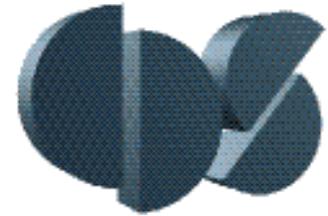


Feedback and Control in Biological Circuit Design



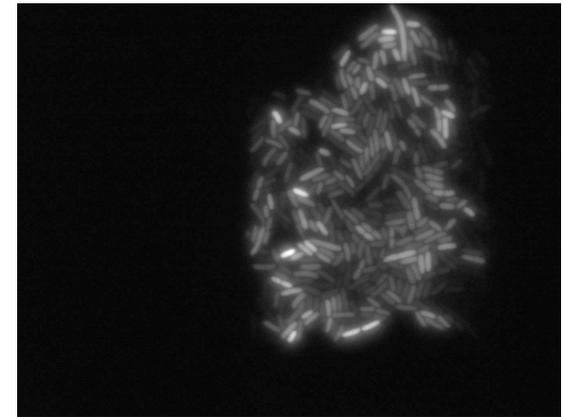
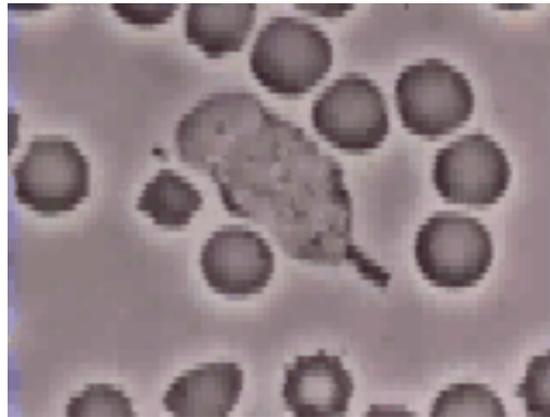
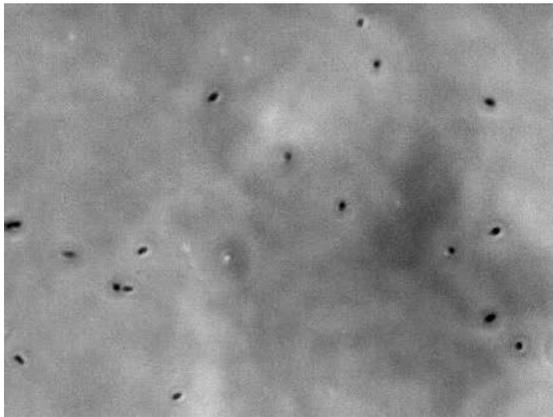
Richard M. Murray

Control and Dynamical Systems

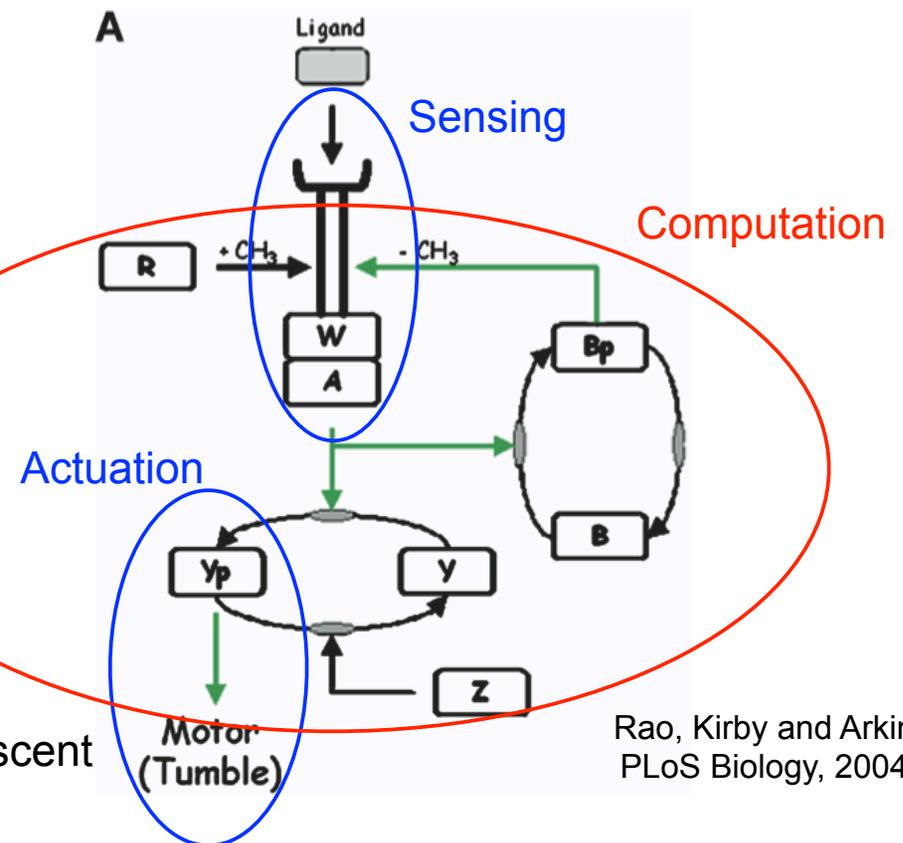
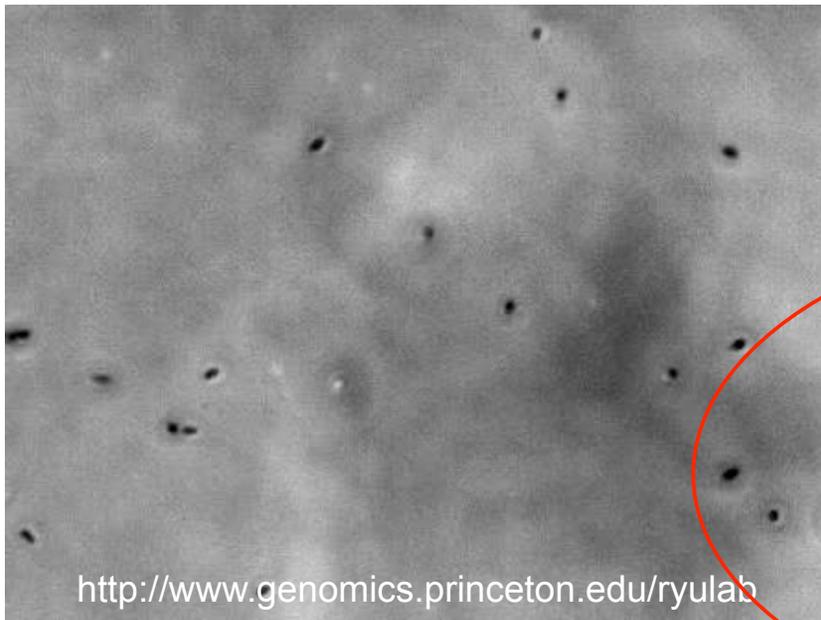
California Institute of Technology

Mary Dunlop (ME) Elisa Franco (CDS) Johan Ugander (Lund)

Dionysios Barmpoutis (CNS) Ophelia Venturelli (BMB)



Example: Chemotaxis



Rao, Kirby and Arkin
PLoS Biology, 2004

Implements key principles of feedback

- Dynamics allow exploration, gradient ascent
- Robust with respect to nutrient levels (adaptation via methylation of receptors)

Can we do better?

- Make use of modular sensors and actuators
- Modify dynamics to provide different types of behavior, robustness properties, ...

