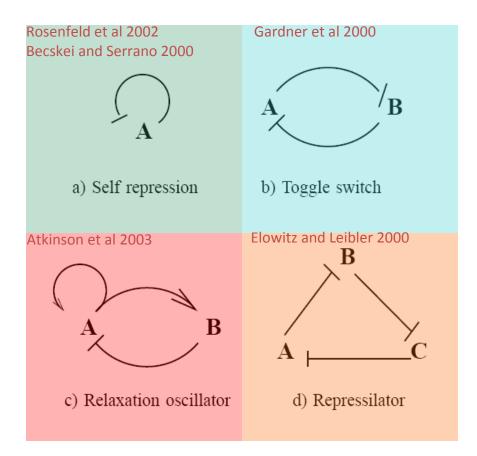


Extraneous DNA

Early modules fabricated in vivo

Monostable modules

Relaxation oscillators

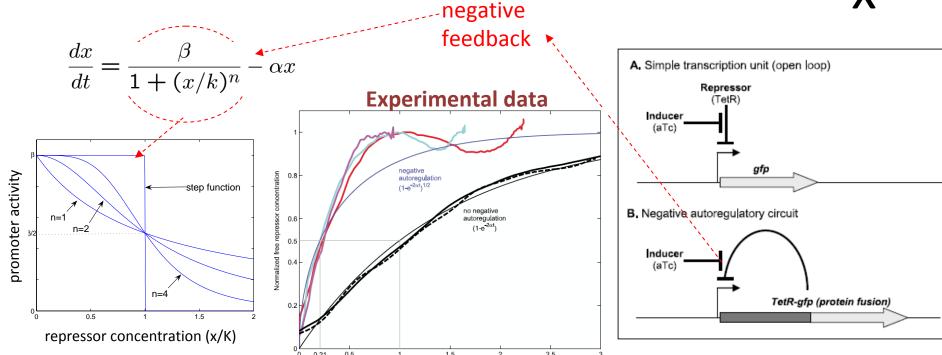


Bistable modules

Loop oscillators

A self repressed gene: Dynamics





For n=1:

Without negative feedback

$$\frac{x_1(t)}{x_1^{\text{st}}} = 1 - e^{-\alpha t}$$

With negative feedback

cell cycles

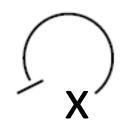
$$\frac{x_2(t)}{x_2^{\text{st}}} = \sqrt{1 - e^{-2\alpha t}}, \qquad \beta_2/\alpha \gg k \qquad x_2^{\text{st}} = \sqrt{k\beta_2/\alpha}$$

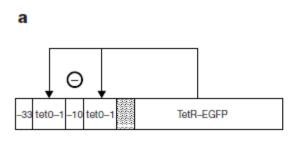
$$\beta_2/\alpha \gg k$$

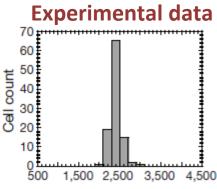
 $x_1^{\rm st} = \beta_1/\alpha$

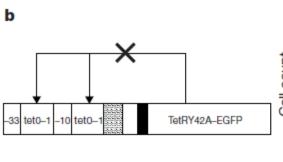
$$x_2^{\rm st} = \sqrt{k\beta_2/\alpha}$$

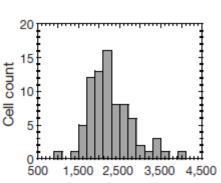
A self repressed gene: robustness

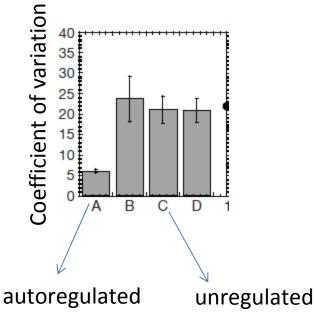


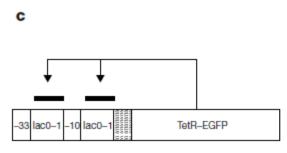


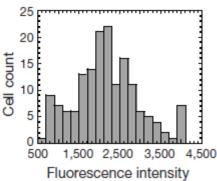












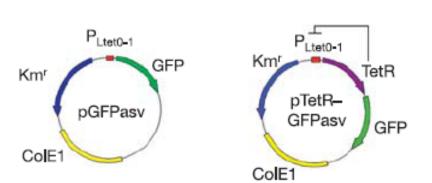
(arbitrary units)

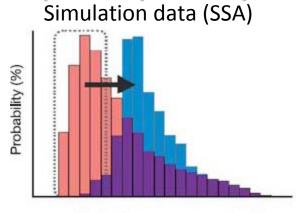
Becskei and Serrano, Nature 2000:

Negative autoregulation decreases noise

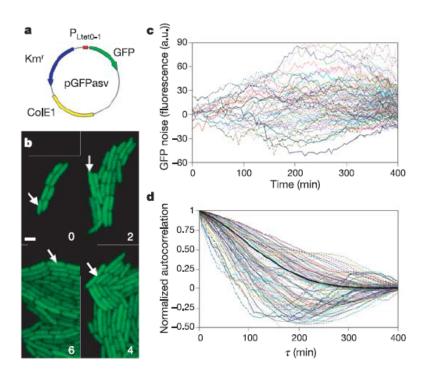


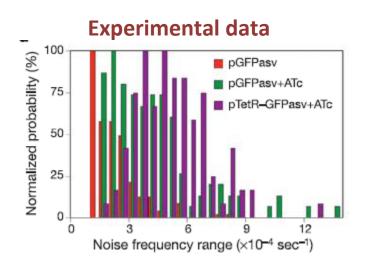
Self repressed gene: Frequency analysis





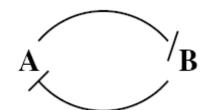
Noise frequency range (sec-1)





