

# W3-Note

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## W3. Bistability in Biochemical Signaling Models

### Q1. Some biological background

- Biological importance of bistability
- Qualitative requirements of bistability

**Def of biastability:** a situation where two possible steady-states are both stable.

In general, these correspond to a “**low activity**” state and a “**high activity**” state.

- **classic experiment:** add progesterone to Xenopus (frog) oocytes, measure MAPK activity
- **Population response:** gradual increase in MAPK with progesterone

What happens when MAPK activity is measured in each cell?

- With increasing progsterone, oocytes switch from **low** state to **high** state.
- An intermediate [progesterone] both high and low states are present.

**Biology: generally monostable and analog** response depends directly on level of stimulus.

When stimulus removed, response returns to prior level.

**When is analog not good enough?** Ex: fertilization, action potentials, cell division, apoptosis, differentiation, learning.

For these processes, a graded response is inadequate.

These phenomena also require persistence.

**Biochemically, how does bistability arise?**

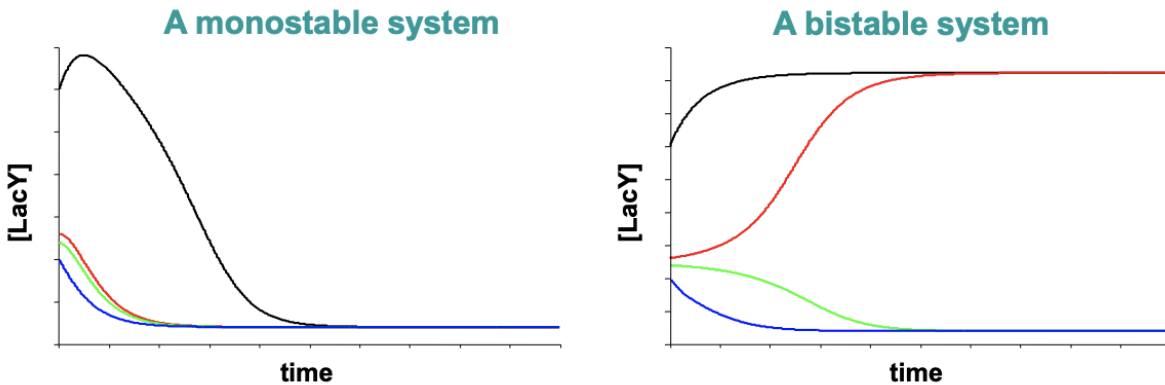
1. Mutual activation
2. Mutual inhibition

These types of circuits CAN produce bistability, but they do NOT guarantee bistability.

=> so we need quantitative analysis.

**Bistability in terms of dynamical behaviour**

stable & unstable, fixed points & limit cycles



Multiple steady-states are possible.

IC determine which steady-state is reached.

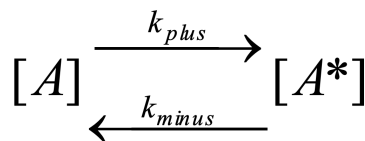
Thus.

- Bistability can be a useful property for biological processes that require persistence.
- Bistability means that a biological response will be essentially digital, or all-or-none, rather than graded.
- In the language of dynamical systems, bistability means that two fixed points are possible, with initial conditions determining which fixed point is reached.

## Q2. How to predict if bistability will be present?

- A simple, 1D sample
- Rate-balance plots
- Ultrasensitive positive feedback can create bistability
- rate-balance plots
- Example of rate-balance plots in MATLAB

## Quantitative analyses of bistability



Example 1. A simple “Michaelian” system

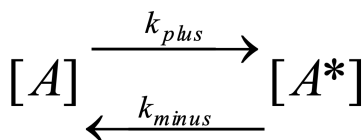
- $A^*$  = phosphorylated A
- Total amount of [A] is constant:  $[A]_{TOTAL} = [A] + [A^*]$
- To solve for  $[A^*]$  in the steady-stat:

$$\frac{d[A^*]}{dt} = k_{plus}([A]_{TOTAL} - [A^*]) - k_{minus}[A^*] = 0$$

$$[A^*] = \frac{k_{plus}[A]_{TOTAL}}{k_{plus} + k_{minus}} \quad \text{or} \quad \frac{[A^*]}{[A]_{TOTAL}} = \frac{1}{1 + \frac{k_{minus}}{k_{plus}}}$$

\* Rate balance plots

Instead of solving equations, find solution graphically:



