The Continuous π-Calculus: A Process Algebra for Biochemical Modelling

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Overview

The *continuous π-calculus* ($c\pi$) is a process algebra for modelling behaviour and variation in molecular systems.

It has a structured operational semantics that captures system behaviour as trajectories through a continuous process space, by generating familiar differential-equation models.

We have existing biochemical systems expressed in $c\pi$; the aim is to use this to investigate evolutionary properties of biochemical pathways.

Marek Kwiatkowski and Ian Stark.

The Continuous π-Calculus: A Process Algebra for Biochemical Modelling. In *Computational Methods in Systems Biology: Proc. CMSB 2008*

Lecture Notes in Computer Science 5307, pages 103–122. Springer 2008

Overview

Contents

- Systems Biology and Process Algebras
- The Continuous π-Calculus
- Example: Circadian Rhythms in *Synechococcus Elongatus*

Systems Biology

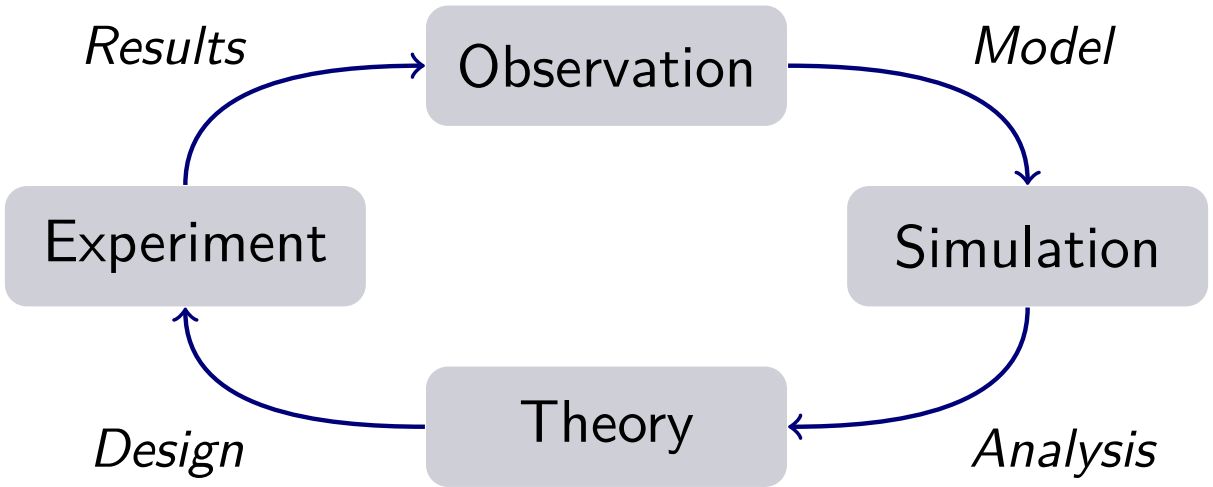
Biology is the study of living organisms; Systems Biology is the study of the dynamic processes that take place within those organisms.

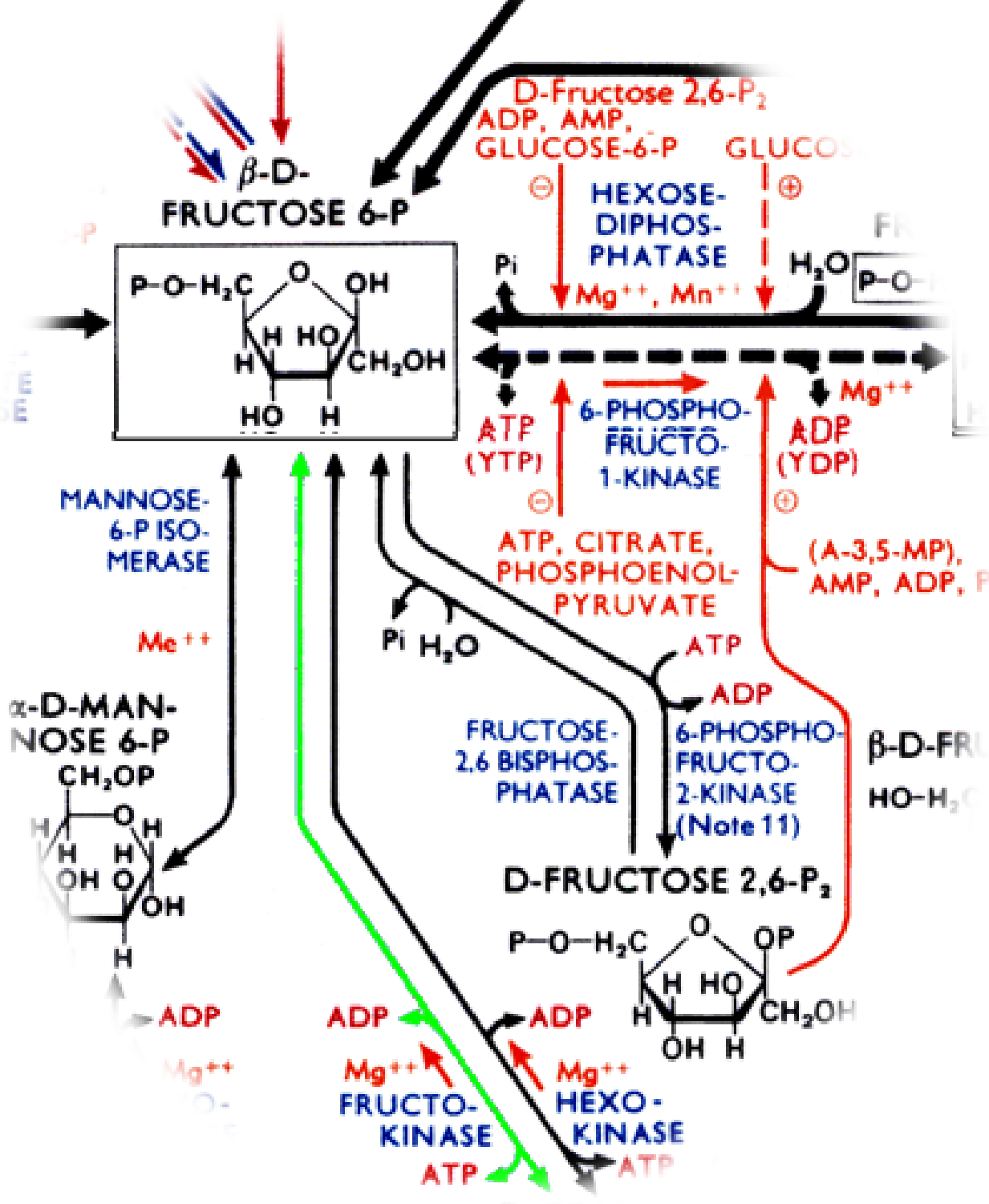
In particular:

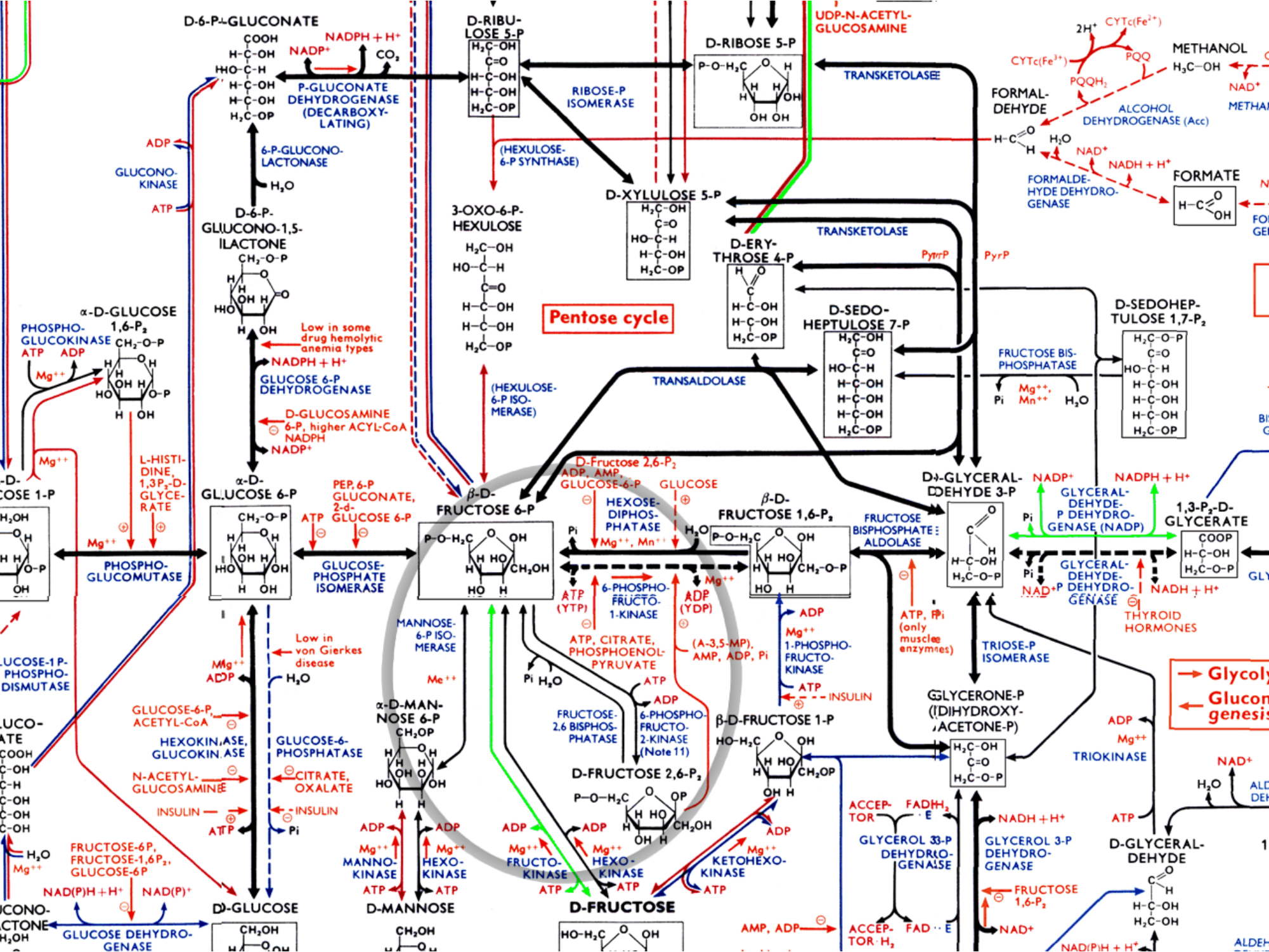
- Interaction between processes;
- Behaviour emerging from such interaction; and
- Integration of component behaviours.

