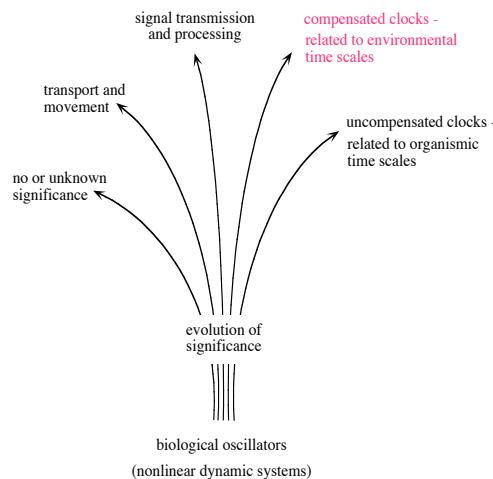


Dynamics, adaptation and fluctuations in bio-networks
March 24-27, Santa Barbara, California

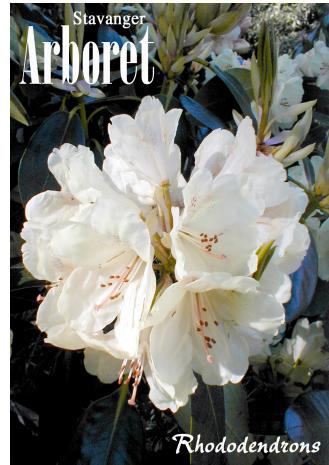
Modelling circadian clocks and temperature-compensation

Peter Ruoff
Dartmouth Medical School, USA
Stavanger University Center, Norway

Biological Oscillators



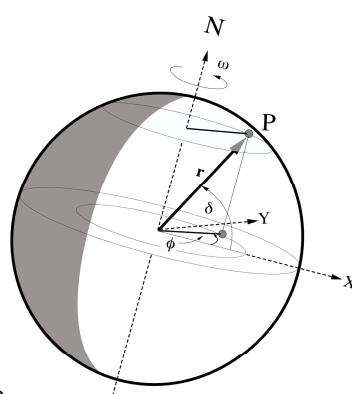
Circadian rhythms



- are important for the adaptation of organisms to their environment
 - for example to anticipate day/night changes as well as seasonal changes (important for many physiological processes).
- act as *clocks*, they have *temperature-compensation* as well as other *homeostatic regulation mechanisms* of their period against environmental influences.

Describing circadian clocks purely mathematically

The rotation/oscillation of point P along the coordinates X, Y describes an harmonic oscillator, which is used, for example, as the basis of the (harmonic) *Cosinor* or *Van der Pol* description of circadian rhythms.



$$\begin{aligned}\frac{dX}{dt} &= \square Y \\ \frac{dY}{dt} &= -\square X\end{aligned}\quad \square \quad \ddot{X} + \square^2 X = 0$$

$\square = 1.157 \times 10^{-5} \text{ Hz}$

Describing circadian clocks purely mathematically

Van der Pol oscillator:

$$\begin{aligned}\frac{dX}{dt} &= \square [Y + \square X \square \square X^3)] \\ \frac{dY}{dt} &= \square \square X\end{aligned}$$

Normally \square is chosen to be small (0.1-0.25), which makes the van der Pol oscillator quasilinear and very similar to the harmonic oscillator.

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Vol. 62, No. 4, pp. 1283-1296

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RECONCILING MATHEMATICAL MODELS OF BIOLOGICAL CLOCKS BY AVERAGING ON APPROXIMATE MANIFOLDS*

DANIEL B. FORGER[†] AND RICHARD E. KRONAUER[‡]

Abstract. Although oscillations are typically described in terms of two variables (phase and amplitude), most oscillating systems contain more than two variables. We present a method for approximating the dynamics of the phase and the amplitude of a quasilinear system of three or more variables. This method is used to analyze Goldbeter's five-variable model of the biological circadian (approximately 24-hour period) clock in the fruit fly *Drosophila* [*Proc. Roy. Soc. London B*, 261 (1995), pp. 319-324]. Using this method, we show that Forger, Jewett, and Kronauer's mathematical model [*J. Biol. Rhythms*, 14 (1999), pp. 532-537] of the human circadian system (based on the van der Pol equation) is almost identical to Goldbeter's model, even though these models were developed independently. This leads to (1) a biochemical analogue of the van der Pol equation, (2) biological support for the numerous mathematical models of circadian systems which use the van der Pol equation, and (3) possible evidence of the similarity between the circadian system in *Drosophila* and the circadian system in man.

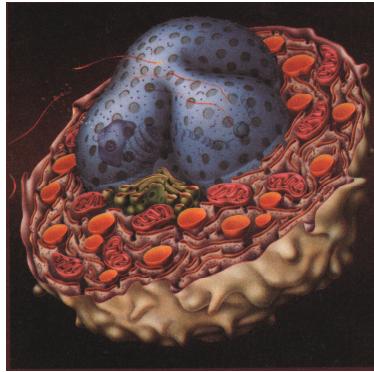
Key words. biological clocks, method of averaging, biochemical oscillations, van der Pol equation, Fitzhugh-Nagumo equations

Physical chemical approach to model circadian rhythms

"..... All living things are physico-chemical machines.....How can you make progress if you do not know physical chemistry?....."

Sinclair Lewis, *Arrowsmith*

Modelling circadian clocks in reaction kinetic terms



artistic view of an eukaryotic cell

All chemical change, no matter how complex, is believed to be due to chemical (*elementary*) processes.

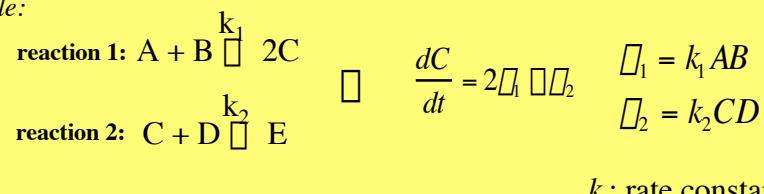
If we would know all processes and their kinetics in a single cell, we would be able to give a *quantitative* description of the cell's physiology through the chemical rate equations.

Chemical Rate Equations

Considering a set of reacting metabolites $\{S_i\}$, the reaction system is described by a set of coupled differential equations:

$\frac{dS_i}{dt} = \sum_j n_{ij} \square_j$ n_{ij} denote stoichiometric coefficients
 \square_j are reaction rates of individual process "j"

Example:

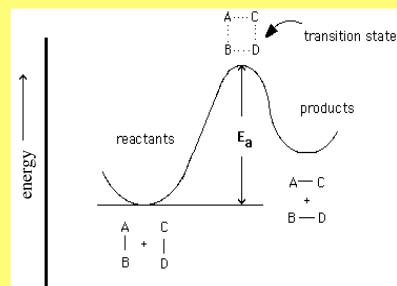


k_i : rate constant

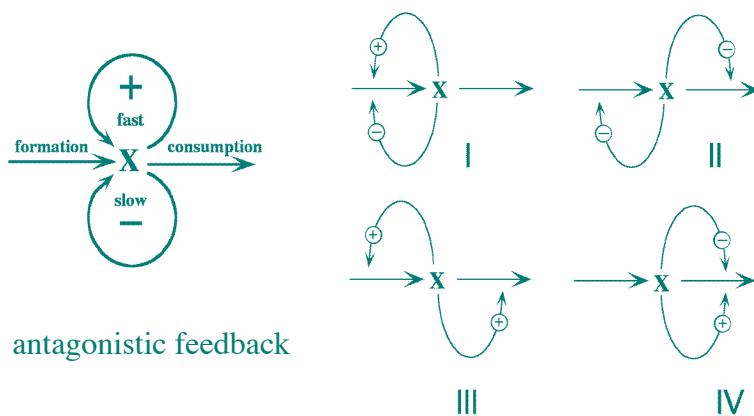
Influence of temperature on reaction rates:
the Arrhenius equation

$$k = A e^{-E_a/RT}$$

rate constant
activation energy
temperature (K)
gas constant
collision factor

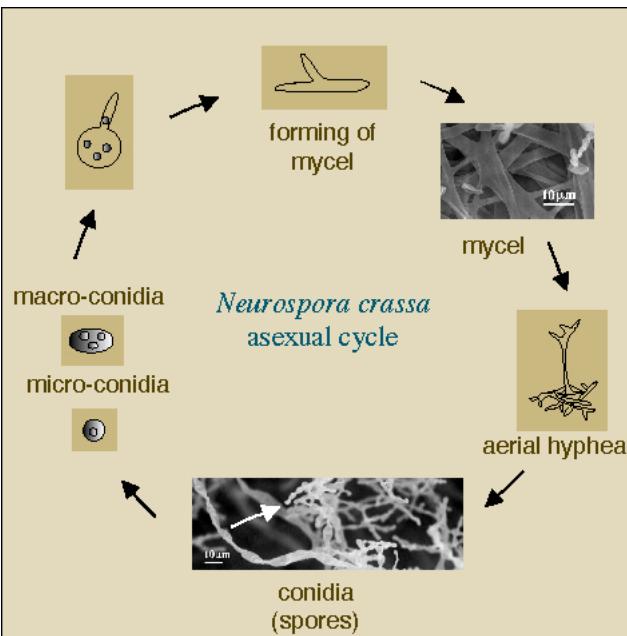


Franck's antagonistic feedback concept
in physico-chemical oscillators

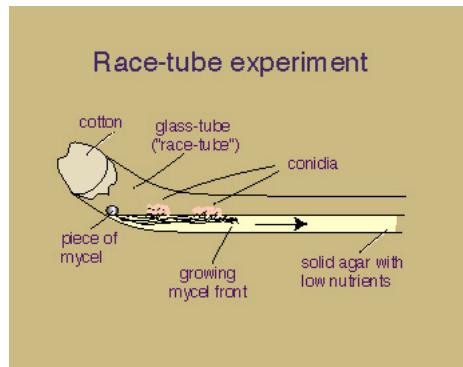


Franck, Angew. Chem. Int. Ed. Engl. 17 (1978) 1-15

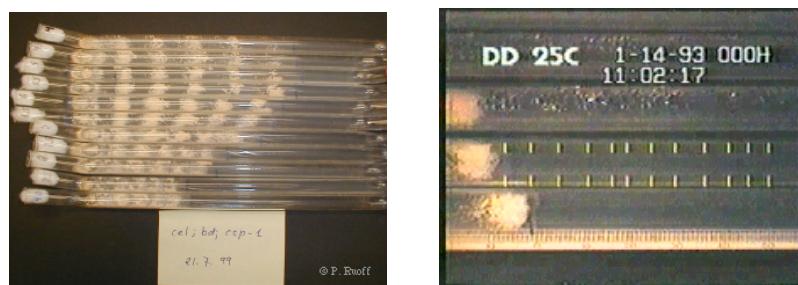
Neurospora crassa: A Model Organism
for the Study of Circadian Rhythms



Racetube assay of clock rhythm in *Neurospora crassa*



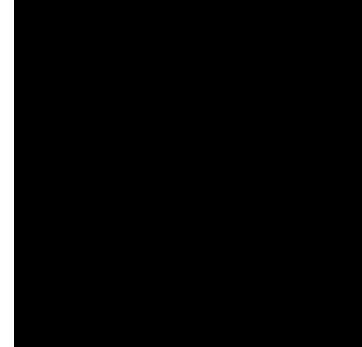
Neurospora crassa: A Model Organism
for the Study of Circadian Rhythms



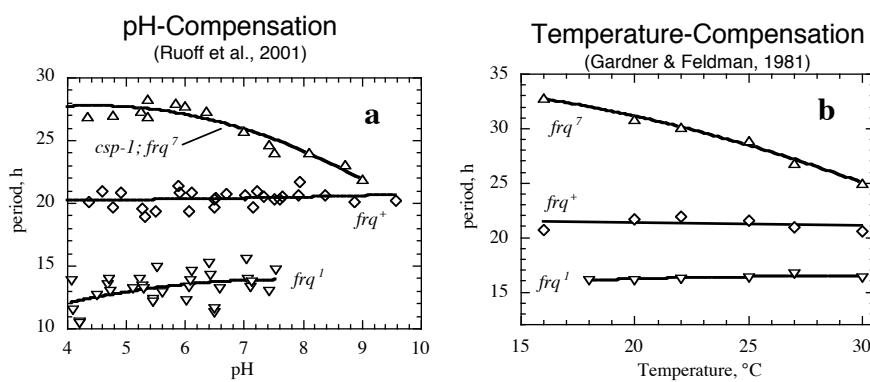
Influence of extracellular pH on
Neurospora's circadian rhythm

