

# ***six lectures on systems biology***

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lecture 3  
5 april 2011

part 2 seminar room, department of genetics

## a rather provisional syllabus

0. why mathematical models?

→ 1. post-translational modification of proteins

2. microscopic cybernetics

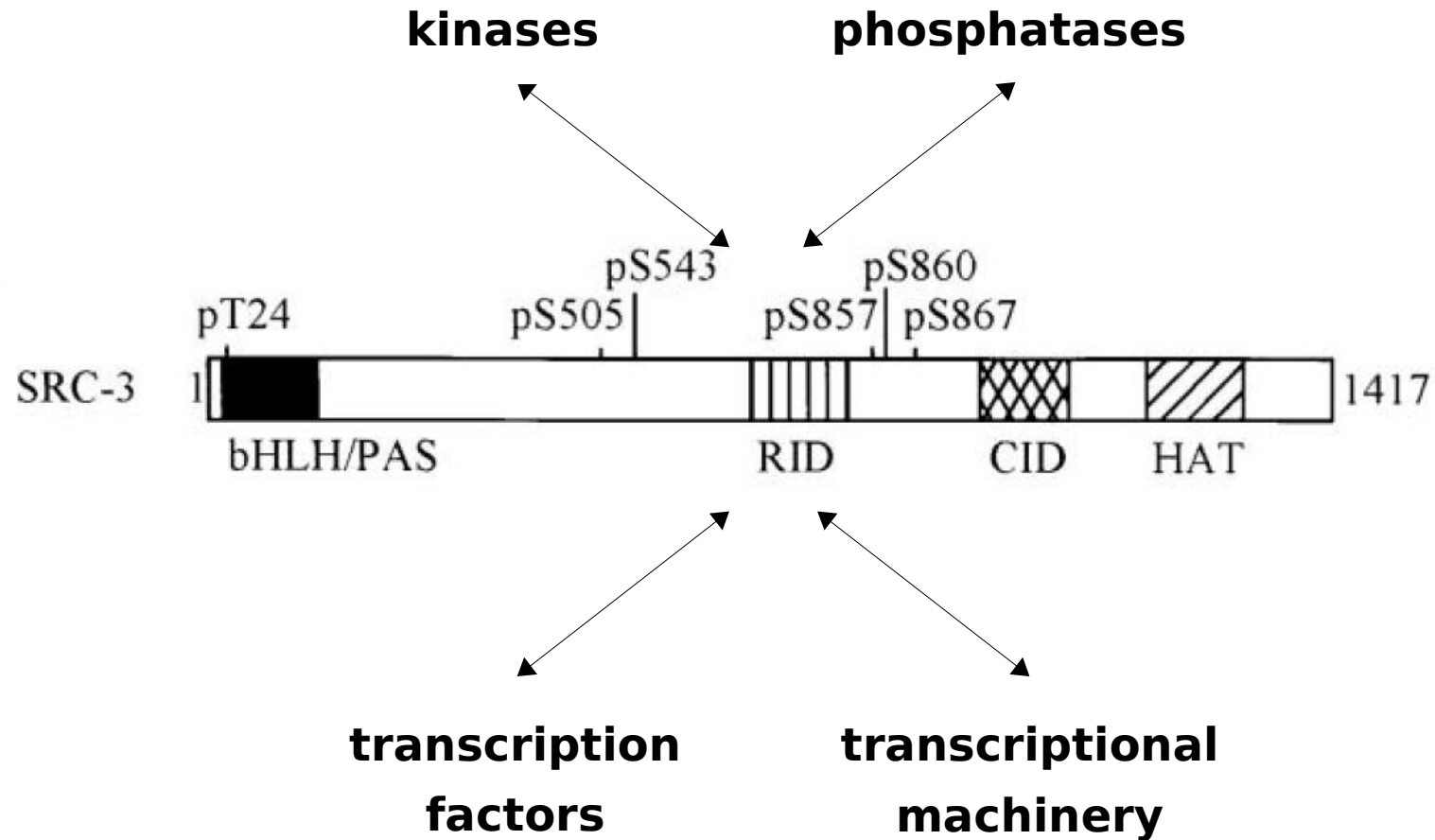
3. development and evolution

# 1. post-translational modification (PTM)

how can we measure mod-form distributions?

 how do PTM networks regulate the distributions?