1. Forward Rate & Backward Rate

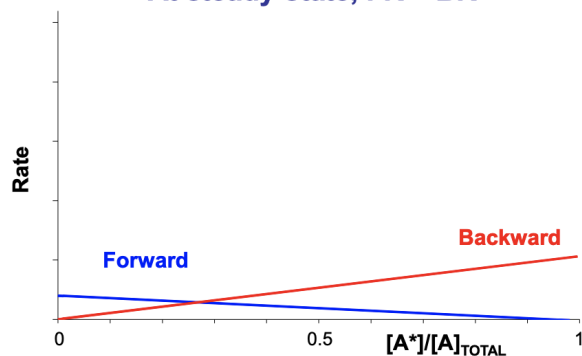
**Forward Rate**

$$FR = k_{plus}([A]_{TOTAL} - [A^*])$$

**Backward Rate**

$$BR = k_{minus}[A^*]$$

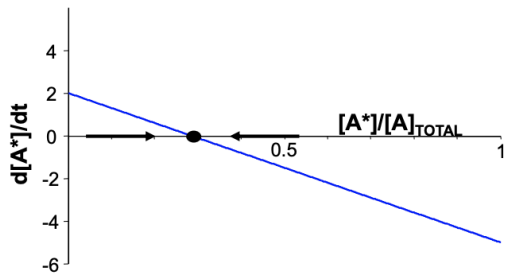
**At steady-state, FR = BR**



$$\frac{d[A^*]}{dt} = FR - BR$$

**At steady-state, FR - BR = 0**

2.  $dA/dt = FR - BR$

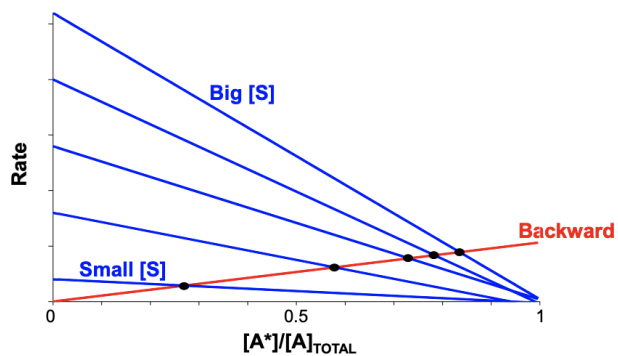


Intuitively, then, this fixed point is stable

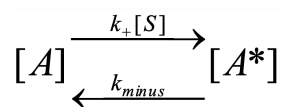
$$k_{plus} = k_+[S]$$

3. Assume FR is a function of stimulus:  $k_{plus} = k_+[S]$

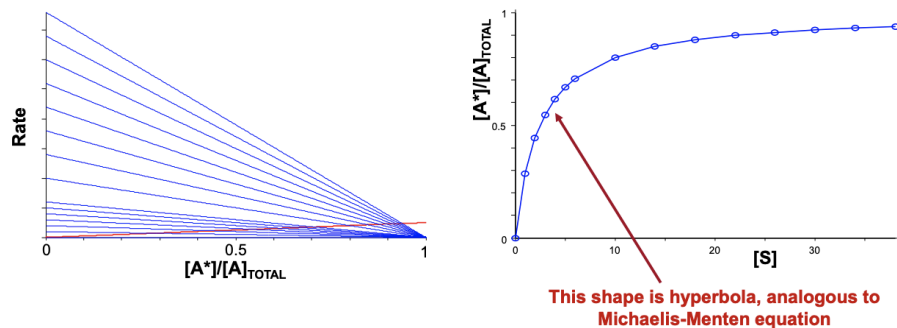
Plot rate balance for different values of stimulus [S]

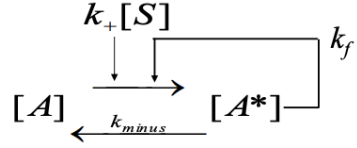


The reaction becomes:



This analysis can be used to plot [S] versus  $[A^*]/[A]_{TOTAL}$



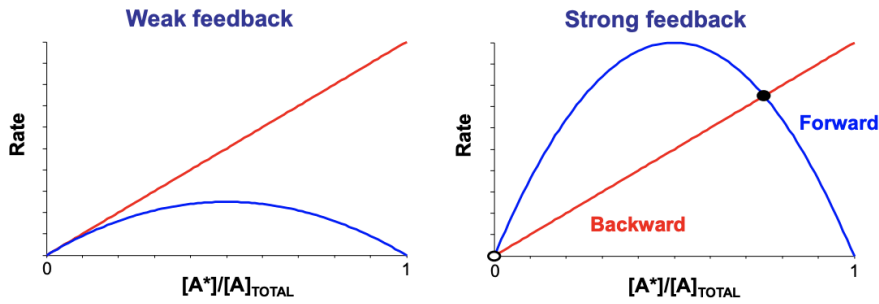


Example 2. Michaelian system with linear feedback

Forward Rates:

$$FR = (k_+[S] + k_f[A^*])([A]_{TOTAL} - [A^*])$$

**$k_f$  determines strength of feedback**



**The right plot "looks" bistable. Is it? Answer: No.**

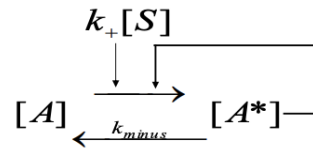
The right plot is not bistable, as:

consider a tiny deviation from  $A^* = 0$ , (a spontaneous phosphorylation).

FR exceeds BR, this leads to a further increase in  $A^*$ , so this steady state is unstable.