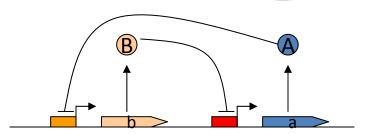
Austin et al. Nature 2006: Negative autoregulation shifts frequency content to high frequency

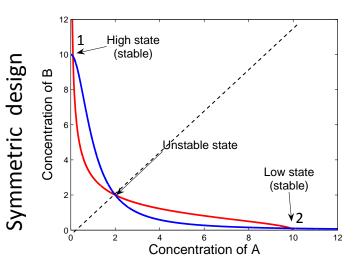
Toggle switch

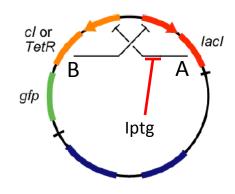


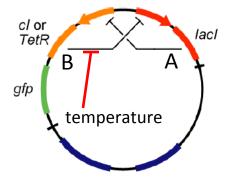
$$\frac{dA}{dt} = \frac{\beta}{1 + (B/K_1)^n} - \alpha_1 A$$

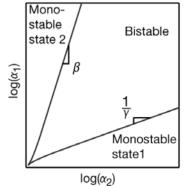
$$\frac{dB}{dt} = \frac{\gamma}{1 + (A/K_2)^n} - \alpha_2 B$$

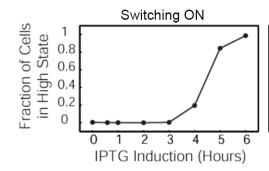


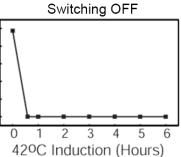




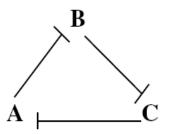


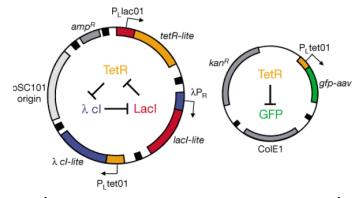




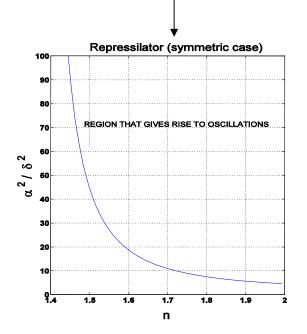


Loop oscillators: The repressilator





(Hastings 1977, Mallet-Paret 1990)



$$f(p) = \frac{\alpha^2}{1 + p^n}$$

$$\frac{dr_A}{dt} = f(C) - \delta_r r_A$$

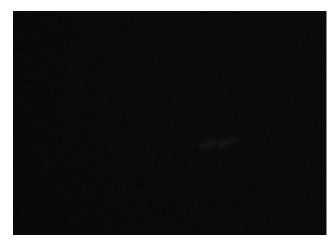
$$\frac{dA}{dt} = r_A - \delta A$$

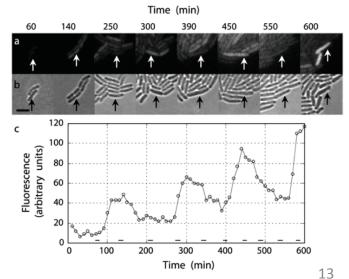
$$\frac{dr_B}{dt} = f(A) - \delta_r r_B$$

$$\frac{dB}{dt} = r_B - \delta B$$

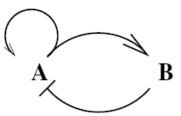
$$\frac{dr_C}{dt} = f(B) - \delta_r r_C$$

$$\frac{dC}{dt} = r_C - \delta C$$





Relaxation oscillators: Atkinson et al. clock



$$\frac{dr_A}{dt} = -\frac{\delta_1}{\epsilon} r_A + F_1(A, B) \qquad F_1(A, B) = \frac{K_1 A^n + K_{A0}}{1 + \gamma_1 A^n + \gamma_2 B^m}$$

$$\frac{dA}{dt} = \nu(-\delta_A A + \frac{k_1}{\epsilon} r_A)$$

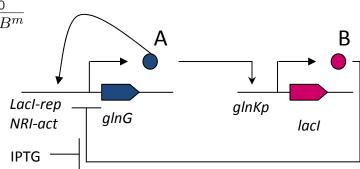
$$\frac{dr_B}{dt} = -\frac{\delta_2}{\epsilon} r_B + F_2(A)$$

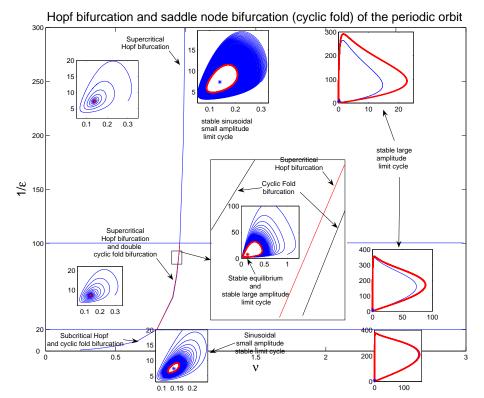
$$\frac{dB}{dt} = -\delta_B B + \frac{k_2}{\epsilon} r_B,$$