# information processing by PTMs

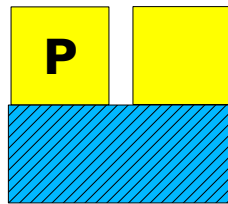


# the cartoon fallacy about PTMs

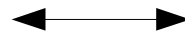
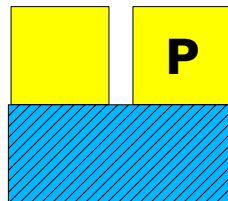
*“a protein with more sites of modification carries more information”*

molecule

downstream process



decision A

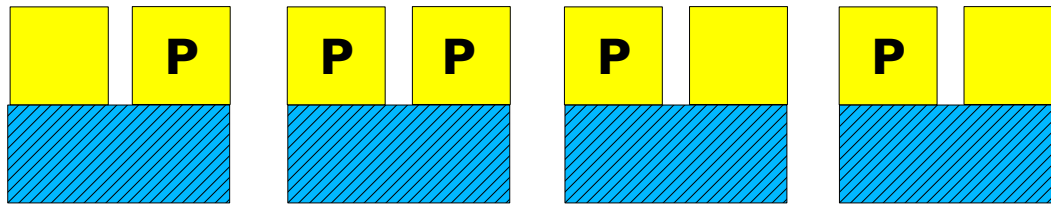


decision B

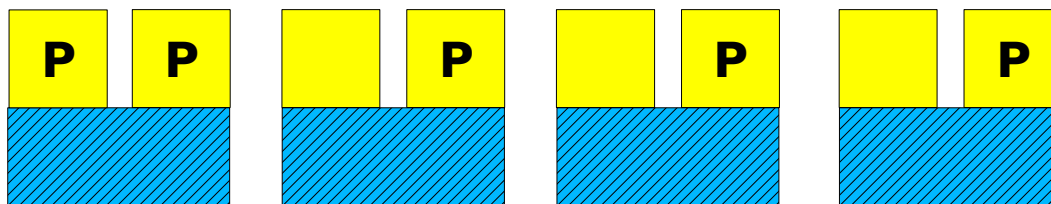
# but under the hood

distribution