Design of Biomolecular Feedback Systems

I. Biological circuit design (synthetic biology)

II. System Identification in Cells

- Regulatory activity revealed by dynamic correlations in gene expression noise (Mary Dunlop [UC Berkeley/U. Vermont])
- Joint work with Michael Elowitz

III. Robustness to Uncertainty

- *In vitro* rate regulators (Elisa Franco, Fei Chen)
- Joint work with Erik Winfree

IV. Design of Dynamics

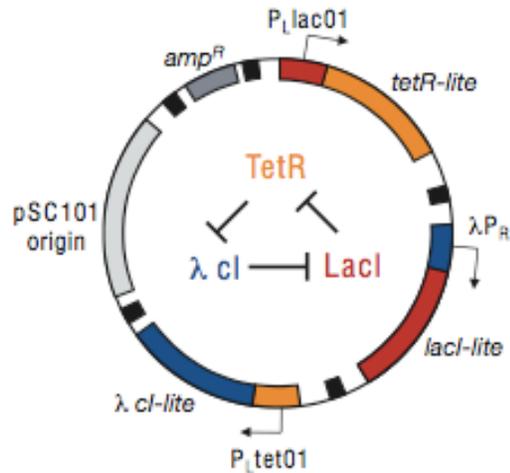
- Using time-delays to tune *in vivo* oscillators (Johan Ugander, Arthur Prindle)

V. Control Design (?)

- Some thoughts on a “framework” for design of biochemical feedback systems



Biological Circuit Design (Synthetic Biology)

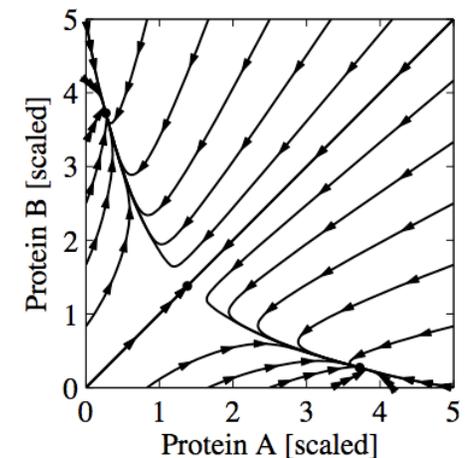
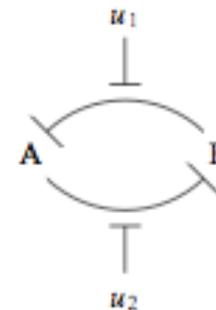


Repressilator (Elowitz & Leibler)

- Ring oscillator with three repressors in a cycle
- Provides oscillations at frequency comparable to cell cycle

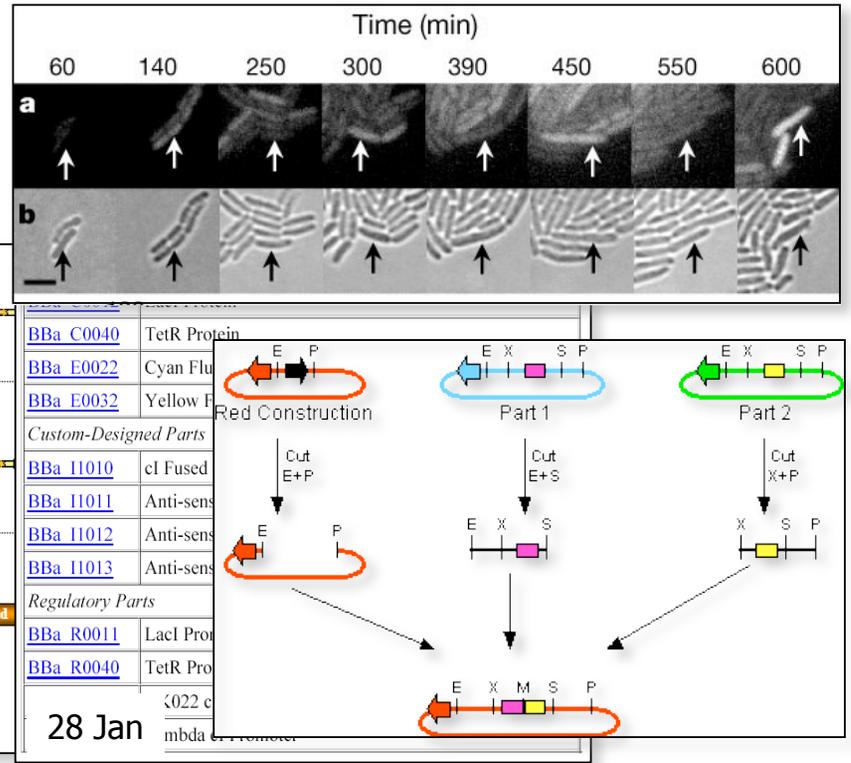
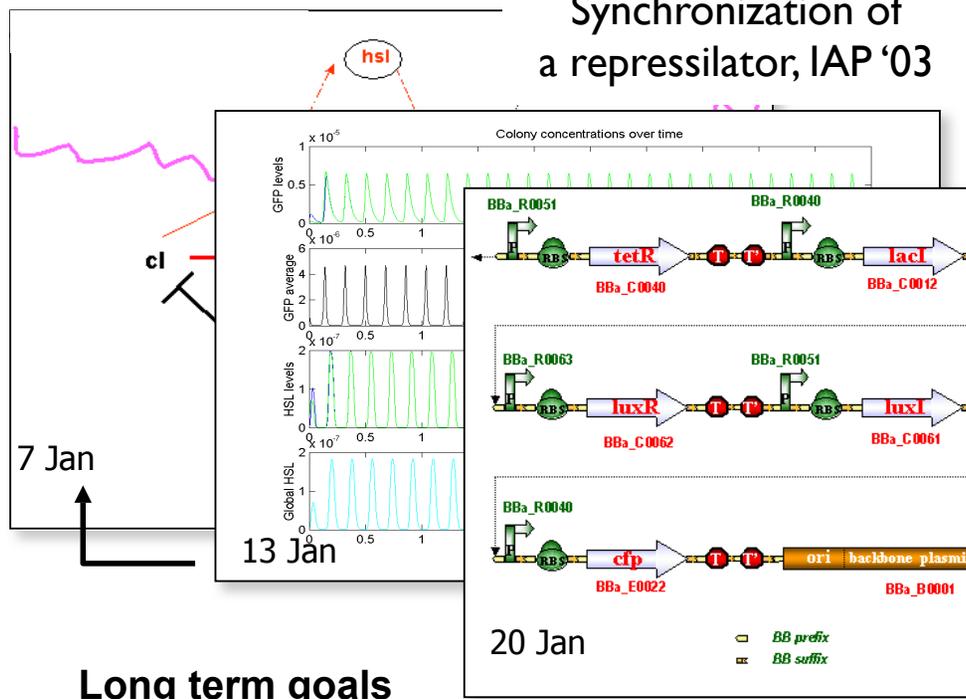
Genetic Switch (Collins and others)

- Interconnect two genes via cross-repression
- Resulting circuit has two states: “(1,0)”, “(0,1)”
- Can analyze robustness, speed of response



Modular Synthetic Biology

MIT Bio-Bricks program



Long term goals

- Better understanding of biological function
- New devices for interfacing with biological systems (diagnosis, medication)
- Novel biological processes: biofuels production, bio-remediation

State of the Art

- DNA synthesis: < \$1 per base pair (simple circuit: 5000 bp), 6-10 weeks delivery time
- Alternative: manual cloning to put together existing components (eg, bio-bricks)

Toward a Control Theory for Synthetic Biology

Differences from traditional systems

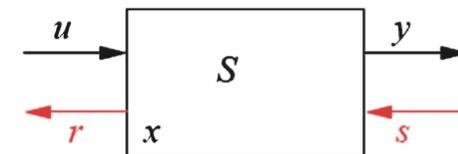
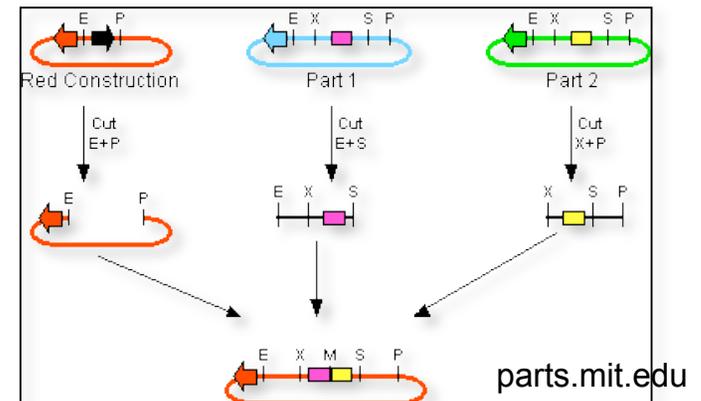
- Complexity - biological systems are much more complicated than engineered systems
- Communications - signal representations are very different (spikes, proteins, etc)
- Uncertainty - very large uncertainty in components; don't match current tools
- Evolvability - mutation, selection, etc