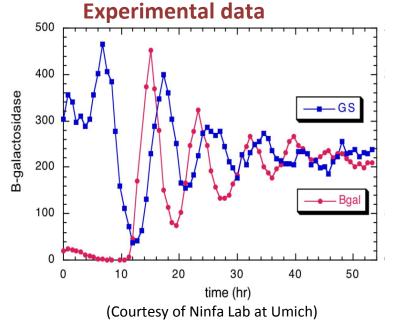
$$F_2(A) = \frac{K_2 A^n + K_{B0}}{1 + \gamma_3 A^n}.$$
Lace the properties of the propertie

$$F_1(A,B) = \frac{K_1 A^n + K_{A0}}{1 + \gamma_1 A^n + \gamma_2 B^m}$$

$$F_2(A) = \frac{K_2 A^n + K_{B0}}{1 + \gamma_3 A^n}.$$







(Cell population measurements)

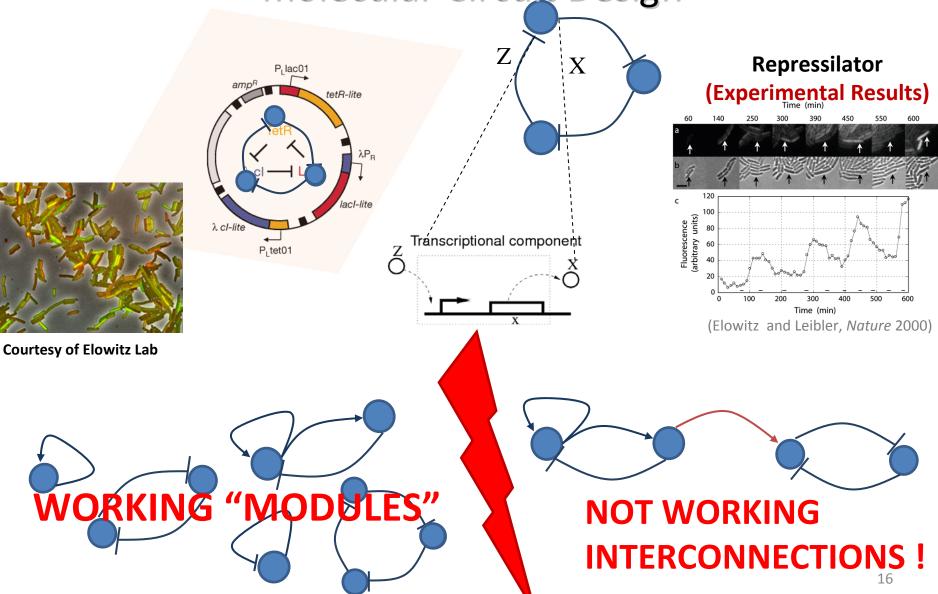
Atkinson et al. Cell 2003

Part 2

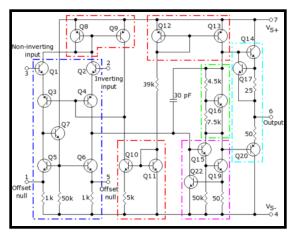
The challenge of composing modules together

- Retroactivity phenomenon and its modeling
- Insulation devices
- Implementation examples

Synthetic Biology: Enabling Technology for Biomolecular Circuit Design



Modularity: A fundamental property

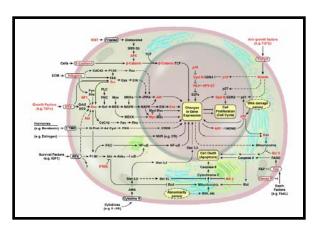


Internal circuitry of an OPAMP: It is composed of well defined modules

Modularity guarantees that the input/output behavior of a component (a module) does not change upon interconnection.

<u>Electronics and Control Systems Engineering</u> rely on modularity to predict the behavior of a complex network by the behavior of the composing subsystems.

Result: Computers, Videos, cell phones...



The Emergent integrated circuit of the cell [Hanahan & Weinberg (2000)]

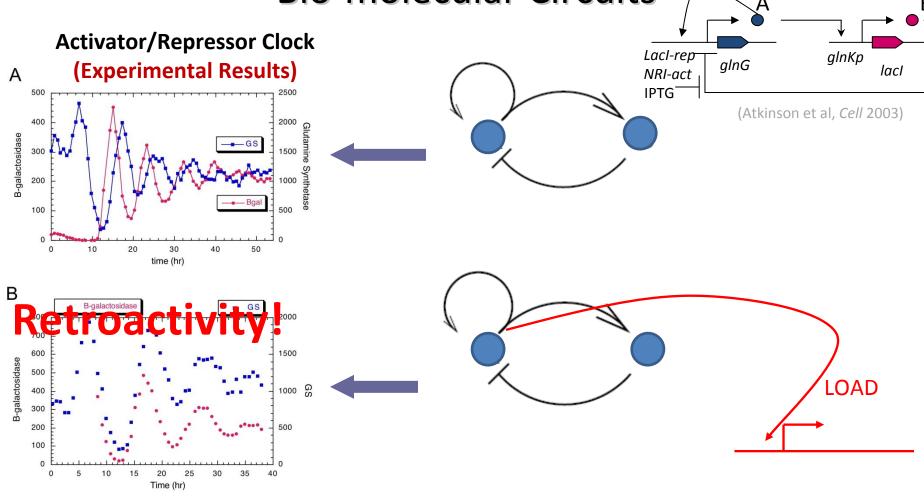
Functional modules seem to recur also in <u>biological</u> <u>networks</u> (e.g. Alon (2007)). But...

