

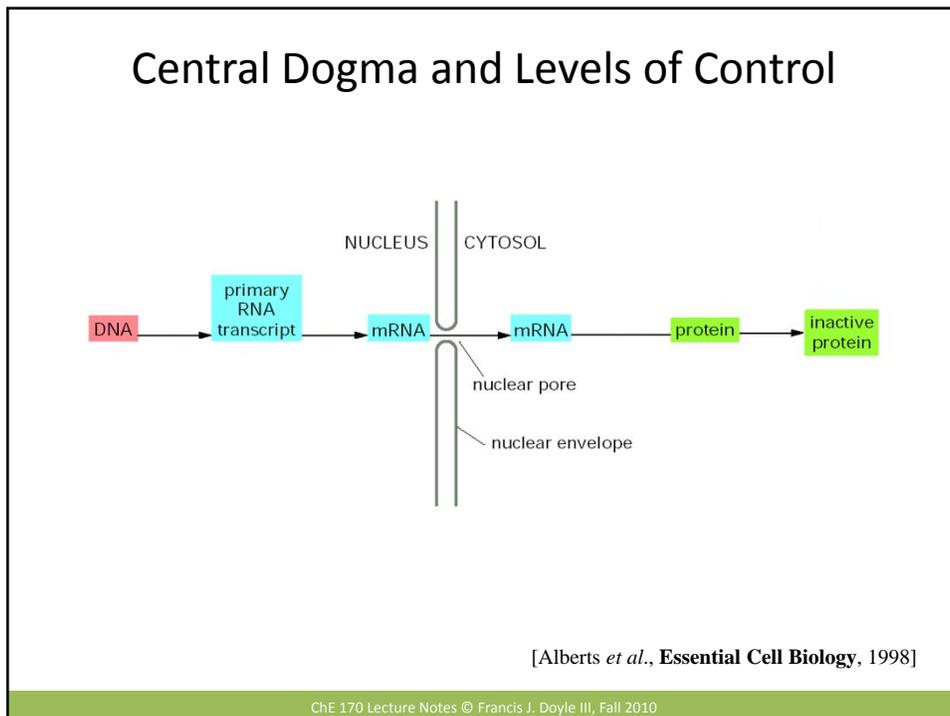
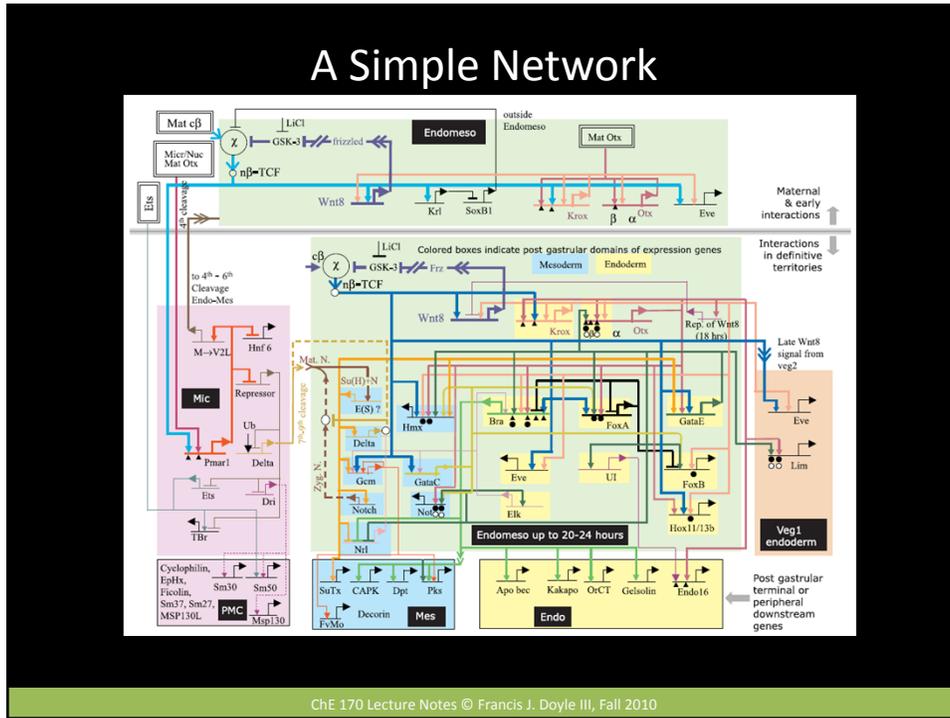
What is “Systems Biology”?

[WTEC Benchmark Study (2005): M. Cassman, A. Arkin, F. Doyle, F. Katagiri, D. Lauffenburger, C. Stokes]
[also: *Nature*, Dec 22, 2005]

- **Primary Definition:** *The understanding of biological **network** behavior through the application of modeling and simulation, tightly linked to experiment*
- **Related Ideas**
 - Identification and validation of networks
 - Creation of appropriate datasets
 - Development of tools for data acquisition and software
- **Motivation:** Phenotype is governed by the behavior of **networks**, rather than the operation of single genes. Understanding the dynamics of even the simplest biological networks requires the application of modeling and simulation.



