

4 A differential equation model of the lac operon

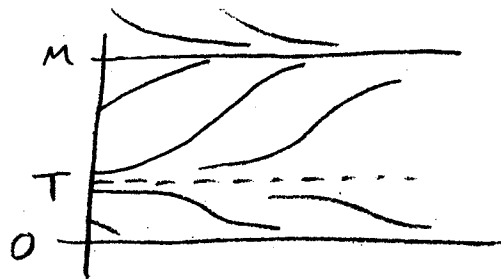
A system is bistable if it is capable of resting in two stable steady states separated by an unstable steady state.

Ex: Recall the threshold population model

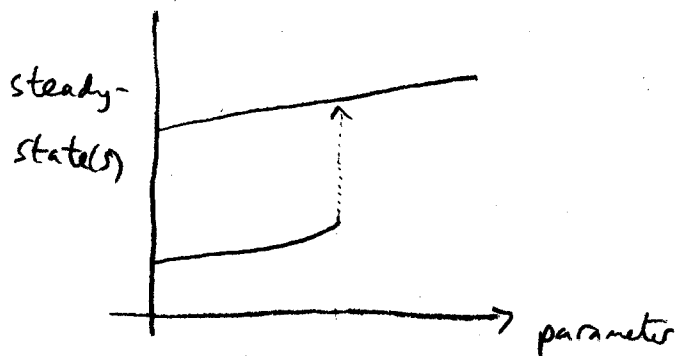
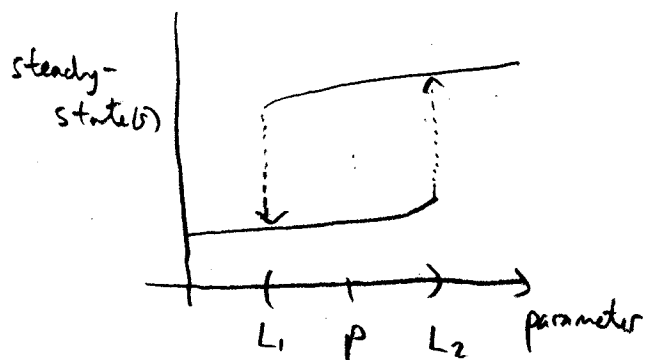
$M$  = carrying capacity (stable)

$T$  = extinction threshold (unstable)

$0$  = extinct (stable)



Bistability can be reversible or irreversible:



One can plot the steady-state as a function of (tuning) some parameter (see above).

Note that the "up-switch"  $L_2$  is higher than the "down-switch"  $L_1$ .

This is hysteresis - a dependence of a state on its current state and past state.

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The lac operon exhibits a hysteresis effect & bistability:

If lactose concentration:  $[L] < L_1$ , operon is off

$[L] > L_2$ , operon is on.

$L_1 < [L] < L_2$ , operon could be on or off.

This "region of bistability",  $(L_1, L_2)$ , has both induced and uninduced cells.

Cells that were grown in a lactose environment likely "on"

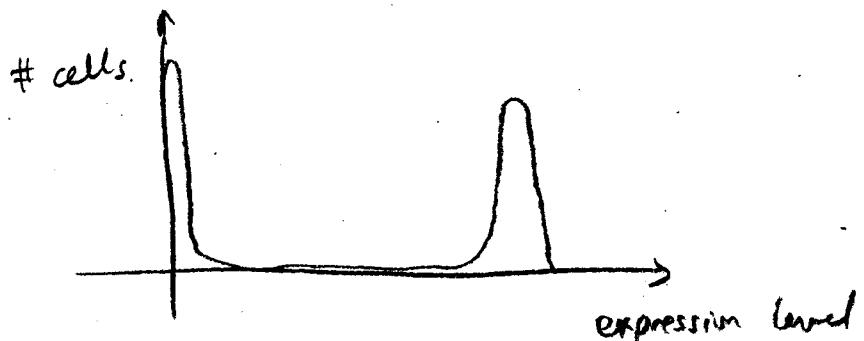
Cells grown in an environment with low levels of lactose likely to be "off."

Analogy: Thermostat turns heater: on if temp  $< 67$   
off if temp  $> 72$

x If temp is  $70^\circ$ , it's impossible to know if heater is on or off without knowing past data.

Distribution of gene expression levels of cells for some

$[L] \in (L_1, L_2)$ :



The simple Boolean network models we've seen can't capture bistability.

We'll see an ODE model, & a Boolean model that can.

Bistable systems often have positive feedback loops, or a double-negative feedback loop.

Model by Yildirim & Mackey, 2004

3 variables:  
 $M = \text{mRNA}$   
 $B = \beta\text{-galactosidase}$   
 $A = \text{allolactose}$

Assumption: Internal lactose (L) is available & is a parameter.