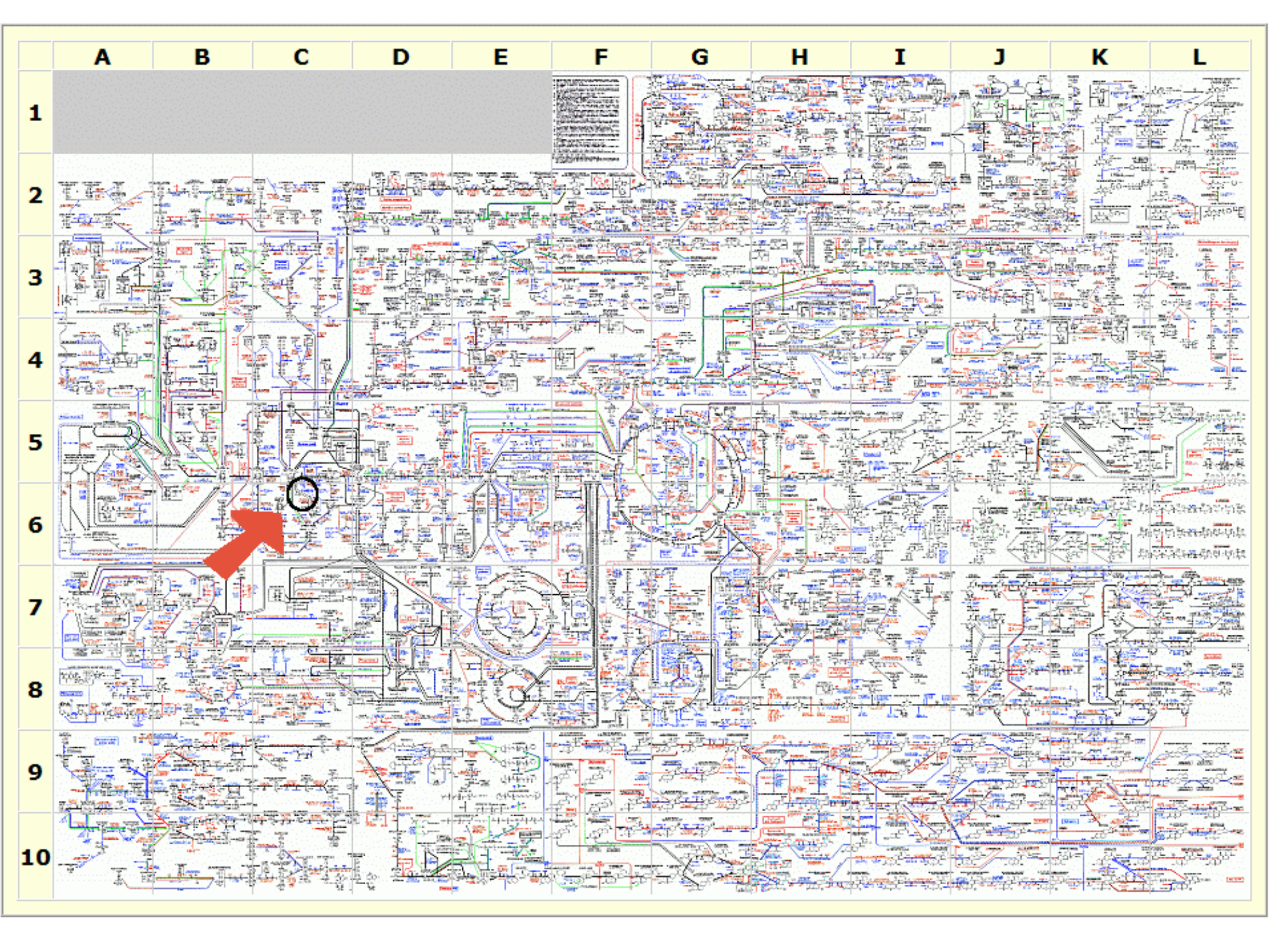
Systems Biology

Biology is the study of living organisms; Systems Biology is the study of the dynamic processes that take place within those organisms.




β-D-
FRUCTOSE 6-P
D-Fructose 2,6-P2
ADP, AMP,
GLUCOSE-6-P
GLUCOSE
HEXOSE-
DIPHOS-
PHATASE
Mg++, Mn++
Pi
H2O
ATP
(YTP)
6-PHOSPHO-
FRUCTO-
1-KINASE
ADP
(YDP)
Mg++
ATP, CITRATE,
PHOSPHOENOL-
PYRUVATE
(A-3,5-MP),
AMP, ADP,
MANNOSE-
6-P ISO-
MERASE
Me++
α-D-MAN-
NOSE 6-P
Pi H2O
ATP
ADP
FRUCTOSE-
2,6 BISPHOS-
PHATASE
6-PHOSPHO-
FRUCTO-
2-KINASE
(Note 11)
β-D-FR
D-FRUCTOSE 2,6-P2
ADP
ADP
ADP
Mg++
Mg++
Mg++
FRUCTO-
KINASE
HEXO-
KINASE
ATP
ATP


Pentose cycle
D-6-P-GLUCONATE
NADP+
NADPH + H+
CO2
P-GLUCONATE DEHYDROGENASE (DECARBOXY-LATING)
D-RIBULOSE 5-P
D-RIBOSE 5-P
RIBOSE-P ISOMERASE
TRANSKETOLASE
UDP-N-ACETYL-GLUCOSAMINE
METHANOL
FORMALDEHYDE
ALCOHOL DEHYDROGENASE (Acc)
FORMALDEHYDE DEHYDROGENASE
FORMATE
6-P-GLUCONO-LACTONASE
GLUCONO-KINASE
ADP
ATP
H2O
D-6-P-GLUCONO-1,5-LACTONE
(HEXULOSE-6-P SYNTHASE)
3-OXO-6-P-HEXULOSE
D-XYLULOSE 5-P
TRANSKETOLASE
D-ERYTHROSE 4-P
PyrP
D-SEDOHEPTULOSE 7-P
D-SEDOHEPTULOSE 1,7-P2
FRUCTOSE BISPHOSPHATASE
Mg++, Mn++
Pi
TRANSALDOLASE
(HEXULOSE-6-P ISOMERASE)
α-D-GLUCOSE 1,6-P2
PHOSPHOGLUCOKINASE
Low in some drug hemolytic anemia types
NADPH + H+
GLUCOSE 6-P DEHYDROGENASE
D-GLUCOSAMINE 6-P, higher ACYL-CoA
NADP+
L-HISTIDINE, 1,3P2-D-GLYCERATE
α-D-GLUCOSE 6-P
β-D-FRUCTOSE 6-P
PEP, 6-P GLUCONATE, 2-d-GLUCOSE 6-P
D-Fructose 2,6-P2
ADP, AMP, GLUCOSE-6-P
GLUCOSE
HEXOSE-DIPHOSPHATASE
β-D-FRUCTOSE 1,6-P2
FRUCTOSE BISPHOSPHATE ALDOLASE
D-GLYCERALDEHYDE 3-P
GLYCERALDEHYDE-P DEHYDROGENASE (NADP)
1,3-P2-D-GLYCERATE
PHOSPHOGLUCOMUTASE
GLUCOSE-PHOSPHATE ISOMERASE
6-PHOSPHOFRUCTO-1-KINASE
ATP (YTP)
ADP (YDP)
ATP, CITRATE, PHOSPHOENOLPYRUVATE
(A-3,5-MP), AMP, ADP, Pi
1-PHOSPHOFRUCTOKINASE
INSULIN
GLYCERALDEHYDE-P DEHYDROGENASE
THYROID HORMONES
ATP, Pi (only muscle enzymes)
TRIOSE-P ISOMERASE
GLYCERONE-P (DIHYDROXYACETONE-P)
Low in von Gierkes disease
MANNOSE-6-P ISOMERASE
α-D-MANNOSE 6-P
FRUCTOSE-2,6 BISPHOSPHATASE
6-PHOSPHOFRUCTO-2-KINASE (Note 11)
D-FRUCTOSE 2,6-P2
β-D-FRUCTOSE 1-P
TRIOKINASE
GLUCOSE-6-P, ACETYL-CoA
HEXOKINASE, GLUCOKINASE
GLUCOSE-6-PHOSPHATASE
N-ACETYL-GLUCOSAMINE
CITRATE, OXALATE
INSULIN
FRUCTOSE-6P, FRUCTOSE-1,6P2, GLUCOSE-6P
NAD(P)H + H+
NAD(P)+
D-GLUCOSE
D-MANNOSE
D-FRUCTOSE
MANNOKINASE
HEXOKINASE
FRUCTOKINASE
KETOHEXOKINASE
ACCEPTOR
FADH2
GLYCEROL 3-P DEHYDROGENASE
FRUCTOSE 1,6-P2
AMP, ADP
D-GLYCERALDEHYDE
GLUCOSE DEHYDROGENASE
Glycoly
Glucon genesis


A
B
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D
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1
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10

What can Computer Science do for Systems Biology?

Machines

Large Databases: Semistructured data; data integration; data mining

Large Simulations: Experiments *in silico*; parameter scans; folding search

Ideas

Language: Abstraction; modularity; semantics; formal models

Reasoning: Logics; behavioural description; model checking

Scope of Study

Processes

- Metabolic networks
- Regulatory systems: promotion, inhibition
- Signalling pathways
- Gene expression: translation, transcription

Models

- Discrete time, continuous time
- Discrete space, continuous space
- Deterministic, nondeterministic, probabilistic
- Qualitative, quantitative

Biochemical Simulation

Biologists routinely use one of two alternative approaches to computational modelling of biochemical systems:

- Stochastic simulation
  - Discrete behaviour: tracking individual molecules
  - Randomized: Gillespie's algorithm
- Ordinary Differential Equations
  - Continuous behaviour: chemical concentrations
  - Deterministic: Numerical ODE solutions

The classical approach is to use the mathematics directly as the target formal system. However, experience in Computer Science suggests the value of an intermediate *language* to describe a system. An expression in this language can then be analysed as it stands, or further mapped into (one or more) mathematical representations.

Process Algebras in Systems Biology

- Petri nets
- π-calculus; stochastic π; BioSPI; SPiM
- Beta binders; BlenX
- Ambients, bioAmbients
- Brane calculi; Bitonal systems
- PEPA, bioPEPA
- Kappa
- PRISM
- Pathway Logic
- ...

The Continuous π-Calculus

The *Continuous π-Calculus* ($c\pi$) is a process algebra for modelling behaviour and variation in molecular systems.

Based on the π-calculus, it introduces continuous variability in:

- rates of reaction;
- affinity between interacting names; and
- quantities of processes.

while retaining classic process-algebra features of:

- compositional semantics (modular, not monolithic);
- abstraction (separating language and semantics);
- specifying interaction (taking behaviour as it emerges).

Motivated by Fontana's work on evolutionary change, neutral spaces and the "topology of the possible".

Basics of $c\pi$

Continuous π has two levels of system description:

- Species
  - Individual molecules (proteins)
  - Transition system semantics
- Processes
  - Bulk population (concentration)
  - Differential equations

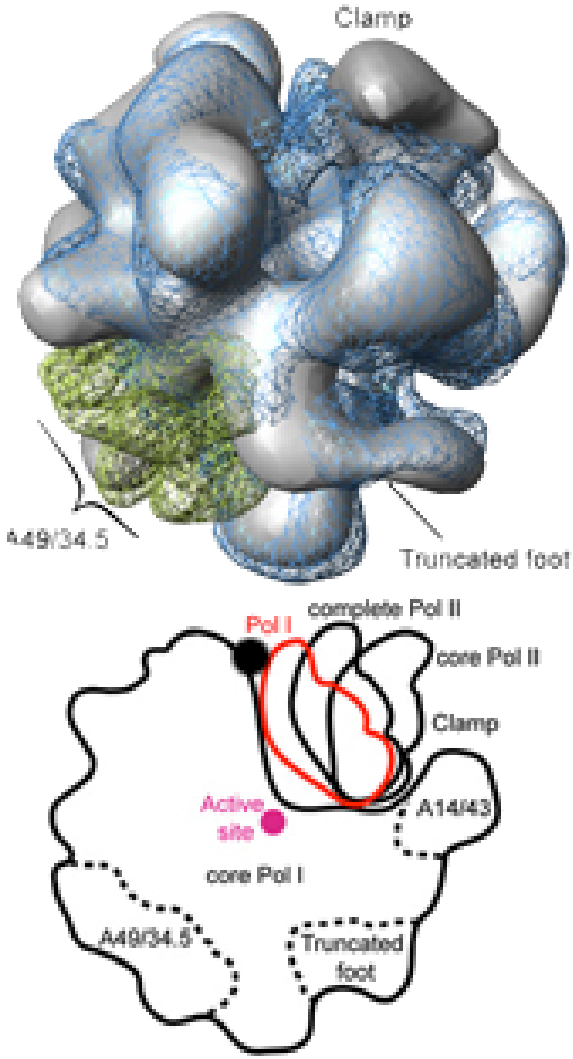
Process space arises as a real-valued vector space over species, with each point the state of a system and behaviours as trajectories through that.

Names in $c\pi$

As in standard π-calculus, *names* indicate a potential for interaction: for example, the docking sites on an enzyme where other molecules may attach.

These sites may interact with many different other sites, to different degrees.

This variation is captured by an *affinity network*: a graph setting out the interaction potential between different names.