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Circadian Rhythms

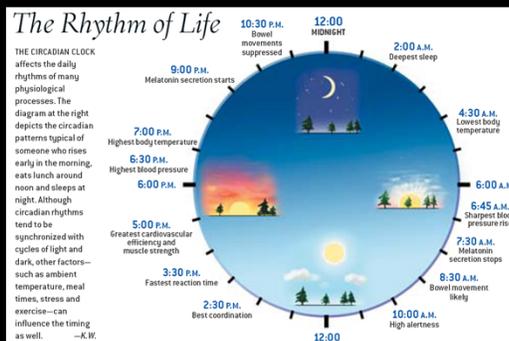
Circadian rhythms = self-sustained biological rhythms characterized by a *free-running period* of about 24h (*circa diem*)

Circadian rhythms characteristics:

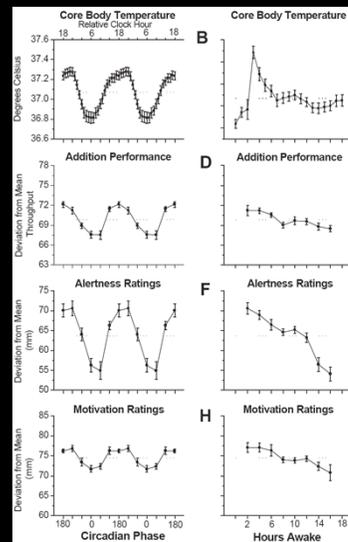
- General – bacteria, fungi, plants, flies, fish, mice, humans, etc.
- Entrainment by light-dark cycles (**zeitgeber**)
- Phase shifting by light pulses
- Temperature compensation

Circadian rhythms emerge at the molecular level but are reinforced at cellular and tissue scales

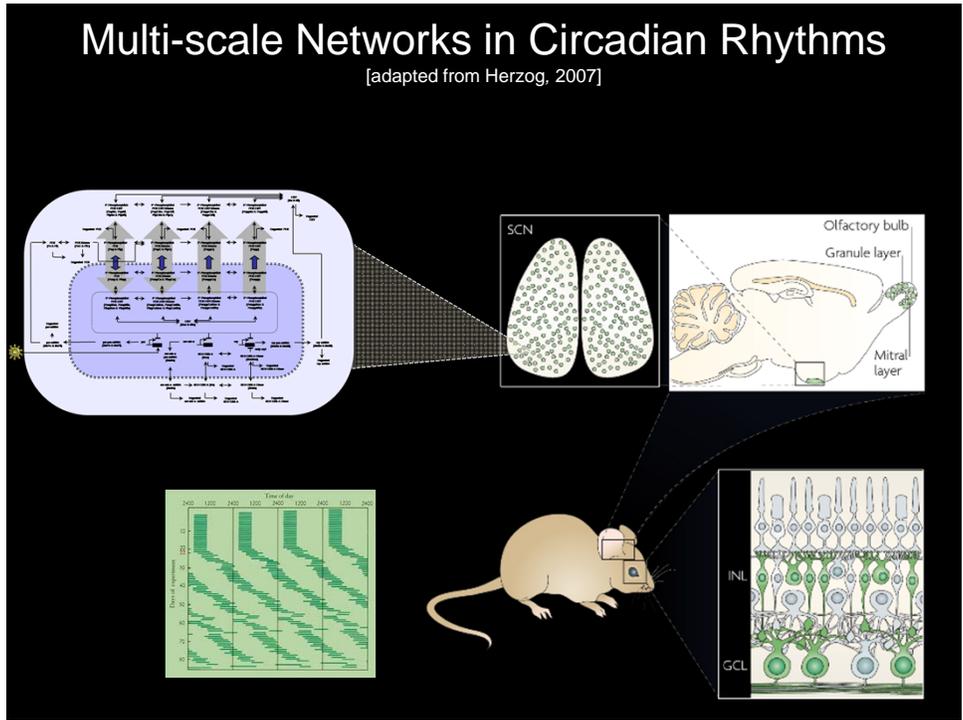
Circadian Component to Performance



[The Body Clock Guide to Better Health, 2000]