Jason C. Thoen and Van Gooch:
"Time Lapse Video Showing an Internal
Circadian Clock in Mold (*Neurospora*) Growth"

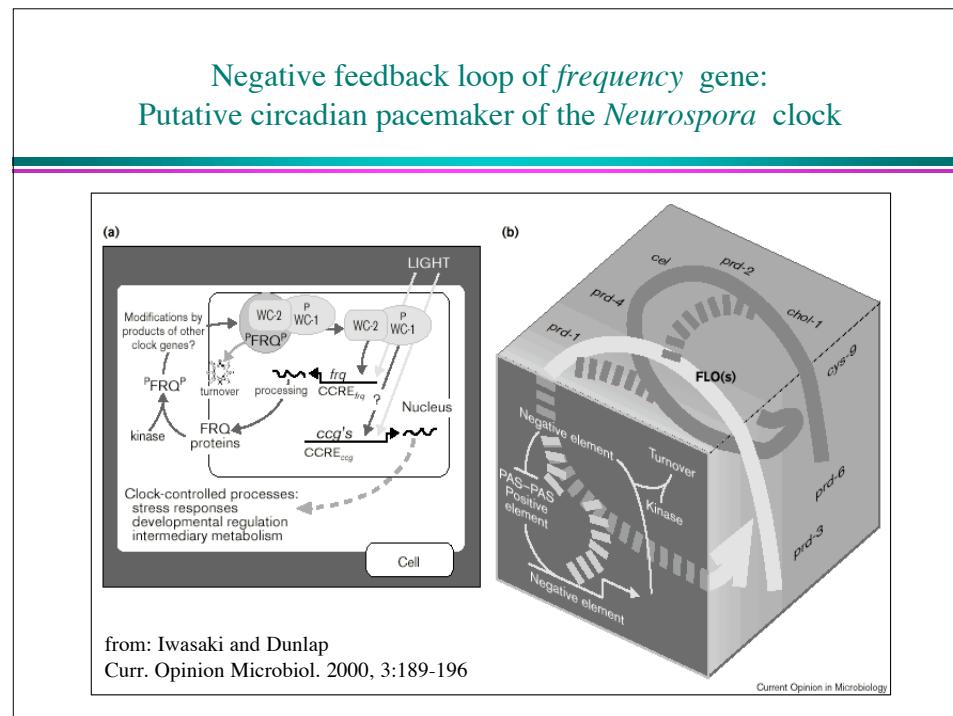
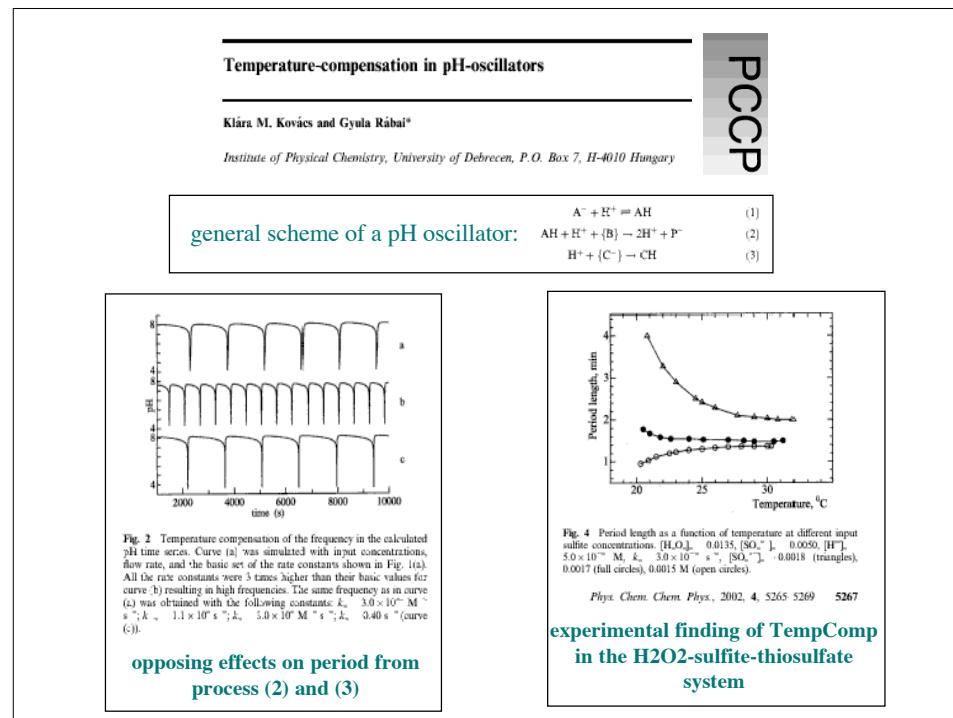
Neurospora crassa growth and banding
in a Petri dish

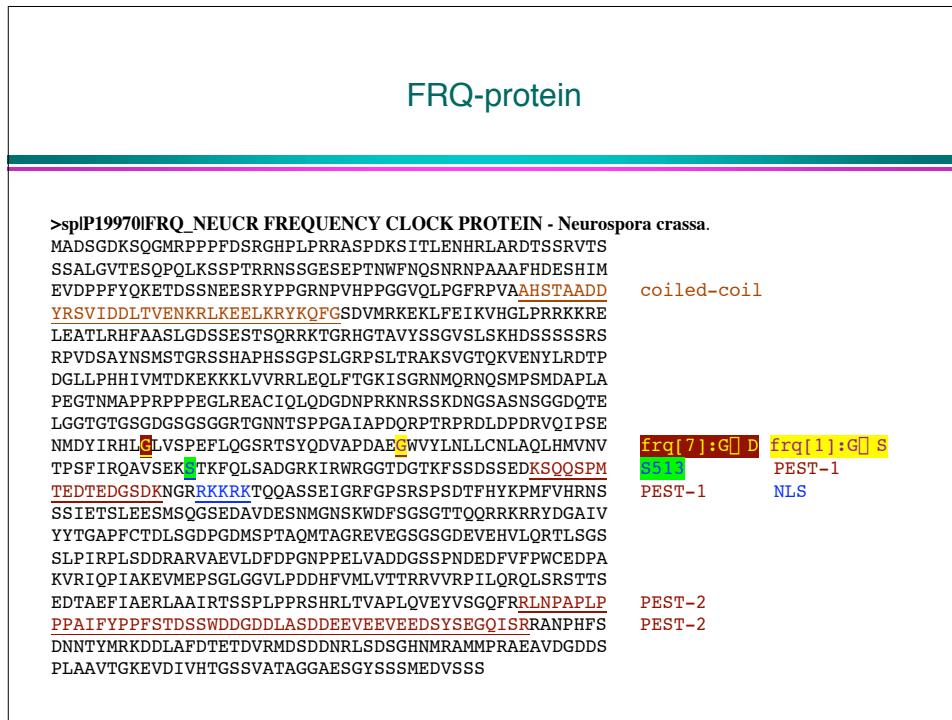


pH- and temperature-compensation in
mutants of *Neurospora crassa*

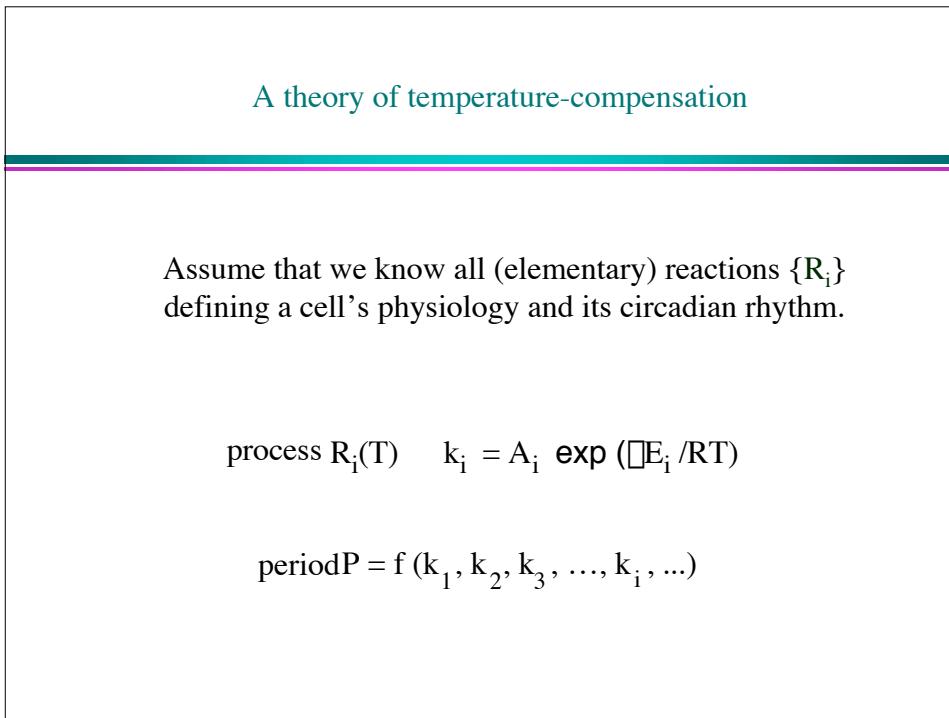
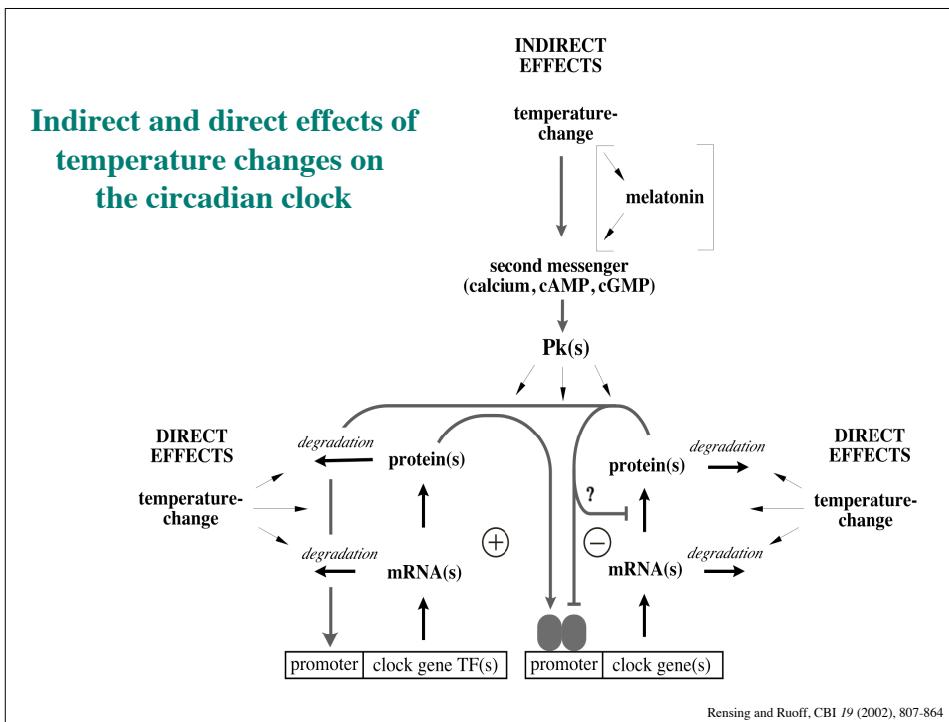


A Theory for Temperature-Compensation and Period Homeostasis in Reaction Kinetic Models of Biological Oscillators





modelling temperature-compensation



Antagonistic balance in temperature

$$P = f(k_1, k_2, k_3, \dots, k_i, \dots)$$

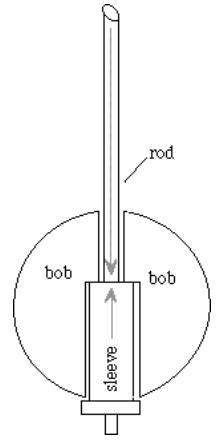
$$\frac{dP}{dT} = \sum_i \frac{\partial f}{\partial k_i} \left(\frac{\partial k_i}{\partial T} \right) = 0$$

$$\sum_i \frac{\partial \ln(f)}{\partial \ln(k_i)} \Delta E_i = 0$$

Concept of opposing reactions
(antagonistic balance in temperature)

$$\sum_j \frac{\partial \ln f}{\partial \ln k_j} \Delta E_j = - \sum_i \frac{\partial \ln f}{\partial \ln k_i} \Delta E_i$$

(P-increasing) (P-decreasing)



Mechanical analogy:
Temp-Comp Invar pendulum

Infinite number of possibilities to realize temperature-compensation!