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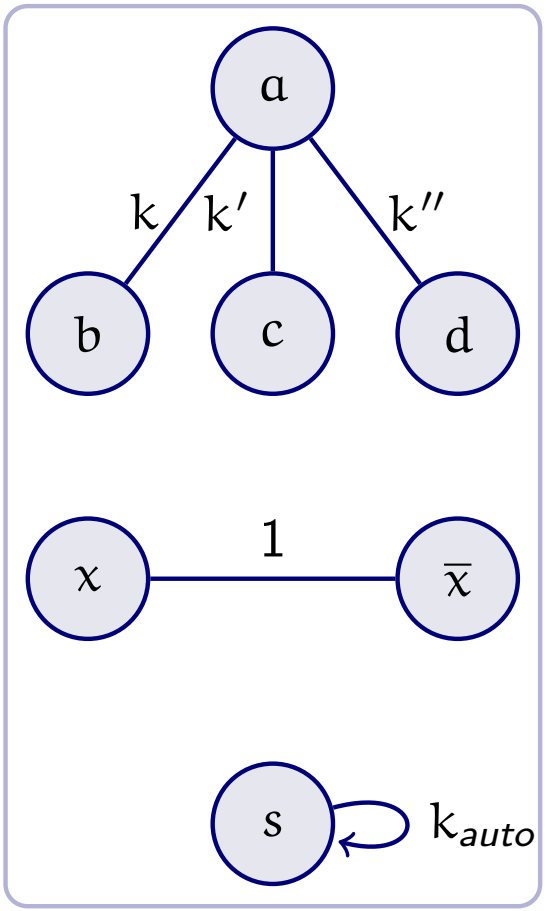

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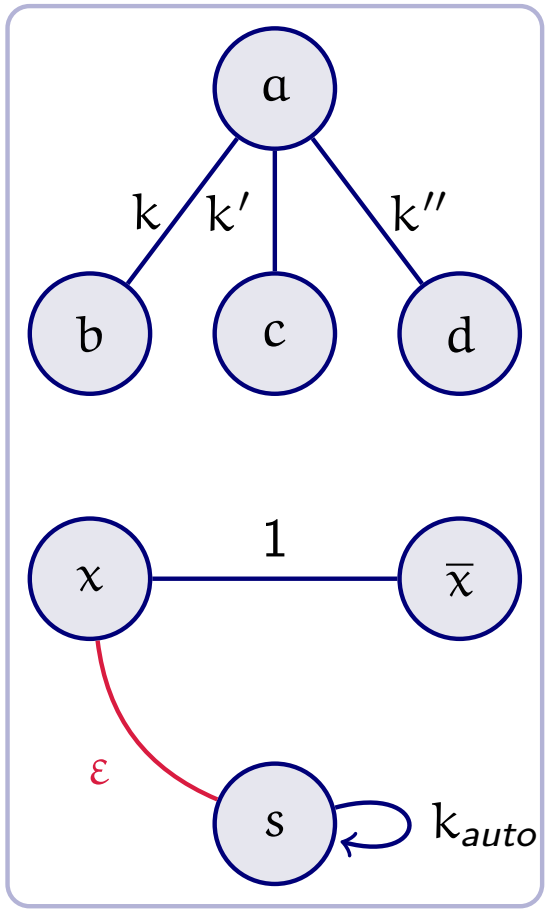


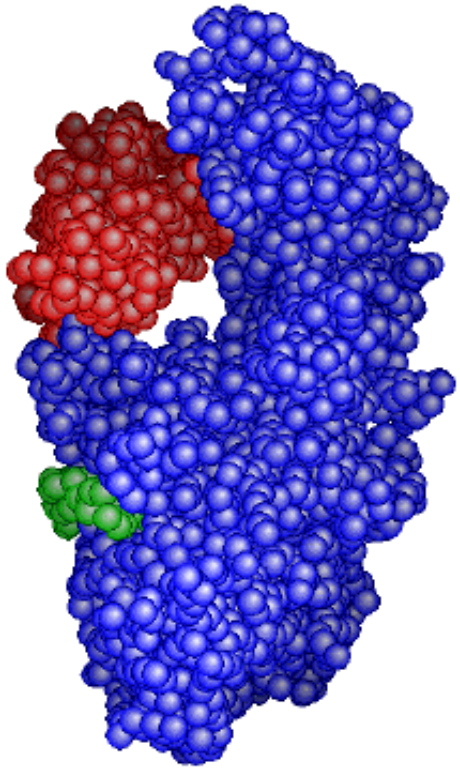
Restriction in $c\pi$

Name restriction $\nu x(A \mid B)$ captures molecular *complexes*, with local name x mediating further internal modification, or decomplexation.

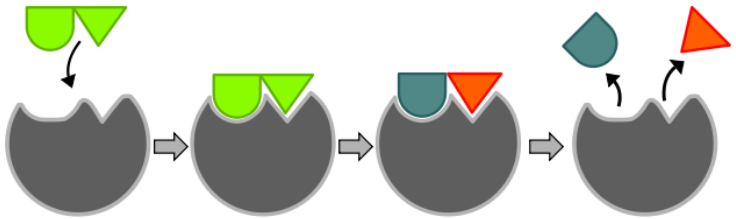
The binder can be a single local name $(\nu x.-)$, or several names with their own affinity network $(\nu M.-)$.

As in the classic π-calculus "cocktail party" model, interacting names can communicate further names, allowing further interactions.

In particular, we use name *extrusion* to model complex formation.



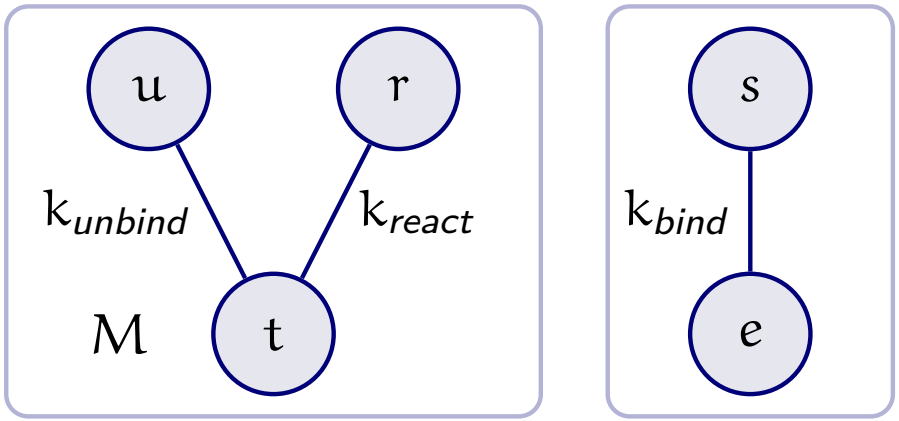
Example Species: Enzyme Catalysis

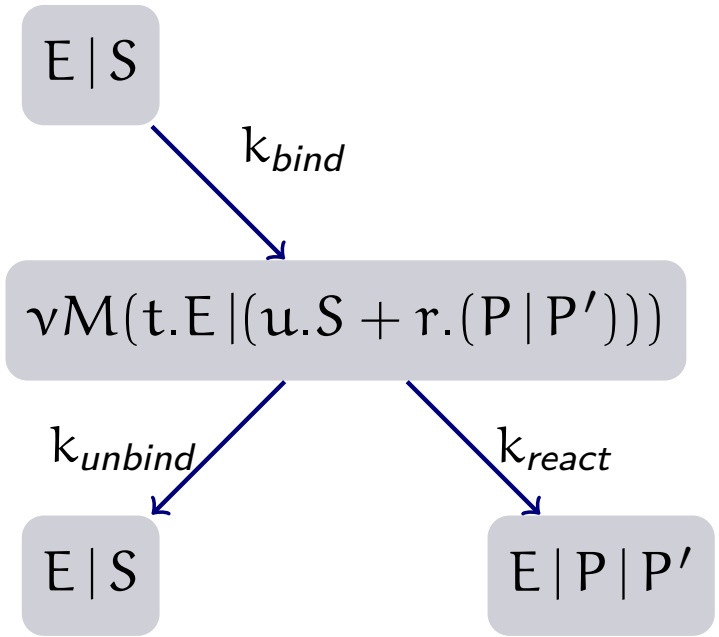


$$S = s(x, y).(x.S + y.(P|P'))$$

$$E = \nu(u, r, t{:}M).(e\langle u, r\rangle.t.E)$$

$$P = P' = \tau@k_{degrade}.0$$





Species

Species $\quad A, B ::= 0 \mid S(\vec{a}) \mid \Sigma a(\vec{b}; \vec{y}).A \mid \tau@k.A \mid A \mid B \mid \nu M.A$

Symmetric prefix	$a(b, c; x, y).A$ for two-way communication.
Guarded sums	$\Sigma_i \alpha_i.A$ or $\alpha.A + \alpha'.A'$ for alternative choices.
Silent transition	$\tau@k.A$ for constitutive reactions at rate $k \in \mathbb{R}_{\geqslant 0}$.
Parallel composition	$A \mid B$ within complexes.
Restriction	$\nu M.A$ for intra-complex reaction.
Recursion	via guarded species definitions $S(\vec{x}) = \ldots$

Set $\mathcal{S}$ of species up to structural congruence, and $\mathcal{S}^{\#}$ of *prime* species.

Operational Semantics for Species

The behaviour of a species is given by transitions:

$$A \xrightarrow{a} (\vec{b};\vec{y})B \qquad \text{Potential interaction}$$

$$A \xrightarrow{\tau@k} B \qquad \text{Immediate action (fixed rate)}$$

$$A \xrightarrow{\tau\langle x,y\rangle} B \qquad \text{Internal action (rate tbd)}$$

Here $(\vec{b};\vec{y})B$ is a *concretion* representing potential interaction; the result of actual interaction is given by pseudo application:

$$(\vec{a};\vec{x})A \circ (\vec{b};\vec{y})B \;=\; A\{\vec{b}/\vec{x}\} \mid B\{\vec{a}/\vec{y}\}$$

Rules for deriving transitions give a structural operational semantics:

$$\frac{A \xrightarrow{a} F \quad B \xrightarrow{b} G}{A \mid B \xrightarrow{\tau\langle a,b\rangle} F \circ G} \qquad \frac{A \xrightarrow{\tau\langle a,b\rangle} B \quad a,b \in M}{\nu M.A \xrightarrow{\tau@M(a,b)} B} \qquad \ldots$$

Processes

Processes	$P, Q ::= 0 \mid c \cdot A \mid P \parallel Q$
Component	species $c \cdot A$ at concentration $c \in \mathbb{R}_{\geqslant 0}$.
Mixture	of processes $P \parallel Q$.

We can identify processes, up to structural congruence, with elements of *process space* $\mathcal{P} = \mathbb{R}^{\mathcal{S}^\#}$.

Species embed in process space $\langle - \rangle : \mathcal{S} \to \mathcal{P}$ at unit concentration.

Operational Semantics for Processes

The behaviour of a process over time is a trajectory through process space.

Immediate behaviour	$\frac{dP}{dt} \in \mathbb{R}^{\mathcal{S}^\#}$	vector in process space
Interaction potential	$\partial P \in \mathbb{R}^{\mathcal{S}\times\mathcal{N}\times\mathcal{C}} = \mathcal{D}$	interaction space

Space $\mathcal{D}$ has basis $\langle A \xrightarrow{a} F \rangle$ for species A, name a, concretion F.