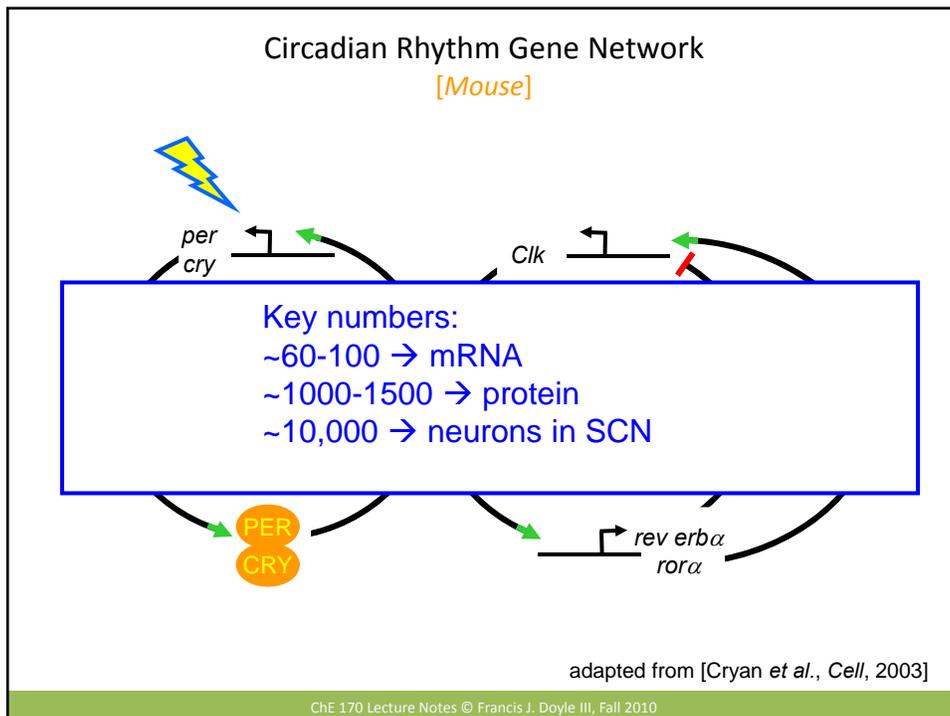
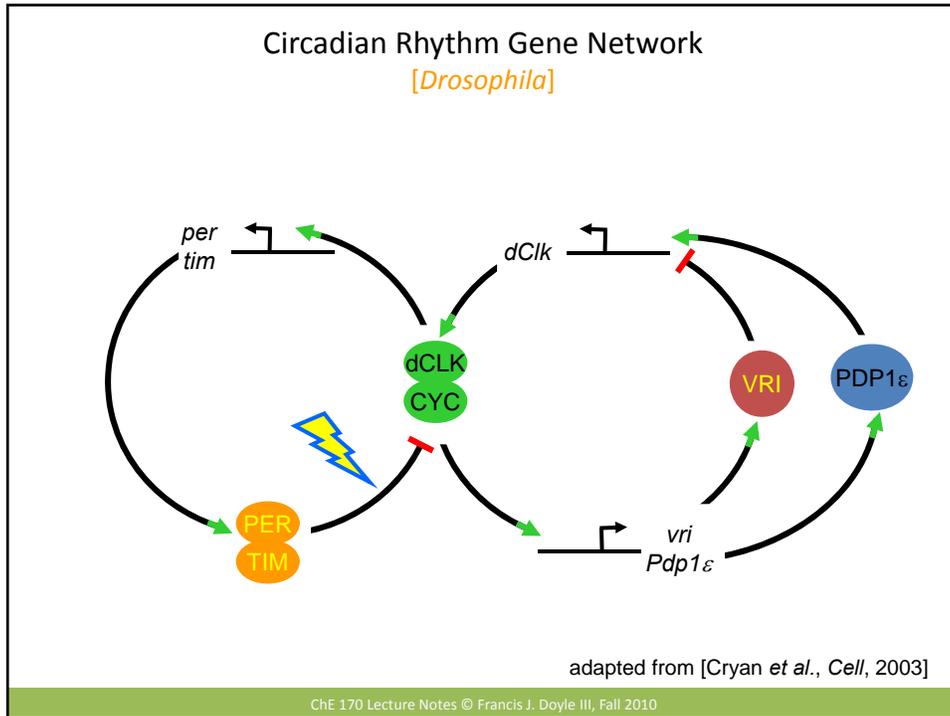
[Hull et al., 2003]



Molecular Components of Circadian Clock

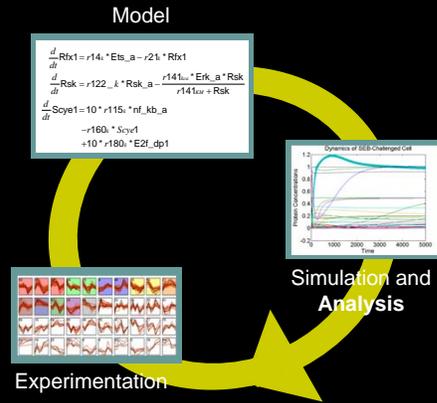
[Forger et al., 2003]

Clock Protein Families		Clock Gene Mutations		
Protein	Features	Gene	Type	Behavioral Phenotype
CLOCK		<i>Clock</i>	Deletion	Long period to arrhythmic
MOP4		<i>BMAL1</i>	Null	Arrhythmic
BMAL1				
MOP9				
PER1		<i>Per1</i>	Null	Var. period to arrhythmic
PER2		<i>Per2</i>	Null	Var. period to arrhythmic
PER3		<i>Per3</i>	Null	Short period
		<i>Per1/2</i>	Null/Null	Arrhythmic
CRY1		<i>Cry1</i>	Null	Short period
CRY2		<i>Cry2</i>	Null	Long period
		<i>Cry1/2</i>	Null/Null	Arrhythmic
CK1ε		<i>CK1ε</i>	Missense	Short period (<i>tau</i> hamster)
CK1δ		<i>CK1δ</i>	Not Done	Unknown



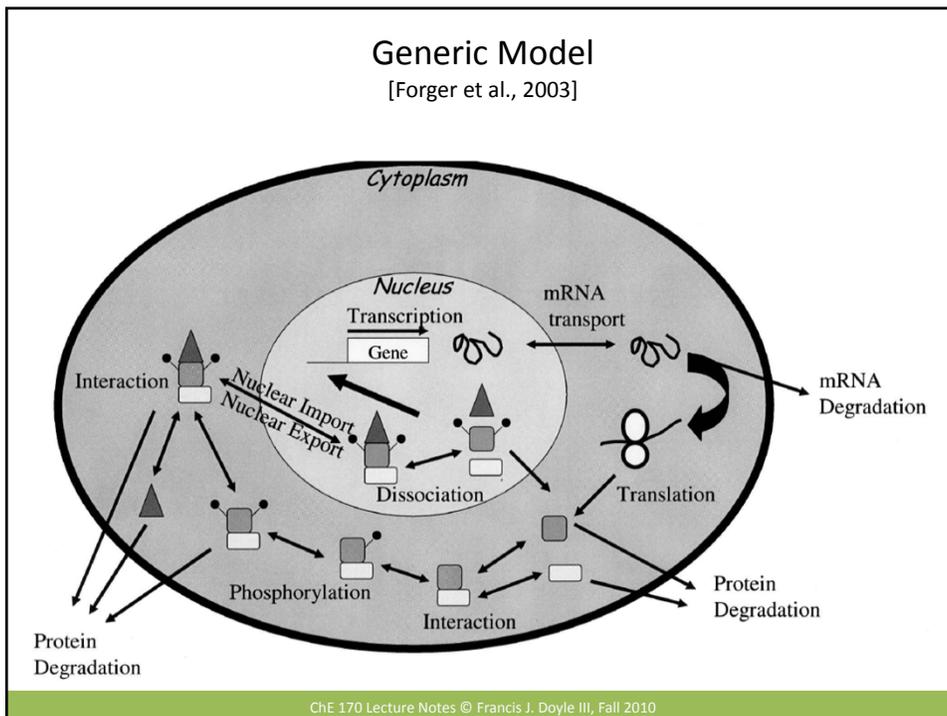
Model Analysis

- Test hypotheses
- *In silico* clinical trials
- Refine design of experiment
- Probe for understanding
- Determine best control input