- For a given reaction kinetic oscillator model
- there is an infinite number of activation energy combinations that will lead to temperature compensation. Evolution has "realized" some of them.

$$\sum_j \frac{\partial \ln f}{\partial \ln k_j} \Delta E_j = - \sum_i \frac{\partial \ln f}{\partial \ln k_i} \Delta E_i$$

(P-increasing) (P-decreasing)

- Analogous homeostasis conditions may be formulated for other physico-chemical properties, as for example, pH or salinity:

salinity: $k_i = k_{oi} \exp(\pm \sqrt{I})$, I = ionic strength
pH: $k_i = k_{oi} \exp(-k_i (\text{pH} - \text{pH}_{i,\text{opt}})^2)$

Euler's summation theorem

$$P = f(k_1, k_2, k_3, \dots, k_i, \dots)$$

$(\frac{\partial \ln P}{\partial \ln k_i})$ are called *control coefficients* C_i , *period elasticities*, or *sensitivity coefficients*.

The period function $P = f$ is called homogenous to degree -1 if:

$$f(tk_1, tk_2, tk_3, \dots, tk_i, \dots) = t^{-1}f(k_1, k_2, k_3, \dots, k_i, \dots) = t^{-1}P$$

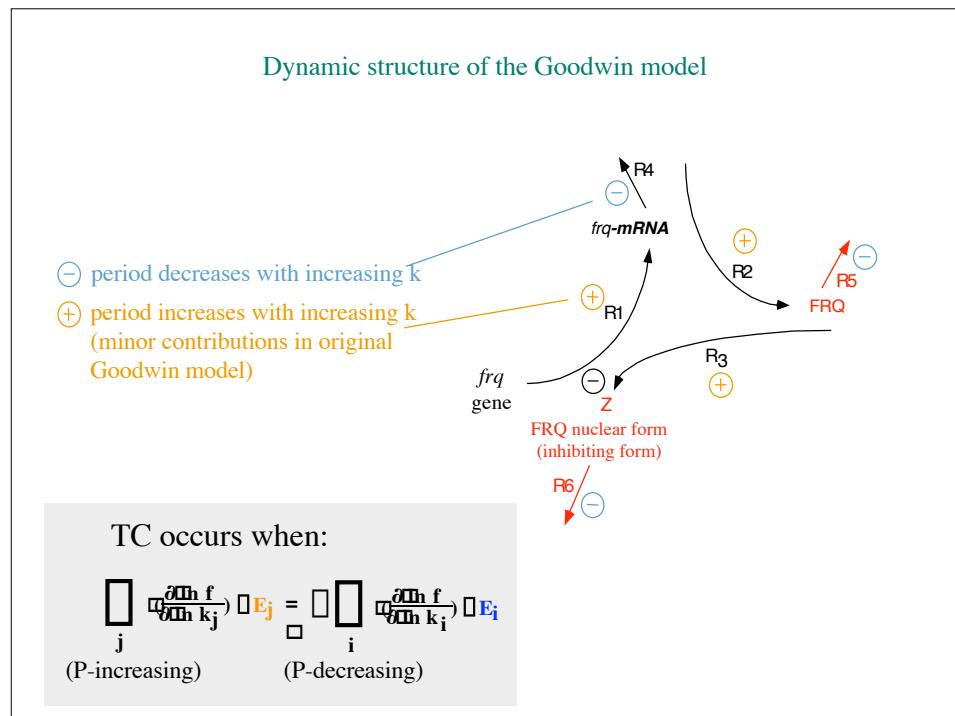
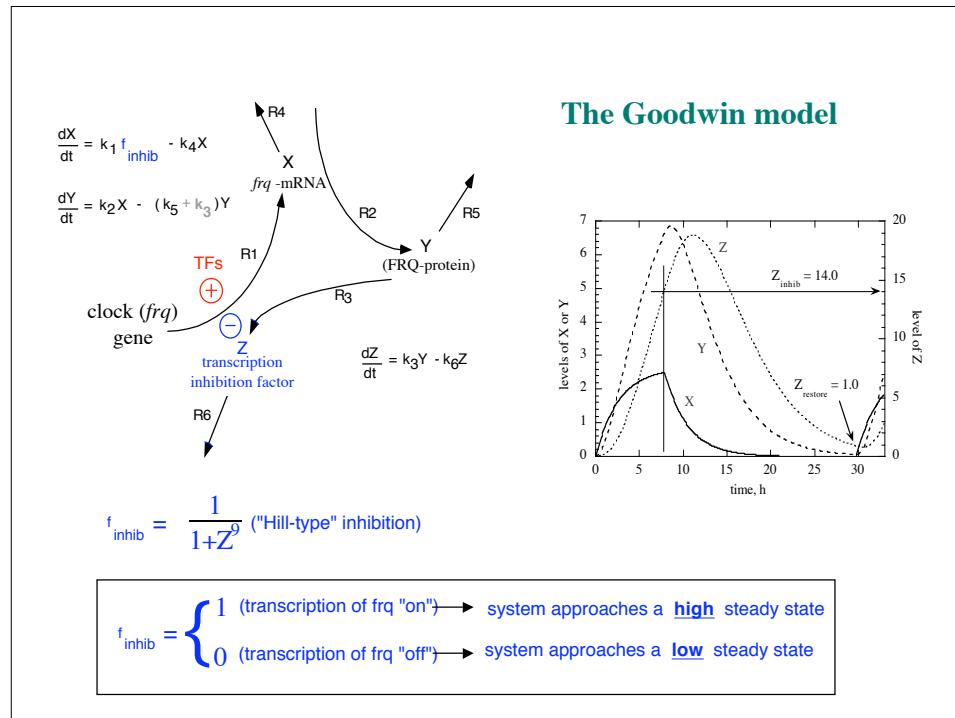
$$\boxed{\frac{\partial \ln P}{\partial \ln k_i}} = \boxed{1}$$

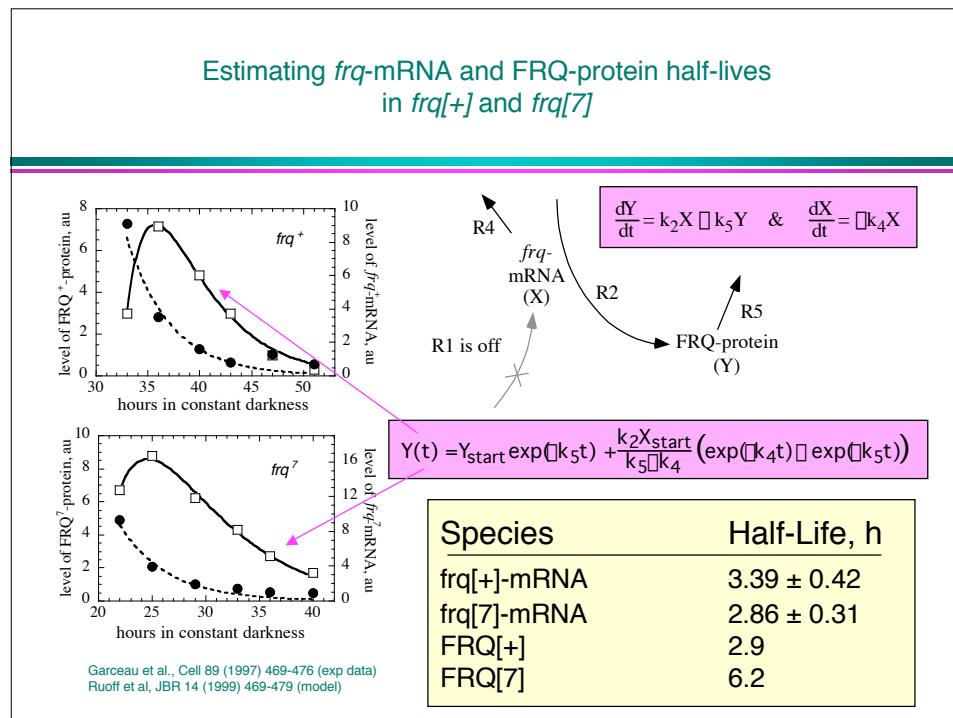
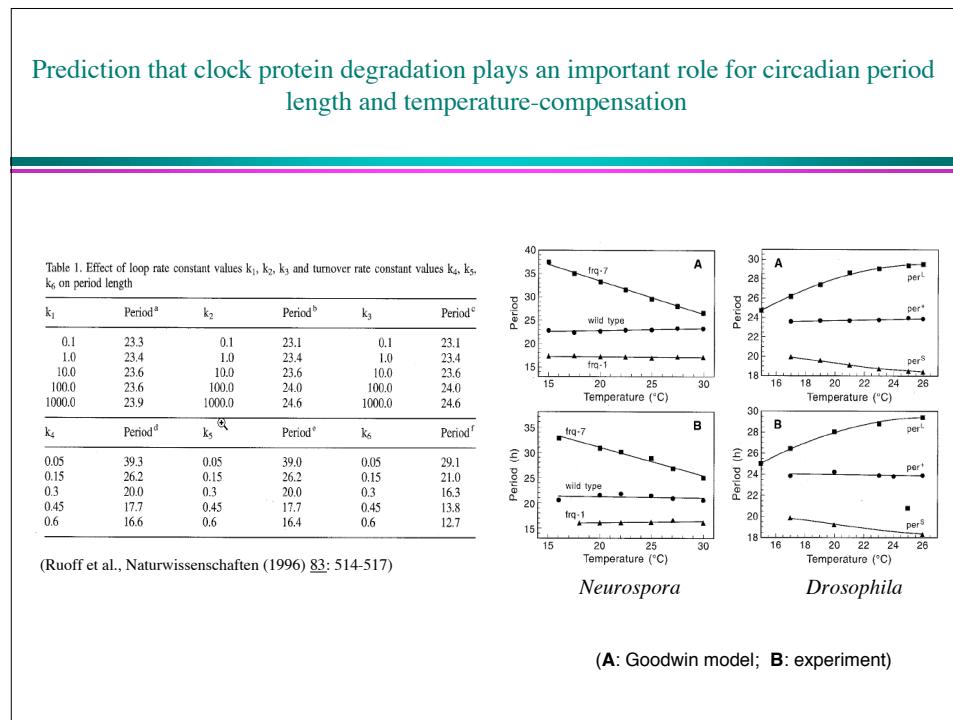
A reaction kinetic modelling strategy: working with minimal models