But can they be interconnected and still maintain their behavior unchanged?

If not, what mechanism can be used to interconnect modules without altering their behavior?

Does nature already employ such mechanisms?

Modularity is **not** a Natural Property of Bio-molecular Circuits



Courtesy of Ninfa Lab at Umich

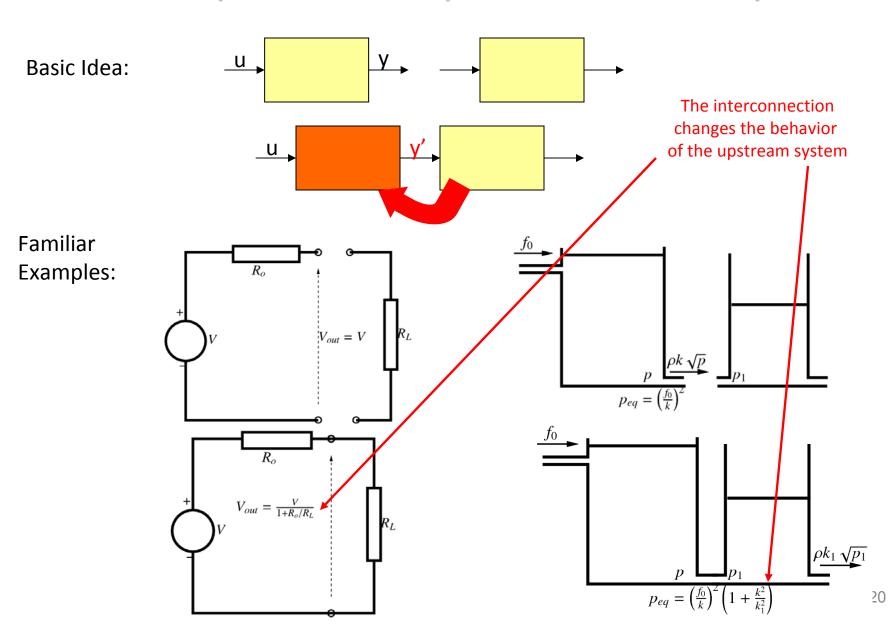
How do we model these effects? How do we prevent them?

Part 2

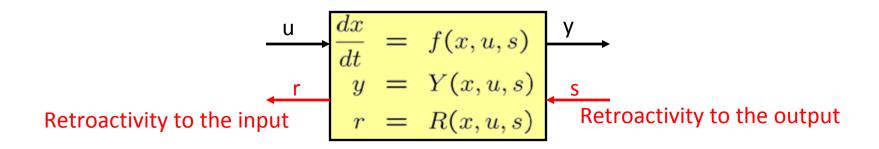
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A systems theory with retroactivity

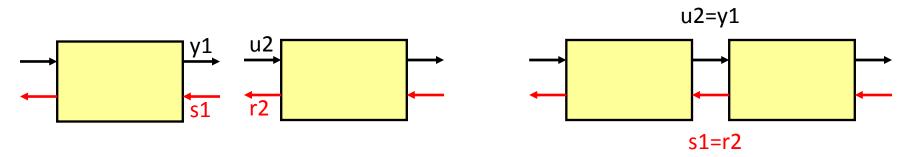


A systems theory with retroactivity

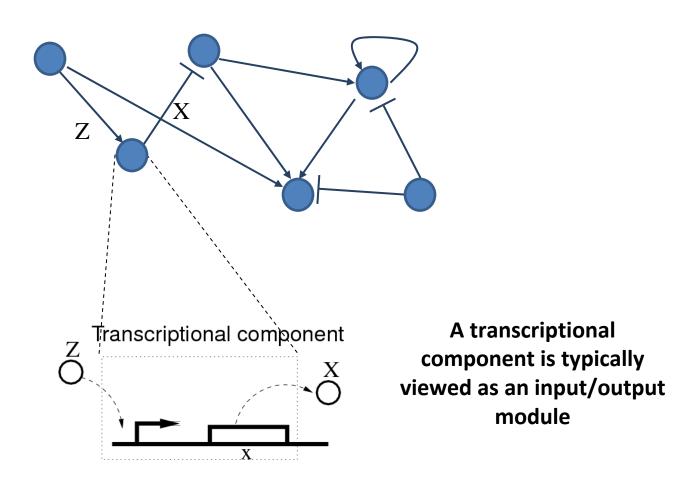


<u>Def:</u> The I/O model of the **isolated system** is obtained when s=0 and when r is not an additional output

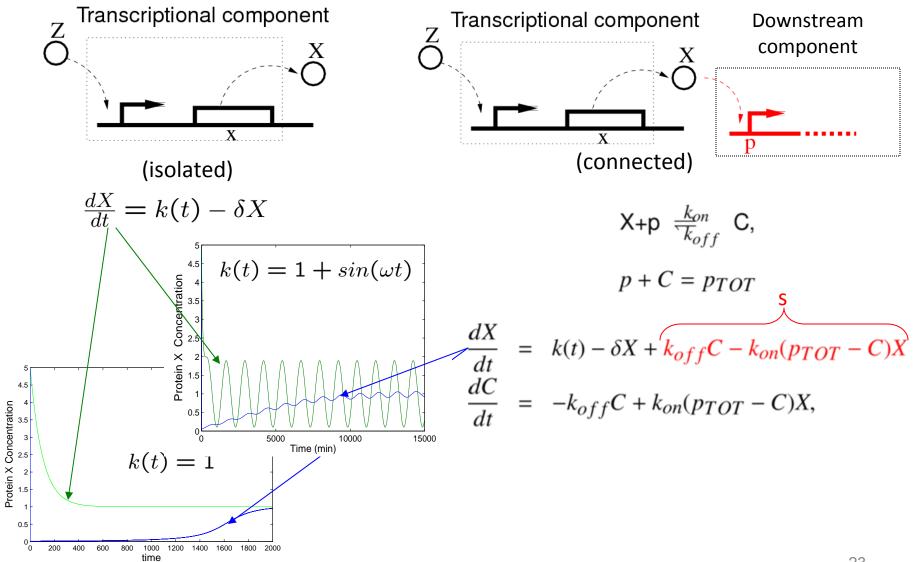
The interconnection of two systems is possible only when the internal state variable sets are disjoint:



Gene regulatory circuitry: A network of transcriptional modules



Retroactivity in transcriptional networks has dramatic effects on the dynamics



Measure of the retroactivity

We seek to quantify the difference in the dynamics of the state X between the connected and isolated system

$$\frac{dX}{dt} = k(t) - \delta X$$

$$\frac{dX}{dt} = k(t) - \delta X$$

$$\frac{dC}{dt} = -k_{off}C + k_{on}(p_{TOT} - C)X$$

$$\frac{dC}{dt} = -k_{off}C + k_{on}(p_{TOT} - C)X,$$

To compare the X dynamics we seek a 1D approximation for the connected system:

$$\frac{d\bar{X}}{dt} = k(t) - \delta\bar{X} + \overline{s}$$

Measure of retroactivity will be given by \overline{s}

Calculation of s

We exploit the time-scale separation between the output X dynamics and the dynamics of the input stage of the downstream component

$$\frac{dX}{dt} = k(t) - \delta X + k_{off}C - k_{on}(p_{TOT} - C)X$$

$$\frac{dC}{dt} = -k_{off}C + k_{on}(p_{TOT} - C)X,$$

$$\epsilon = \delta/k_{off} \quad y = X + C$$

$$\epsilon = 0 \Rightarrow C = \gamma(y) \text{ (slow manifold)}$$

$$\epsilon \frac{dC}{dt} = -\delta C + \frac{\delta}{k_d}(p_{TOT} - C)(y - C),$$

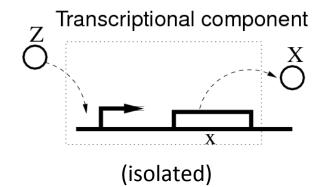
$$\epsilon = \delta/k_{off}$$

$$\epsilon = \delta/k_$$

$$\frac{d\gamma(y)}{dy} = \frac{1}{1 + \frac{(1 + X/k_d)^2}{p_{TOT}/k_d}} =: \mathcal{R}(X)$$
 The value of the retroactivity measure for the interconnection through transcriptional regulation

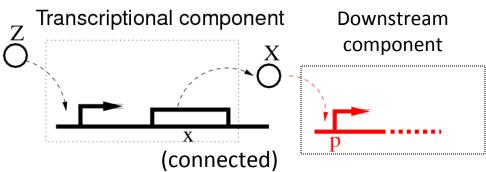
The value of the retroactivity

Effect of R(X) on the dynamics

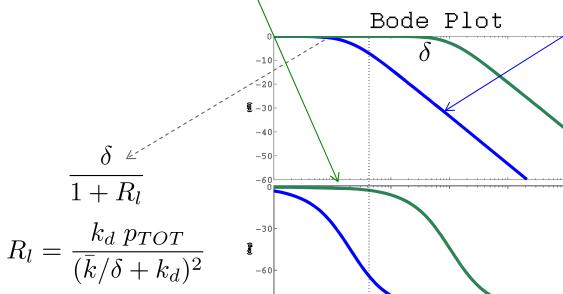


$$\frac{dX}{dt} = k(t) - \delta X$$

 $k(t) = \bar{k} + A_0 \sin(\omega t)$



$$\frac{dX}{dt} = (k(t) - \delta X) (1 - \mathcal{R}(X))$$



90<u>L</u> 10⁻⁵

10-3

10-1

10°

 10^{-4}

Retroactivity shifts the poles of the transfer function of the linearized system toward low frequency

Is this finding experimentally relevant?

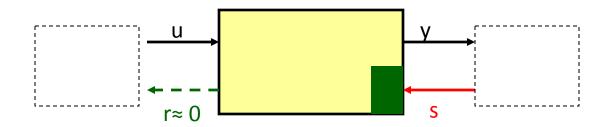
Part 2

The challenge of composing modules together

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Dealing with retroactivity: Insulation devices

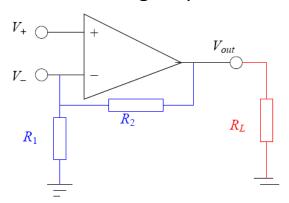
In general, we cannot <u>design</u> the downstream system (the load) such that it has low retroactivity. But, we can design an <u>insulation system</u> to be placed between the upstream and downstream systems.