### Preliminaries

- Modeling dilution in protein concentration due to bacterial growth

E-coli population grows fast! (can double in 20 minutes!)

Let  $V = \text{ave. volume of bacterial cell}$

$x = \# \text{ molecules of protein } X \text{ in that cell.}$

Assume cell volume increases exponentially in time:  $\frac{dV}{dt} = \mu V$

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Assume degradation of  $X$  is exponential:  $\frac{dx}{dt} = -\beta x$

Concentration  $[x] = \frac{x}{V}$

$$\begin{aligned} \Rightarrow \frac{d[x]}{dt} &= (x'V + V'x) \frac{1}{V^2} = (-\beta xV - \mu Vx) \frac{1}{V^2} \\ &= -(\beta + \mu) \frac{x}{V} = -(\beta + \mu)[x]. \end{aligned}$$

• Modeling of lactose repressor dynamics

Assume: lac repressor protein is produced at const. rate (regardless of external lactose).

\* Repressor binds to allolactose:  $R + nA \xrightleftharpoons[1]{K_1} RA_n$

$$\frac{d[RA_n]}{dt} = K_1 [R][A]^n - [RA_n]$$

Assume reaction is at equilibrium:  $\frac{d[RA_n]}{dt} = 0 \Rightarrow K_1 = \frac{[RA_n]}{[R][A]^n}$

\* Repressor binds to operator region (O) if no allolactose:



At equilibrium:  $K_2 = \frac{[OR]}{[O][R]}$

Let  $O_{tot}$  = total operator concent. (const.),

↓ eliminate

Then  $O_{tot} = [O] + [OR] = [O] + K_2 [O][R]$   
 $= [O](1 + K_2)[R]$

$\Rightarrow \frac{[O]}{O_{tot}} = \frac{1}{1 + K_2 [R]}$  "proportion of free (unbound) operator sites."

Let  $R_{tot}$  be total concentration of repressor protein.

Assume const. (since not regulated by extracellular lactose)

$\Rightarrow R_{tot} = [R] + [OR] + [RA_n]$

Assume only a few molecules of operator site/cell

$\Rightarrow [OR] \ll \max\{[R], [RA_n]\}$

$\Rightarrow R_{tot} = [R] + [RA_n] = [R] + K_1 [R][A]^n$   
 eliminate  $\uparrow$

$\Rightarrow [R] = \frac{R_{tot}}{1 + K_1 [A]^n}$

Plug into  $\frac{[O]}{O_{tot}} = \frac{1}{1 + K_2 \left( \frac{R_{tot}}{1 + K_1 [A]^n} \right)} \cdot \frac{1 + K_1 [A]^n}{1 + K_1 [A]^n}$   
 $= \frac{1 + K_1 [A]^n}{K + K_1 [A]^n}$  where  $K = 1 + K_2 R_{tot}$   
 $\underbrace{\hspace{10em}}_{:= f([A])}$

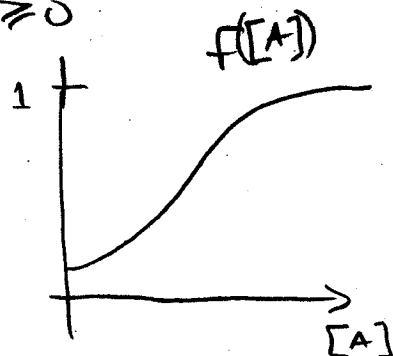
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Remarks: •  $f([A]=0) = \frac{1}{K} > 0$  "minimal basal level of gene expression"

•  $f$  is increasing in  $[A]$ , when  $[A] \geq 0$

•  $\lim_{[A] \rightarrow \infty} f([A]) = 1$

"With abundance of allolactose,  
gene expression level is max."



• How to incorporate  $f'(t) = -\mu p(t - \tau_m)$  [exp. decay w/ time delay]

Suppose Protein P decays exponentially,  $\text{concent.} = p$ .

$$\frac{dp}{dt} = -\mu p \Rightarrow \int_{t-\tau_m}^t \frac{dp}{p} = -\mu \int_{t-\tau_m}^t dt$$

$$\Rightarrow \ln p(t) \Big|_{t-\tau_m}^t = -\mu t \Big|_{t-\tau_m}^t dt$$

$$= \ln \frac{p(t)}{p(t-\tau_m)} = -\mu [t - (t-\tau_m)] = -\mu \tau_m$$

$$\frac{p(t)}{p(t-\tau_m)} = e^{-\mu \tau_m}$$

$$\Rightarrow p(t) = e^{-\mu \tau_m} p(t-\tau_m)$$

↑ accounts for time delay  $\tau_m$

Yildirim-Mackey model for lac operon (variables  $M, B, A$ ).

