

Understanding Intracellular Signalling in Bacterial Chemotaxis

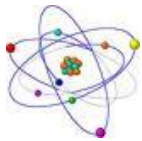
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Motivation

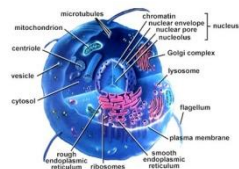
- Biological and biomedical sciences are today driven by discoveries at the subcellular level.
- Impact of changes at subcellular level is felt mostly on the multicellular or tissue level cell, e.g. drug discovery.
- Understanding many biological systems requires integrating information across various length scales and also timescales.
- Integrating such information is a major challenge for Systems Biology and related fields.



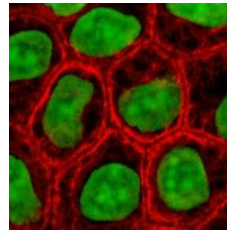
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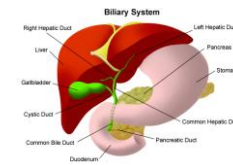
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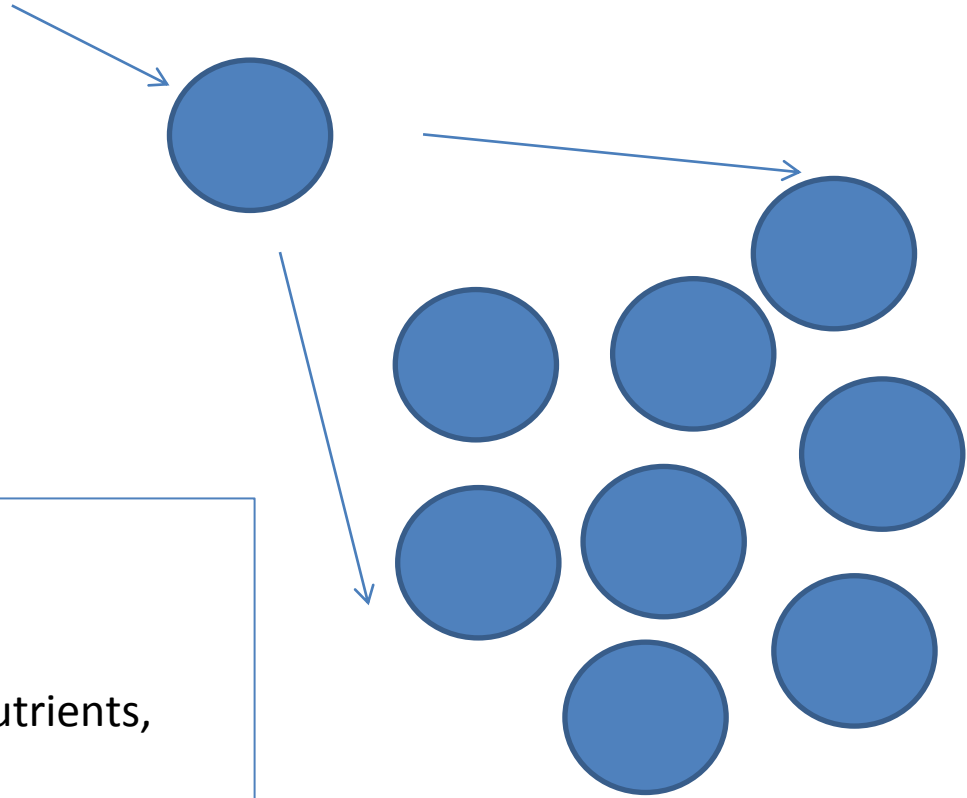
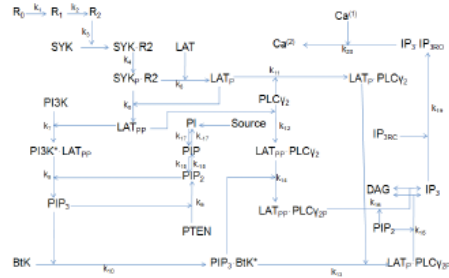