- 1. The retroactivity to the input is approx zero: r≈0
- 2. The retroactivity to the output s is attenuated
- 3. The output is proportional to the input: y=c u

Attenuation of the retroactivity to the output "s": Large feedback and large amplification

Non-inverting amplifier:



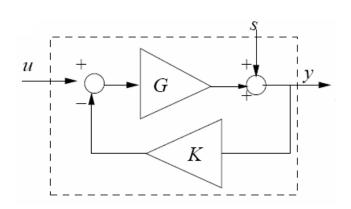
$$V_{out} = G(V_+ - V_-)$$

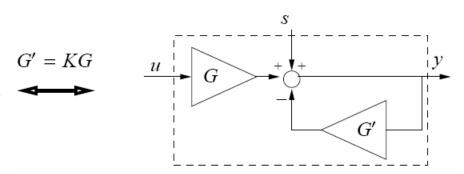
For G large enough:

$$V_{out} = \frac{V_{+}}{K}, K = \frac{R_1}{R_1 + R_2}$$

Conceptually:

$$y = G(u - Ky) + s \implies y = u \frac{G}{1 + KG} + \frac{s}{1 + KG}$$





Attenuation of the retroactivity to the output "s" in the transcriptional component

Connected system approximated dynamics

Isolated system

$$\frac{d\bar{X}}{dt} = (k(t) - \delta \bar{X})(1 - \frac{d\gamma(\bar{y})}{d\bar{y}})$$

$$\frac{dX}{dt} = k(t) - \delta X$$

Apply large input amplification G and large output feedback G'

$$\frac{d\bar{X}}{dt} = (Gk(t) - G'\bar{X} - \delta\bar{X})(1 - \frac{d\gamma(\bar{y})}{d\bar{y}}) \qquad \frac{dX}{dt} = Gk(t) - G'X - \delta X$$

$$\frac{dX}{dt} = Gk(t) - G'X - \delta X$$

Lemma. Consider the system

$$\frac{dX}{dt} = G(t)(u(t) - KX)$$

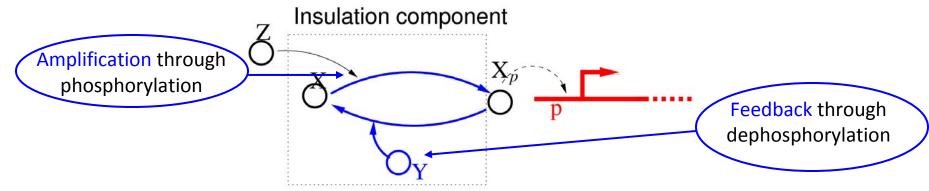
in which $G(t) \ge G_0 > 0$ and $|u'(t)| \le V$ uniformly in t. Then,

$$|X(t) - \frac{u(t)}{K}| \le \exp(-tG_0K)|X(0) - \frac{u(0)}{K}| + \frac{V}{G_0K^2}.$$

Let G'=GK, then as G grows the signals $\bar{X}(t)$ and X(t) become close to each other

How do we realize a large input amplification and a large negative feedback?

A phosphorylation-based design for a biomolecular insulation device



$$Z + X \stackrel{k_1}{\rightarrow} Z + X_p$$
, $Y + X_p \stackrel{k_2}{\rightarrow} Y + X$

with conservation law $X + X_p + C = X_{TOT}$

$$\frac{dX_p}{dt} = k_1 X_{TOT} Z(t) \left(1 - \frac{X_p}{X_{TOT}} - \boxed{\frac{C}{X_{TOT}}} \right) - k_2 Y X_p + \boxed{k_{off} C - k_{on} X_p (p_{TOT} - C)}$$

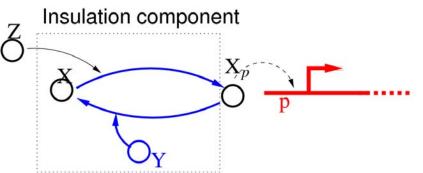
$$\frac{dC}{dt} = -k_{off} C + k_{on} X_p (p_{TOT} - C)$$

$$\epsilon := k_1 X_{TOT} / k_{off}$$
, $p_{TOT} \ll X_{TOT}$, $X_p \ll X_{TOT}$

$$\frac{d\bar{X}_p}{dt} = (GZ(t) - G'\bar{X}_p)(1 - \mathcal{R}(\bar{X}_p)) \qquad G = k_1 X_{TOT}, \ G' = k_2 Y$$

As G and G' grow, $\bar{X}_p(t)$ tends to $X_p(t)$ given by the isolated system

Simulation results for the pho/depho insulation device



Slow time-scale

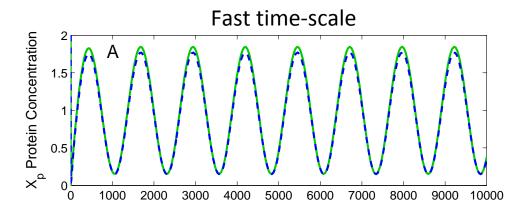
1.5

A

O

1.000 2000 3000 4000 5000 6000 7000 8000 9000 10000

The fast time-scale of the phosphorylation cycle allows to reach insensitivity to very large loads (p=100)