**How to enable the state stable** 1). Non-linear ("ultrasensitive") feedback

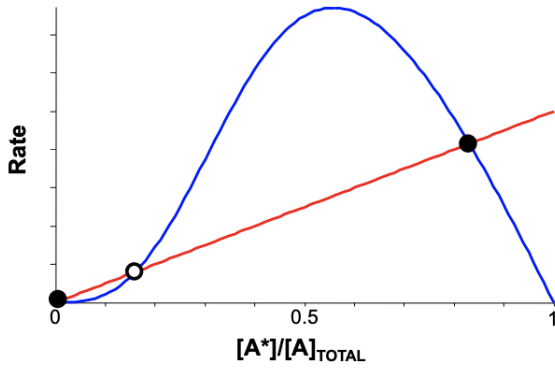
2). Partial saturation of the back reaction



Example 3. Michaelian system with ultrasensitive feedback

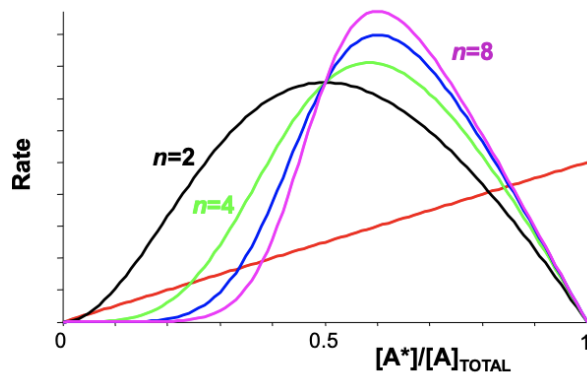
FR:

$$FR = \left( k_+[S] + k_f \frac{[A^*]^n}{[A^*]^n + K_{mf}^n} \right) ([A]_{TOTAL} - [A^*])$$



Change in hill exponent  $n$ :  $n \uparrow$  bistability likely & robust  $\uparrow$

### Effects of changes in hill exponent $n$



A larger hill exponent makes bistability more likely and more robust

- Rate-balance plots are useful for assessing whether bistability may occur in one-variable systems.
- Ultrasensitive positive feedback can produce bistability in a onevariable system.

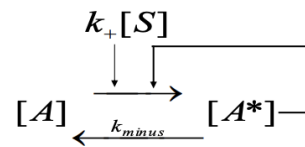
Example 4. Linear feedbacks plus saturating back reaction

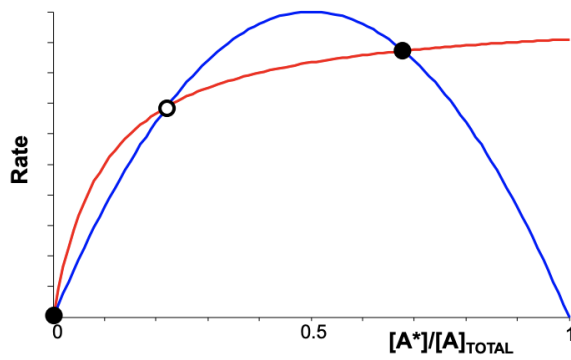
FR:

$$FR = (k_+[S] + k_f[A^*])([A]_{TOTAL} - [A^*])$$

BR:

$$BR = k_{minus} \left( \frac{[A^*]}{[A^*] + K_{mb}} \right)$$



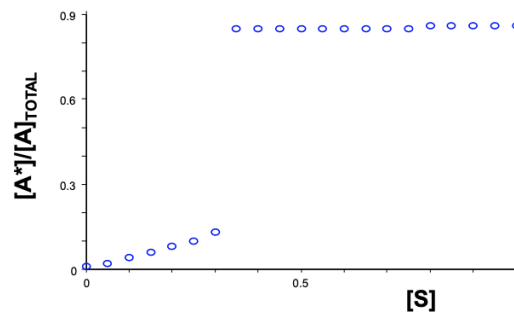
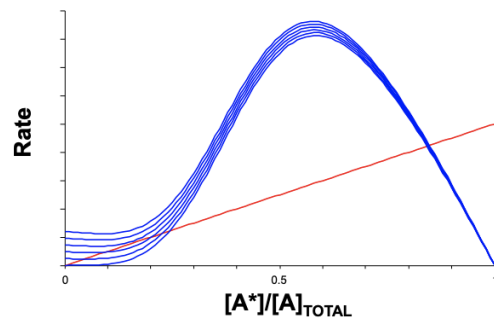


How can the cell change states?