downstream process



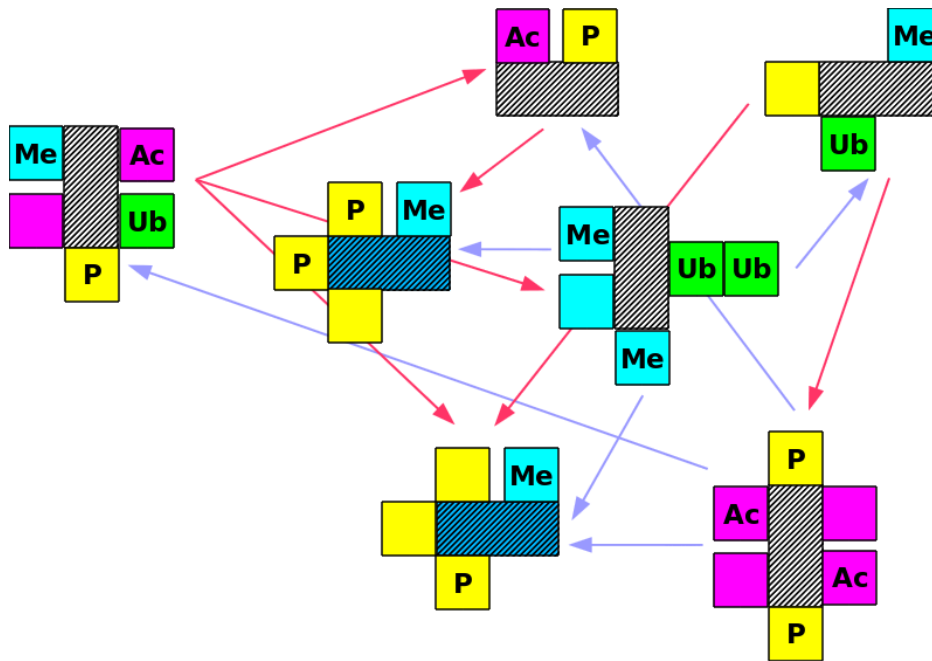
**decision A**



**decision B**

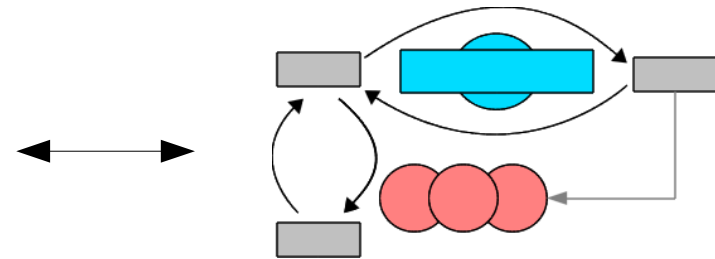
# separation of time scales

network of enzymes and substrates



**fast sub-system**

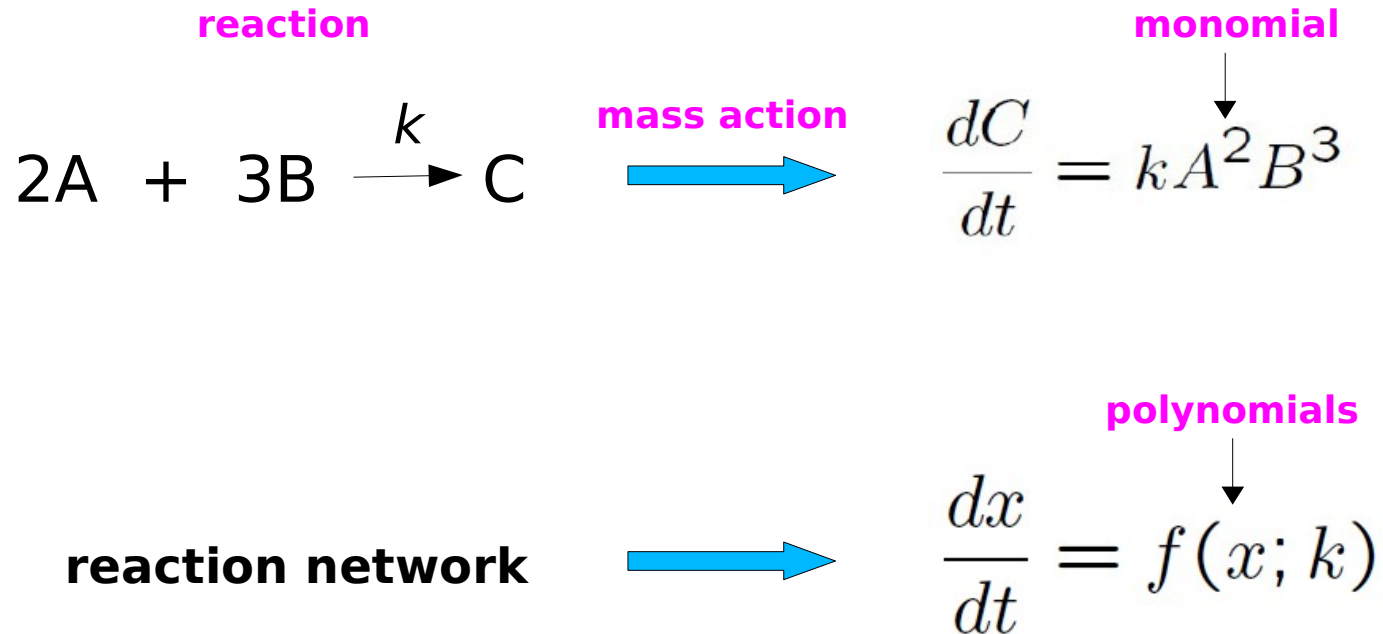
downstream process



**slow sub-system**

the fast sub-system comes to steady state quickly and the slow sub-system “reads” that steady state