Generic Model

[Forger et al., 2003]



Model Validation Criteria

[Forger et al., 2003]

Criterion	Expected model performance
1. Period	Molecular oscillations occur with a free-running period (tau) of approximately 24 h.
2. Phase	(a) Molecular oscillations occur with appropriate phase relationships to each other in free-running conditions (appropriate = consistent with experimental data or plausible). (b) Molecular oscillations occur with appropriate phase relationships to each other and to the light-dark cycle.
3. Entrainment	(a) Input repeated at 24-h intervals results in 24-h periodicity of the molecular oscillations. (b) The molecular basis for the input to influence molecular oscillations should be based on experimental data.
4. Phase response	(a) Single stimuli lead to alterations in the phase of molecular oscillation. (b) The response to a stimulus depends on the phase at which it is administered. (c) The molecular basis for the input to influence molecular oscillations should be based on experimental data.
5. Mutations	Mutations affecting the level or activity of circadian-relevant genes <i>in vivo</i> should produce similar effects on oscillations <i>in silico</i> .

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5-State Model

[Goldbeter, 1996; Gonze et al., 2002]

5 states, 18 parameters

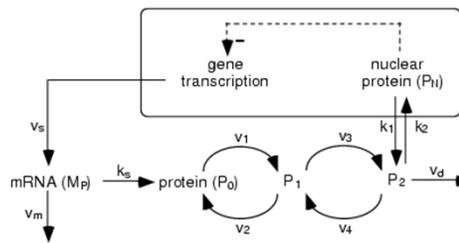
$$\frac{dM_p}{dt} = v_s \frac{K_I^n}{K_I^n + P_N^n} - v_m \frac{M_p}{K_m + M_p}$$

$$\frac{dP_0}{dt} = k_s M_p - v_1 \frac{P_0}{K_1 + P_0} + v_2 \frac{P_1}{K_2 + P_1}$$

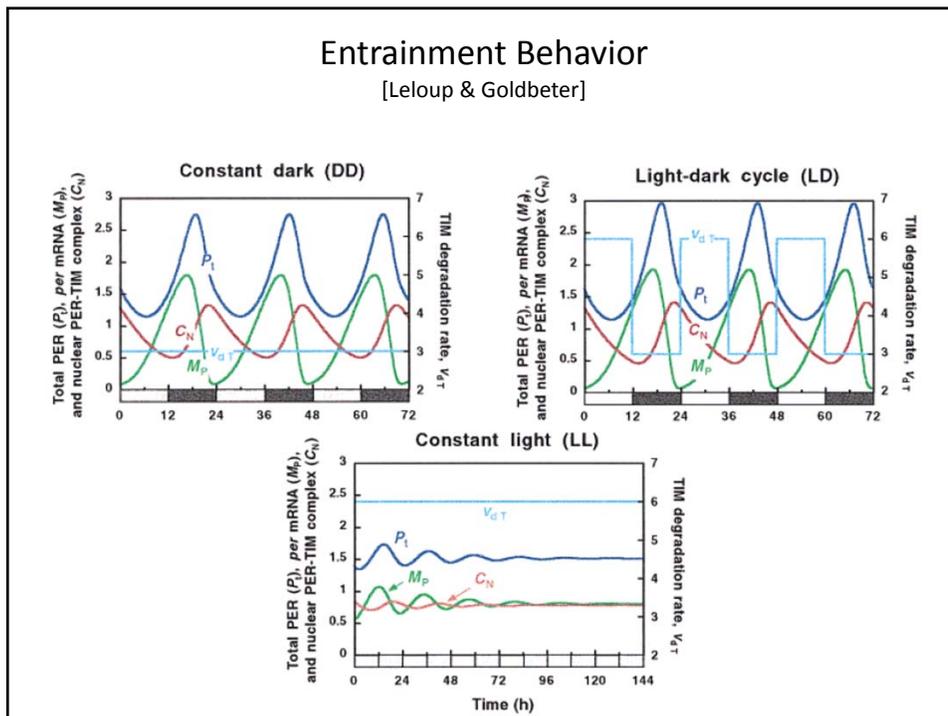
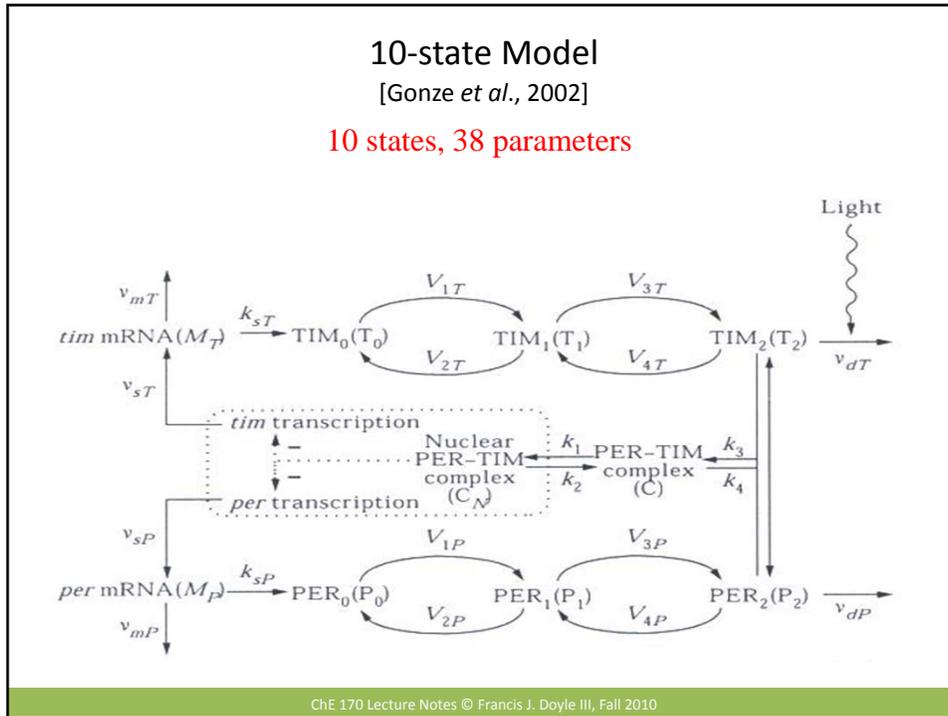
$$\frac{dP_1}{dt} = v_1 \frac{P_0}{K_1 + P_0} - v_2 \frac{P_1}{K_2 + P_1} - v_3 \frac{P_1}{K_3 + P_1} + v_4 \frac{P_2}{K_4 + P_2}$$