Vary the amount of stimulus  $[S]$

Most plots have assumed  $[S] = 0$

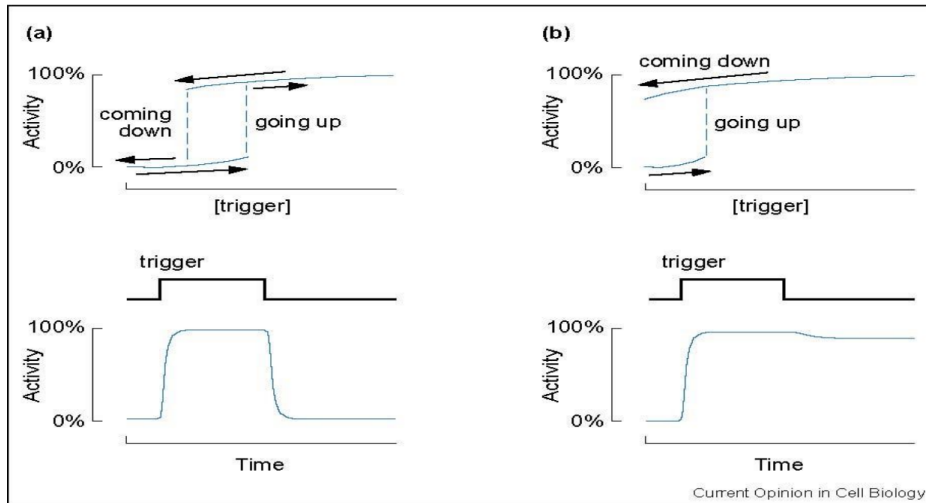
$$FR = \left( k_+[S] + k_f \frac{[A^*]^n}{[A^*]^n + K_{mf}^n} \right) ([A]_{TOTAL} - [A^*])$$



**Where the system switches between 3 and 1 steady-states is a bifurcation**

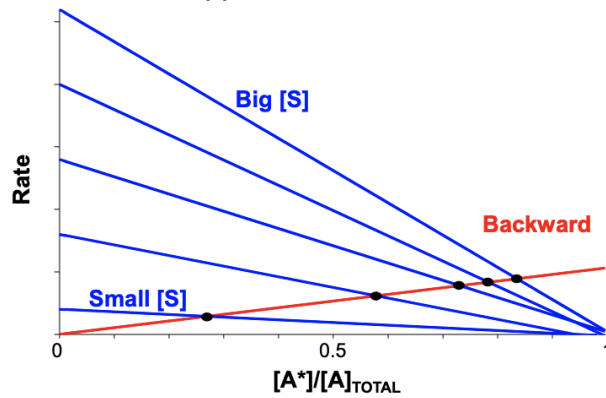
Switching can be reversible or irreversible

In either case, transition on the way up is higher than transition on the way down.

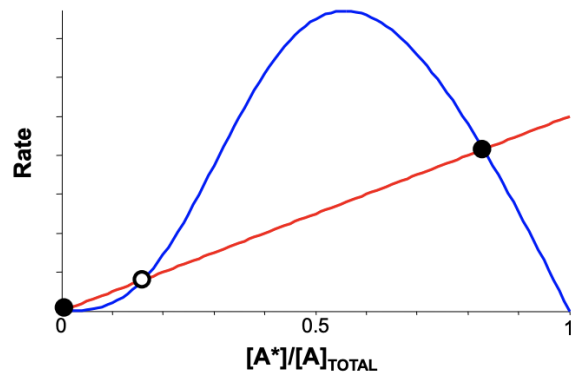


### --- rate-balance plots in MATLAB ---

1. Condition 1. No feedback



2. Condition 2. Ultrasensitive positive feedback



- In a one-variable system, bistability can be produced by:
  - ultrasensitive positive feedback
  - a back reaction that saturates
- Analysis of rate-balance plots can generate a bifurcation diagram showing a transition from monostability



to bistability.

- Array arithmetic in MATLAB can be used to produce helpful rate balance plots.

### Q3. Bistability in 2 variables systems

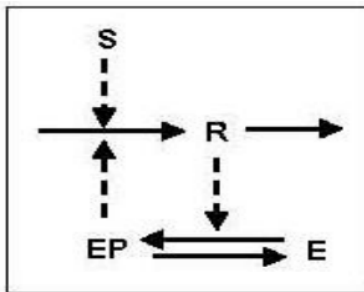
- Can occur by mutual activation or mutual repression
- Dynamic simulations can demonstrate bistability
- Bifurcation plots establish bistable regime

How to predict where bistability will be present?

-- 1). Plot nullclines in the phase planes

Analysis of 2 variables systems

R = response  
S = stimulus  
E = enzyme



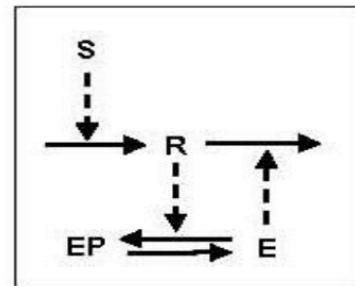
- R causes phosphorylation of E
- EP leads to synthesis of R

$$\frac{d[R]}{dt} = k_{1R}([E]_{TOTAL} - [E]) + k_{1R}[S]$$

$$\frac{d[E]}{dt} = -k_{2E}[R] \frac{[E]}{[E] + K_{m2E}} + k_{1E} \frac{[I]}{[E]_{TO}}$$

For [S] less than ~11, two steady states are possible.