# polynomial dynamical systems



$$x = (x_1, \dots, x_n) \quad \text{species concentrations}$$

$$k = (k_1, \dots, k_p) \quad \text{rate constants (parameters)}$$



## seeing the wood for the trees

**simulation requires the numerical values of all parameters to be known beforehand**

*for PTM systems, the numbers of sites of modification need to be known and the combinatorial complexity increases exponentially ...*

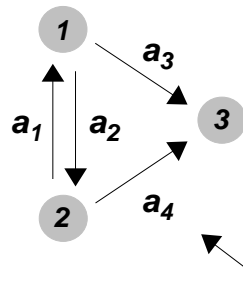
however, at steady state,  $\frac{dx}{dt} = 0$

**solutions of a system of polynomial algebraic equations**  $f(x; k) = 0$

and this can be analysed with the parameters treated as symbols

# linear chemistry

directed graph



mass action



Laplacian matrix

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -(a_2 + a_3) & a_1 & 0 \\ a_2 & -(a_1 + a_4) & 0 \\ a_3 & a_4 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

symbolic label

each edge is a unimolecular reaction, with the label as rate constant

Gunawardena, "A linear elimination framework for nonlinear biochemical systems", submitted, 2011.

Gustav Kirchhoff, "Über die Auflösung der Gleichungen, auf welche man bei der Untersuchung der linearen Verteilung galvanischer Ströme geführt wird", Ann Phys Chem, **72**:497-508 1847

## with nonlinear applications

labelled  
directed graph

$G$



Laplacian dynamics

$$\frac{dx}{dt} = \mathcal{L}(G).x$$

the labels in  $G$  may be complicated algebraic expressions

$$a_1 = \left( \frac{k_1 + k_2}{k_3} \right) (L_1)^2 L_2$$

**provided no concentration -  $L_1$  or  $L_2$  - is that of a node in  $G$**

(although this constraint can be weakened ... )

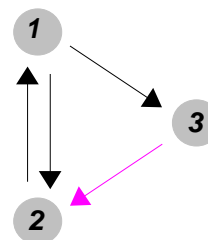
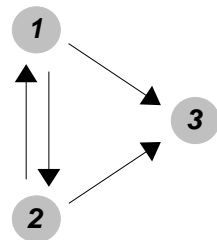
## steady states in the linear framework

$$\mathcal{L}(G).x = 0 \quad x \in \ker \mathcal{L}(G)$$

if  $G$  is **strongly connected** with positive labels, then there is a unique steady state, up to a constant factor

$$\ker \mathcal{L}(G) = \langle \rho \rangle$$

**strongly connected** - *any two distinct nodes are connected by a contiguous path of edges pointing in the same direction*



**strongly connected**

## the matrix-tree theorem

$$\ker \mathcal{L}(G) = \langle \rho \rangle$$

$$\rho_i = \sum_{T \in \Theta_i(G)} \left( \prod_{j \xrightarrow{a} k \in T} a \right) \quad \text{polynomial in the symbolic labels}$$