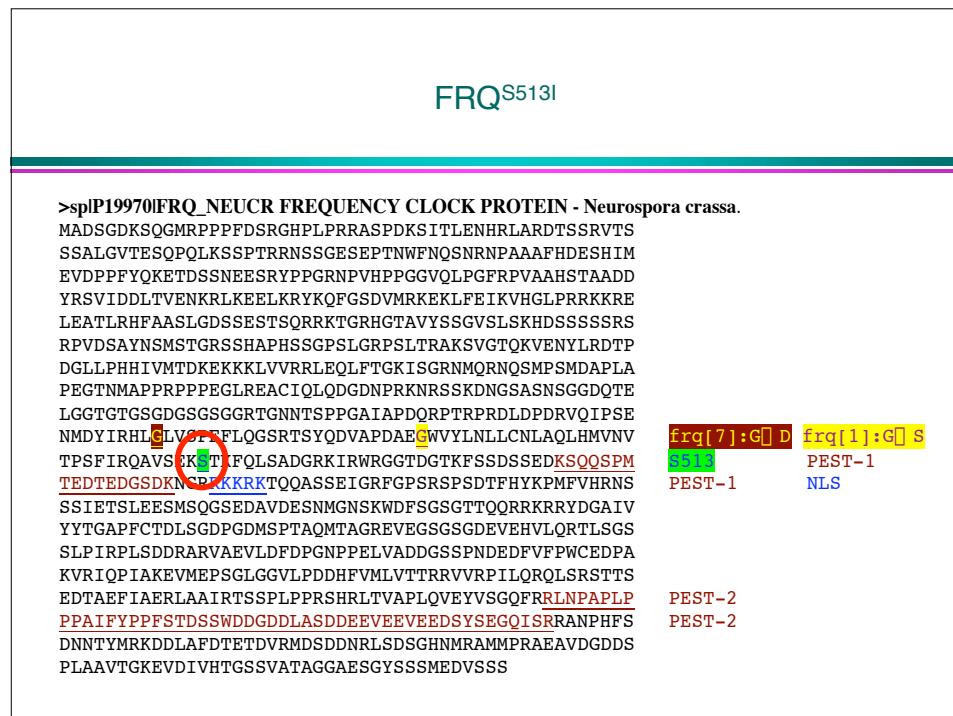
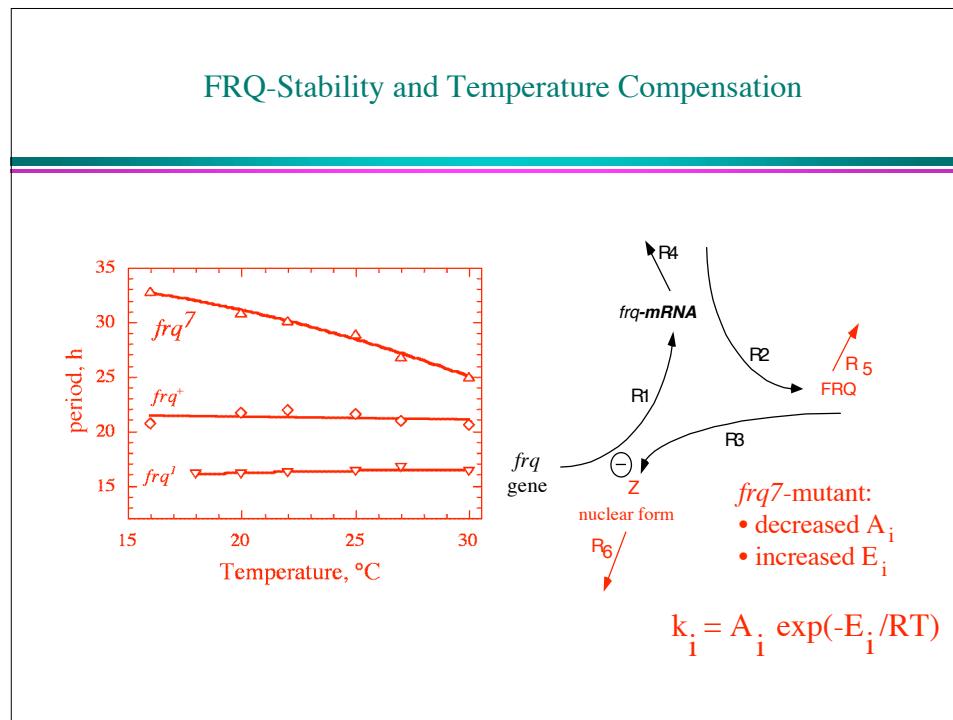
Ockham's razor:

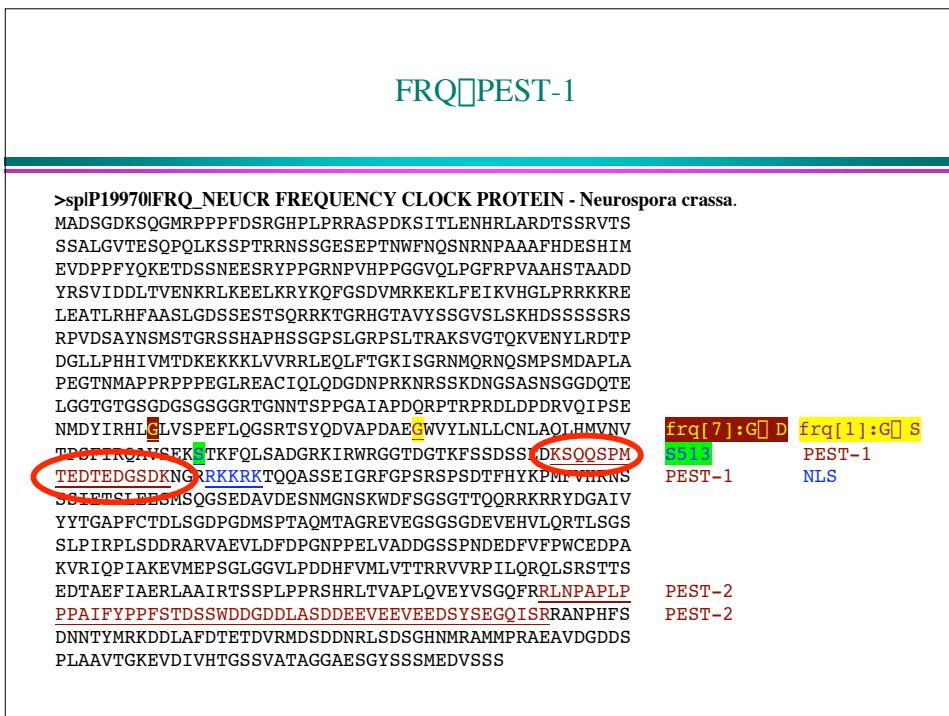
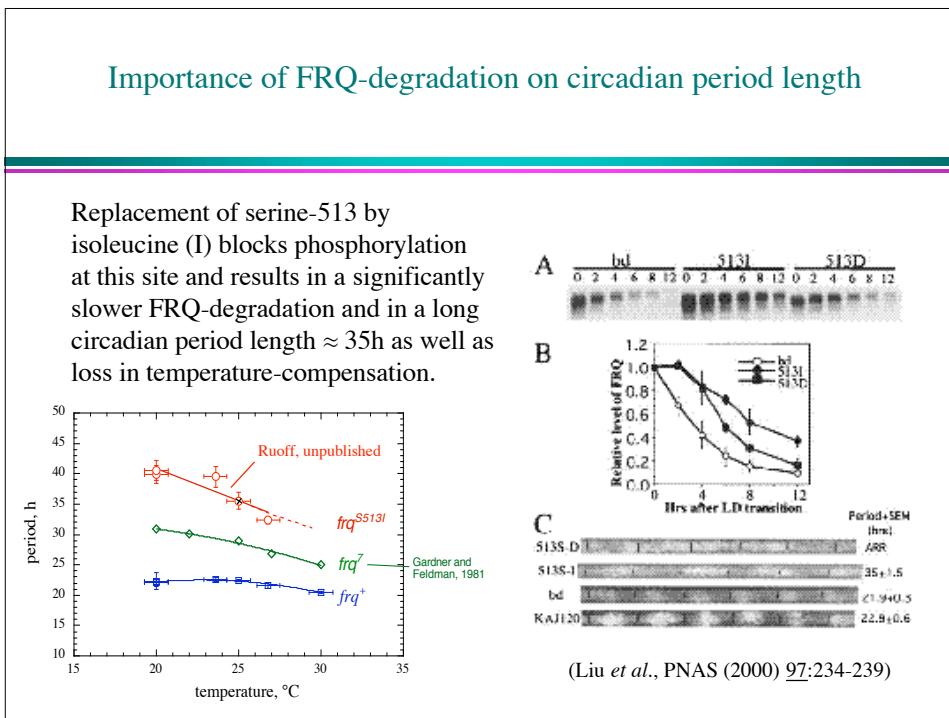
“Plurality is not to be assumed without necessity”

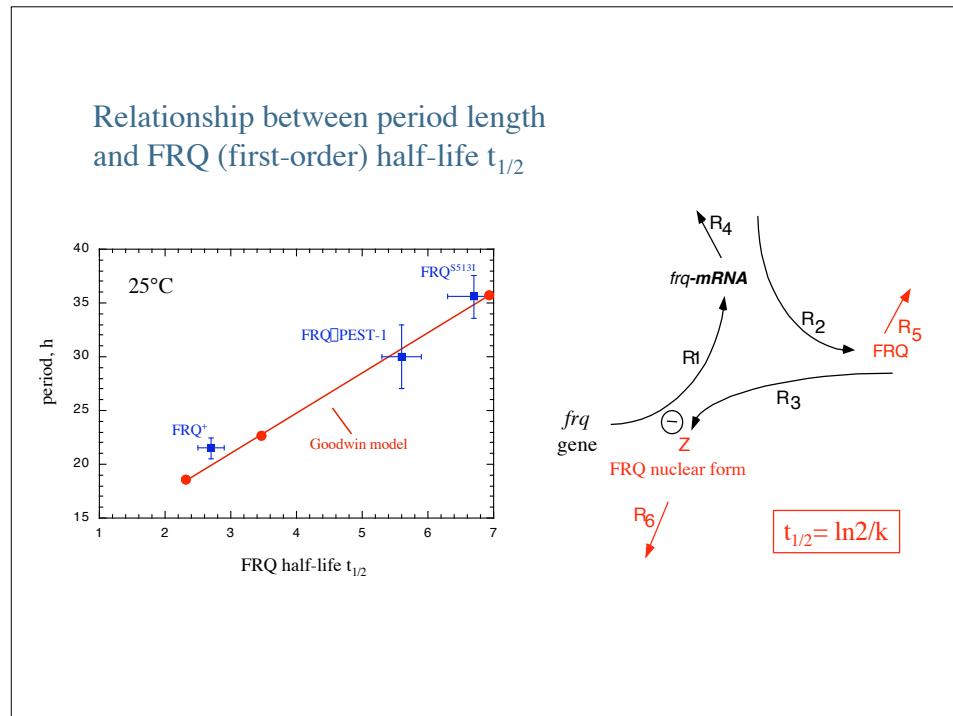
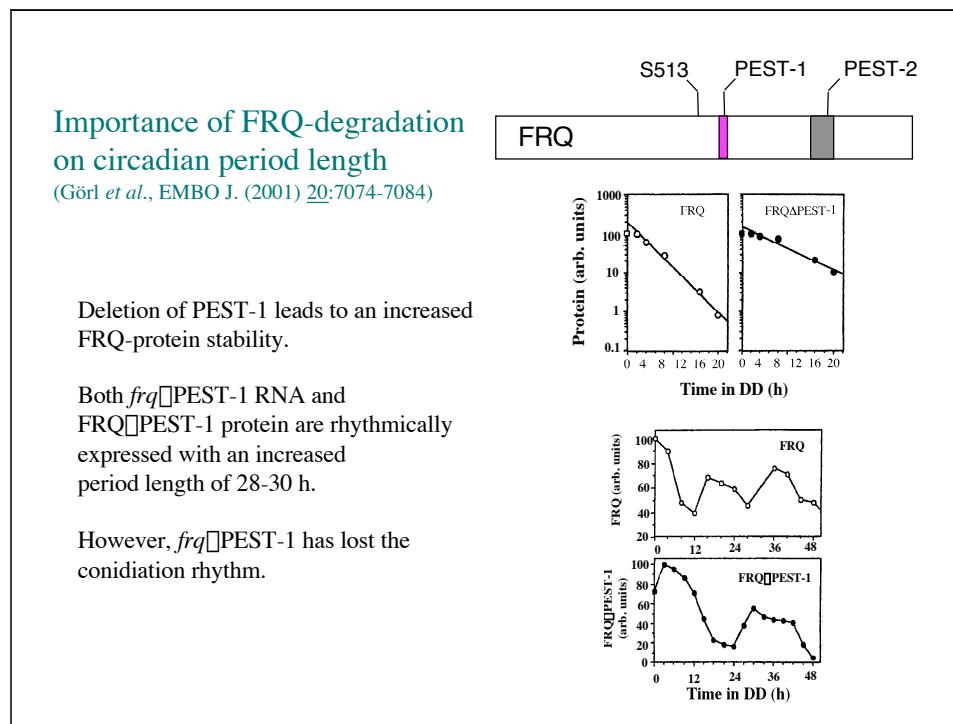
William of Ockham, 1285-1349