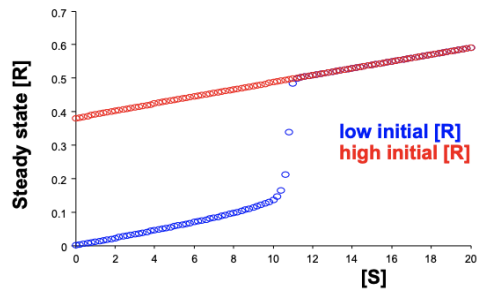
R = response  
S = stimulus  
E = enzyme



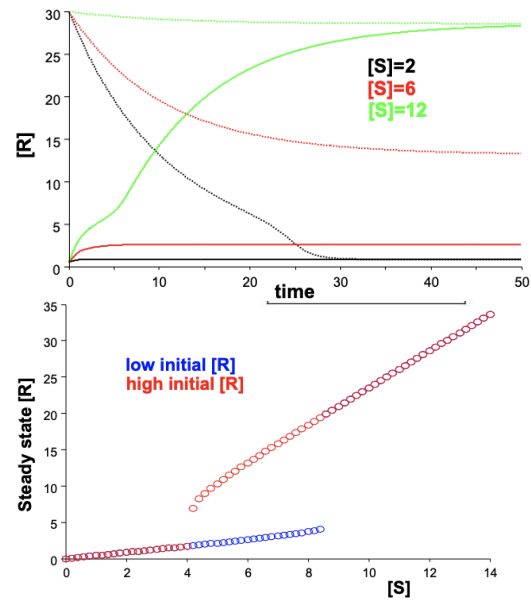
- R causes phosphorylation of E
- E (not EP) leads to degradation of R

$$\frac{d[R]}{dt} = k_0 + k_1[S] - (k_2 + k_2'[E])[R]$$

Time course of [R] at different values of [S]



Initial Conditions determine which steady-state is reached



Nullclines: How to determine that bistability will occur at only some values of [S]?

=> Plot **nullclines**, i.e. points where either  $\frac{d[R]}{dt} = 0$  or  $\frac{d[E]}{dt} = 0$

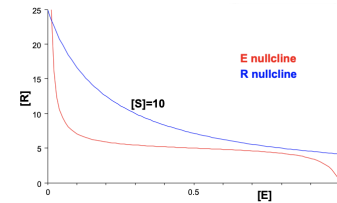
$$\frac{d[R]}{dt} = k_0 + k_1[S] - (k_2 + k_2'[E])[R] = 0$$

$$\frac{d[E]}{dt} = -k_{2E}[R]\frac{[E]}{[E] + K_{m2E}} + k_{1E}\frac{[E]_{TOTAL} - [E]}{[E]_{TOTAL} - [E] + K_{m1E}} = 0$$

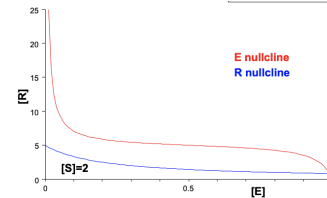
**First equation: equally easy to solve for [E] in terms of [R] or vice-versa**

**Second equation: MUCH easier to solve for [R] as function of [E]**

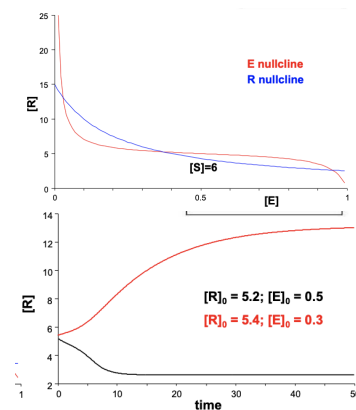
$[S] = 10$   
One intersection: monostable.



$[S] = 2$   
One intersection: monostable.



$[S] = 6$   
Two intersections: this suggests (but does not prove) the middle fixed point is unstable



- In a two-variable system, bistability can be produced by:
  - mutual activation
  - mutual repression
- Bifurcation diagrams summarize which regions of particular
  - parameters are associated with bistability.
- Plotting nullclines in the phase plane is the first step towards
  - predicting whether bistability is present.

-- 2). **Mathematically rigorous: Jacobian & eigenvalues** To understand stable & unstable fixed points mathematically:

- Step 1. we compute the **Jacobian matrix**.

Based on:

$$\frac{d[E]}{dt} = -k_{2e}[R] \frac{[E]}{[E] + K_{m2e}} + k_{1e} \frac{[E]_{TOTAL} - [E]}{[E]_{TOTAL} - [E] + K_{m1e}} = f$$

$$\frac{d[R]}{dt} = k_{0r} + k_{1r}[S] - (k_{2r} + k_{3r}[E])[R] = g$$

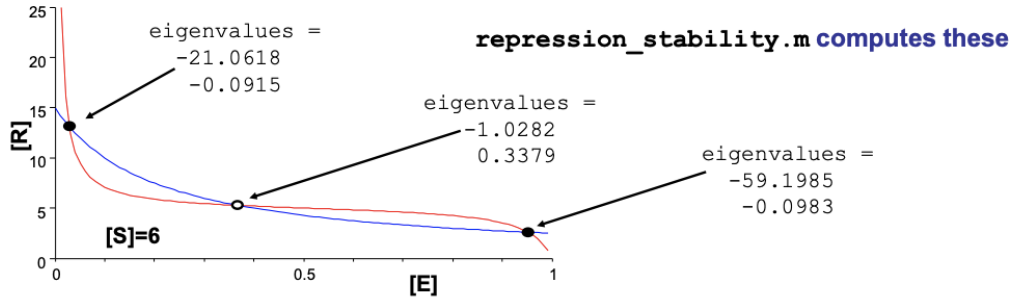
Jacobian matrix:

$$J = \begin{bmatrix} \frac{\partial f}{\partial [E]} & \frac{\partial f}{\partial [R]} \\ \frac{\partial g}{\partial [E]} & \frac{\partial g}{\partial [R]} \end{bmatrix} = \begin{bmatrix} \frac{-k_{2e}[R]K_{m2e}}{([E] + K_{m2e})^2} - \frac{k_{1e}K_{m1e}}{([E]_{TOTAL} - [E] + K_{m1e})^2} & \frac{-k_{2e}[E]}{[E] + K_{m2e}} \\ -k_{3r}[R] & -(k_{2r} + k_{3r}[E]) \end{bmatrix}$$

- Step 2. Evaluate this at the fixed points defined by  $[E^*]$ ,  $[R^*]$

$$J = \begin{bmatrix} \frac{-k_{2e}[R^*]K_{m2e}}{([E^*] + K_{m2e})^2} - \frac{k_{1e}K_{m1e}}{([E]_{TOTAL} - [E^*] + K_{m1e})^2} & \frac{-k_{2e}[E^*]}{[E^*] + K_{m2e}} \\ -k_{3r}[R^*] & -(k_{2r} + k_{3r}[E^*]) \end{bmatrix}$$

- Step 3. The **eigenvalues** of the Jacobian (at the fixed points) determine stability:
  - The real part of either is positive: **the fixed point is unstable**
  - Real parts of both are negative: **the fixed point is stable**



-- 3). Qualitative and graphical: direction arrows In 2D phase plane, direction determined by:

$$\begin{bmatrix} d[E]/dt \\ d[R]/dt \end{bmatrix}$$

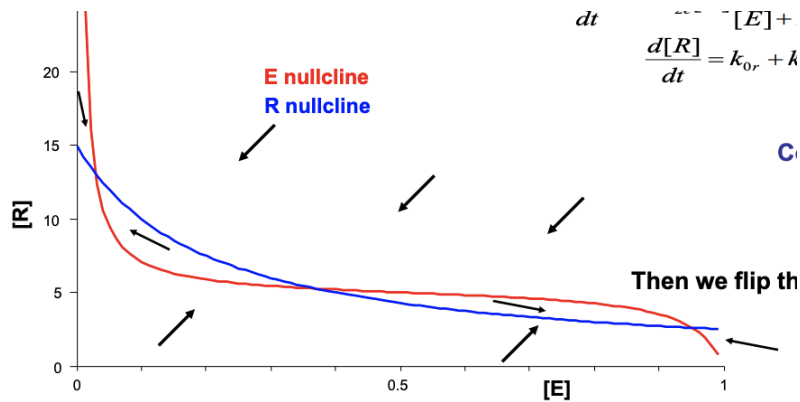
$$\frac{d[E]}{dt} = -k_{2e}[R] \frac{[E]}{[E] + K_{m2e}} + k_{1e} \frac{[E]_{TOTAL} - [E]}{[E]_{TOTAL} - [E] + K_{m1e}}$$

$$\frac{d[R]}{dt} = k_{0r} + k_{1r}[S] - (k_{2r} + k_{3r}[E])[R]$$

Consider  $[E]$  big and  $[R]$  big,

$$\frac{d[E]}{dt} < 0; \frac{d[R]}{dt} < 0$$

Then we flip the arrow each time we cross a nullcline.



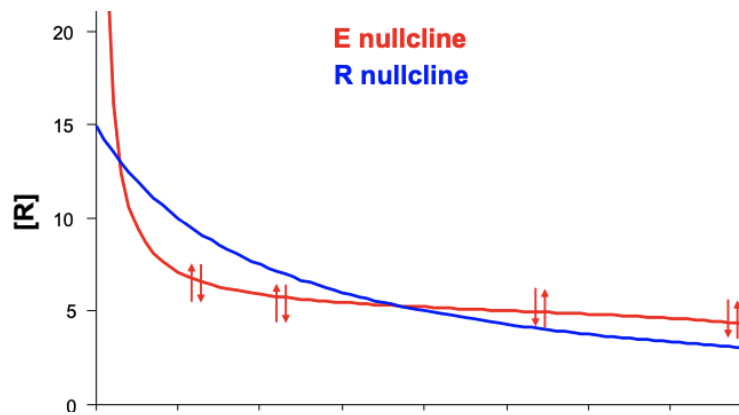
With these simple rules, we can often determine stability.