Interaction tensor $\quad \leftthreetimes : \mathcal{D} \times \mathcal{D} \to \mathcal{P}$

Bilinear function generated by

$$\langle A \xrightarrow{a} F \rangle \leftthreetimes \langle B \xrightarrow{b} G \rangle = Aff(a, b)(\langle F \circ G \rangle - \langle A \rangle - \langle B \rangle)$$

Process Semantics

$\frac{dP}{dt}$: Immediate behaviour

- Vector field $\frac{d}{dt}$ over process space $\mathcal{P}$
- Equivalent to an ODE system

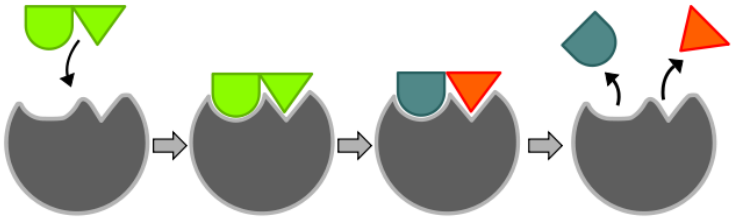
∂P: Interaction potential

- Element of $\mathbb{R}^{\mathcal{S}\times\mathcal{N}\times\mathcal{C}}$
- Equivalent to transition system

$$\partial(P \parallel Q) = \partial P + \partial Q$$

$$\frac{d(P \parallel Q)}{dt} = \frac{dP}{dt} + \frac{dQ}{dt} + \partial P \oplus \partial Q$$

Example Process: Enzyme Catalysis



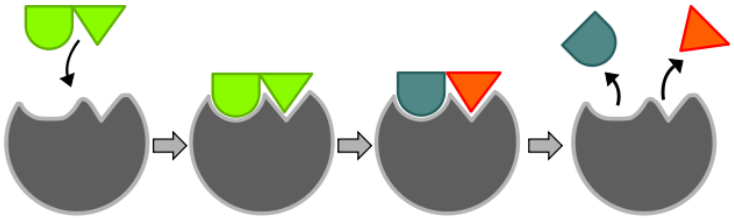
$$S = s(x, y).(x.S + y.(P|P'))$$

$$E = \nu(u, r, t{:}M).(e\langle u, r\rangle.t.E)$$

$$P = P' = \tau@k_{degrade}.0$$

$$c_S \cdot S \,\|\, c_E \cdot E$$

Example Process: Enzyme Catalysis



$$S = s(x, y).(x.S + y.(P|P'))$$

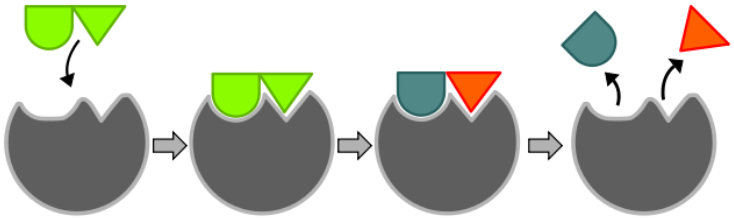
$$E = \nu(u, r, t{:}M).(e\langle u, r\rangle.t.E)$$

$$P = P' = \tau@k_{degrade}.0$$

$$c_S \cdot S \,\|\, c_E \cdot E$$

```
enzyme.cpi
...
species E() = {
  site t, u, r;
...
```

Example Process: Enzyme Catalysis



$$S = s(x, y).(x.S + y.(P|P'))$$

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$$c_S \cdot S \,\|\, c_E \cdot E$$

Cpi tool

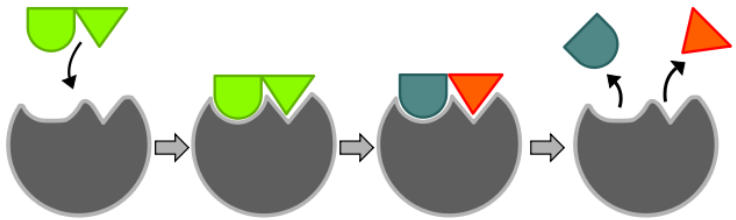
```
enzyme.cpi
...
species E() = {
  site t, u, r;
...
```

ODEs

$$x_2' = -k_1 x_4 x_2 + \ldots$$

$$\vdots$$

Example Process: Enzyme Catalysis



$$S = s(x, y).(x.S + y.(P|P'))$$

$$E = \nu(u, r, t{:}M).(e\langle u, r\rangle.t.E)$$

$$P = P' = \tau@k_{degrade}.0$$

$$c_S \cdot S \,\|\, c_E \cdot E$$

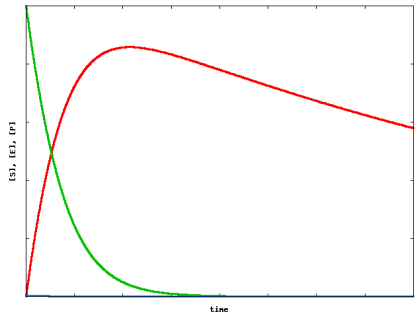
Cpi tool → *Octave*

```
enzyme.cpi
...
species E() = {
  site t, u, r;
...
```

ODEs

$$x_2' = -k_1 x_4 x_2 + \ldots$$

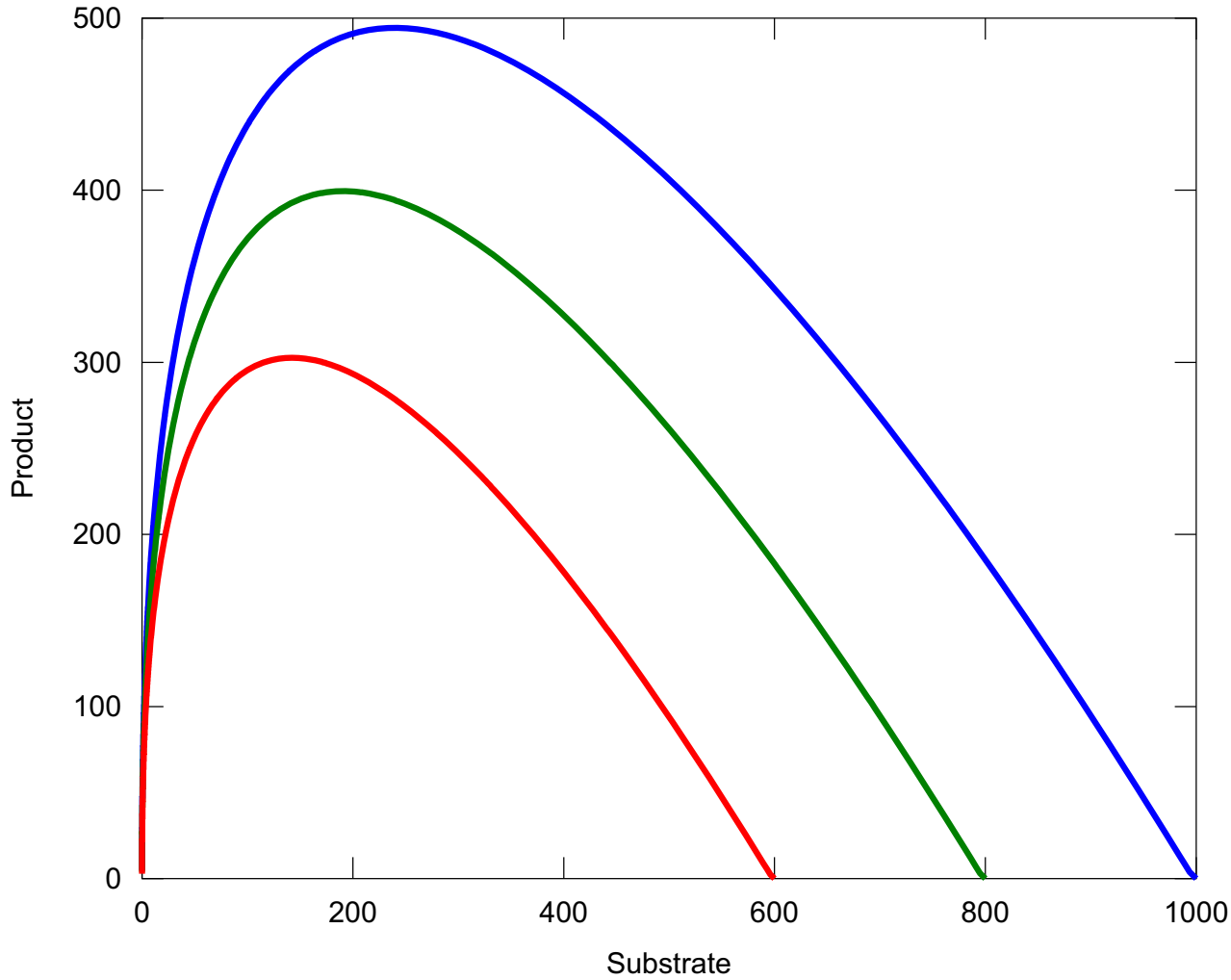
$$\vdots$$



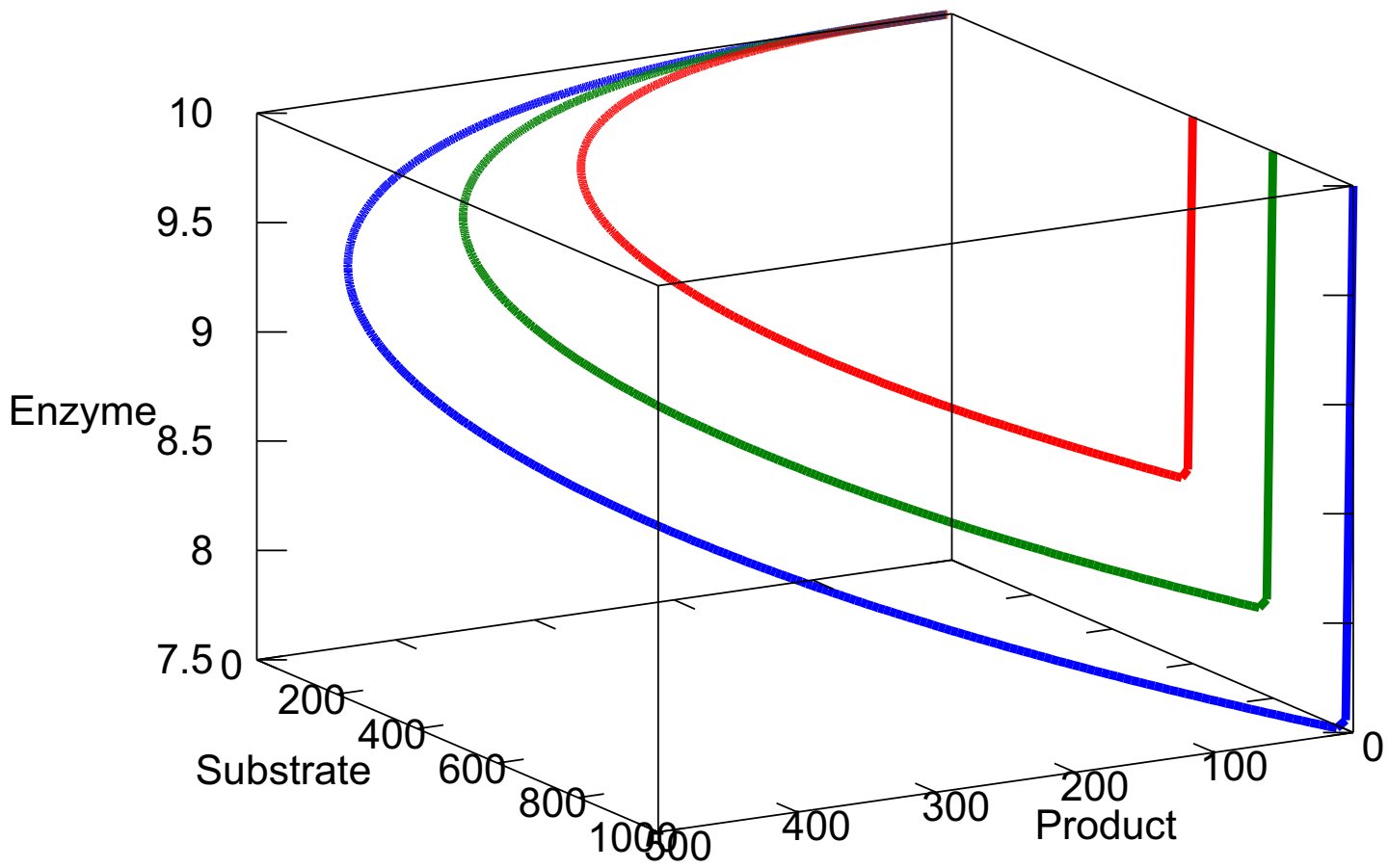
Cpi tool input syntax

```
const kbind=1e-3;    const kreact=2.0;
const kunbind=1.0;   const kdegrade=3e-4;

site e,s; react (e,s)@kbind;

species S() = { body s(;x,y).(x(;).S() + y(;).P());   init 1000.0; }

species E() = { site u,r,t;
                react (u,t)@kunbind;
                react (r,t)@kreact;
                body e(u,r;).act(;).E();
                init 10.0; }

species P() = { body tau<kdegrade>.0; init 0.0; }
```