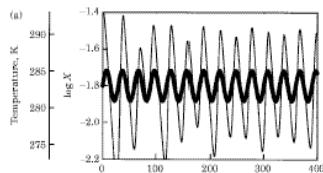


modelling entrainment by temperature cycles

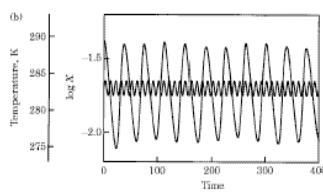
&

temperature-pulse phase response curves

entrainment by temperature cycles in the Goodwin model

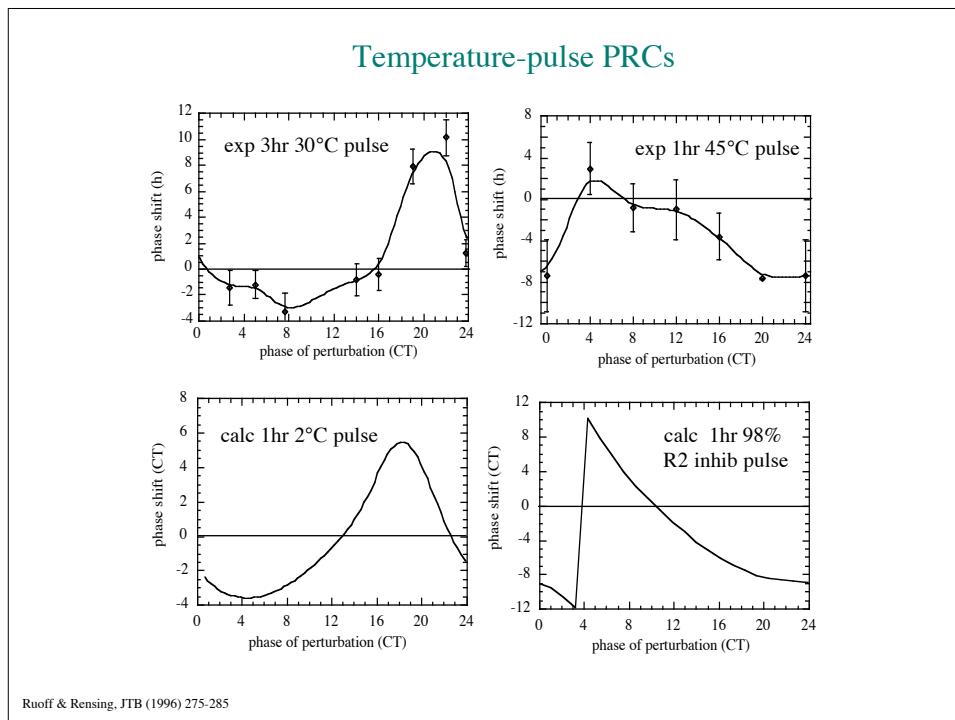


(a) Temperature entrainment using a sinusoidal temperature perturbation with period =30 and a temperature amplitude of 1°C. The inner oscillation is the entraining temperature cycle while the outer oscillation is the responding log X value.



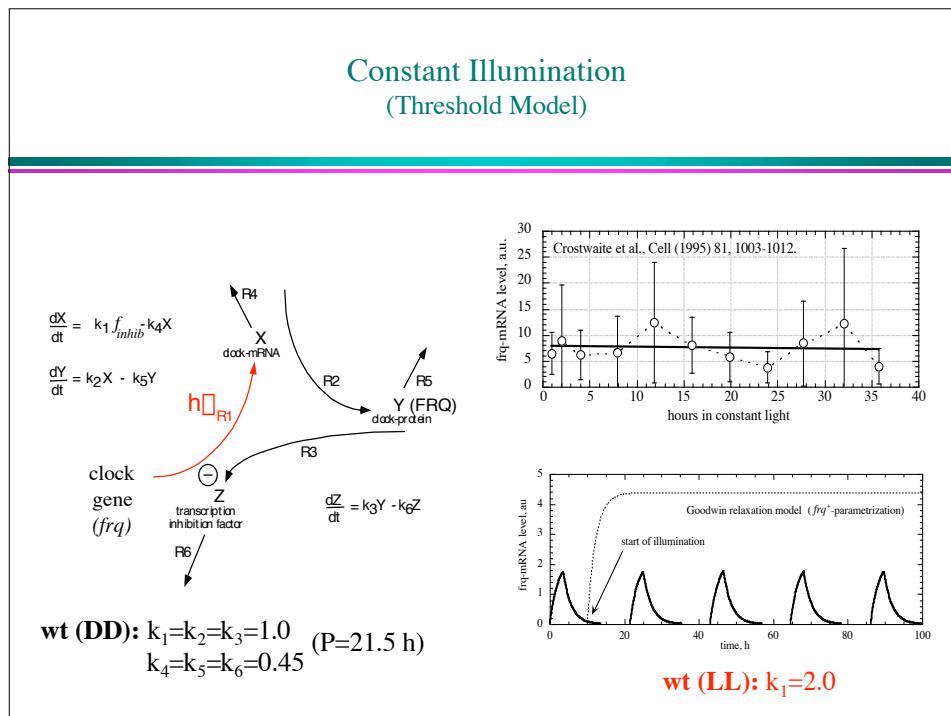
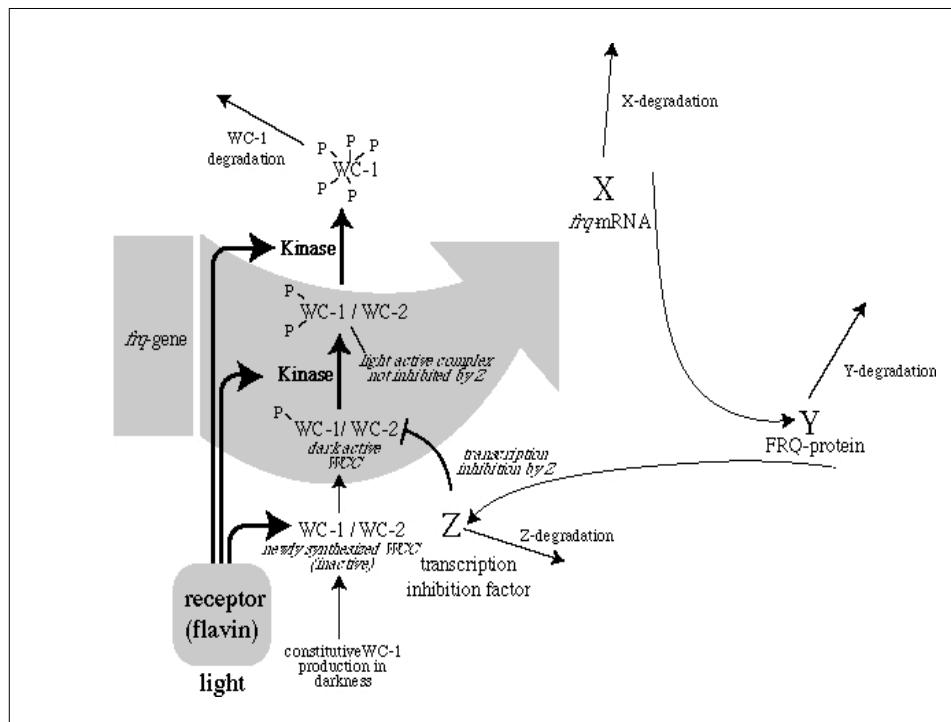
(b) When the period is decreased to 10 the oscillator is no longer able to follow the perturbing rhythm and resets to its original unperturbed period of 38.

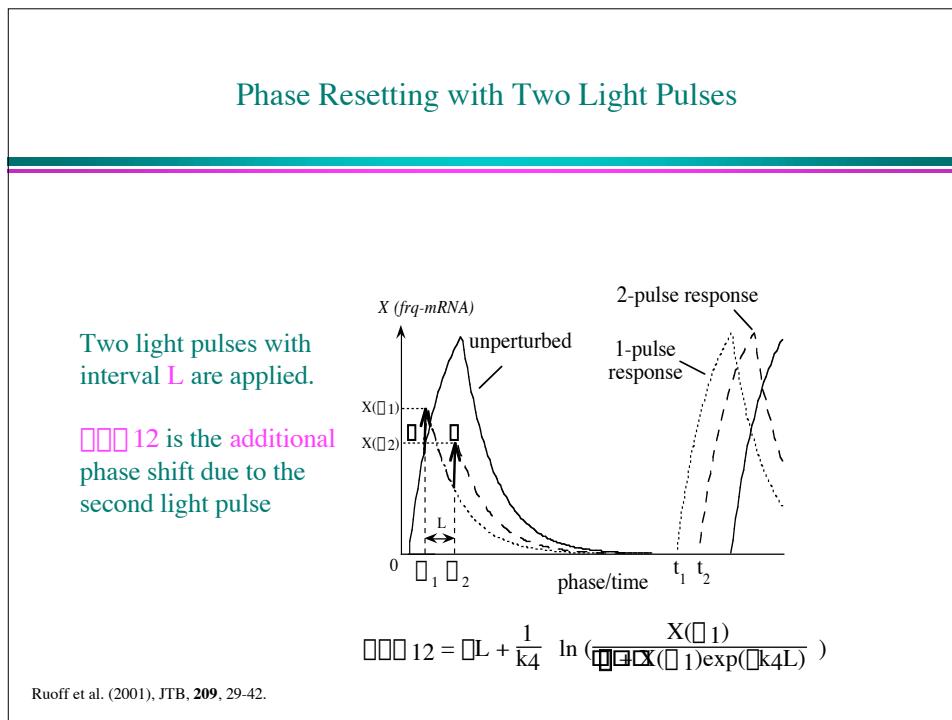
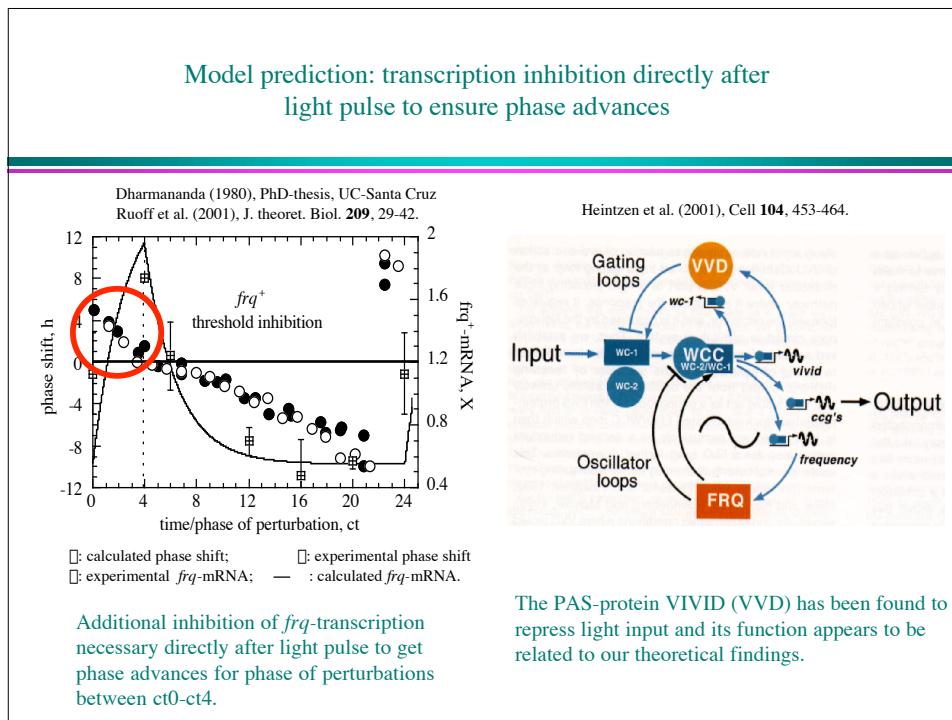
Ruoff & Rensing, JTB (1996) 275-285