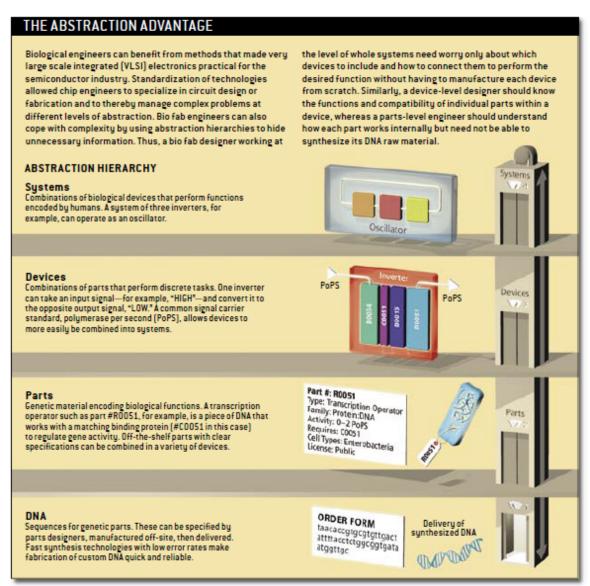
Xp for the isolated system

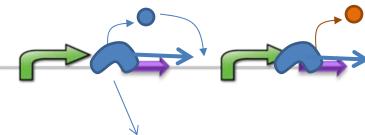
Xp for the connected system

Part 4

Fabrication Technology

Parts, Devices, Systems



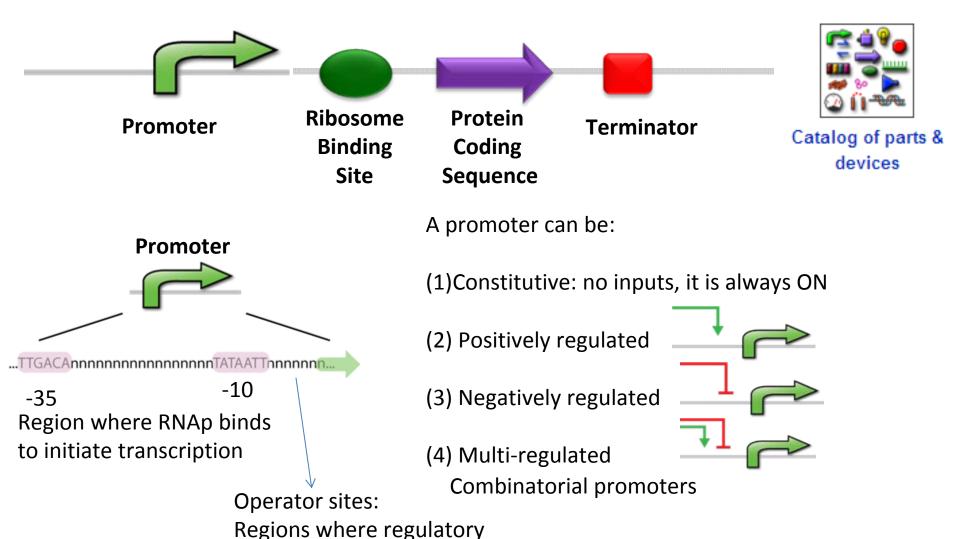


RnaPolymerase

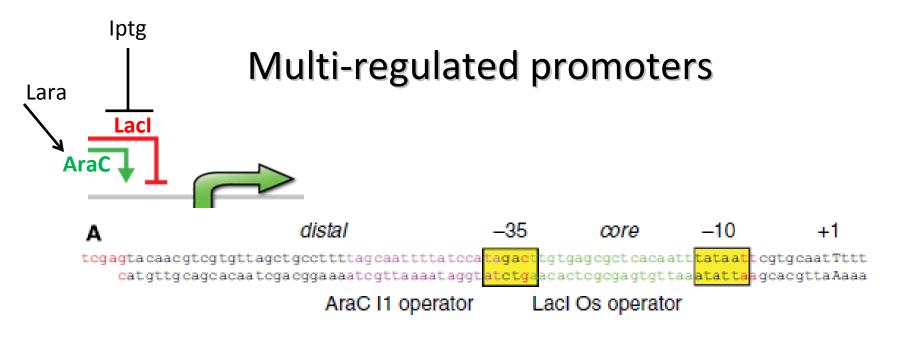
PoPS= Polymerase Per Second

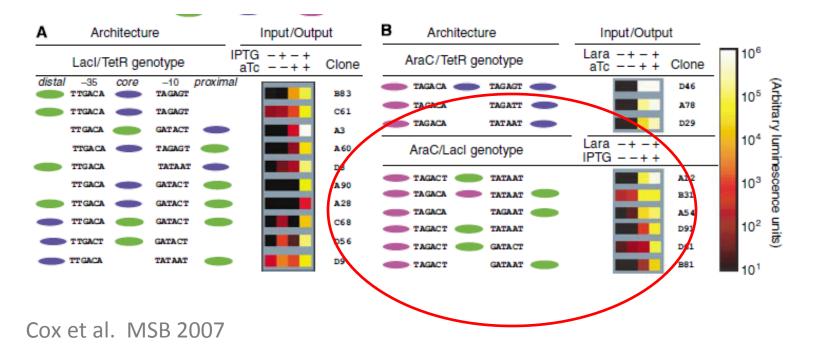
David Baker,
George Church,
Jim Collins,
Drew Endy,
Joseph Jacobson,
Jay Keasling,
Paul Modrich,
Christina Smolke
and Ron Weiss
Scientific American
2006