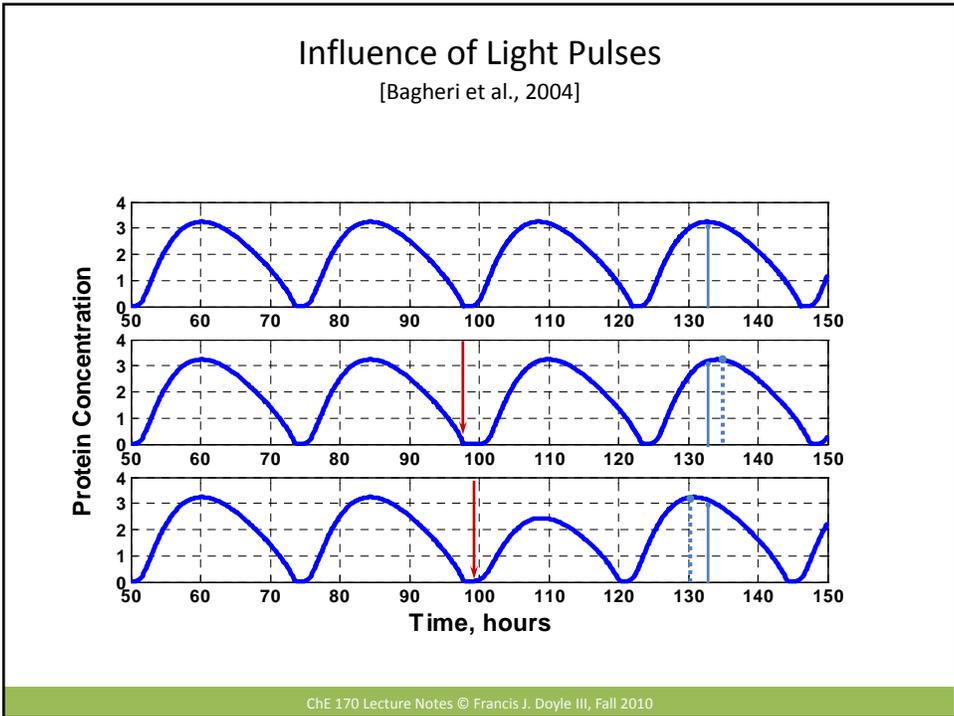
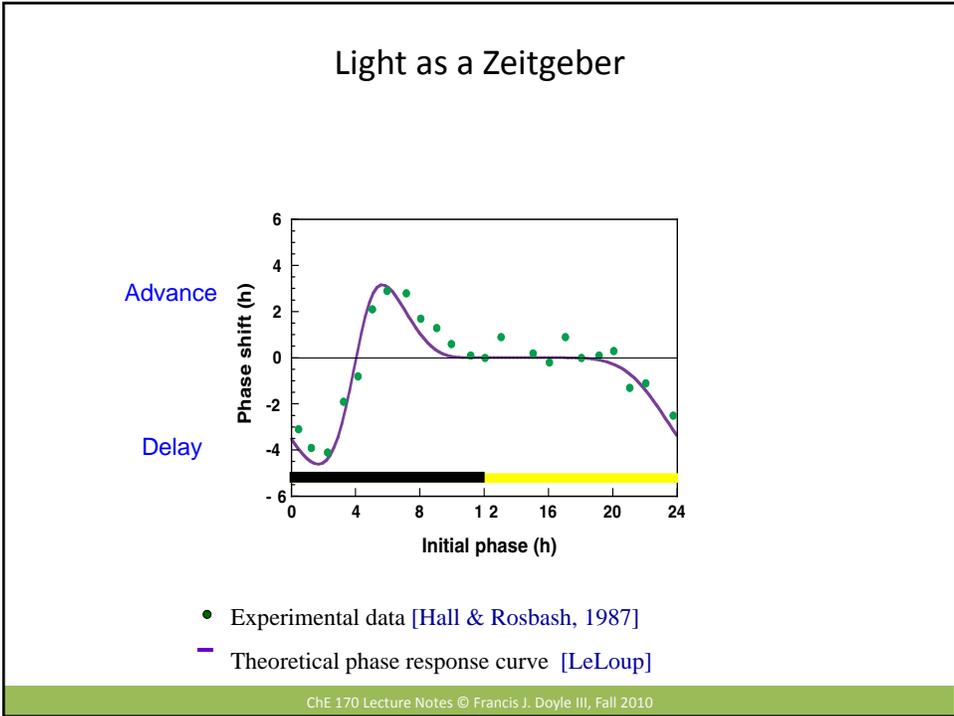
$$\frac{dP_2}{dt} = v_3 \frac{P_1}{K_3 + P_1} - v_4 \frac{P_2}{K_4 + P_2} - v_d \frac{P_2}{K_d + P_2} - k_1 P_2 + k_2 P_N$$