Process Space: Substrate & Product



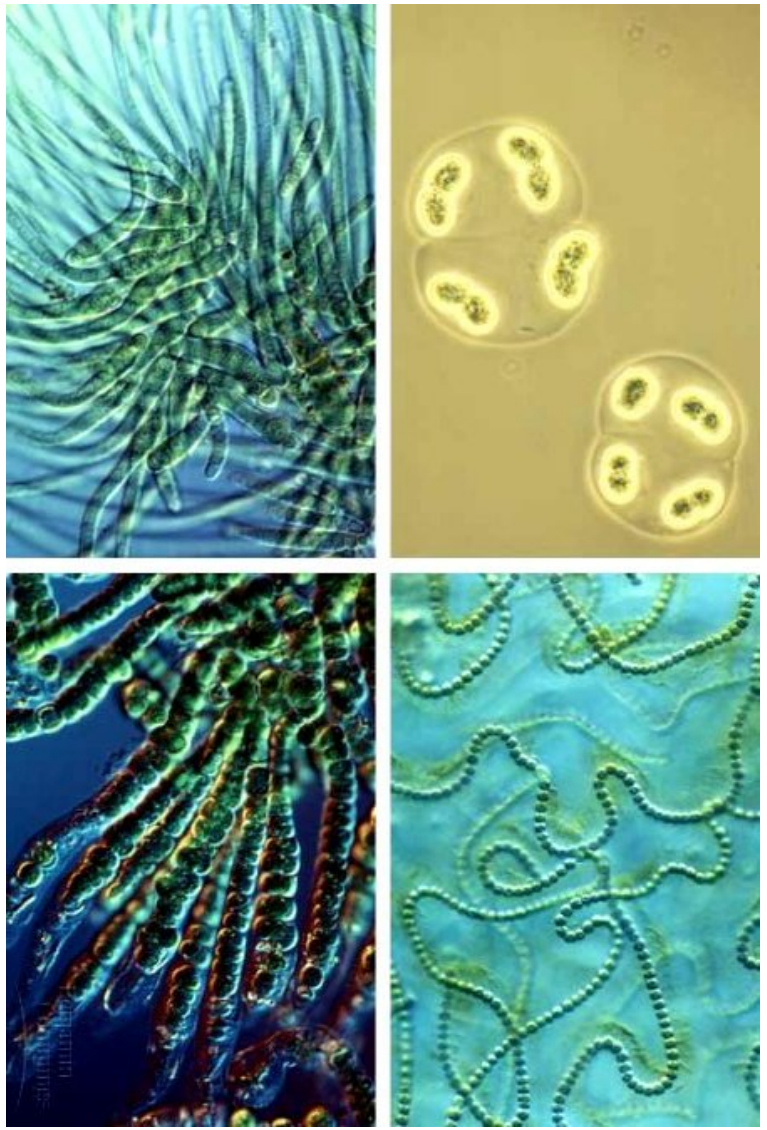
Process Space: Substrate & Product & Enzyme



Example: Synechococcus Elongatus

Synechococcus is a genus of cyanobacteria (blue-green algae): single-celled photosynthesising plankton that provide a foundation for the aquatic food chain.

S. Elongatus is a species of Synechococcus that is particularly abundant: some estimates suggest that it contributes 25% of marine nutrient primary production.

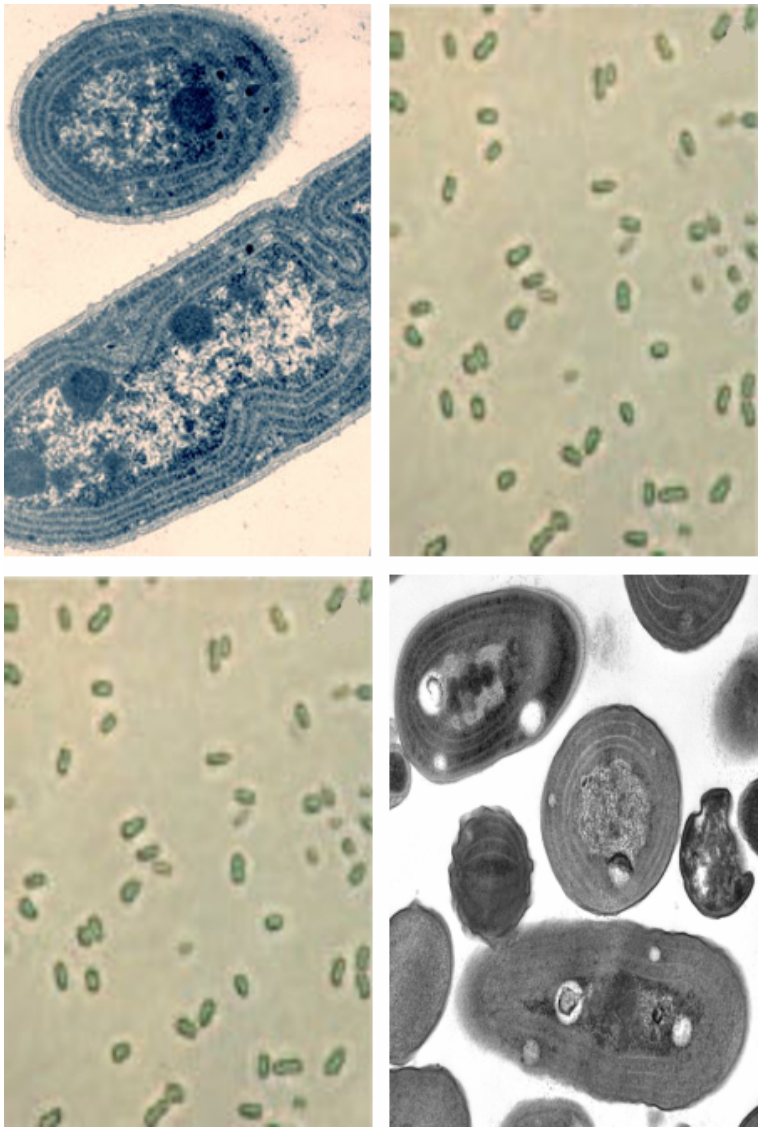


Circadian Clock in S. Elongatus

S. Elongatus has an internal clock, that turns genes on and off through day and night.

The cycling mechanism does not require gene transcription, and will operate in a test tube (*in vitro*).

Although it is entrained by light, it will also run for weeks without external stimulus.



Tomita, Nakajima, Kondo, Iwasaki.
No transcription-translation feedback in circadian rhythm of KaiC phosphorylation.
Science **307**(5707) (2005) 251–254

Proposed Mechanism

The S. Elongatus clock requires three proteins: KaiA, KaiB and KaiC (for *kaiten*). One proposed mechanism is the following:

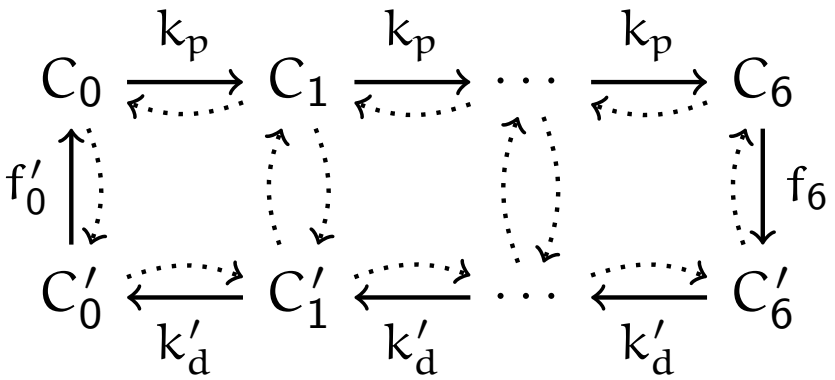
- KaiC forms hexamers, with six phosphorylation sites.
- KaiC also has two conformations; it preferentially phosphorylates in one and dephosphorylates in the other,
- KaiA catalyses phosphorylation of the first (active) conformation.
- KaiB dimers stabilise the second (inactive) conformation.
- A KaiB dimer bound to KaiC will bind a further two KaiA, removing them from other possible interactions.
- Cyclic phosphorylation of individual KaiC gives the basic mechanism; interaction with varying KaiA and KaiB coordinates this across the cell.

van Zon, Lubensky, Altena, ten Wolde.

An allosteric model of circadian KaiC phosphorylation.

PNAS **104**(18) (2007) 7420–7425

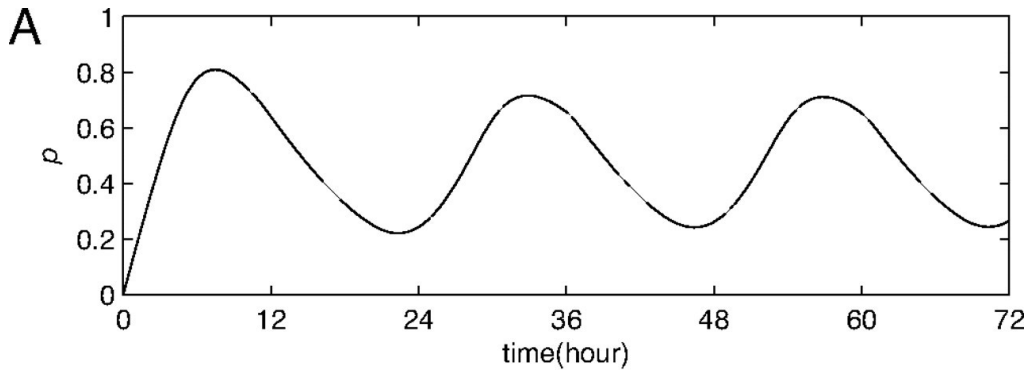
ODE Model



Active forms

Inactive forms

van Zon et al. give an ODE model of this mechanism, and show that it cycles. They conjecture that *differential affinities* are a key feature.



Continuous π Model

$$C_i = \nu(u, r, t{:}M_i).((\tau@k_p.C_{i+1}) + (\tau@f_i.C'_i) + (\tau@k_d.C_{i-1}) + (a_i\langle t\rangle.(u.C_i + r.C_{i+1})))$$

$$C'_i = (\tau@k'_p.C'_{i+1}) + (\tau@f'_i.C_i) + (\tau@k'_d.C'_{i-1}) + (b'_i.b'_i.BC'_i)$$

$$BC'_i = (\tau@k'_p.BC'_{i+1}) + (\tau@k'_d.BC'_{i-1}) + (\tau@k_i^{uB}.(C'_i \mid B \mid B)) + (a'_i.a'_i.ABC'_i)$$