Influence of light on the *Neurospora* clock:

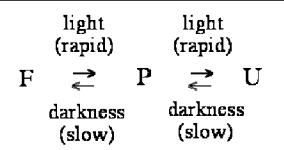
phase response curves





2-Pulse Light PRC: Experimental Results and Interpretation

Kinetics of light-signal in 2-pulse light perturbations:

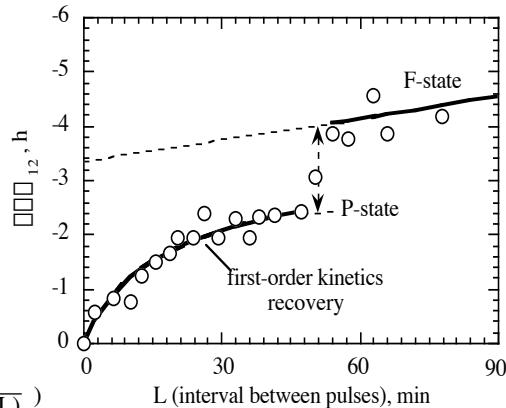


$$P = P_0(1 - \exp(-\frac{t}{T_p}))$$

$$\begin{aligned} \frac{dX}{dt} &= \frac{dP}{dt} = \frac{dP_0}{dt}(1 - \exp(-\frac{t}{T_p})) \\ &= \frac{dP_0}{dt}(1 - \exp(-\frac{t}{T_p})) \end{aligned}$$

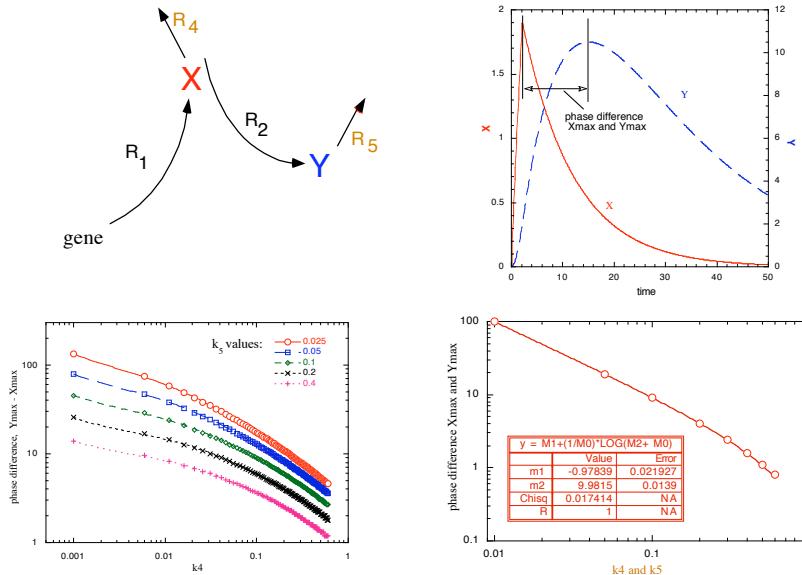
$$\frac{dX}{dt} = \frac{1}{k_4} L + \frac{1}{k_4} \ln \left(\frac{X(0)}{X(0) \exp(-\frac{L}{k_4})} \right)$$

Exp. Result: Dharmananda, S. (1980) PhD thesis.

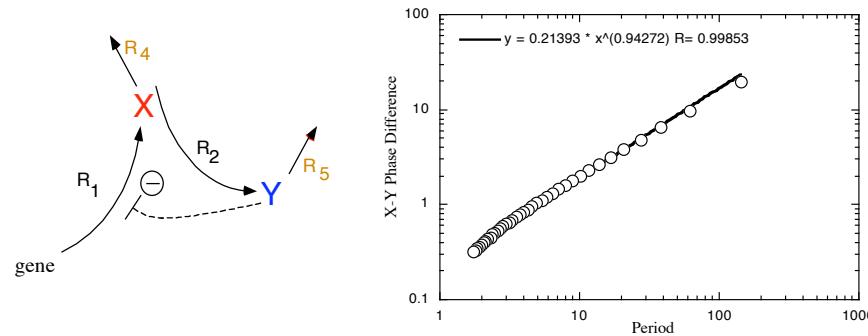


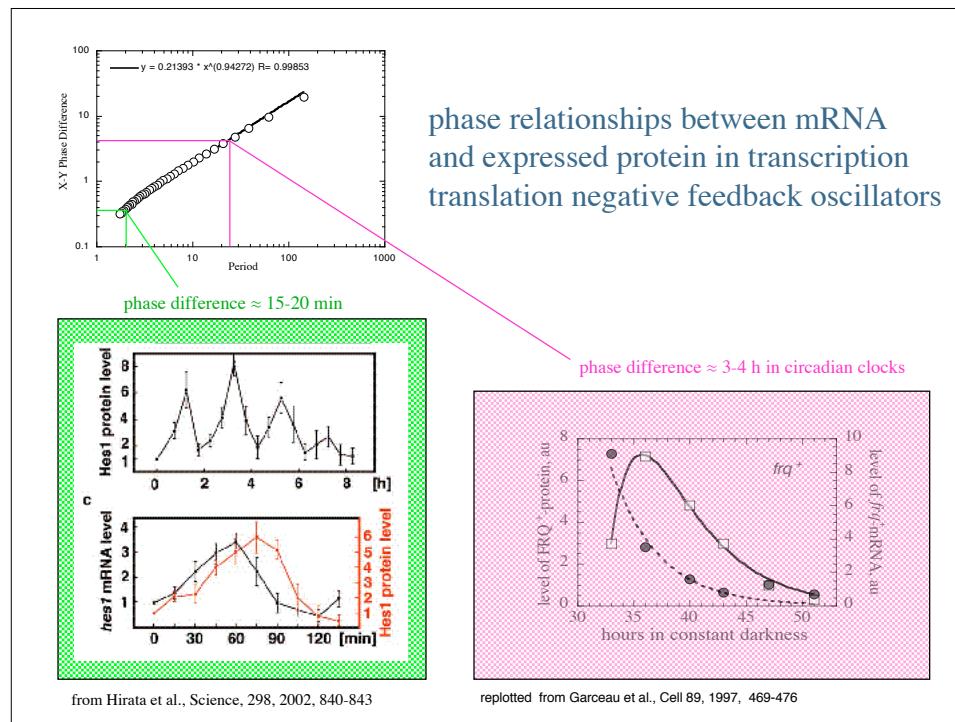
phase relationship between transcript and protein levels

phase relationship between transcript and protein levels in the transcription-translation sequence



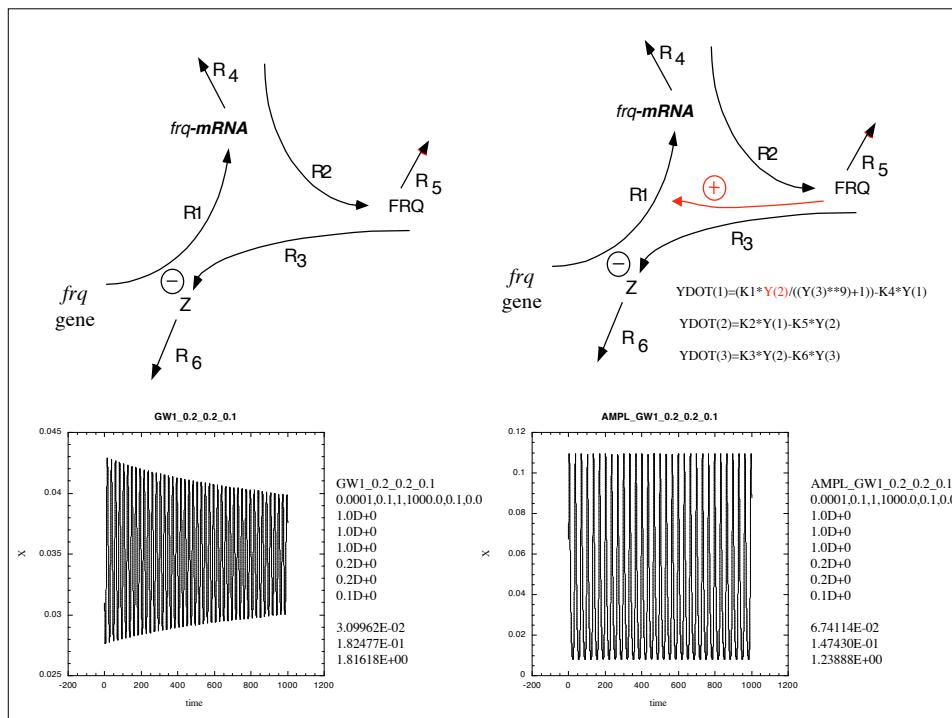
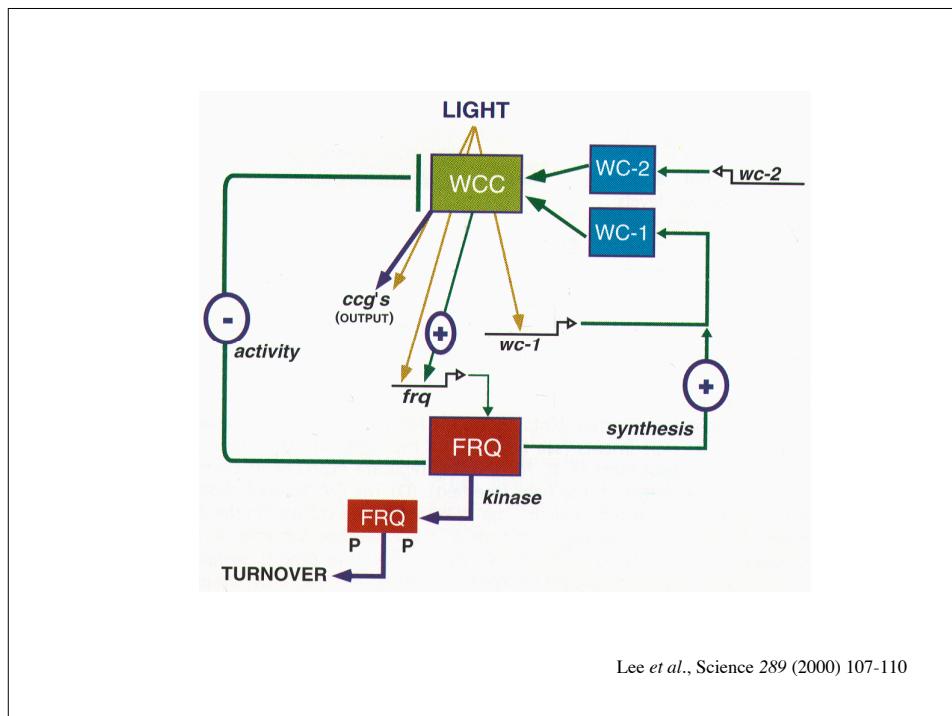
phase relationships between transcript and protein levels in the transcription-translation negative feedback loop as a function of period length