$$\frac{dP_N}{dt} = k_1 P_2 - k_2 P_N$$



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Computational Models of Chemical Reacting Systems

- **Continuous and deterministic** – (*rate equations*) Described by ordinary differential equations (ODE). **Huge numbers of molecules.**
- **Continuous and stochastic** – (*Langevin regime*) Valid under certain conditions. Described by Stochastic Differential Equations (SDE). **Large numbers of molecules.**
- **Discrete and stochastic** – Finest scale of representation for well stirred molecules. Exact description via Stochastic Simulation Algorithm (SSA) [Gillespie, 1977]. The *only* algorithm for **small numbers of molecules.**

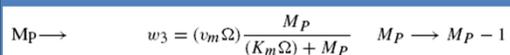
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Enzymatic Degradation of mRNA

$$\frac{dM_P}{dt} = v_s \frac{K_I^n}{K_I^n + P_N^n} - v_m \frac{M_P}{K_m + M_P}$$

$$v_m = 0.3 \text{ nMh}^{-1}$$

$$K_m = 0.2 \text{ nM}$$



$$v_m = 0.3 \text{ mol h}^{-1}$$

$$K_m = 0.2 \text{ mol}$$



$$k_{m1} = 165 \text{ mol}^{-1} \text{ h}^{-1},$$

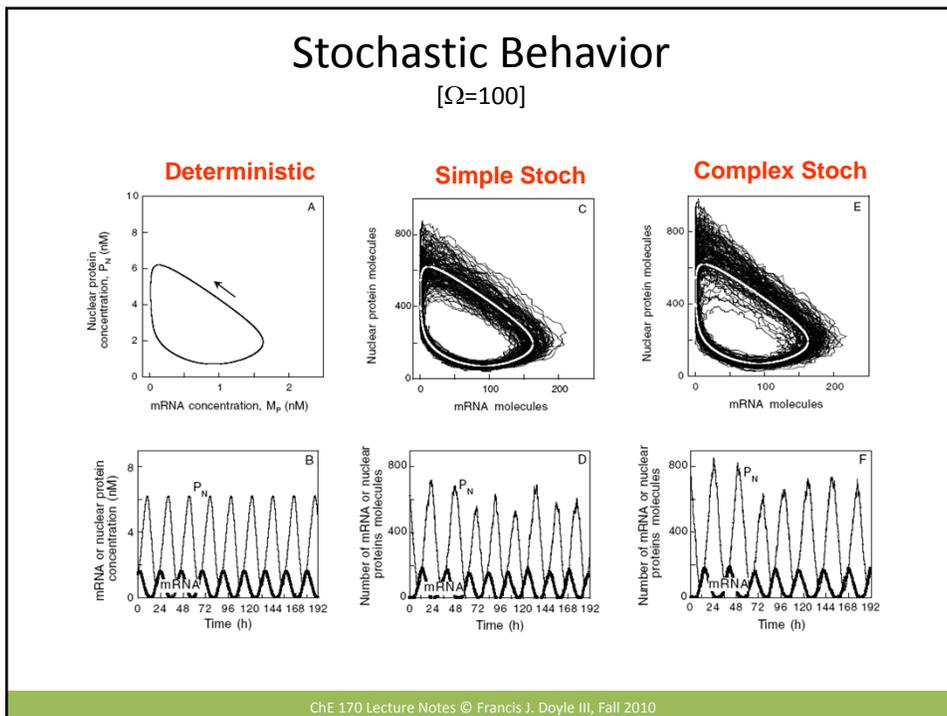
$$k_{m2} = 30 \text{ h}^{-1}, k_{m3} = 3 \text{ h}^{-1},$$