(Engineered) Modularity would be very useful

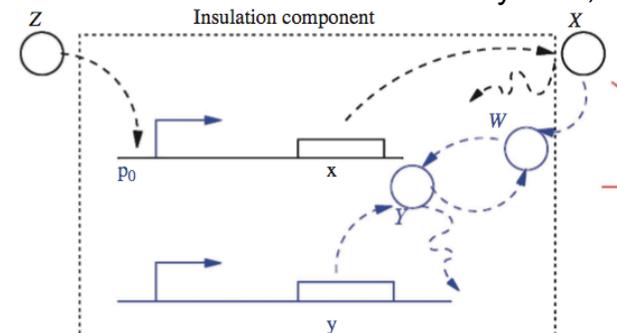
- To build complex systems, we need to be able to isolate subsystems (probably)
- Biobricks: modularity at DNA + device level
- Retroactivity (DDV et al): candidate methods for minimizing effects of loading by downstream devices

Stochasticity and robustness are critical

- Program time-evolving distributions to achieve desired function
- Make use of heterogeneous redundancy to provide robustness (?)



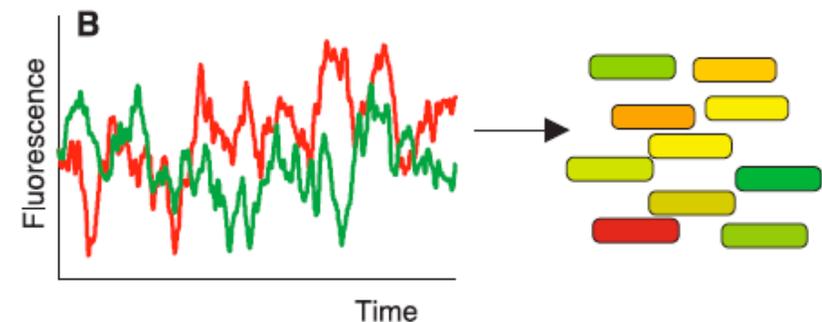
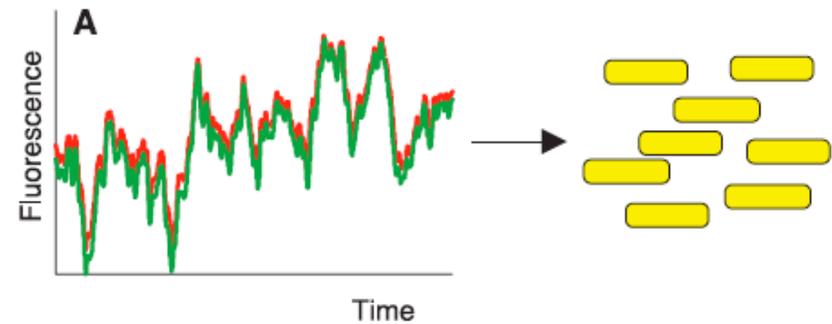
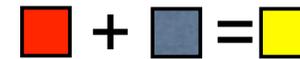
Del Vecchio et al, Mol Sys Bio, 2008



Cell Noise (Elowitz et al, 2002)

Noise in cells

- Experiments by Elowitz, Levine, Siggia, Swain. *Science* 2002
- Put RFP and GFP under identical promoters; *should* get yellow
- Results: get range of colors



Extrinsic Noise:

- global to a single cell, but varies from one cell to the next (e.g. cell volume, plasmid copy number)

Intrinsic Noise:

- inherent stochasticity in gene expression (e.g. what order reactions occur in)

$$\dot{x}_i = E(t) \cdot f_i(x_i, I_i(t))$$

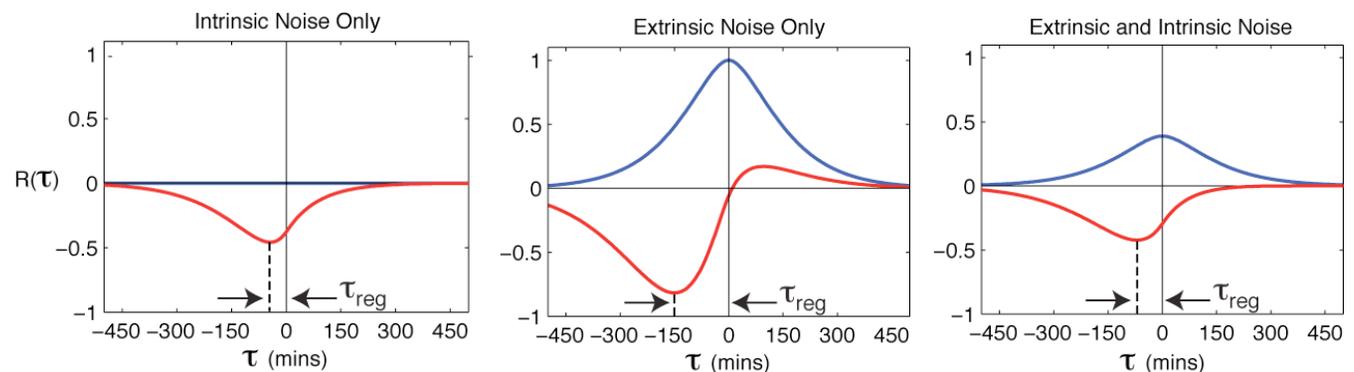
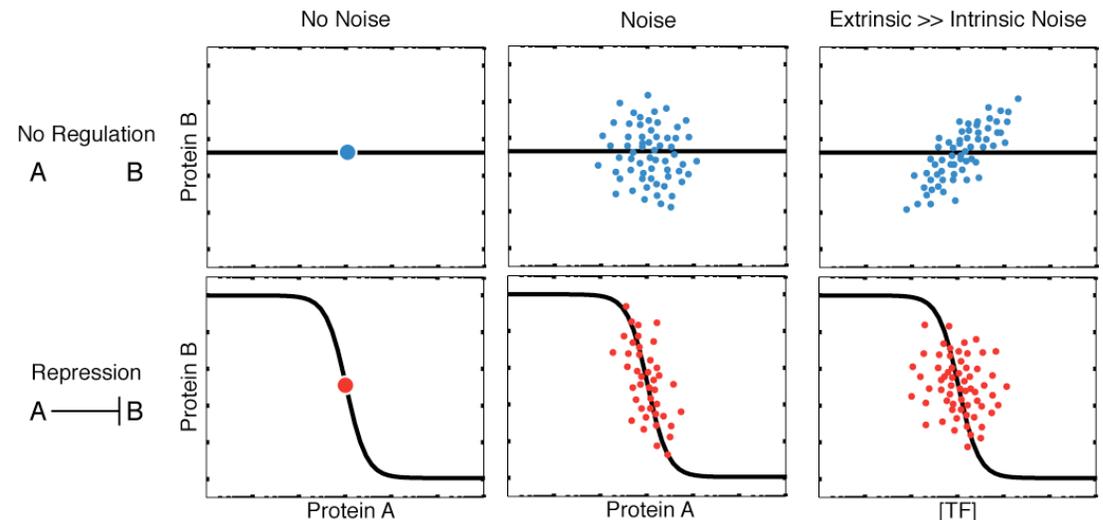
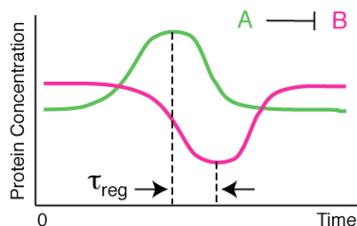
System Identification Using Cell Noise