$\Theta_i(G)$  = set of spanning trees rooted at  $i$

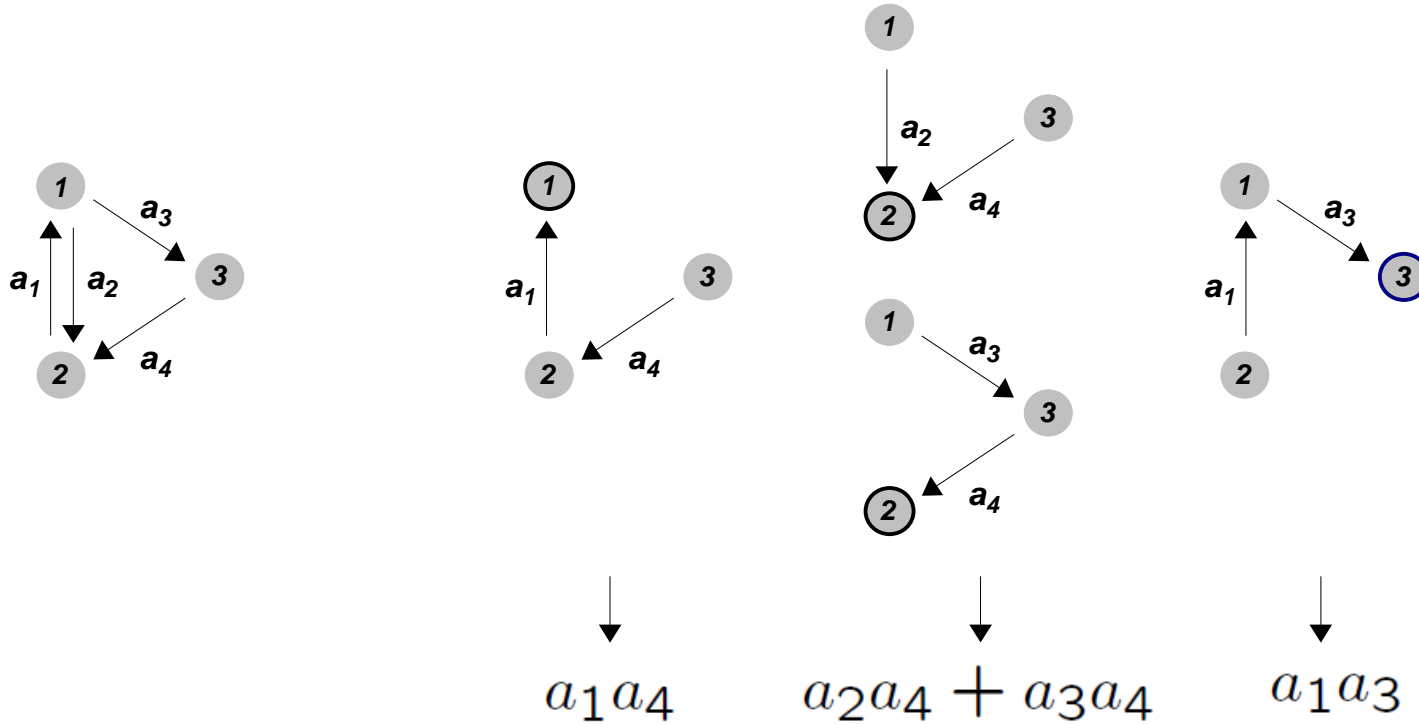
W T Tutte, "The dissection of equilateral triangles into equilateral triangles", Proc Camb Phil Soc **44**:463-82 1948

## spanning trees

**rooted spanning tree** - a sub-graph  $T$  of  $G$  that

1. spans  $G$  - every node of  $G$  is also a node of  $T$
2. is a TREE -  $T$  has no cycles, ignoring edge directions
3. is rooted at  $i$  -  $i$  is the only node of  $T$  with no outgoing edges

# spanning tree formula



$$\begin{pmatrix} -(a_2 + a_3) & a_1 & 0 \\ a_2 & -a_1 & a_4 \\ a_3 & 0 & -a_4 \end{pmatrix} \begin{pmatrix} a_1 a_4 \\ (a_2 + a_3) a_4 \\ a_1 a_3 \end{pmatrix} = 0$$

Laplacian

$\rho$

## elimination of internal complexity

if there is a steady state  $\frac{dx}{dt} = 0$      $\mathcal{L}(G).x = 0$

$$x = \lambda \rho \quad \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \lambda \begin{pmatrix} \rho_1 \\ \vdots \\ \rho_n \end{pmatrix}$$

then each  $x_i$  can be eliminated

$$x_i = \left( \frac{\rho_i}{\rho_1 + \dots + \rho_n} \right) x_{tot} \quad x_i = \left( \frac{\rho_i}{\rho_1} \right) x_1$$