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Motivation



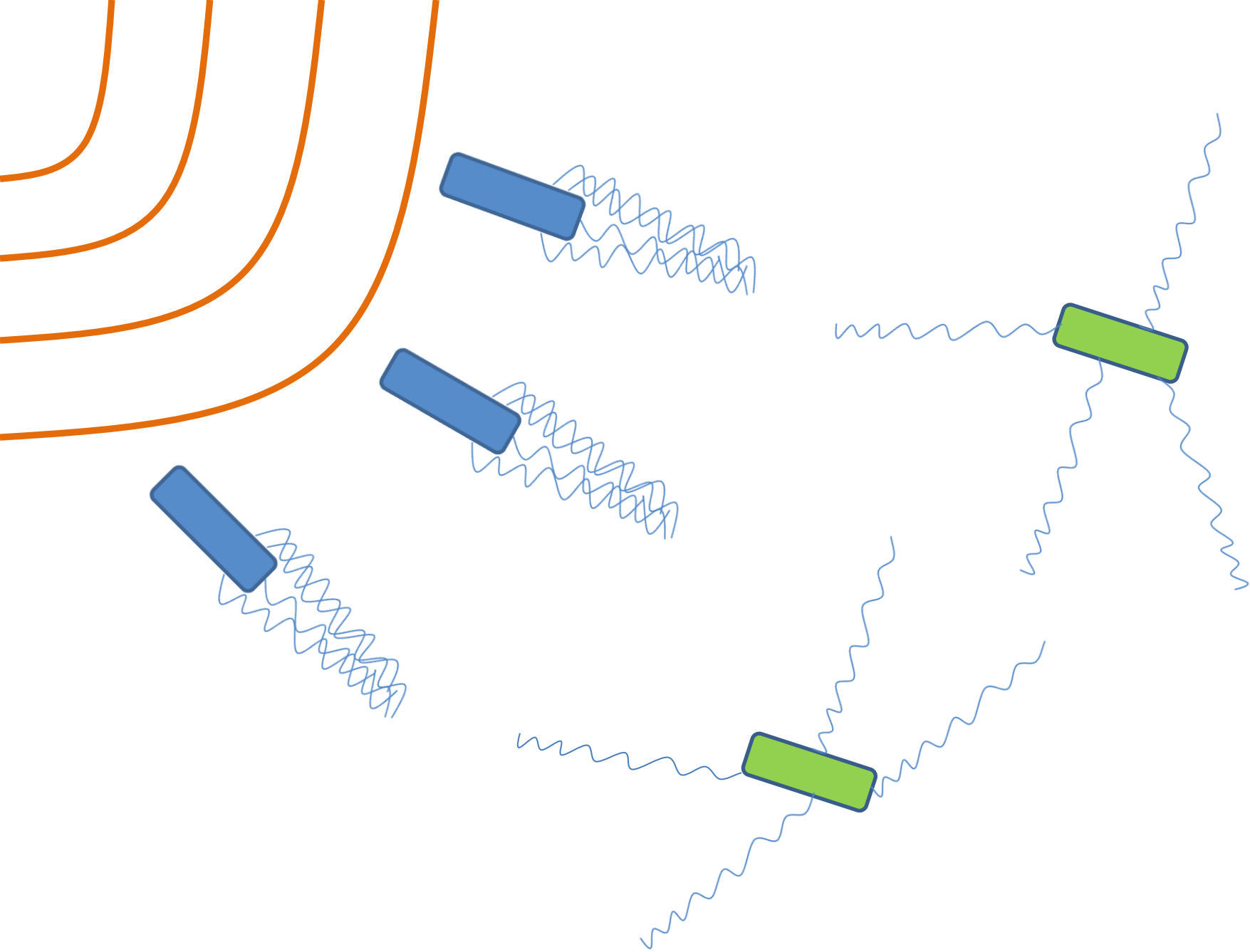
- Cellular biochemistry
- Cell mechanics
- Extracellular matrix
- Extracellular factors (e.g. Nutrients, growth factors)

Mathematical Methods

- Deterministic ordinary and partial differential equations (ODEs and PDEs).
- Asymptotic methods.
- Numerical methods for solving ODEs and PDEs.
- Network theory.
- Inverse methods.
- Parameter estimation.
- Agent based modelling.
- Hybrid continuum/discrete methods.
- Multiscale modelling.