#### Arrows on the nullclines:

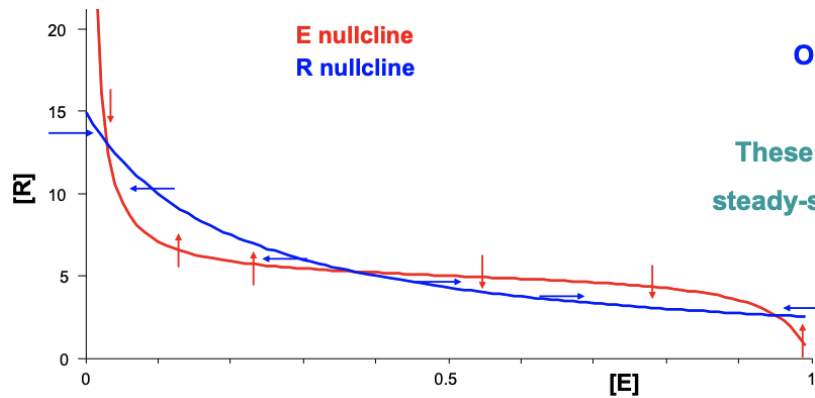
- On the E nullcline,  $dE/dt = 0$ , direction of movement is up/down
- On the R nullcline,  $dR/dt = 0$ , direction of movement is left/right

The direction changes when a nullcline is crossed

$dR/dt = 0$ : on E nullcline



$dE/dt = 0$  on R nullcline



These considerations suggest that

- middle steady-state is unstable.
- left and right steady-states are stable.

### Summary

- In a two-variable system mutual activation or mutual repression can produce bistability.
- When nullclines intersect 3 times, bistability may be present.
- Stability of fixed points can be determined graphically by:
  - plotting direction arrows for extreme values of the two variables
  - flipping arrows in one direction each time a nullcline is crossed

Bistable systems produce digital, all or none, rather than graded responses.

Bistability is biologically useful when persistence is required: apoptosis, cell division, differentiation, etc.

Bistability is produced by complex regulation, eg: mutual activation or inhibition

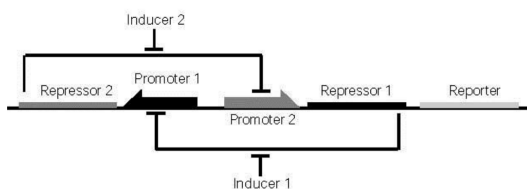
The presence or absence of bistability can be assessed mathematically or graphically (rate balance plots, nullclines in the phase plane).

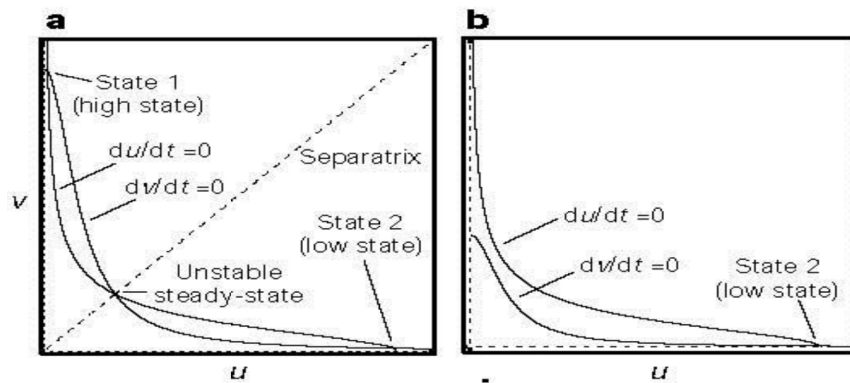
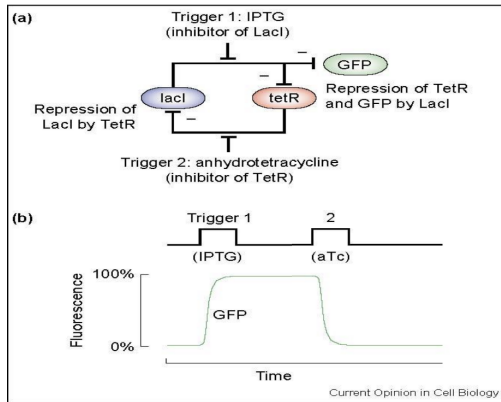
### Q4. Example of bistability:

**Example 1. An artificial genetic “toggle switch”**

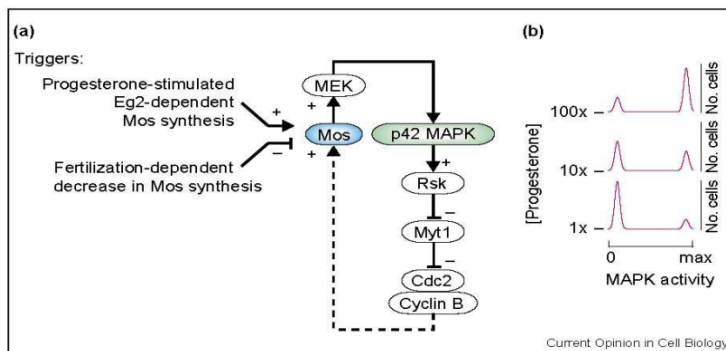
$$\frac{du}{dt} = \frac{\alpha_1}{1+v^\beta} - u$$