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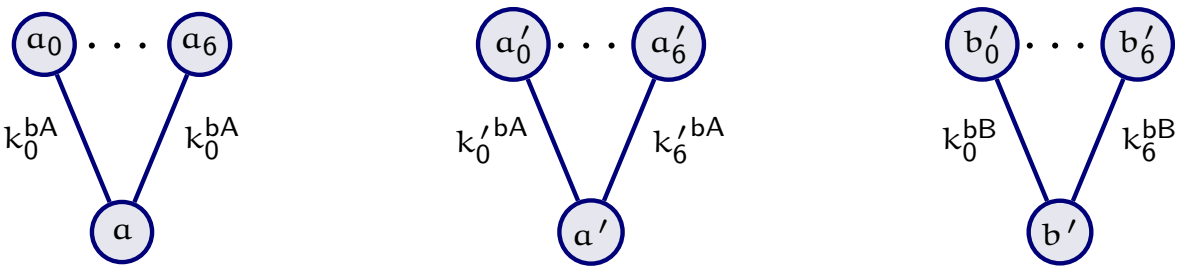
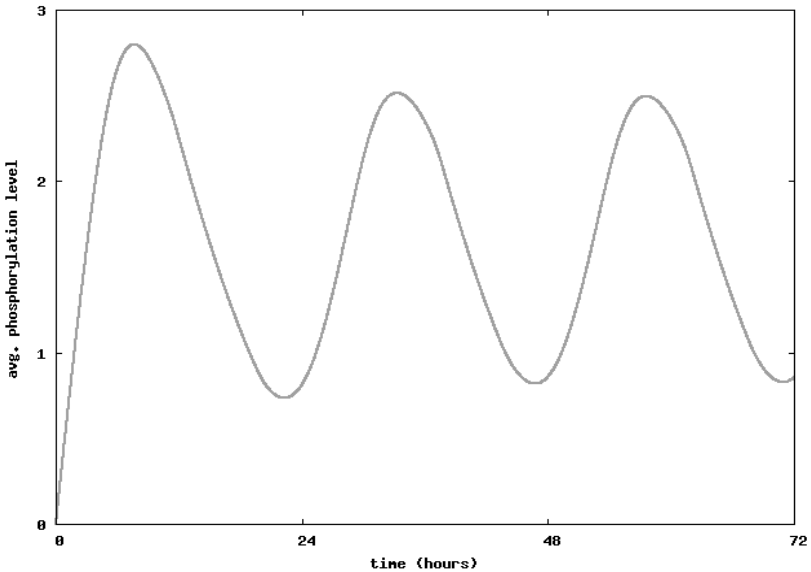
$$A = a(x).x.A + a'.0$$

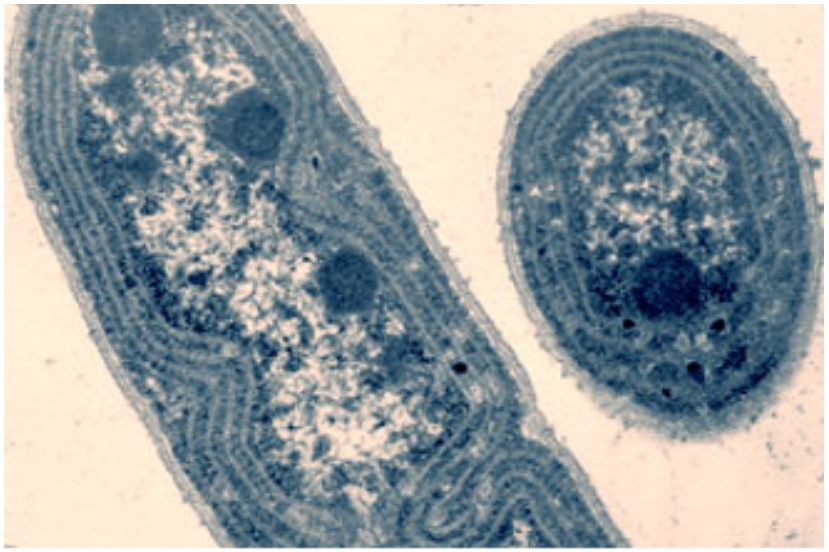
$$B = b'.0$$

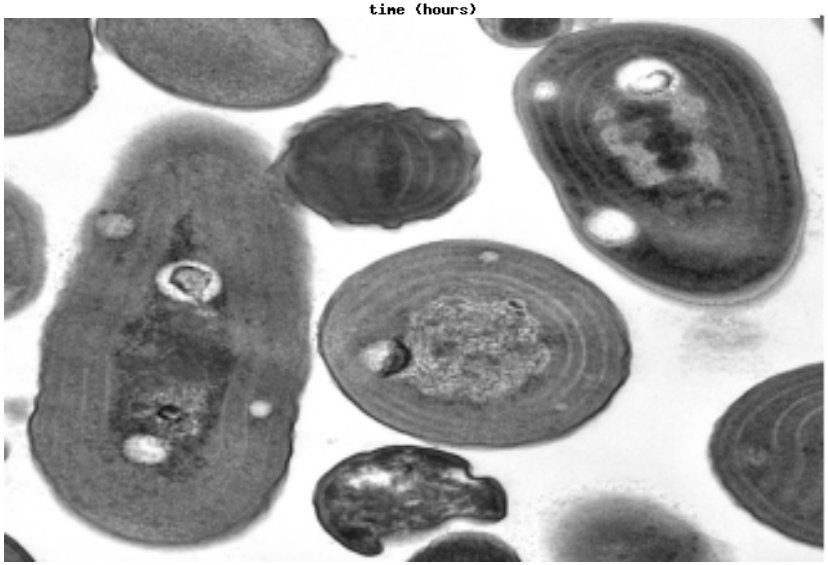
$$P = 0.58 \cdot A \parallel 0.58 \cdot B \parallel 1.72 \cdot C_0$$

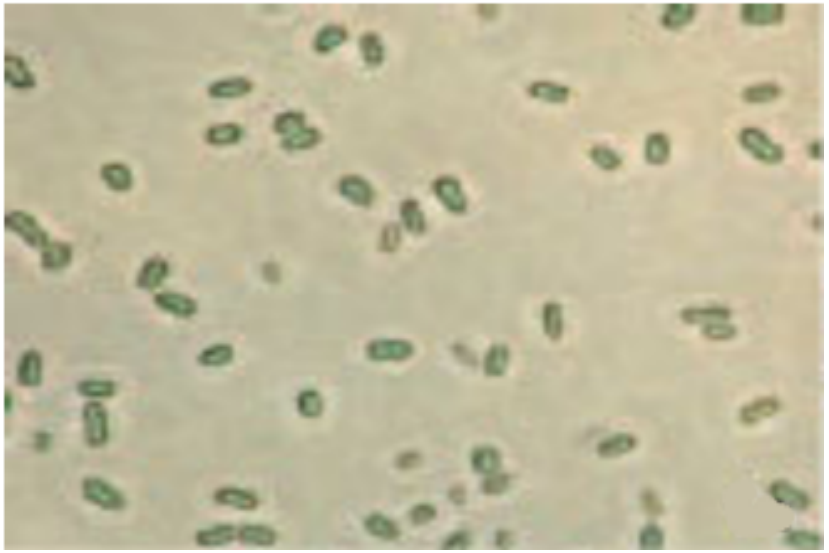


Running π









Modification: Remove autonomous phosphorylation

$$C_i = \nu(u, r, t{:}M_i).((\tau@k_p.C_{i+1}) + (\tau@f_i.C'_i) + (\tau@k_d.C_{i-1}) + (a_i\langle t\rangle.(u.C_i + r.C_{i+1})))$$

$$C'_i = (\tau@k'_p.C'_{i+1}) + (\tau@f'_i.C_i) + (\tau@k'_d.C'_{i-1}) + (b'_i.b'_i.BC'_i)$$

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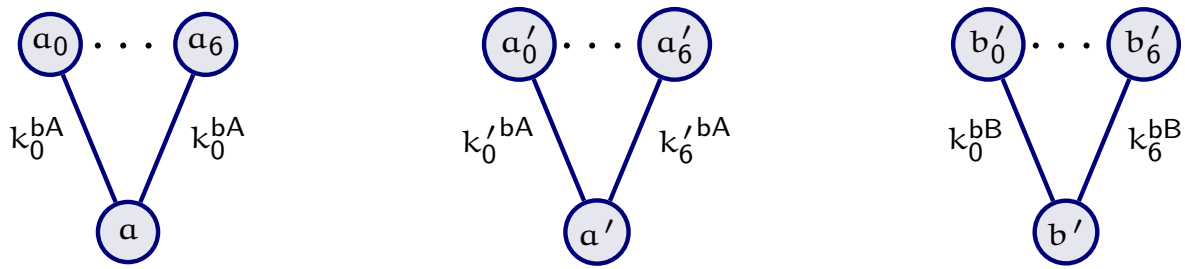

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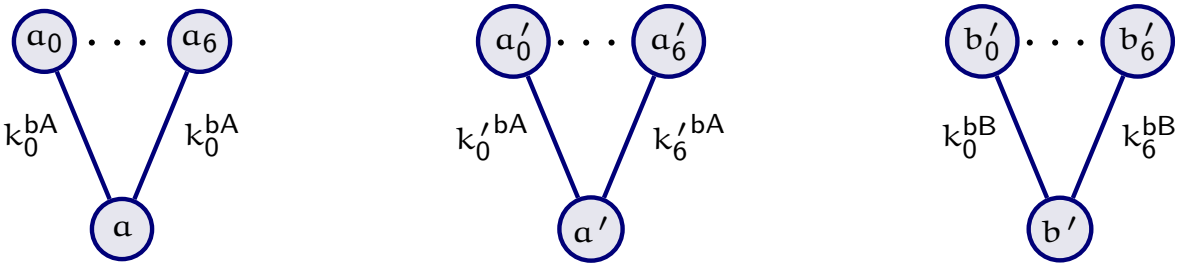

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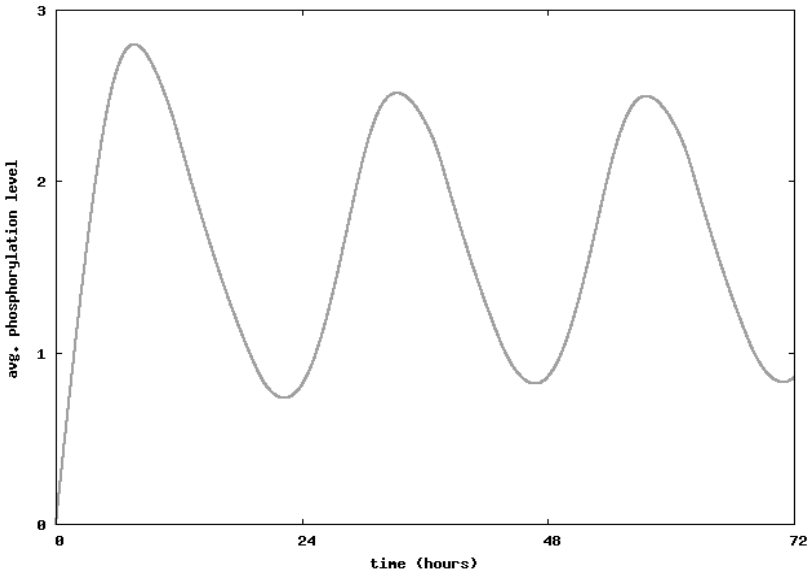