$$(1) \frac{dM}{dt} = \alpha_M \frac{1 + K_1 (e^{-\mu \tau_M} A_{\tau_M})^n}{K + K_1 (e^{-\mu \tau_M} A_{\tau_M})^n} - \tilde{\gamma}_M M \quad (\text{mRNA})$$

where  $A_{\tau_M} = A(t - \tau_M)$  (time delay  $\tau_M$ )

$\tilde{\gamma}_M = \gamma_M + \mu$  (degradation + dilution constants)

Justification:

\* Production rate of mRNA,  $\frac{dM}{dt}$ , proportional to fraction of free operator sites:  $\frac{[O]}{O_{tot}} = \frac{K_1 [A]^n}{K + K_1 [A]^n}$

\*  $e^{-\mu \tau_M} A_{\tau_M}$  accounts for the concentration of  $A$  at time  $t - \tau_M$ , which is proportional to  $\frac{dA}{dt}$ .

\*  $\tilde{\gamma}_M M = \gamma_M M + \mu M$   
 $\uparrow$  loss due to mRNA degradation       $\uparrow$  loss due to dilution from bacterial growth.

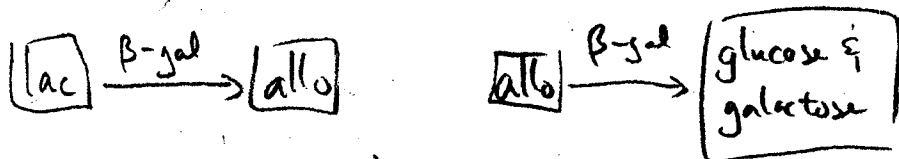
$$(2) \frac{dB}{dt} = \alpha_B e^{-\mu \tau_B} M_{\tau_B} - \tilde{\gamma}_B B \quad (\text{ $\beta$ -galactosidase})$$

$\tau_B =$  time required for mRNA translation,  $\tilde{\gamma}_B = \gamma_B + \mu$ .

$$M_{\tau_B} = M(t - \tau_B)$$

(2)

$$(3) \frac{dA}{dt} = \underbrace{\alpha_A B \frac{L}{K_L + L}}_{\text{Michaelis-Menten eq'n}} - \underbrace{\beta_A B \frac{A}{K_A + A}}_{\text{Michaelis-Menten eq'n}} - \underbrace{\tilde{\tau}_A A}_{\text{degradation \& dilution}} \quad (\text{allolactose})$$



(Michaelis-Menten eq'n)

These are "delay differential eq'ns with discrete timesteps."

Next: Steady-state analysis & numerical simulations.

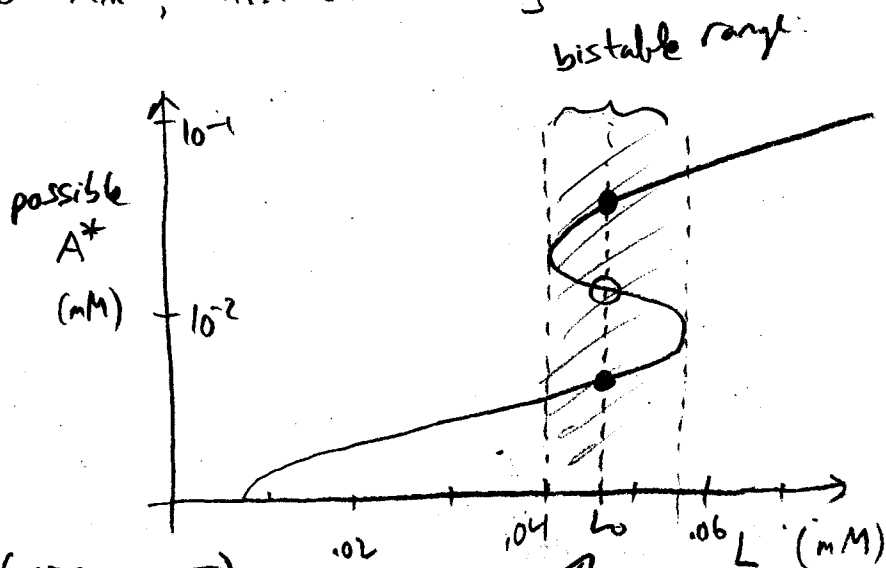
Need to fix unknown parameters, constants.

Approach: Estimate by fitting these models to experimental data.

(e.g.,  $\mu = 3.03 \times 10^{-2} \text{ min}^{-1}$ , Hill coeff.  $n=2$ )

Steady-state analysis:

- Set  $M' = B' = A' = 0$
- Solve for range of  $L$  concentrations (L is the "tuning parameter").



Result Bi-stable range:  $L \in (.039, .0055)$

3 steady-states; 2 stable