reference node  
↓

rational expression in  
the symbolic labels



## **(non-strongly connected graphs)**

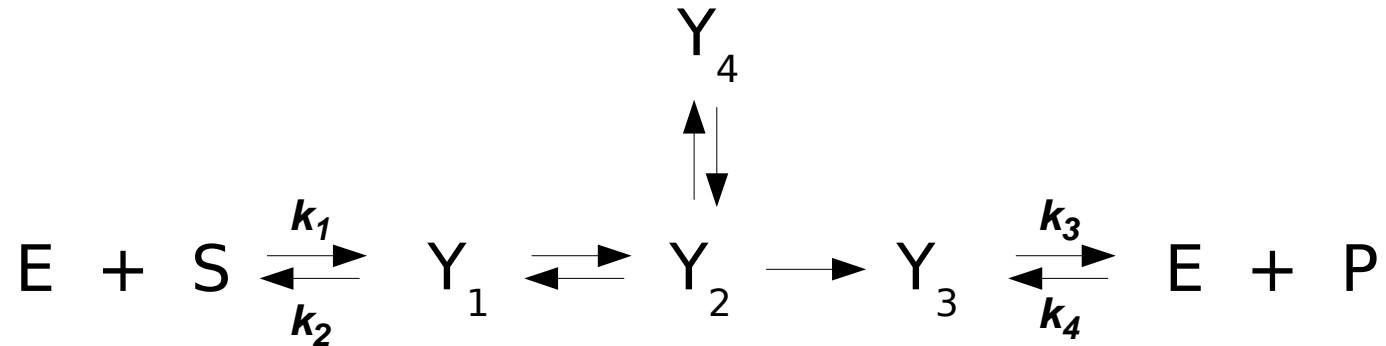
if  $G$  is not strongly connected, there may be multiple independent steady states

they can be completely described but this is not needed here

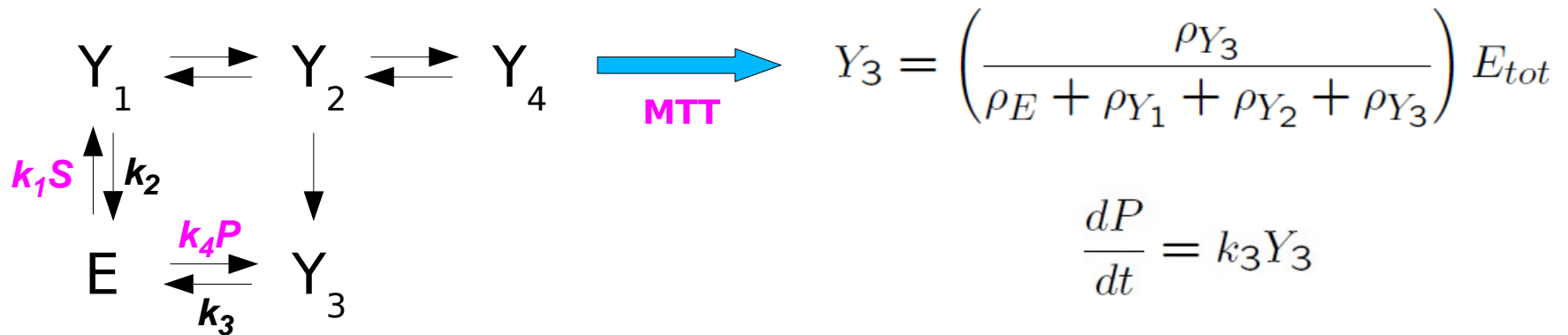
see below for more details

Gunawardena, "*A linear elimination framework for nonlinear biochemical systems*",  
submitted, 2011

## applications - enzyme kinetics



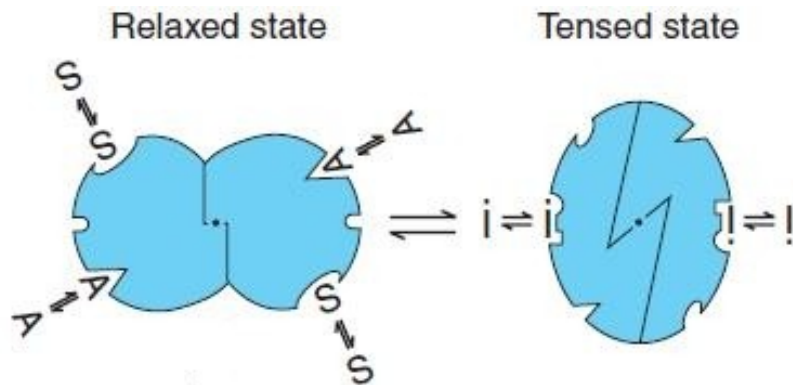
labelled directed graph



King & Altman, "A schematic method of deriving the rate laws for enzyme-catalyzed reactions", J Phys Chem **60**:1375-8 1956