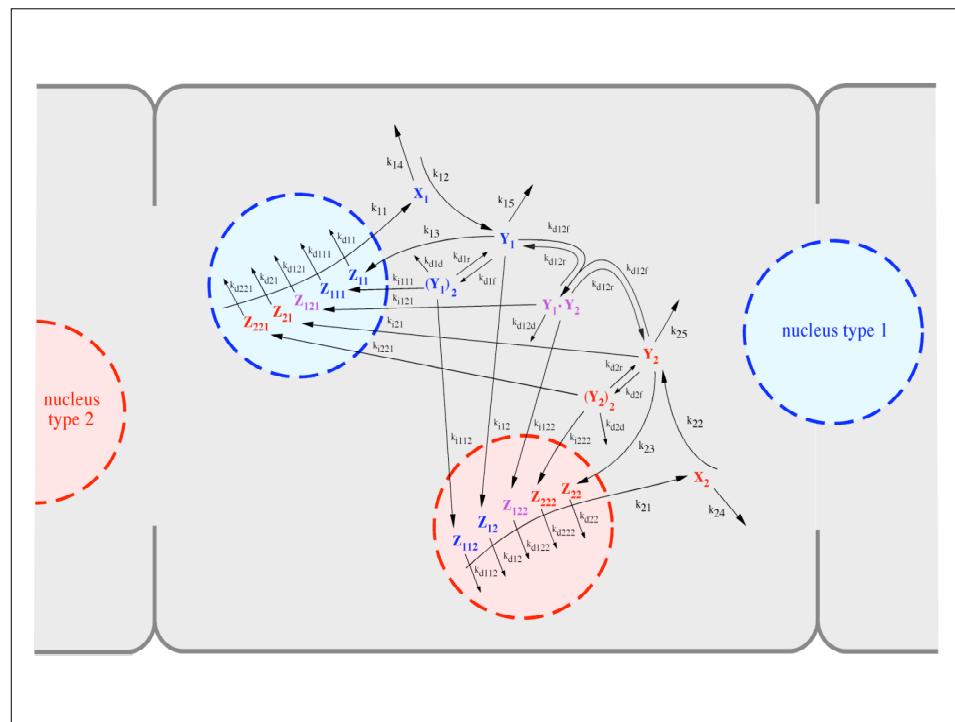
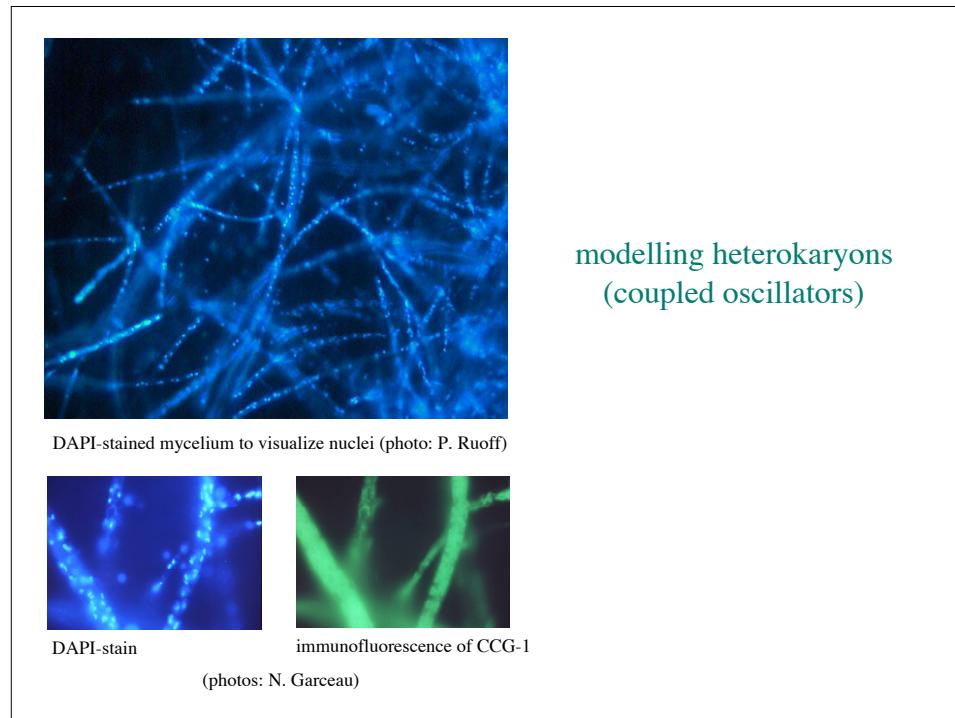




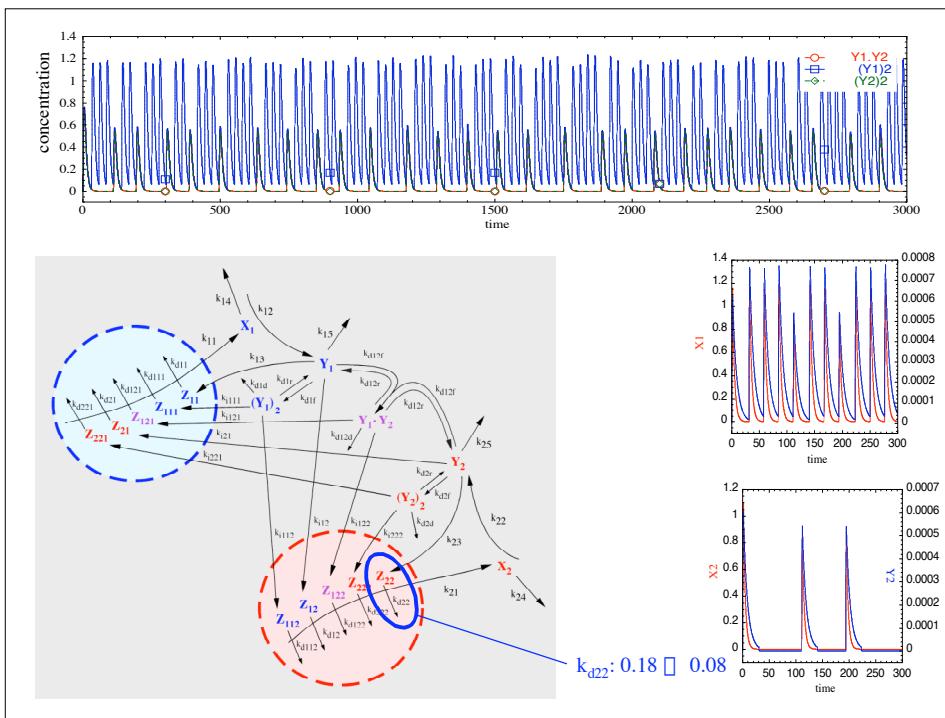
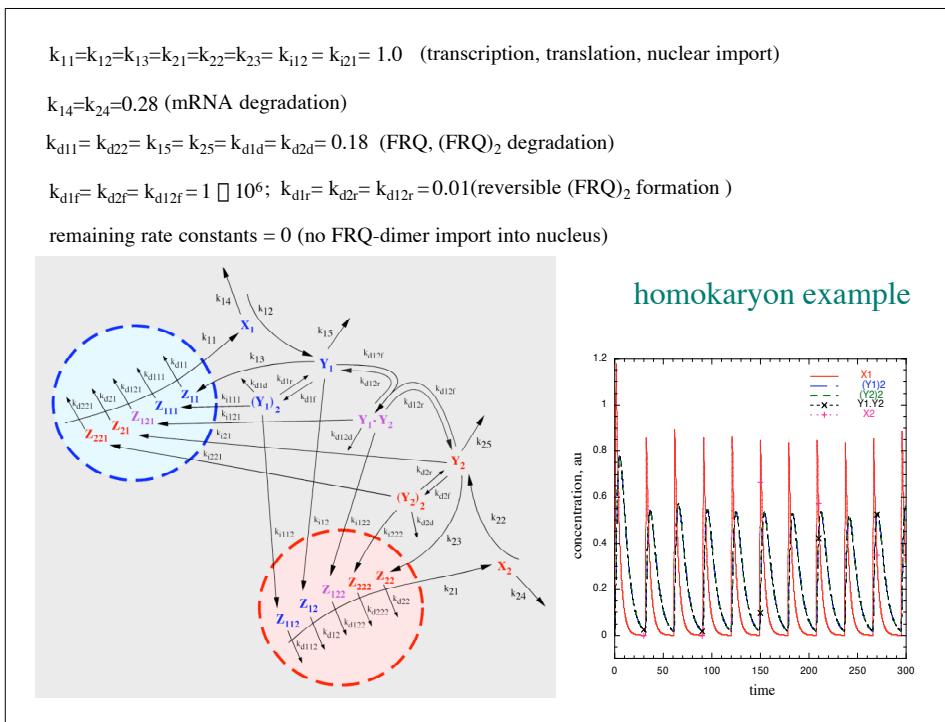
stabilizing role of positive feedback
(autocatalytic loop)

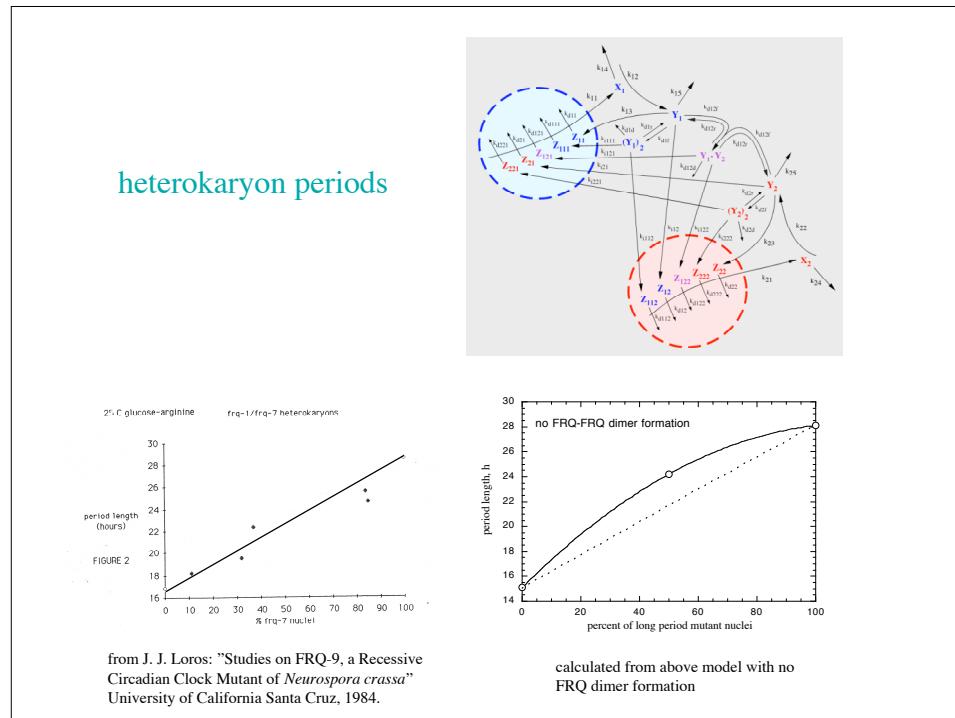
A Theory for Temperature-Compensation and Period Homeostasis in Reaction Kinetic Models of Biological Oscillators





A Theory for Temperature-Compensation and Period Homeostasis in Reaction Kinetic Models of Biological Oscillators





Summary

- Importance of *Ockham's razor* to remove redundancy in models, i.e., create minimal models.
- Minimal models can be further developed in accordance with experimental findings.
- More concerted interactions between experimentalist and theoreticians/modellers are needed in order to do more systematic experimental studies to get data for building models.



Baker Library, Dartmouth College



Lysefjord, near Stavanger

Thanks to

- Prof. Jay Dunlap for support during my stay at the Genetics Department, Dartmouth Medical School.
- Stavanger University College for giving me the possibility for a sabbatical leave.

Appendix

Euler's summation theorem

$$P = f(k_1, k_2, k_3, \dots, k_i, \dots)$$

$\left(\frac{\partial \ln P}{\partial \ln k_i}\right)$ are called *control coefficients* C_i , *period elasticities*, or *sensitivity coefficients*.