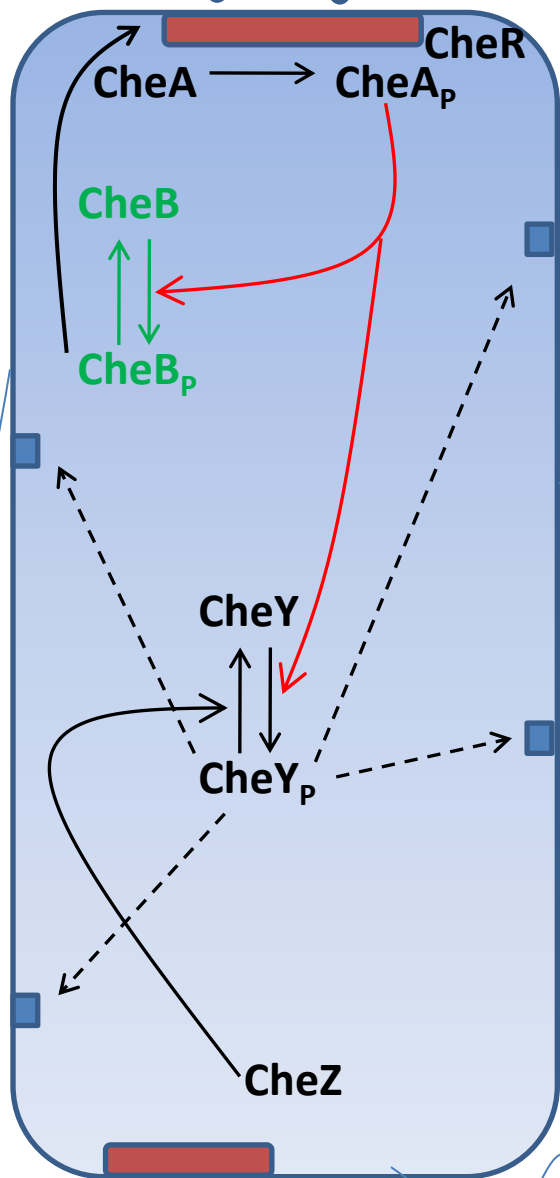
Life Science Application Areas

- Cardiovascular cell biology (platelets & cardiac myocytes).
- Lipoprotein metabolism (including models of endocytosis).
- Tumour growth.
- Bacterial chemotaxis.