# applications - protein allostery

Monod-Wyman-Changeux formula



fractional saturation

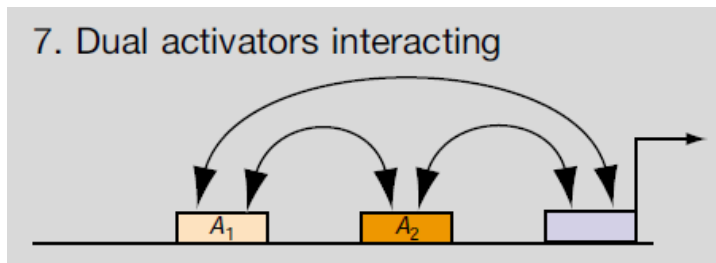
$$Y = \frac{(1 + x)^{n-1}x + cL'(1 + cx)^{n-1}x}{(1 + x)^n + L'(1 + cx)^n}$$

$$\frac{dP}{dt} = V_{max}Y$$

Monod, Wyman, Changeux, "On the nature of allosteric transitions: a plausible model",  
J Mol Biol **12**:88-118 1965

Changeux & Edelstein, "Allosteric mechanisms of signal transduction", Science **308**:1424-8 2005

# applications - transcriptional regulation



rate of gene expression

$$\frac{1 + \frac{[A_1]}{K_{A_1}} f_1 + \frac{[A_2]}{K_{A_2}} f_2 + \frac{[A_1][A_2]}{K_{A_1} K_{A_2}} f_1 f_2 \omega}{1 + \frac{[A_1]}{K_{A_1}} + \frac{[A_2]}{K_{A_2}} + \frac{[A_1][A_2]}{K_{A_1} K_{A_2}} \omega}$$

Bintu, Buchler, Garcia, Gerland, Hwa, Kondev, Phillips, “*Transcriptional regulation by the numbers*” *Curr Opin Gen Dev* **15**:116-24 2005 - see also **15**:125-35 2005

Ackers, Johnson, Shea, “*Quantitative model for gene regulation in lambda phage repressor*”, *PNAS* **79**:1129-33 1982

