Mathematical Cell Biology Graduate Summer Course University of British Columbia, May 1-31, 2012 Leah Edelstein-Keshet

Simple biochemical motifs (1)

www.math.ubc.ca/~keshet/MCB2012/

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Biochemical (and gene) circuits

Switches, oscillators, adaptation, and amplification circuits

Production-decay at constant rates



$$\frac{dx}{dt} = I - \gamma x$$

$$I, \gamma > 0$$
 constants.

Unique positive Steady state



Signal-induced Production



$$\frac{dR}{dt} = k_0 + k_1 S - k_2 R.$$

Note typical "1- $\exp(-k_2 t)$ " rise and exponential decay tail



Feedback to production



I is now a function of x

Michaelian Feedback to production



Michaelian Feedback to production



Sigmoidal Feedback to production



Sigmoidal cont'd

$$\frac{dx}{dt} = f(x) = \frac{x^2}{1+x^2} - mx + b$$

Actual number of steady states depends on parameters, e.g. on slope m (decay rate of x)



Generic bistability



Bifurcation Diagram



Hysteresis





Adaptation

$$\frac{dR}{dt} = k_1 S - k_2 X R,$$
$$\frac{dX}{dt} = k_3 S - k_4 X.$$



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Simple biochemical motifs (2)

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Genetic toggle switch

Construction of a genetic toggle switch in *Escherichia coli*

Timothy S. Gardner*+, Charles R. Cantor* & James J. Collins*+

NATURE VOL 403 20 JANUARY 2000 www.nature.com

An actual "engineered genetic circuit" based on the concepts and models of biochemical switches.



Genetic toggle switch

"Here we present the construction of a genetic toggle switch: a synthetic, bistable gene-regulatory network in E. coli and provide .. theory that predicts conditions for bistability."

Production-decay of two proteins

Gene U



Gene V

 $\frac{du}{dt} = I_u - d_u u,$ $\frac{dv}{dt} = I_v - d_v v.$

Negative feedback

Gene U



Gene V

du dt $=I_u-d_u u,$ $=I_{v}-d_{v}v.$ dv dt α

Negative feedback function

$$I_x=\frac{\alpha}{1+x^n}.$$

Higher *n* means sharper response with increasing *x*



Mutual inhibition



Each gene product inhibits the other gene.

"... the toggle equations have 2 fundamental aspects: cooperative repression and degradation .. of the repressors"

Switch-like behaviour

$$\frac{du}{dt} = \frac{\alpha_1}{1+v^n} - u,$$
$$\frac{dv}{dt} = \frac{\alpha_2}{1+u^m} - v.$$



Plasmid circuit

a synthetic, bistable gene-regulatory network in E. coli



Cells switching can be induced



fluorescence

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Simple biochemical motifs (2.5)

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Noise-based switches and amplifiers for gene expression

Jeff Hasty*[†], Joel Pradines*, Milos Dolnik*[‡], and J. J. Collins*

PNAS | February 29, 2000 | vol. 97 | no. 5 | 2075-2080

Dimerization and the phage lambda



- The phage λ gene encodes for protein (conc *x*)
- Protein dimerizes (conc of dimers y).
- Dimers bind to regulatory sites on the gene.
- Binding to OR2 activates transcription.
- Biding to OR3 inhibits transcription.

Reaction scheme

Dimerization: $2X \xleftarrow{K_1}{K_1} X_2$

Binding to DNA (OR2): $D + X_2 \xleftarrow{K_2} DX_2$

Binding to DNA (OR3): $D + X_2 \xleftarrow{K_3} DX_2^*$

Double binding (OR2 and OR3): $DX_2 + X_2 \xleftarrow{K_4} DX_2X_2$

 DX_2 = the dimerized repressor bound to site OR2 DX_2 * = the dimerized repressor bound to site OR3, DX_2X_2 = both OR2 and OR3 are bound by dimers



$$y = K_1 x^2,$$

$$u = K_2 dy = K_1 K_2 dx^2,$$

$$v = \sigma_1 K_2 dy = \sigma_1 K_1 K_2 dx^2,$$

$$z = \sigma_2 K_2 uy = \sigma_2 (K_1 K_2)^2 dx^4.$$

The "fast variables" assumed to equilibrate rapidly with the variable *x*.

Slower timescale

Protein synthesis: $DX_2 + P \xrightarrow{k_1} DX_2 + P + nX$ Protein degradation: $X \xrightarrow{k_d} A$

QSS and scaling the equations: system collapses to one variable, amt of synthesized protein, *x*:

$$\frac{dx}{dt} = \frac{\alpha x^2}{1 + (1 + \sigma_1)x^2 + \sigma_2 x^4} - \gamma x + 1.$$



(a)

(b)

Bifurcation:

$$\frac{dx}{dt} = \frac{\alpha x^2}{1 + (1 + \sigma_1)x^2 + \sigma_2 x^4} - \gamma x + 1.$$



Comments

Combination of scaling, time scale considerations, and various simplifications can often reduce larger networks to effective dynamics of simpler systems.

Other examples will be provided.

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Simple biochemical motifs (3)

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Activation-inactivation



GTPase cycle



Without feedback: Fast equilibration



Time (seconds)

System has a single biologically relevant steady state

Eliminate R, rescale

$$\frac{dR_p}{dt} = \frac{k_1 SR}{K_{m1} + R} - \frac{k_2 R_p}{K_{m2} + R_p}.$$

Use

$$R_T = R + R_p = \text{ constant.}$$

$$r_p = R_p/R_T$$



Steady states

$$\frac{dr_p}{dt} = \frac{k_1 S(1-r_p)}{K'_{m1} + (1-r_p)} - \frac{k_2 r_p}{K'_{m2} + r_p} = 0$$

The steady states can be shown to be solutions to a quadratic equation. Only one is positive and is called the "Goldbeter-Koshland function" of the stimulus.

"Zero order ultrasensitivity"

Steady state response



response is minimal for low signal level, until some threshold. Then there is steep rise to full response. – *Goldbeter and Koshland*