Traditional systems identification

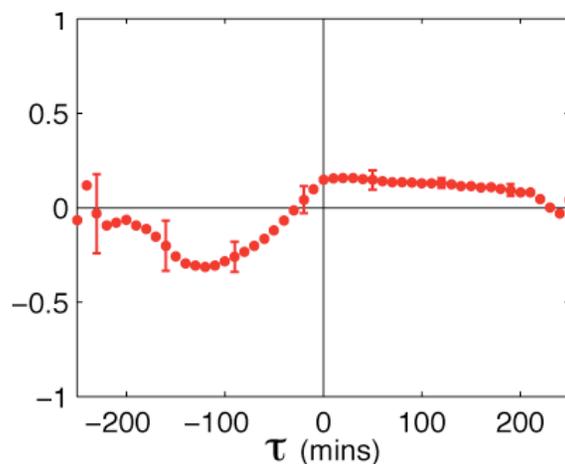
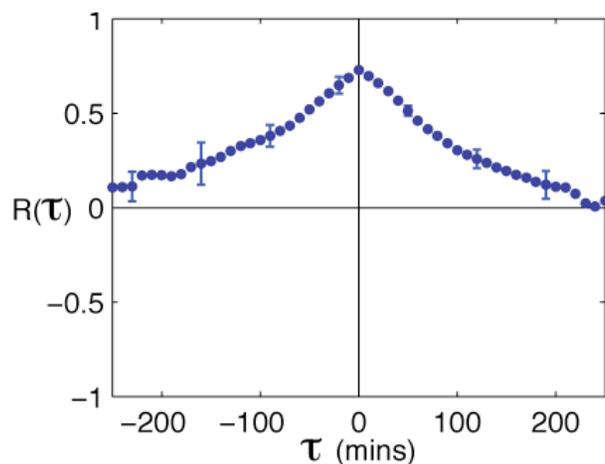
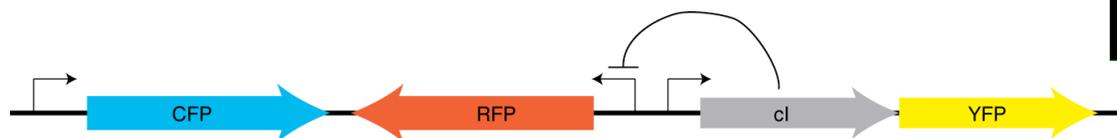
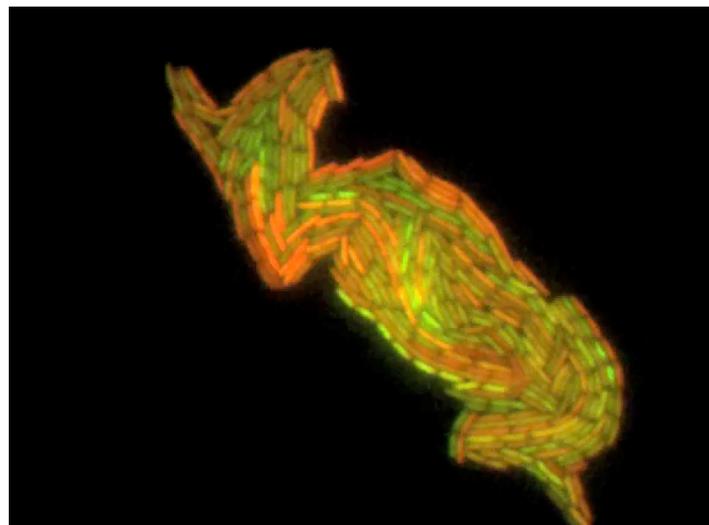
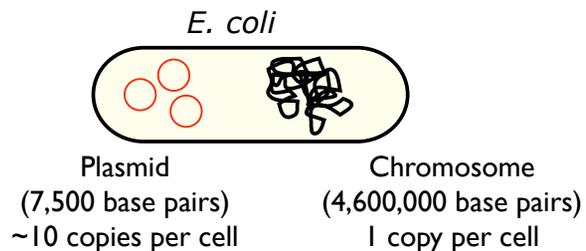
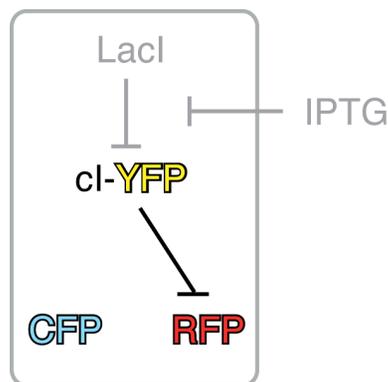
- Engineering: forced response. Difficult to do in in vivo (eg, sinusoids are tricky)
- Biology: gene knockouts; steady state measurements using gene arrays

Idea: use noise as a forcing function

- Steady state distributions are not enough if extrinsic noise is present
- Need to use correlation data instead



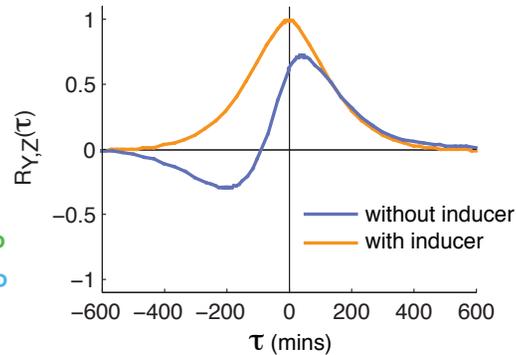
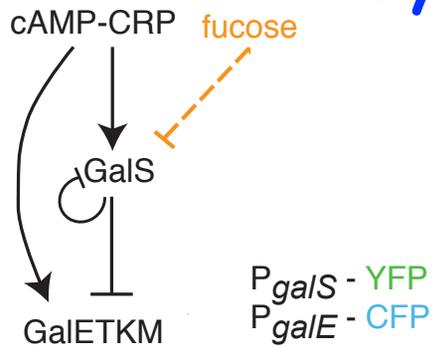
System ID of a Synthetic Circuit (Dunlop, Elowitz & M)



Results to date

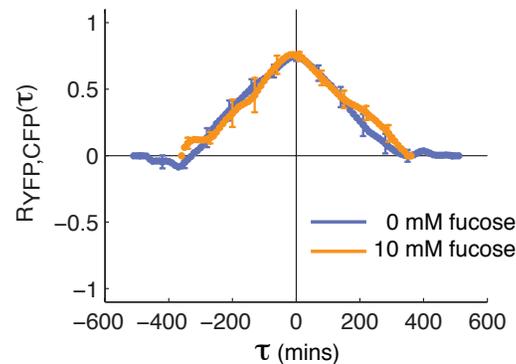
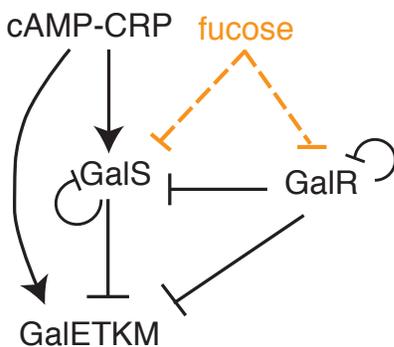
- Synthetic circuit demonstrates viability of approach
- Implemented on natural circuit (using promoter fusion)

System ID of an in vivo circuit



Galactose regulation in *E. coli*

- GalE regulated by CRP via a feedforward loop
- GalR represses feedforward loop when fucose is present
- Promoter fusions measure GalS and GalE concentrations

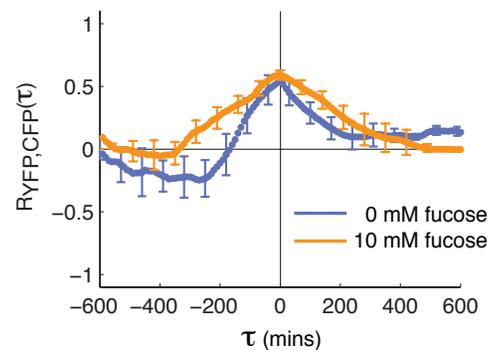
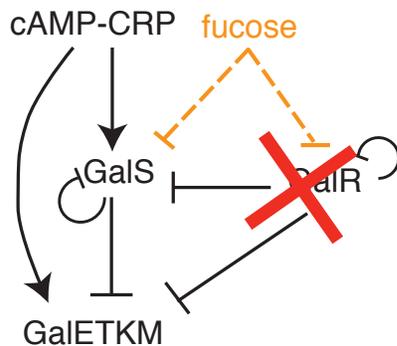