The period function $P = f$ is called homogenous to degree -1 if:

$$f(tk_1, tk_2, tk_3, \dots, tk_i, \dots) = t^{-1}f(k_1, k_2, k_3, \dots, k_i, \dots) = t^{-1}P$$

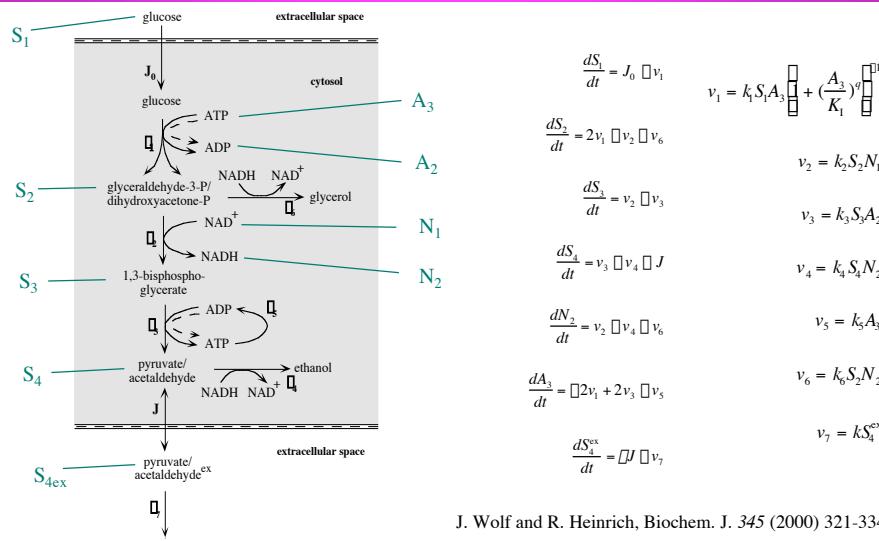
$$\boxed{\square} \quad \boxed{\square \frac{\partial \ln P}{\partial \ln k_i}} = \boxed{\square} 1$$

The temperature-compensated glycolytic oscillator:

testing Euler's summation theorem

and

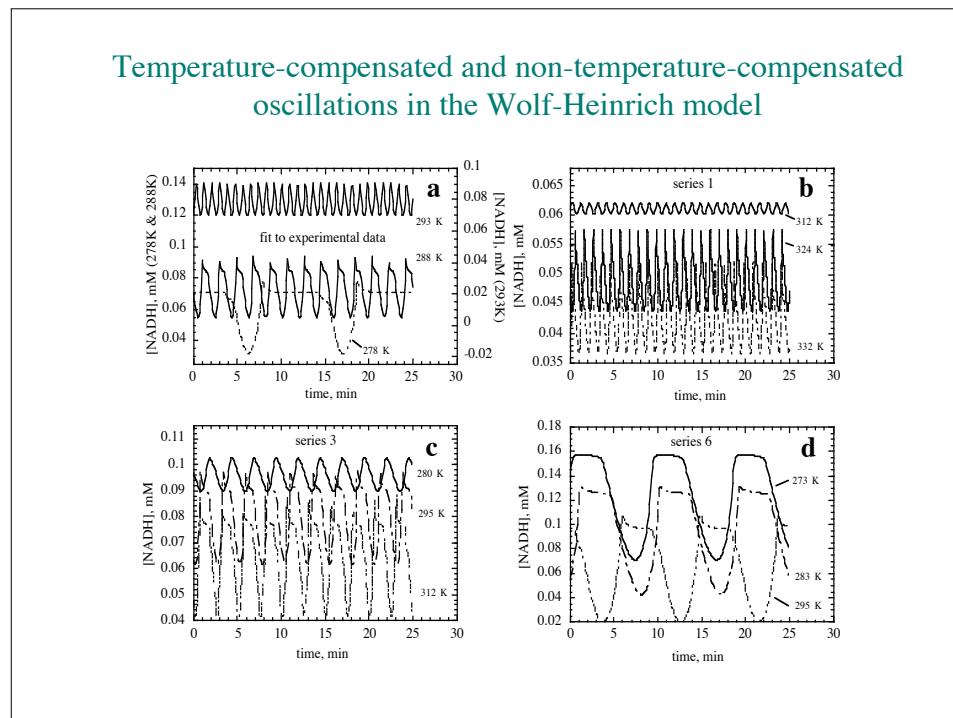
antagonistic balance condition for temperature-compensation

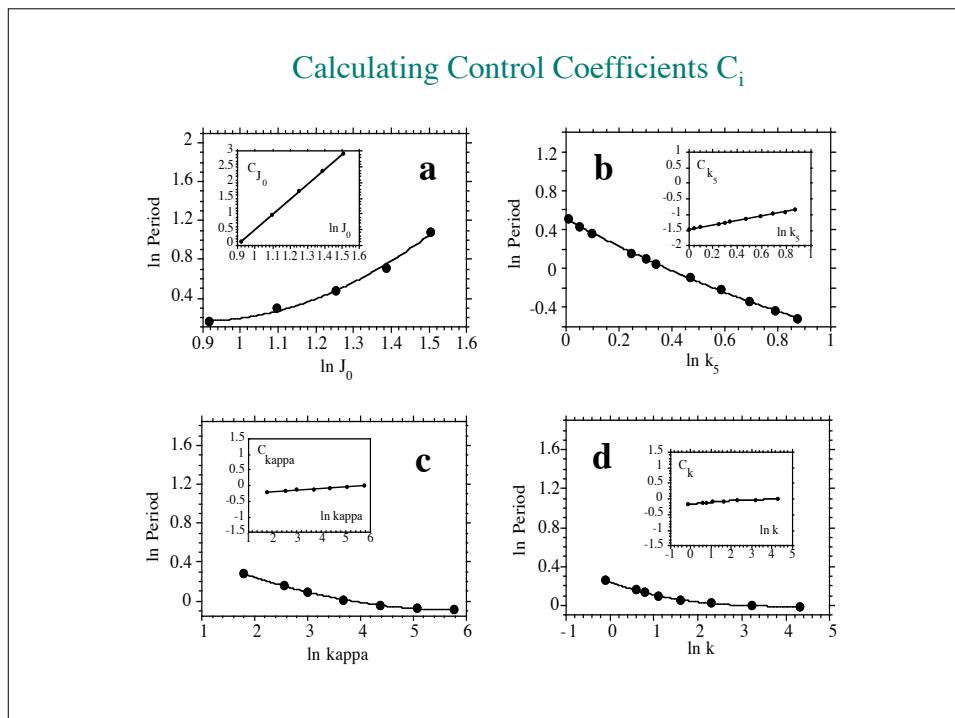
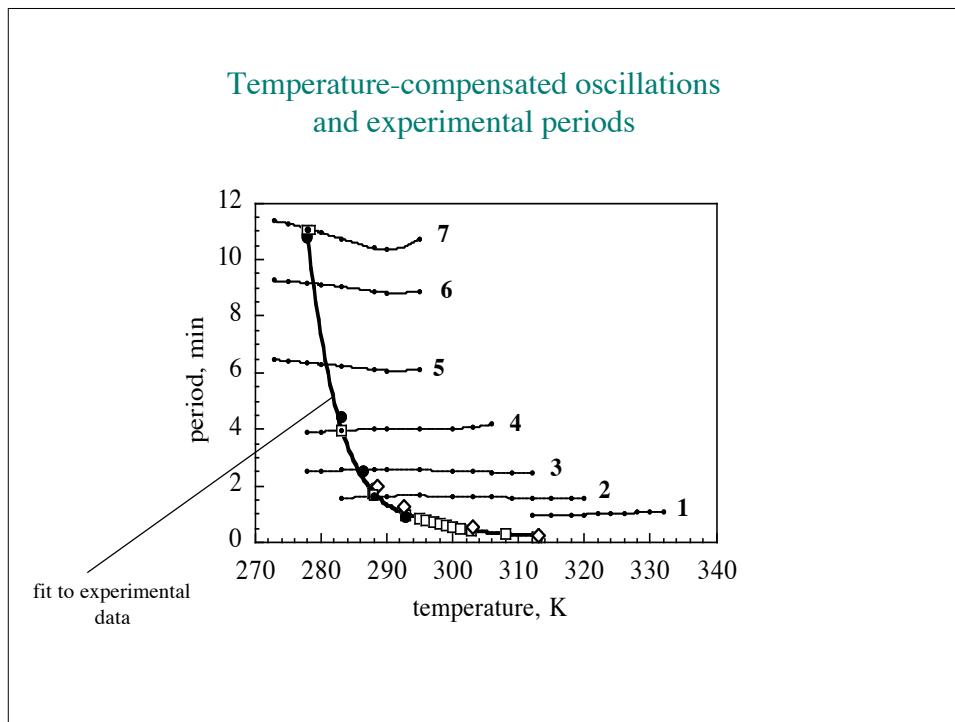


A Theory for Temperature-Compensation and Period Homeostasis in Reaction Kinetic Models of Biological Oscillators

Parameter Values of the Reference State ^a		Initial Concentrations for Reference State Oscillations	
Parameter	Value	Compound	Concentration
J_0	2.5 mM min ⁻¹	glucose	1.187 mM
k_1	100.0 mM ⁻¹ min ⁻¹	glyceraldehyde-3-P/dihydroxyacetone-P	0.193 mM
k_2	6.0 mM ⁻¹ min ⁻¹		
k_3	16.0 mM ⁻¹ min ⁻¹	1,3-bisphosphoglycerate	0.050 mM
k_4	100.0 mM ⁻¹ min ⁻¹		
k_5	1.28 min ⁻¹	pyruvate/acetaldehyde	0.115 mM
k_6	12.0 mM ⁻¹ min ⁻¹	external pyruvate/acetaldehyde	0.077 mM
k	1.8 min ⁻¹		
\bar{J}	13.0 min ⁻¹	ADP	1.525 mM
q	4.0	ATP	2.475 mM
K_1	0.52 mM		
N	1.0 mM	NAD ⁺	0.923 mM
A	4.0 mM		
$\bar{\square}$	0.1	NADH	0.077 mM

^a resulting in a period length of 1.17 min.





Control coefficients C_i for the oscillation period in the reference state		
	control coefficient	value ($\pm 1\%$ variation)
$\frac{dS_1}{dt} = J_0 \square v_1$	C_{J_0}	+ 0.69
$v_1 = k_4 S_1 A_3 \square + \left(\frac{A_3}{K_1}\right)^q \square$	C_{k_4}	0.21
$\frac{dS_2}{dt} = 2v_1 \square v_2 \square v_6$	C_{k_2}	+ 0.04
	C_{k_3}	0.01
$v_2 = k_2 S_2 N_1$	C_{k_4}	+ 0.26
$\frac{dS_3}{dt} = v_2 \square v_3$	C_{k_5}	1.24
$v_3 = k_3 S_3 A_2$	C_{k_6}	0.26
$\frac{dS_4}{dt} = v_3 \square v_4 \square J$	C_k	0.13
$v_4 = k_4 S_4 N_2$	C_J	0.17
$\frac{dN_2}{dt} = v_2 \square v_4 \square v_5$	C_{K_1}	0.86
$v_5 = k_5 A_3$	$\square C_i$ (without K_1)	
$\frac{dA_3}{dt} = \square 2v_1 + 2v_3 \square v_5$	1.02	
$v_6 = k_6 S_2 N_2$	$\square C_i$ (with K_1)	
$\frac{dS_4^{\text{ex}}}{dt} = \square J \square v_7$	1.88	
$v_7 = k S_4^{\text{ex}}$	1.89	

Testing the antagonistic balance equation					
Parameter	E_i , kJ/mol (experimental) ^a	E_i , kJ/mol (series 1)	E_i , kJ/mol (series 3)	E_i , kJ/mol (series 6)	E_i , kJ/mol (series 7)
J_0	16.2	43.9	46.1	41.4	51.0
k_3	44.9	31.7	48.7	30.4	39.2
k_4	58.7	44.2	38.0	48.1	42.4
k_1	13.8	16.0	19.1	14.0	23.3
k_2	60.7	10.4	11.8	10.8	11.9
k_5	41.2	13.3	17.9	15.1	19.8
k_6	31.4	23.7	16.2	24.0	15.7
$\square(kappa)$	15.9	22.3	19.0	18.7	19.7
k	24.3	30.0	32.2	34.1	31.5
K_f^b	47.0	12.2	11.4	8.6	10.0
$\square_i C_{k_i} E_i$	-79.0	-2.3	-6.1	-1.8	-3.6

^a fit to Hemker et al. data, Fig. 3 (large solid dots)
^b in case of K_1 , E_i is interpreted as $\square H_{K_1}^0$