$$\frac{dv}{dt} = \frac{\alpha_2}{1+u^\gamma} - v$$

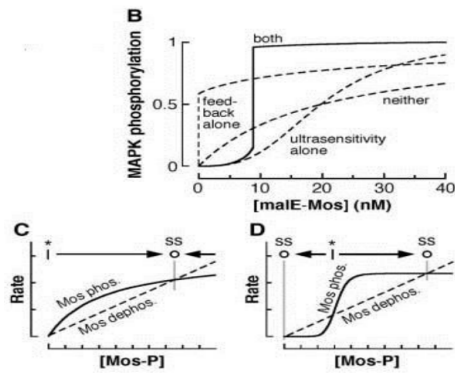




## Example 2. MAPK cascade in oocyte maturation

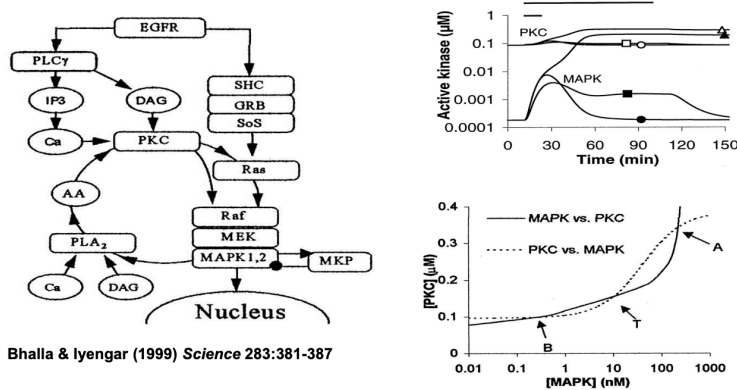


Ferrell (2002) *Curr. Op. Cell Biol.* 14:140–148.



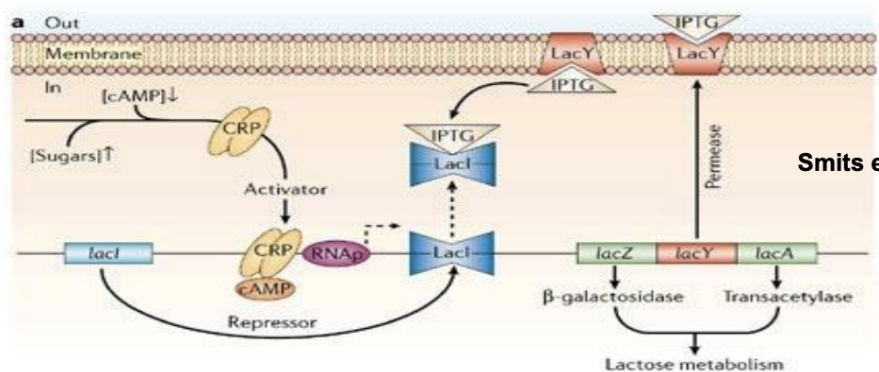
Ferrell & Machleder (1998) *Science* 280:895-898

### Example 3. MAP-kinase pathway in mammalian cells



Bhalla & Iyengar (1999) *Science* 283:381-387

### Example 4. Lac operon in *E. coli*



- With low nutrient levels, LacI will repress transcription of the LacA, LacY, and LacZ genes.
- Lactose, allolactose, or IPTG will bind to LacI, relieve repression.
- LacY encodes a “permease” which allows lactose into the cell.



$$\frac{dl}{dt} = \beta l_{\text{ext}} \text{LacY} - \gamma l$$

$$\frac{d\text{LacY}}{dt} = \delta + p \frac{l^4}{l^4 + l_0^4} - \sigma \text{LacY}$$

$l$  = intracellular lactose

$\text{LacY}$  = expression of LacY/permease

$\beta, \gamma, \delta, \sigma, p, l_0$  = constants

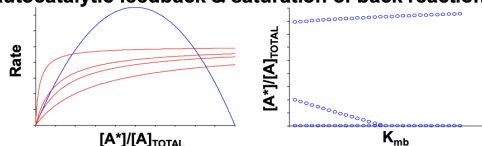
$l_{\text{ext}}$  = external lactose

Example of bistable systems: A minimal model of the lac operon

**(Note: in most models,  $d\text{LacY}/dt$  depends on [lactose] dependence on [lactose]<sup>4</sup> to improve the nullcl**

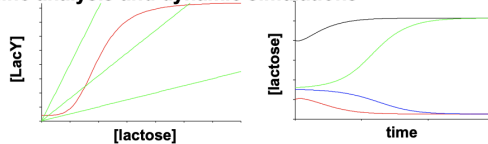
### 1) Rate balance plots

Linear autocatalytic feedback & saturation of back reaction



### 2) Model of *lac* operon

Nullcline analysis and dynamic simulations



- Bistability is observed in biological systems when mutual activation or mutual inhibition is present
  - MAP-kinase signaling
  - The lac operon in *E. coli*
- Mutual activation/inhibition can occur through post-translational modifications (e.g. phosphorylation) or through changes in gene expression.
- Mutual activation/inhibition can be direct or can occur through intermediates.