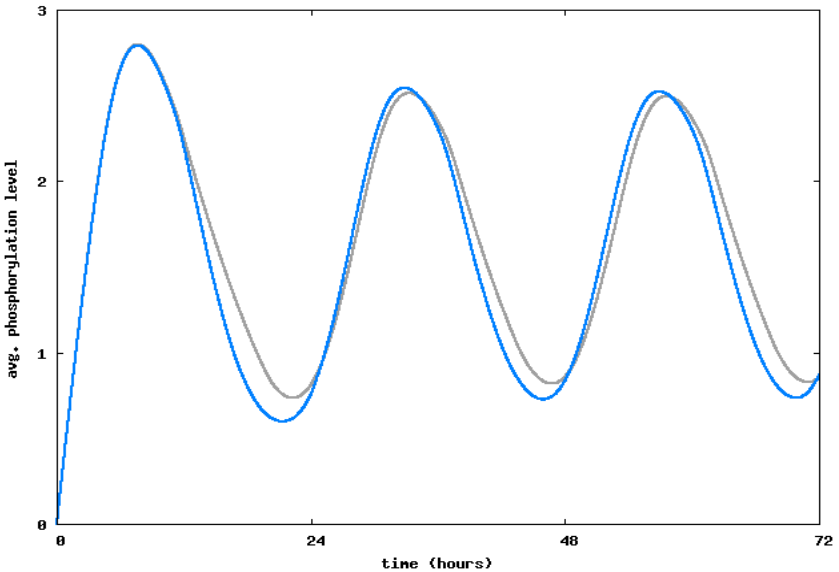

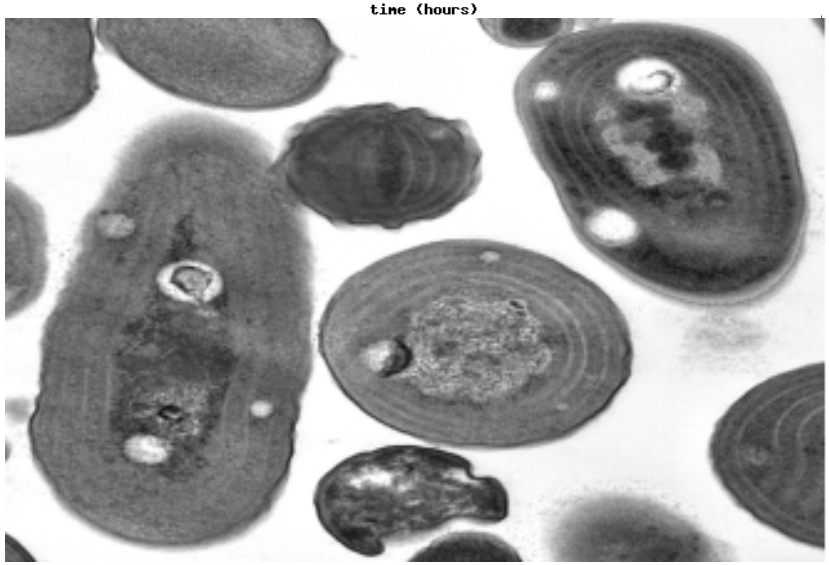

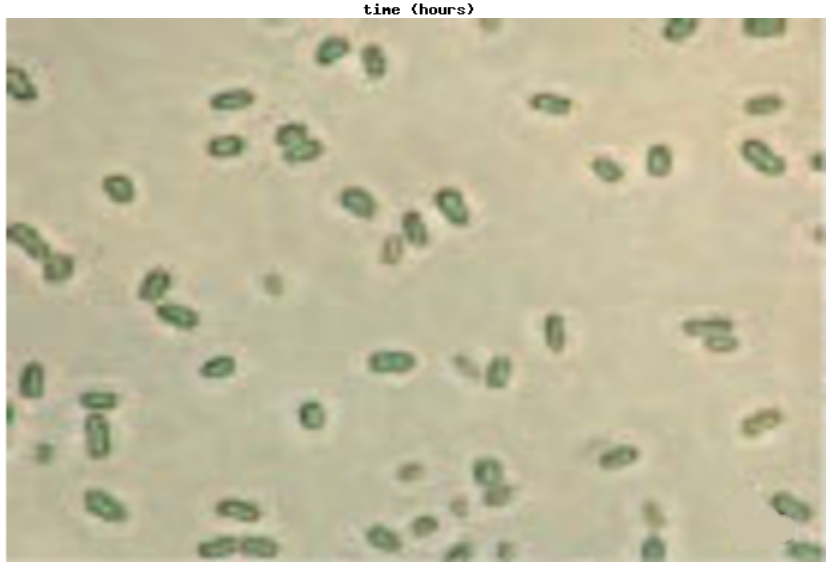

Modification: Weaken KaiA binding

$$C_i = \nu(u, r, t{:}M_i).((\tau@k_p.C_{i+1}) + (\tau@f_i.C'_i) + (\tau@k_d.C_{i-1}) + (a_i\langle t\rangle.(u.C_i + r.C_{i+1})))$$

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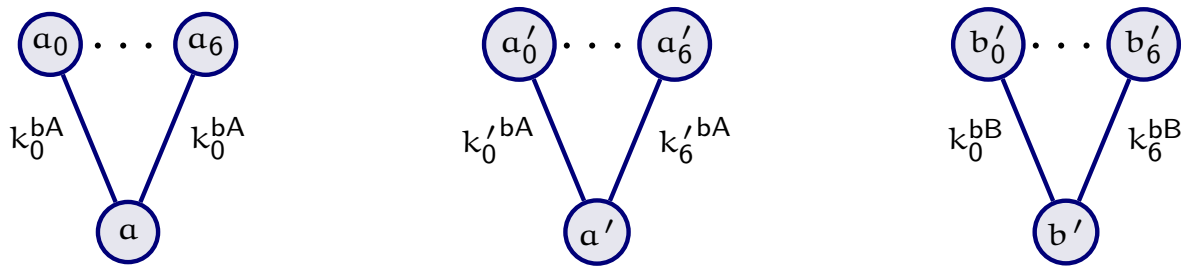

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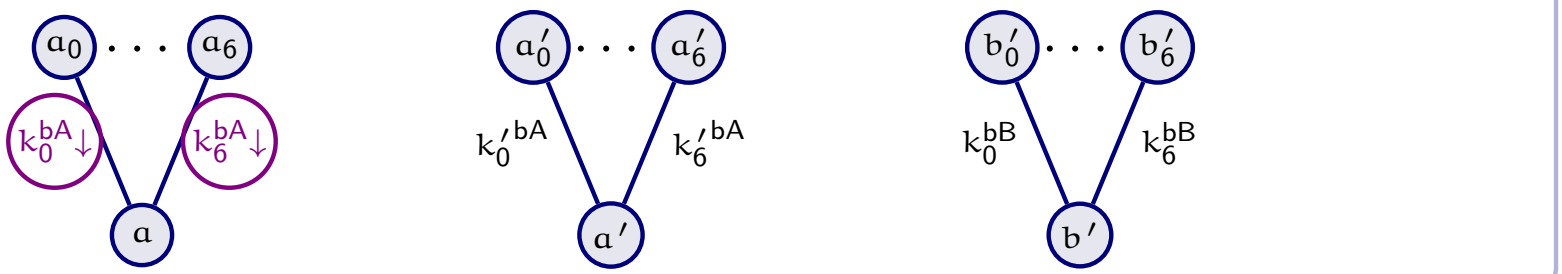

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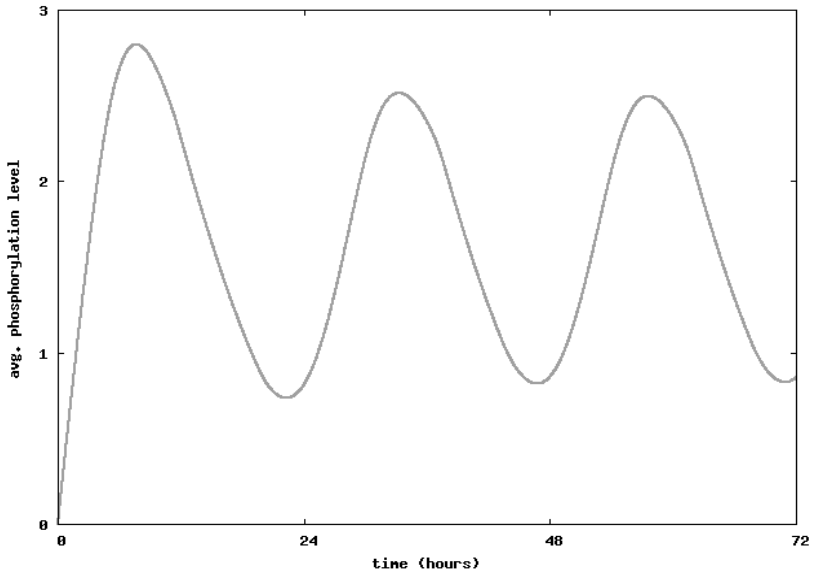

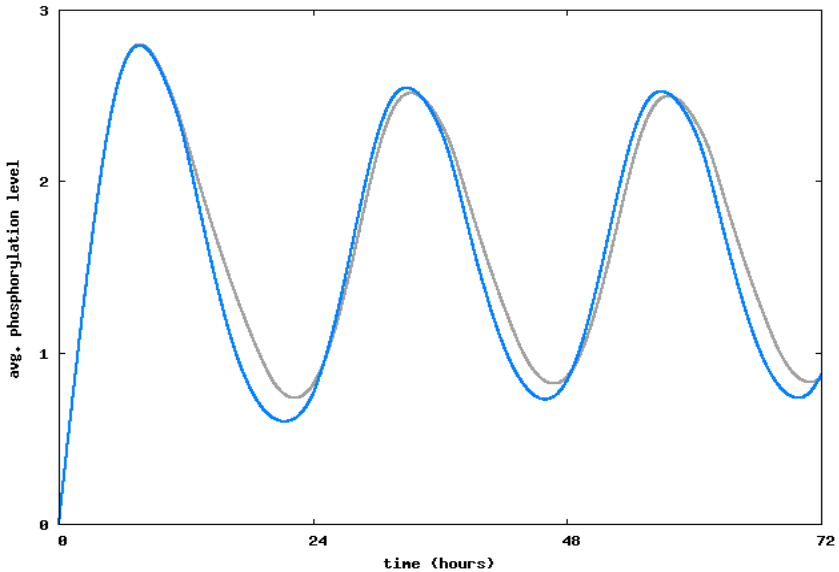

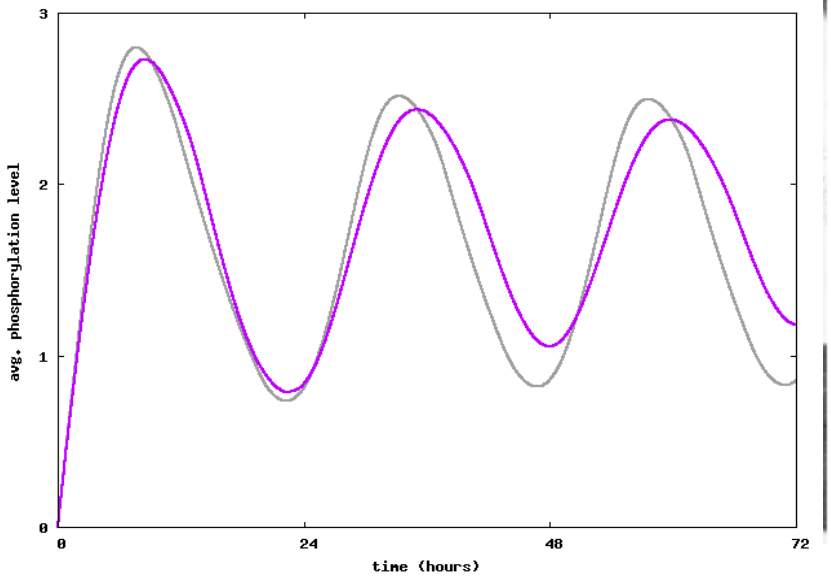

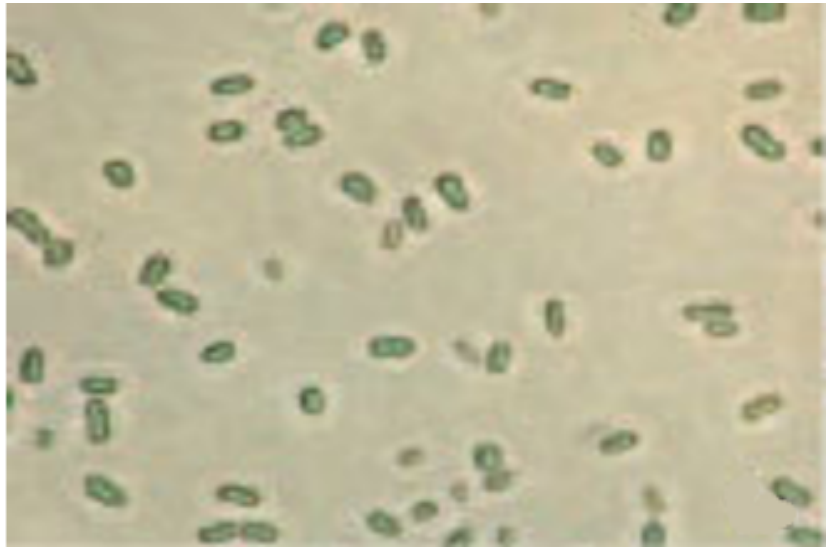

Modification: KaiA-KaiB dimers

$$C_i = \nu(u, r, t{:}M_i).((\tau@k_p.C_{i+1}) + (\tau@f_i.C'_i) + (\tau@k_d.C_{i-1}) + (a_i\langle t\rangle.(u.C_i + r.C_{i+1})))$$