Library of standard Parts: Biobricks

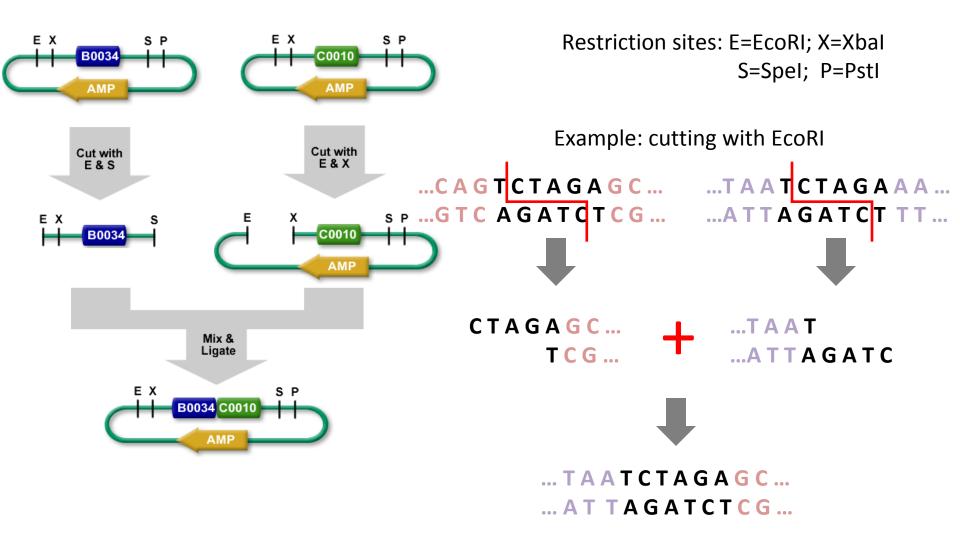


proteins bind



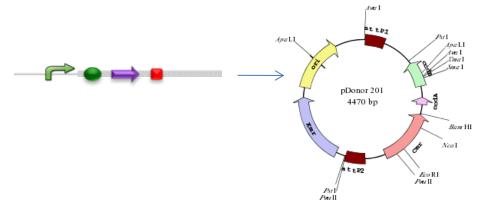


Biobrick standard assembly



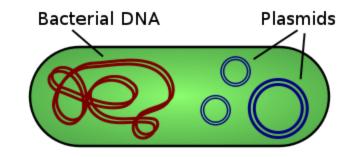
Any two biobricks can be combined in any order to form a new biobrick \rightarrow Modular assembly

In vivo Implementation



Plasmid: circular portion of DNA separate from the chromosomal DNA, which is capable of replicating independently of the chromosomal DNA

Transformation: process of inserting the plasmid within in the cell. It occurs by rendering the cells *competent*, (external membrane becomes permeable).

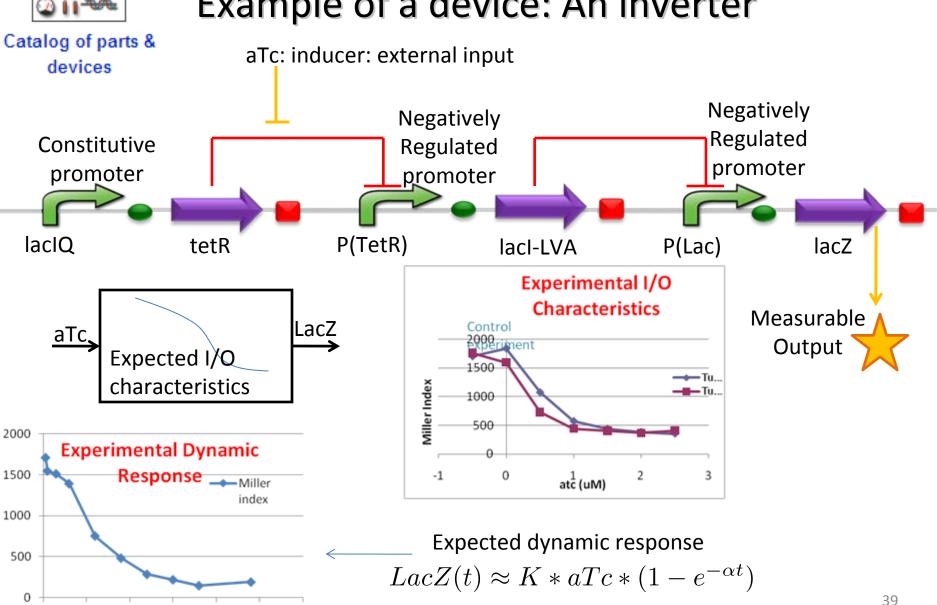


Inducers: signaling molecules that bind to repressors and disable them. The net effect is to start transcription. They can be added to the cell population to provide <u>input forcing</u> to the circuit

Reporter Genes: express proteins that produce an easily observable phenotype, for example, green fluorescent proteins, which causes cells to fluoresce green under blue Light. They are inserted after a gene of interest to measure its production rate. They Provide an easily measurable output to the system.



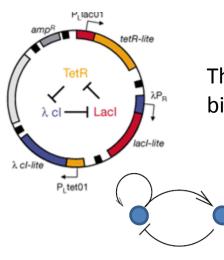
Example of a device: An inverter



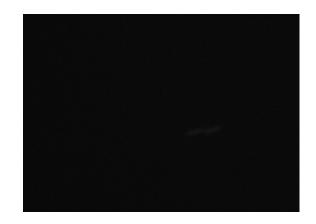


Summary

The ability of fabricating bio-molecular circuits has far reaching applications: medical and energy are two examples

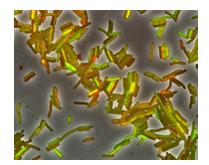


The technology for fabricating synthetic bio-molecular circuits in cells is available



Modular design is a grand challenge

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Stochastic behavior is an integral part of these systems and must be explicitly considered for design

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Wrap-up: Challenges and Opportunities for Control Theory

- <u>Stochastic behavior</u>: Bio-molecular systems are intrinsically stochastic
- <u>Complexity</u>: large state spaces, large number of parameters
- Modularity: tool for analysis/design, input/output descriptions
- <u>Uncertainty</u>: functional systems from uncertain components?
- Redundancy: a way to obtain robustness?
- <u>Crosstalk</u>: any circuit working in the cell environment uses up cellular resources
- …learning to communicate with biologists