System ID shows FFL is not active

- Addition of fucose shows no change in correlations => GalS is not actively regulating GalE

Hypothesis: GalR repression dominant

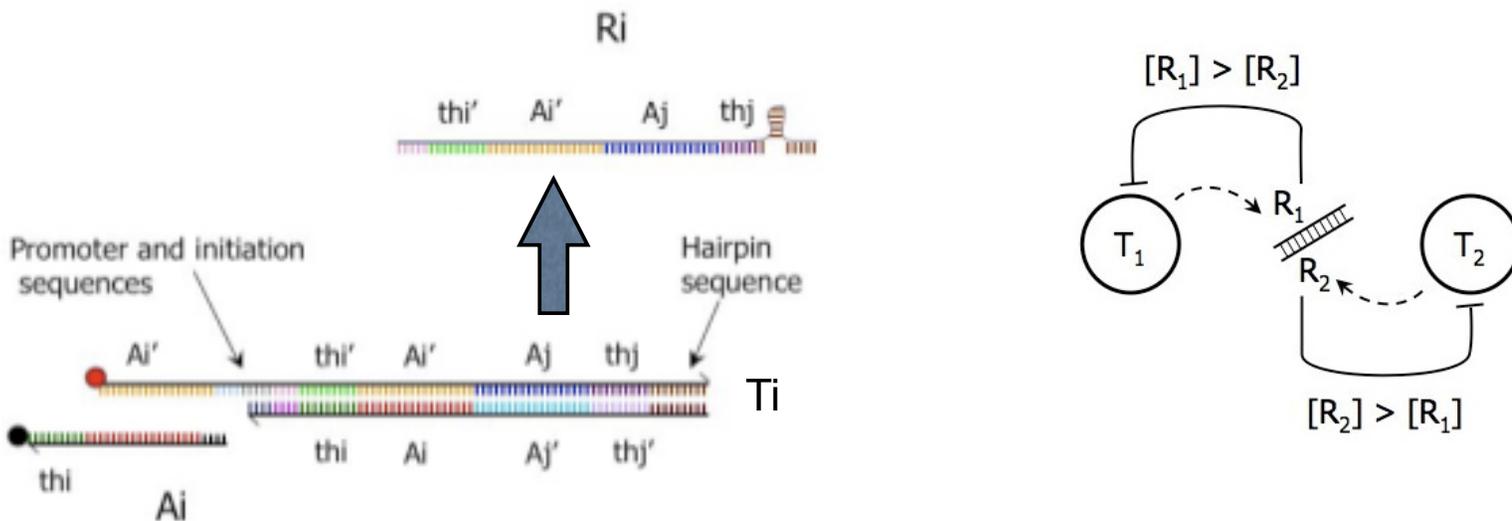
- If repression by GalR is large, GalS is always “off” => no connection
- Removal of GalR recovers expected correlations



In Vitro Rate Regulator (Franco, Winfree & M)

Idea for a circuit: produce two chemicals at same rates

- Common operation for metabolic networks - maintain stoichiometry
- Implemented using *in vitro* technology (test tubes instead of cells)



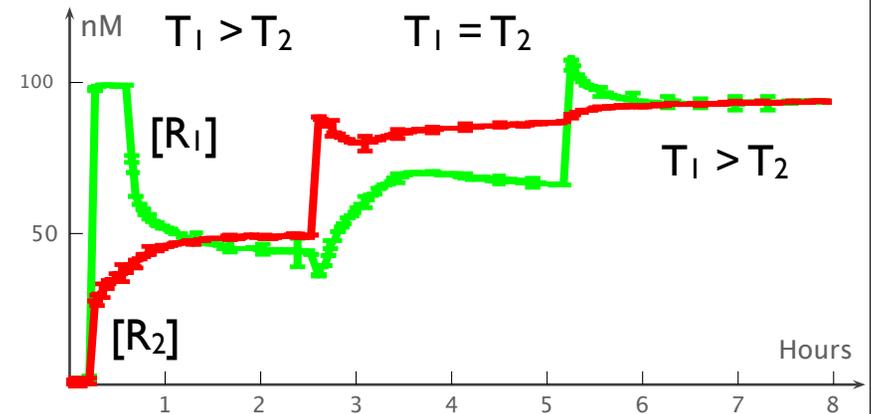
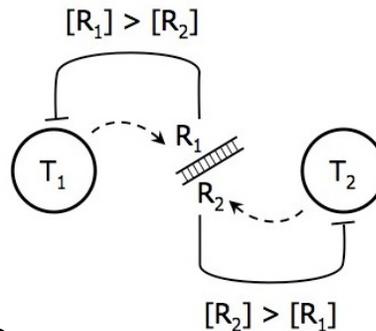
Molecular programming for in vitro systems

- Exploit Watson-Crick base pair binding (A-T, C-G)
- Can “compile” functional specifications into RNA and DNA sequences
- Circuits are biocompatible \Rightarrow some hope of embedding into cells

Rate Regulator Results

In vitro experiments

- Add templates + enzymes to test tube
- Use fluorophors to measure amount of repression

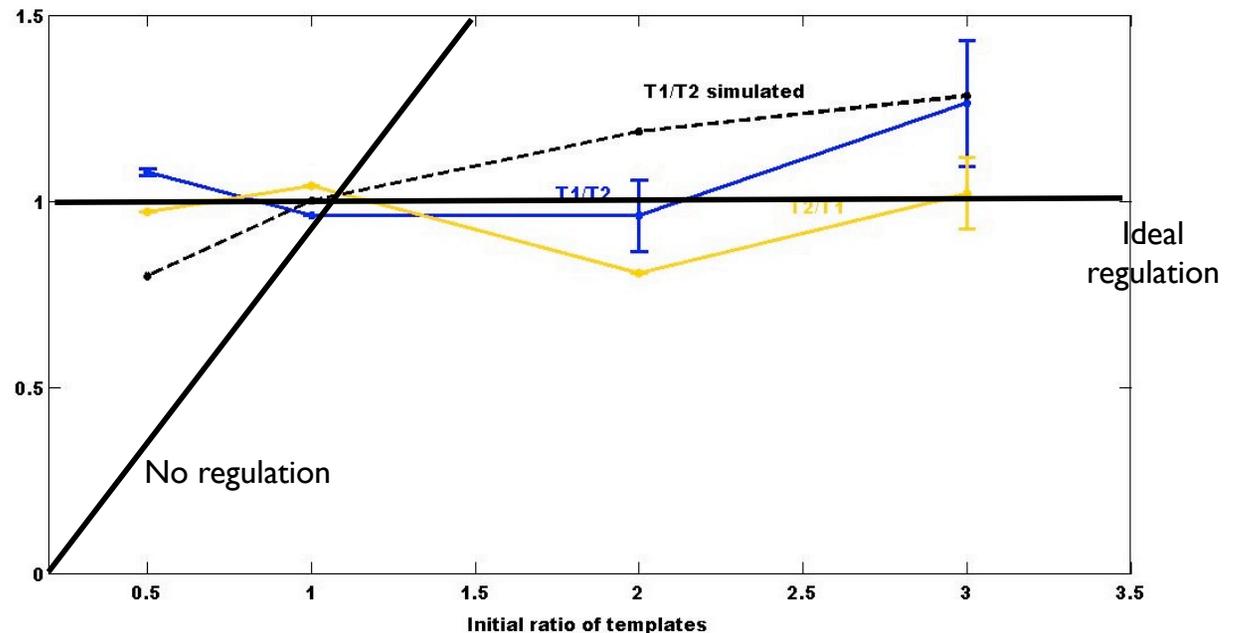


Rate regulator functions correctly

- When T1 is high, get more repression of T1 (to bring R1, R2 into balance)
- Can also use cross activation

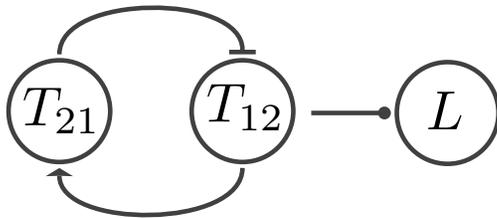
Next steps

- Loading effects
- Sensing/actuation
- Integral feedback (Fei)