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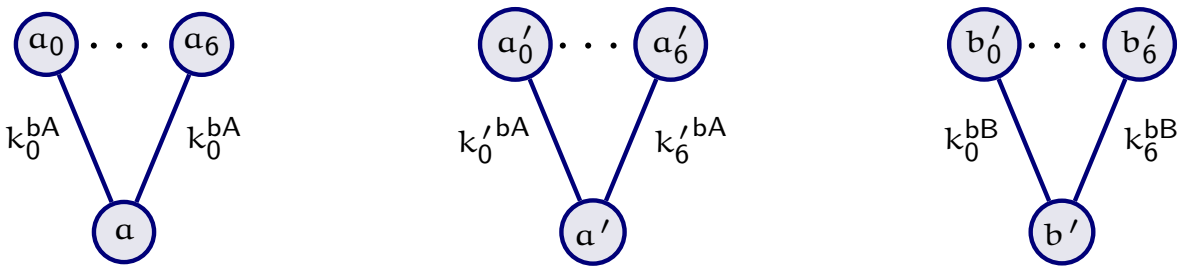

Modification: KaiA-KaiB dimers

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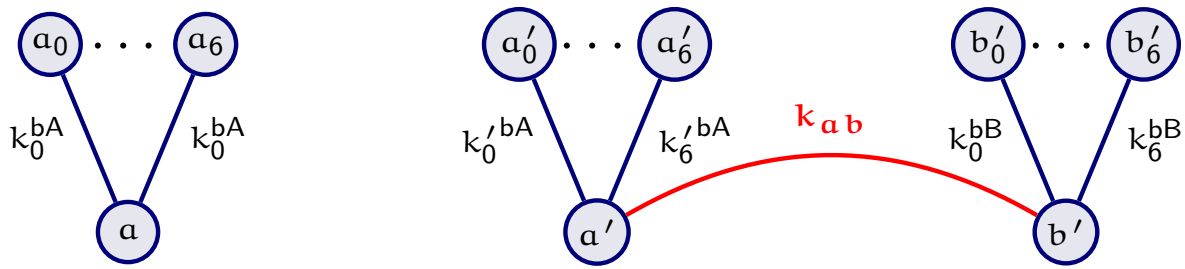
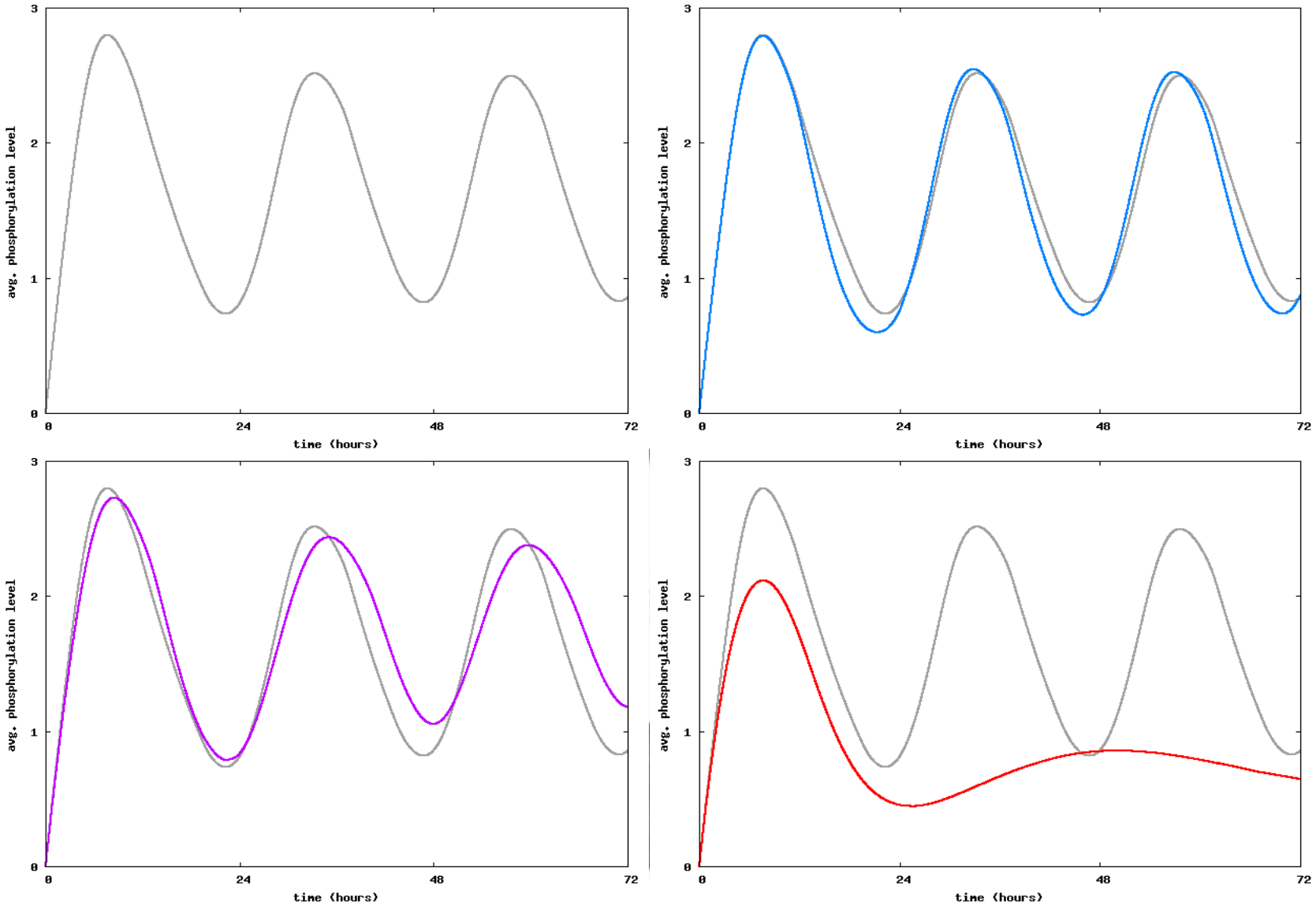


Modification: KaiA-KaiB dimers



Variation

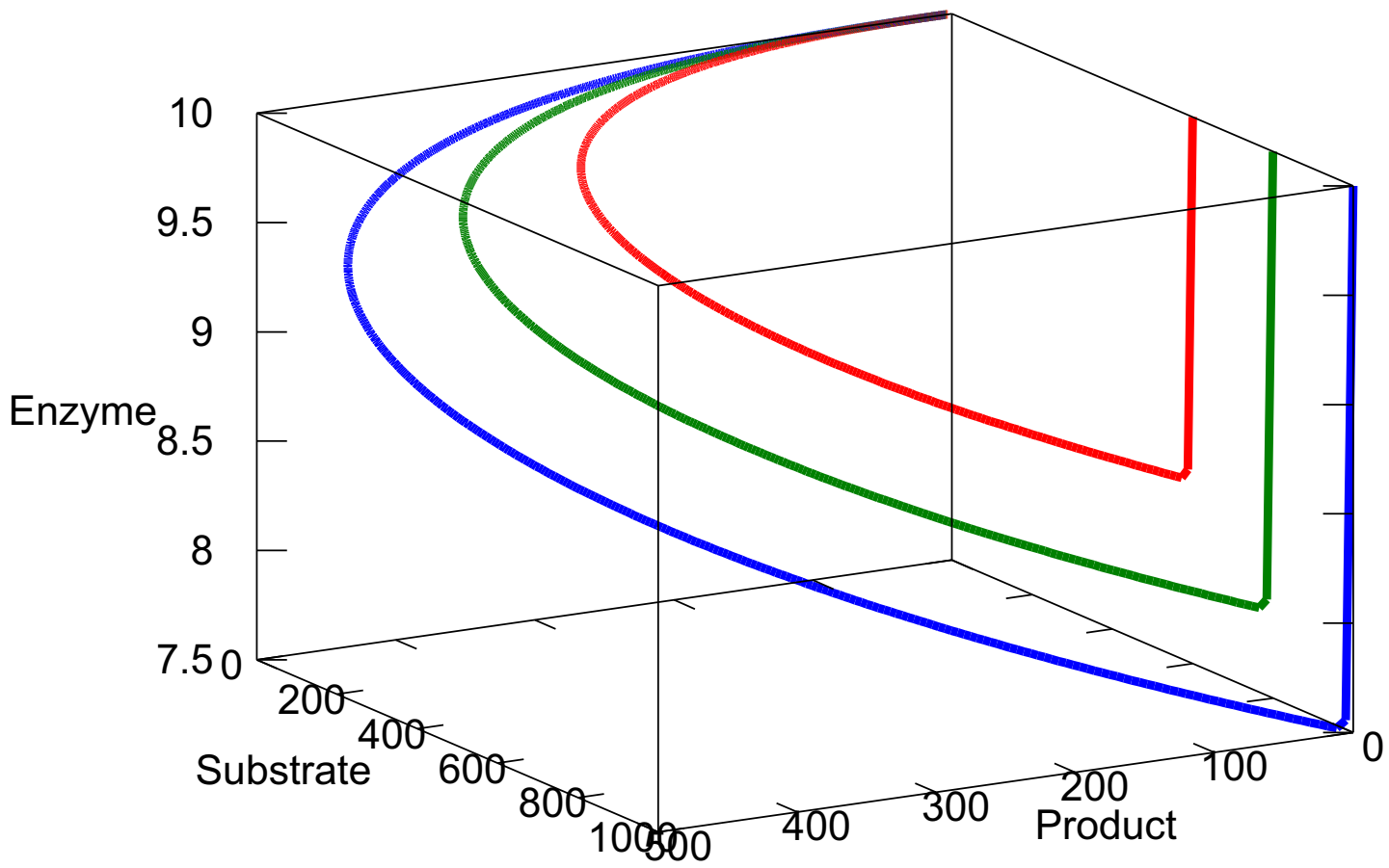
Given a system as a sequence of $c\pi$ species definitions, we can consider perturbing it in various ways.

- Duplication of species definitions
- Duplication of names
- Modification of affinities
- Modification of explicit rates
- Introduction of local names (complexation)

Larger changes can be built up by composing an appropriate basis of variation operators.

While clearly $c\pi$ syntax does not correspond directly to genetics, it is possible to identify syntactically simple variations that are also biologically plausible.

Process Space: Substrate & Product & Enzyme



Further Work: Analysis

To analyse the behaviour of a $c\pi$ system, we can graph its trajectory.

However, one strength of having an intermediate language is the possibility of multiple alternative mathematical models and routes to analysis:

- Numerical solution of ODE simulation; inspection of trajectory.
- Linear (metric) temporal logic over traces from ODE simulation
  [e.g. Fages & Rizk 2007]
- Temporal logic on $c\pi$ processes. Immediate behaviour: $\mathrm{P} \vdash \mathrm{G}_{\leqslant \mathrm{t}}(\phi)$
- Contextual logic for $c\pi$ processes. Interaction potential: $\mathrm{Q} \vdash \mathrm{F}^{\mathrm{c}\cdot\mathrm{a}}_{\leqslant \mathrm{t}}\mathrm{G}(\psi)$

With both variation (genotype) and model-checking (phenotype) we would have the means to explore evolvability, robustness, and neutrality.

Review

- Continuous π-calculus
  - Modular description of biomolecular systems
  - Compositional semantics in real vector spaces
  - Flexible interaction structure
- S. Elongatus circadian clock
  - Protein-protein interaction *in vitro*
  - Candidate mechanism oscillates
  - Behaviour under system variation

Future Work

- Behavioural analysis
  - Temporal logic on numerical simulation traces
  - Model-checking on $c\pi$ terms $P \vdash G_{\leqslant t}(\phi)$
  - Compositional logic of interaction potential $Q \vdash F^{c \cdot a}_{\leqslant t} G(\psi)$
- System Evolution
  - Evolutionary trajectories
  - Variation, evolvability
  - Robustness and neutrality
- Alternative Semantics
  - Markov chains
  - Stochastic simulation
  - Hybrid models, protein/DNA interaction

References

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