$$E_{m \text{ tot}} = E_m + C_m = (0.1 \times \Omega) \text{ mol}$$

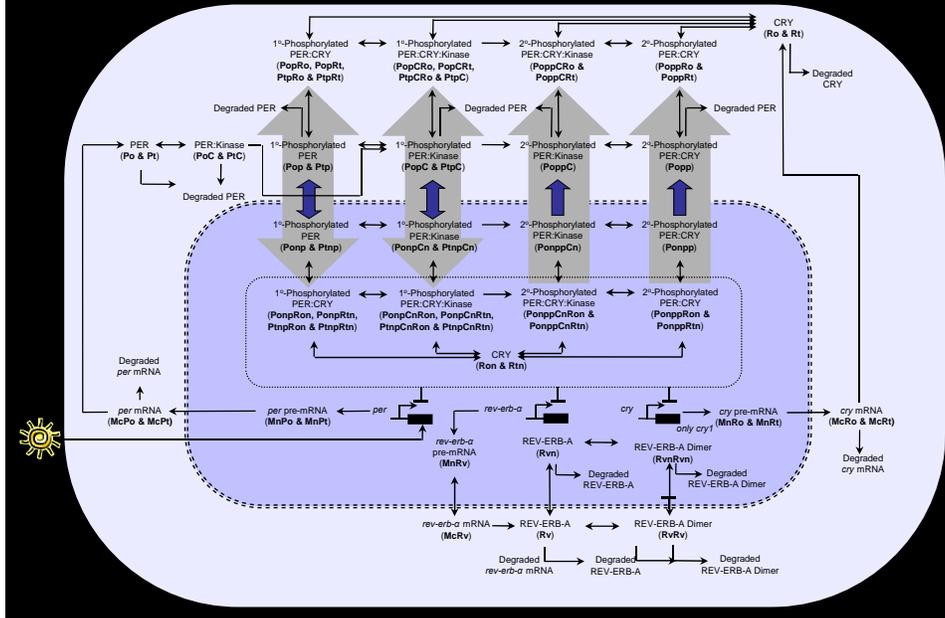
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Reaction number	Reaction step	Probability of reaction step
1a	$G + P_N \xrightarrow{a_1} GP_N$	$w_1 = a_1 \times G \times P_N / \Omega$
1b	$GP_N \xrightarrow{d_1} G + P_N$	$w_2 = d_1 \times GP_N$
1c	$GP_N + P_N \xrightarrow{a_2} GP_{N2}$	$w_3 = a_2 \times GP_N \times P_N / \Omega$
1d	$GP_{N2} \xrightarrow{d_2} GP_N + P_N$	$w_4 = d_2 \times GP_{N2}$
1e	$GP_{N2} + P_N \xrightarrow{a_3} GP_{N3}$	$w_5 = a_3 \times GP_{N2} \times P_N / \Omega$
1f	$GP_{N3} \xrightarrow{d_3} GP_{N2} + P_N$	$w_6 = d_3 \times GP_{N3}$
1g	$GP_{N3} + P_N \xrightarrow{a_4} GP_{N4}$	$w_7 = a_4 \times GP_{N3} \times P_N / \Omega$
1h	$GP_{N4} \xrightarrow{d_4} GP_{N3} + P_N$	$w_8 = d_4 \times GP_{N4}$
1i	$[G, GP_N, GP_{N2}, GP_{N3}] \xrightarrow{v_1} Mp$	$w_9 = v_1 \times (G + GP_N + GP_{N2} + GP_{N3})$
2a	$Mp + E_m \xrightarrow{k_{m1}} C_m$	$w_{10} = k_{m1} \times Mp \times E_m / \Omega$
2b	$C_m \xrightarrow{k_{m2}} Mp + E_m$	$w_{11} = k_{m2} \times C_m$
2c	$C_m \xrightarrow{k_{m3}} E_m$	$w_{12} = k_{m3} \times C_m$
3	$Mp \xrightarrow{k_p} Mp + P_0$	$w_{13} = k_p \times Mp$
4a	$P_0 + E_1 \xrightarrow{k_{11}} C_1$	$w_{14} = k_{11} \times P_0 \times E_1 / \Omega$
4b	$C_1 \xrightarrow{k_{12}} P_0 + E_1$	$w_{15} = k_{12} \times C_1$
4c	$C_1 \xrightarrow{k_{13}} P_1 + E_1$	$w_{16} = k_{13} \times C_1$
5a	$P_1 + E_2 \xrightarrow{k_{21}} C_2$	$w_{17} = k_{21} \times P_1 \times E_2 / \Omega$
5b	$C_2 \xrightarrow{k_{22}} P_1 + E_2$	$w_{18} = k_{22} \times C_2$
5c	$C_2 \xrightarrow{k_{23}} P_0 + E_2$	$w_{19} = k_{23} \times C_2$
6a	$P_1 + E_3 \xrightarrow{k_{31}} C_3$	$w_{20} = k_{31} \times P_1 \times E_3 / \Omega$

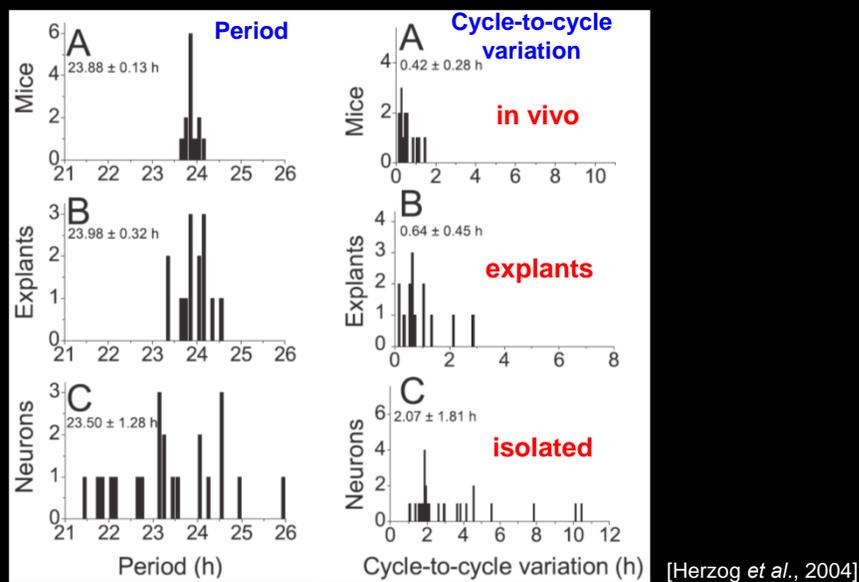
Reaction number	Reaction step	Probability of reaction step
6b	$C_3 \xrightarrow{k_{32}} P_1 + E_3$	$w_{21} = k_{32} \times C_3$
6c	$C_3 \xrightarrow{k_{33}} P_2 + E_3$	$w_{22} = k_{33} \times C_3$
7a	$P_2 + E_4 \xrightarrow{k_{41}} C_4$	$w_{23} = k_{41} \times P_2 \times E_4 / \Omega$
7b	$C_4 \xrightarrow{k_{42}} P_2 + E_4$	$w_{24} = k_{42} \times C_4$
7c	$C_4 \xrightarrow{k_{43}} P_1 + E_4$	$w_{25} = k_{43} \times C_4$
8a	$P_2 + E_d \xrightarrow{k_{d1}} C_d$	$w_{26} = k_{d1} \times P_2 \times E_d / \Omega$
8b	$C_d \xrightarrow{k_{d2}} P_2 + E_d$	$w_{27} = k_{d2} \times C_d$
8c	$C_d \xrightarrow{k_{d3}} E_d$	$w_{28} = k_{d3} \times C_d$
9	$P_2 \xrightarrow{k_1} P_N$	$w_{29} = k_1 \times P_2$
10	$P_N \xrightarrow{k_2} P_2$	$w_{30} = k_2 \times P_N$