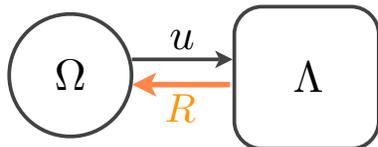
Improving Modularity

Effects of loading

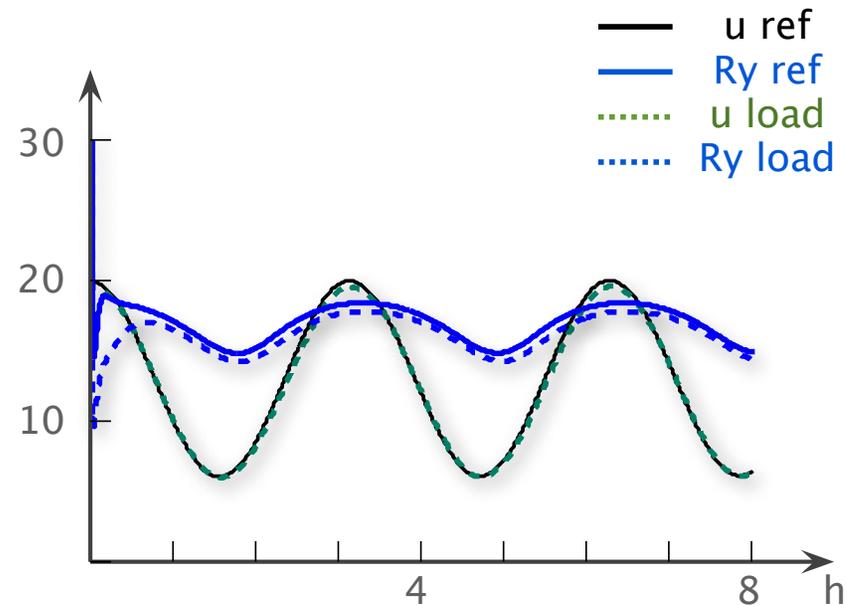
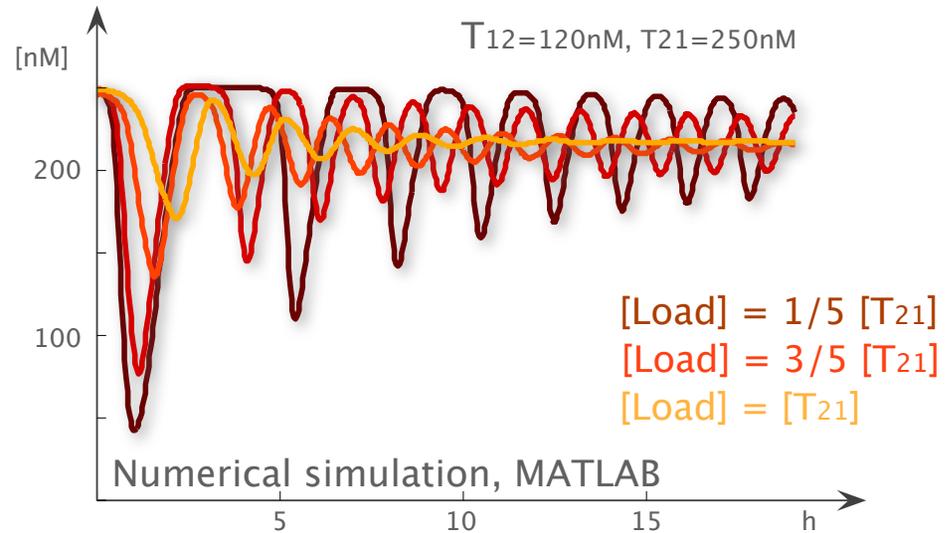
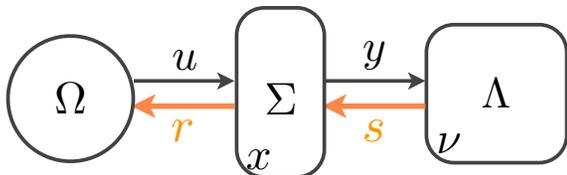


Modeling approach: retroactivity (Del Vecchio and Sontag, 2008)

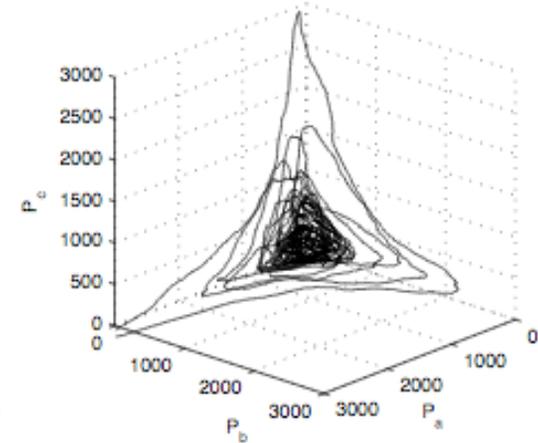
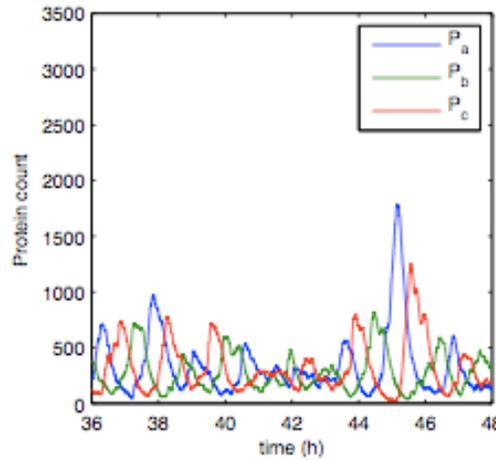
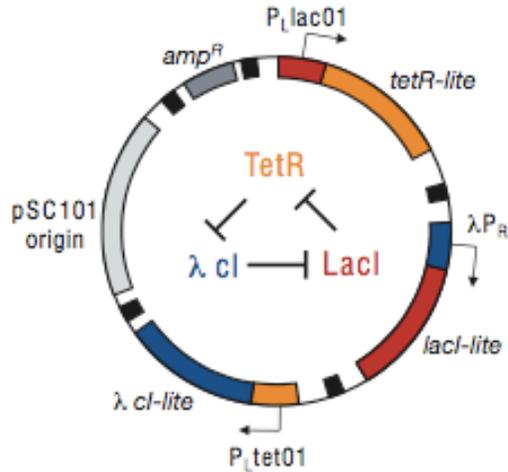
- Keep track of how much downstream load affects circuit



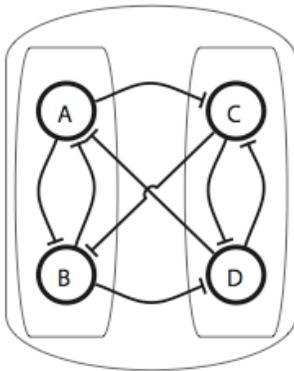
- Use "insulator" to isolate



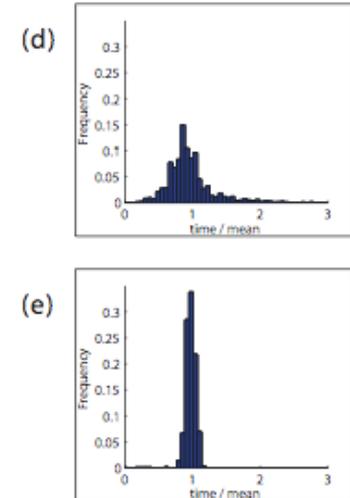
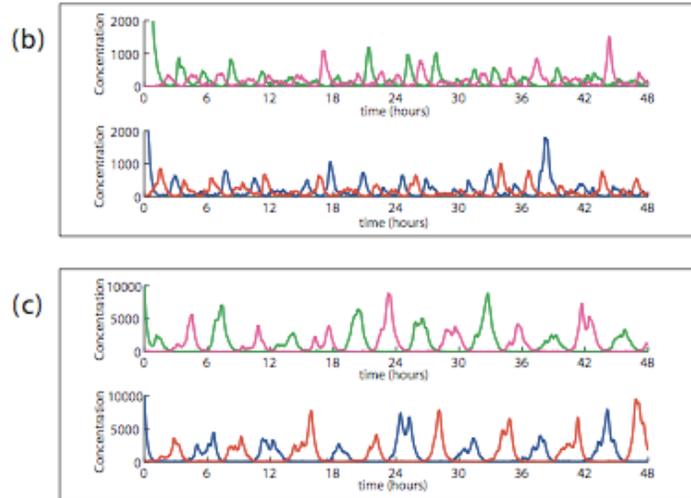
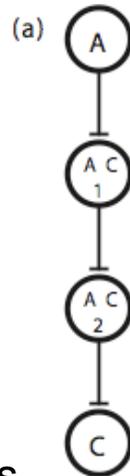
Improving the Performance of Oscillators