## applications - post-translational modification

substrate with multiple types of modifications on multiple sites

$$S_1, \dots, S_N \quad \text{modforms}$$

multiple forward and reverse enzymes

$$E_1, \dots, E_L$$

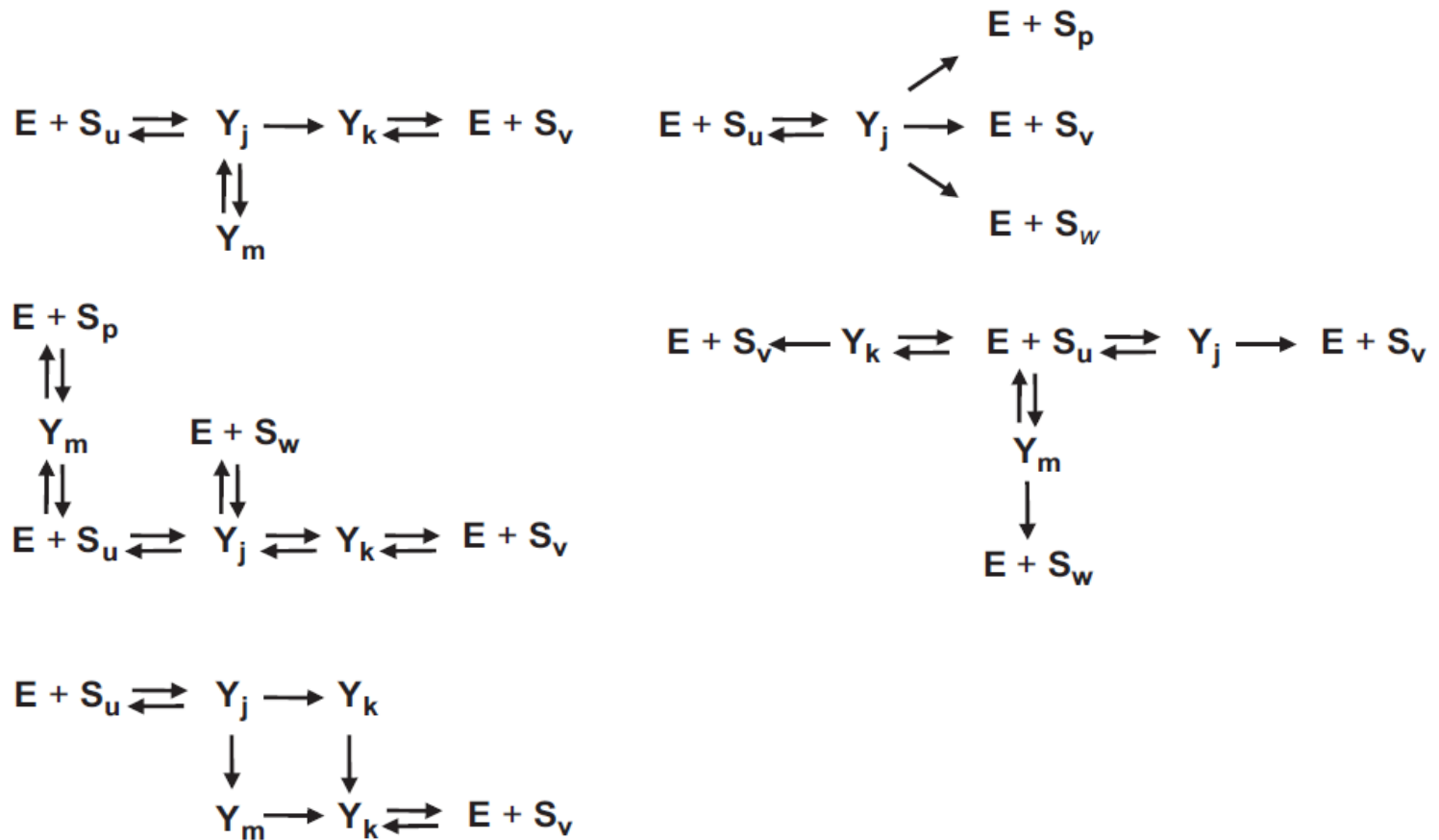
directed “system graph” on the modforms



naturally strongly connected

Thomson, Gunawardena, “The rational parameterisation theorem for multisite post-translational modification systems”, J Theor Biol **261**:626-36 2009

# biochemistry of modification



donor “charge” maintained constant by external mechanisms

no ubiquitin-like modifications

## hierarchical elimination

eliminate the intermediate complexes in favour of substrates and enzymes

*this gives a label for the system graph*

provided no enzyme is also a substrate, the substrates can then be eliminated in favour of the enzymes

$$S_u = r_u(E_1, \dots, E_L) S_0$$

↑  
**rational expression**

$$\begin{array}{rcl} E_{1,tot} & = & \Phi_1(E_1, \dots, E_L) \\ \vdots & & \vdots \\ E_{L,tot} & = & \Phi_L(E_1, \dots, E_L) \end{array}$$

**complete analytical description, with an exponential reduction in the number of independent variables**

## information capacity of PTM systems

*the modform distribution of a single substrate with  $n$  sites has a maximal information storage capacity of at least*

$$\log_2 \left[ \frac{n+2}{2} \right] \text{ bits}$$

Thomson, Gunawardena *“Unlimited multistability in multisite phosphorylation systems”*,  
Nature **460**:274-7 2009



## summing up

1. *separation of time scales allows internal complexity to be eliminated*
2. *this can be done using a linear, graph-theoretic elimination framework*
3. *which underlies the analysis of enzyme kinetics, gene regulation, protein allostery and post-translational modification*
4. *PTM systems can be completely analysed at steady state, symbolically in their parameters, with an exponential reduction in algebraic complexity*
5. *information can be carried in PTM distributions and the information capacity increases logarithmically with the number of sites*