Gene Control Circuit: Mammalian Clock

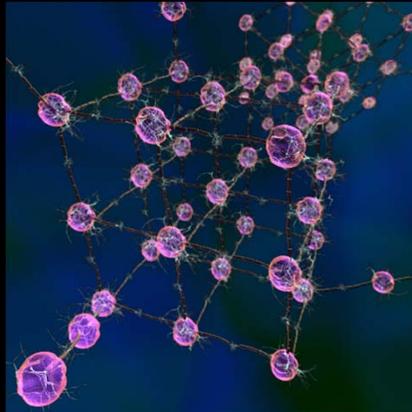


1. Nodes in Network are Stochastic

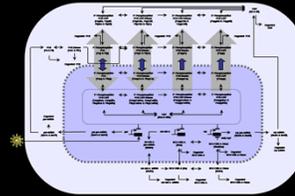


Stochastic Cellular Network Model

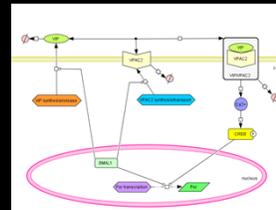
Node: ~25 nonlinear ODES
 Network: 20x20 grid
 Simulation: DASPK



Gene regulation model

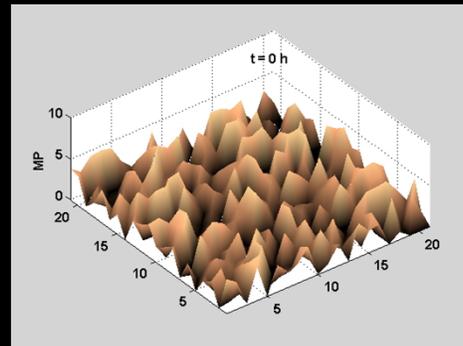
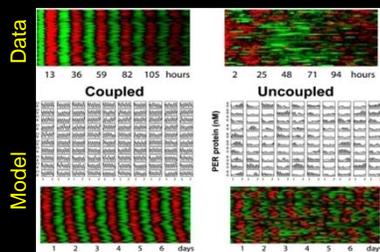


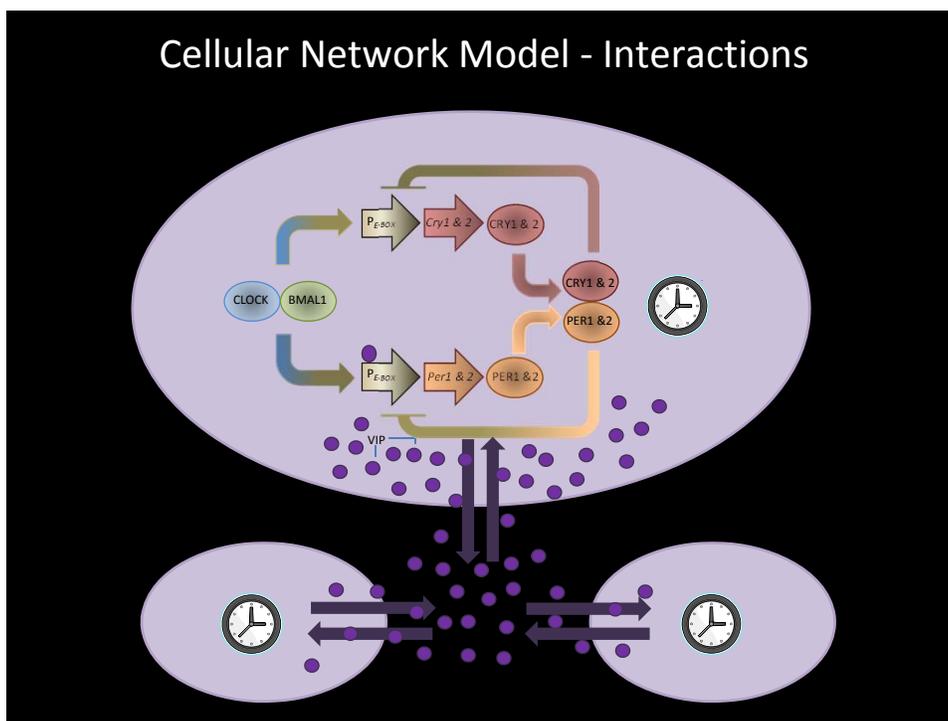
Signaling/coupling model



Stochastic Network Model

[Liu *et al.*, *Cell*, 2007]





Implications from Systems Biology Studies

- Robustness characteristics of feedback architecture under stochastic uncertainty
- Nature of entrainment, and systems characterization
- Possible therapeutic ramifications (sleep disorders, jet lag, shift work performance, etc.)
- General biological oscillator insights