Toggelator



- Coupled oscillators
- Add additional “delay” (ACi)

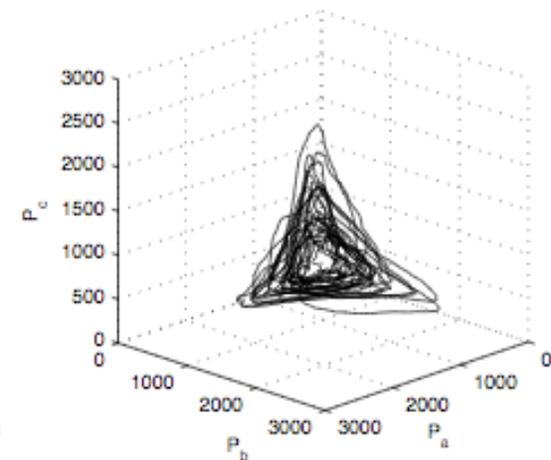
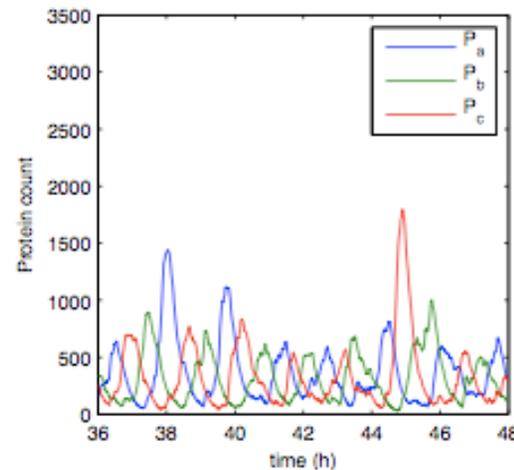
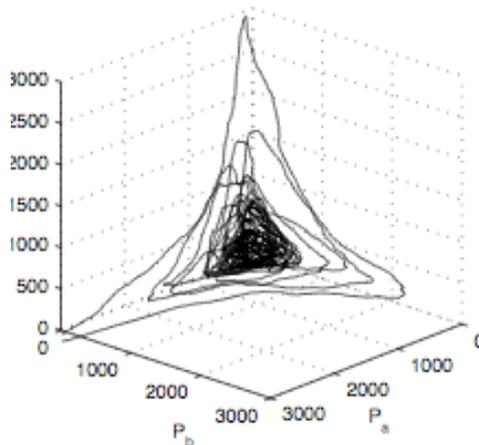


Ugander, Dunlop & M, ACC 07

Improving Oscillator Performance by Adding Delay

Transcription & Translation	Promoter region	Decay
$GeneA_{on}(t) \xrightarrow{k_1} GeneA_{on}(t+\tau_{ps})+M_A(t+\tau_1)$ $GeneA_{off}(t) \xrightarrow{\epsilon \cdot k_1} GeneA_{off}(t+\tau_{ps})+M_A(t+\tau_1)$ $M_A(t) \xrightarrow{k_2} M_A(t+\tau_{rbs})+P_A(t+\tau_2)$	$GeneA_{on}(t)+P_C(t) \xrightarrow{k_3} GeneA_{off}(t)$ $GeneA_{off}(t) \xrightarrow{k_4} GeneA_{on}(t)+P_C(t)$	$P_A(t) \xrightarrow{k_5} \emptyset$ $M_A(t) \xrightarrow{k_6} \emptyset$
$GeneB_{on}(t) \xrightarrow{k_1} GeneB_{on}(t+\tau_{ps})+M_B(t+\tau_1)$ $GeneB_{off}(t) \xrightarrow{\epsilon \cdot k_1} GeneB_{off}(t+\tau_{ps})+M_B(t+\tau_1)$ $M_B(t) \xrightarrow{k_2} M_B(t+\tau_{rbs})+P_B(t+\tau_2)$	$GeneB_{on}(t)+P_A(t) \xrightarrow{k_3} GeneB_{off}(t)$ $GeneB_{off}(t) \xrightarrow{k_4} GeneB_{on}(t)+P_A(t)$	$P_B(t) \xrightarrow{k_5} \emptyset$ $M_B(t) \xrightarrow{k_6} \emptyset$

Junk Delay:

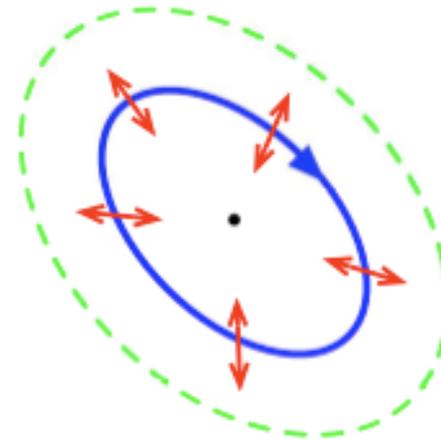


Reduced Order Analysis

$$\dot{P}_A = -\beta P_A(t) + \alpha \left(\frac{1}{1 + P_C(t-\tau)^2} \right) + \alpha_0$$

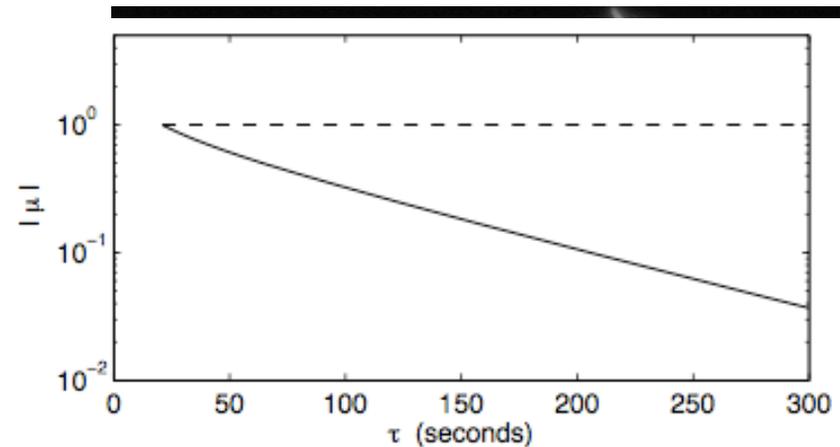
$$\dot{P}_B = -\beta P_B(t) + \alpha \left(\frac{1}{1 + P_A(t-\tau)^2} \right) + \alpha_0$$

$$\dot{P}_C = -\beta P_C(t) + \alpha \left(\frac{1}{1 + P_B(t-\tau)^2} \right) + \alpha_0$$



$$\frac{d\mathbf{x}}{dt} = A_0^*(t)\mathbf{x}(t) + \sum_{i=1}^d A_i^*(t)\mathbf{x}(t - \tau_i)$$

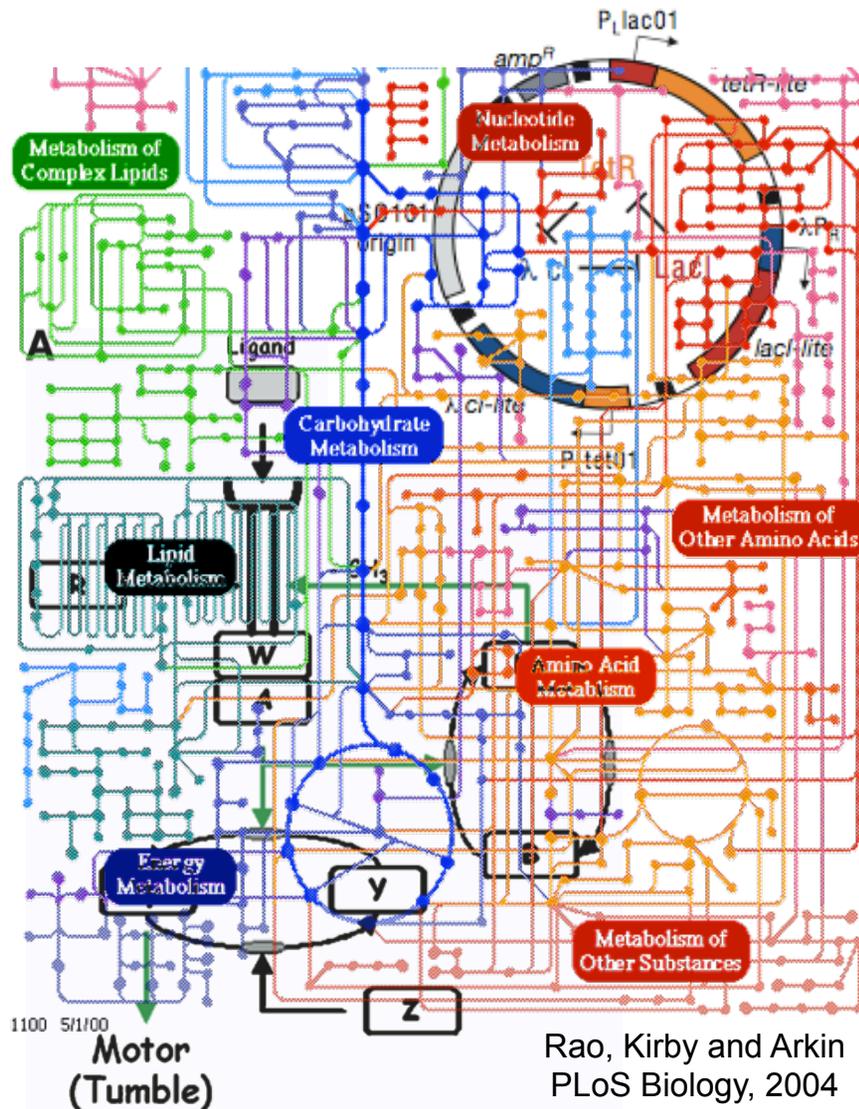
$$\mathcal{M} : \mathcal{C}_0([-\tau, 0], \mathbb{R}^n) \mapsto \mathcal{C}_0([T - \tau, T], \mathbb{R}^n)$$



Curent work: *in vivo* experiments (Ben Prindle)

- Added delay elements to repressilator
- Related experiments: “A fast, robust and tunable synthetic gene oscillator”, Stricker et al (2008).

Control Theory for Biological Systems



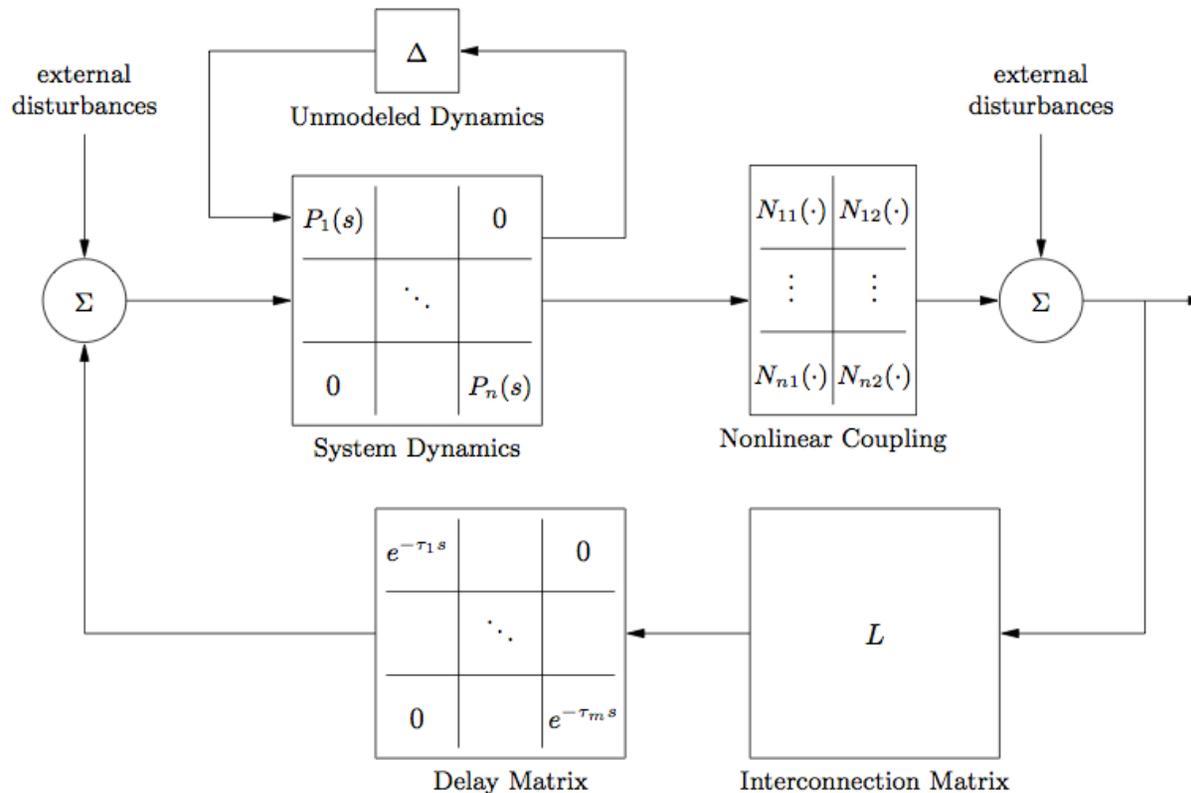
What's different about biological systems

- *Complexity* - biological systems are *much* more complicated than engineered systems
- *Communications* - signal representations are very different (spikes, proteins, etc)
- *Uncertainty* - very large uncertainty in components; don't match current tools
- *Evolvability* - mutation, selection, etc

Potential application areas for control tools

- *System ID* - what are the appropriate component abstractions and models?
- *Analysis* - what are key biological feedback mechanisms that lead to robust behavior?
- *Design* - how to we (re-)design biological systems to provided desired function?
- *Fundamental limits* - what are the limits of performance and robustness for a given biological network topology?

Design of Biomolecular Feedback Systems



Design the easy parts

- Interconnection matrix
- Time delay matrix

Design tools exist for pairwise combinations

- Linear + uncertain = robust control theory
- Linear + nonlinear = describing functions
- Linear + network = formation stabilization
- Linear + delay = Floquet analysis

Open questions

- What is the class of feedback compensators we can obtain using L and τ ?
- How do we specify robustness and performance in highly stochastic settings?
- Can feedback be used to design robust dynamics that implements useful functionality?

Summary and Conclusions

Initial steps in biological circuit design w/ feedback

- System ID - determine active network structure, in vivo
- Feedback circuits - rate regulation, modularity
- Design of dynamics using programmable time delays
- Networked control for biological systems

Next steps: Molecular Programming Project (MPP)

- 5 year goal: create the abstractions, languages and compilers for systematic design of molecular programs
- Explore applications in self-assembly, bio-technology
- Winfree (PI), Bruck, Klavins, M, Pierce, Rothmund

Reading:

- Regulatory activity revealed by dynamic correlations in gene expression noise. MJ Dunlop, RS Cox, JH Levine, RM Murray, MB Elowitz. *Nature Genetics*, 40:1493-1498, 2008
- Design and performance of in vitro transcription rate regulatory circuit. E Franco, RM Murray, CDC 2008 (+2009)
- Stochastic Sensitivity Analysis of Genetic Regulatory Networks, J. Ugander, MS thesis, 2008