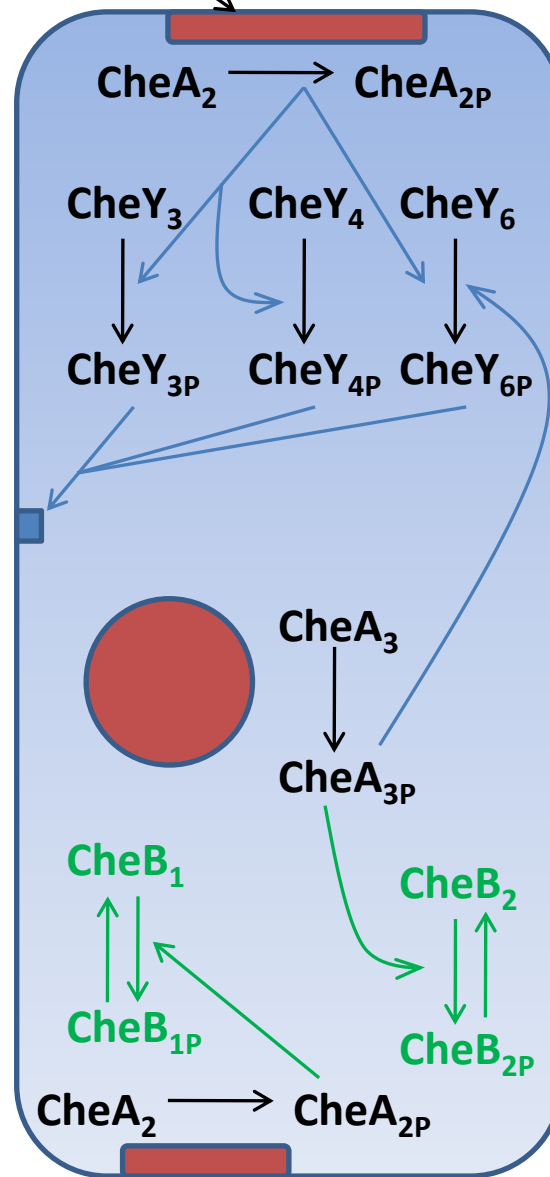


Ligand

Receptor clusters

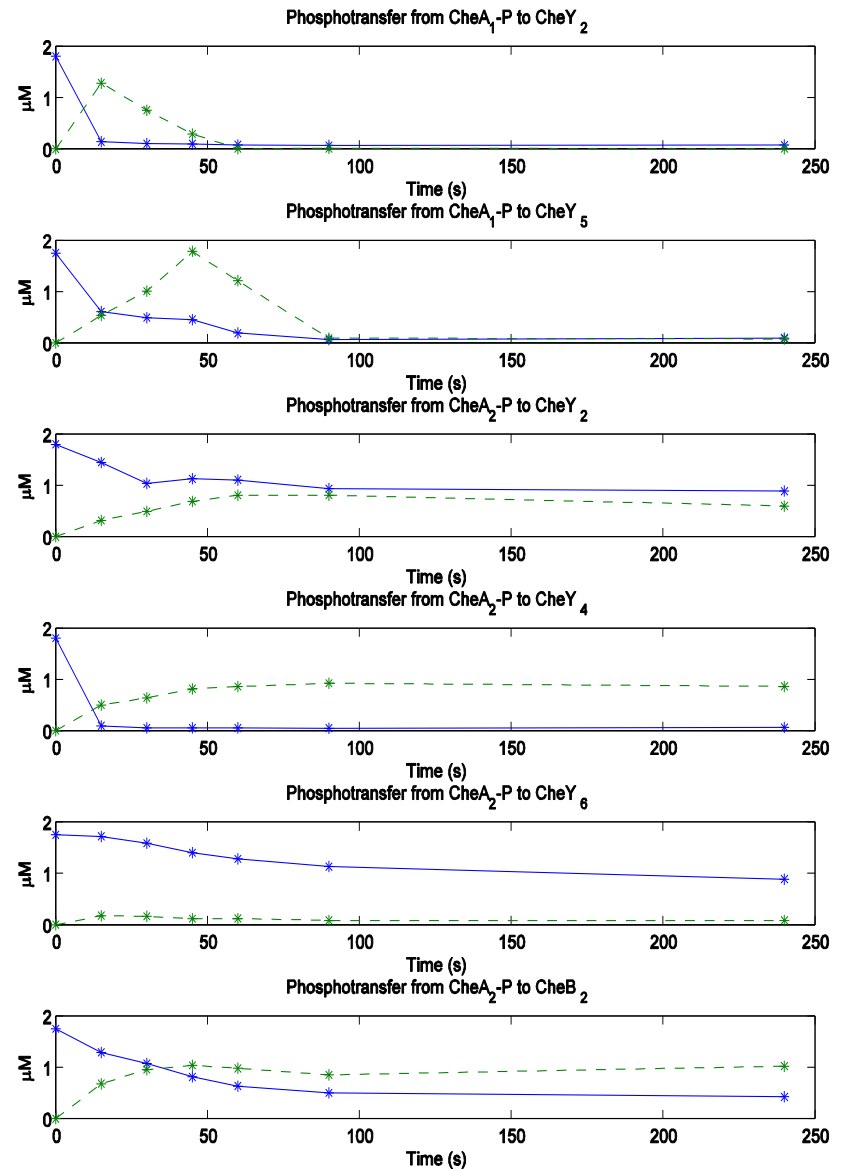
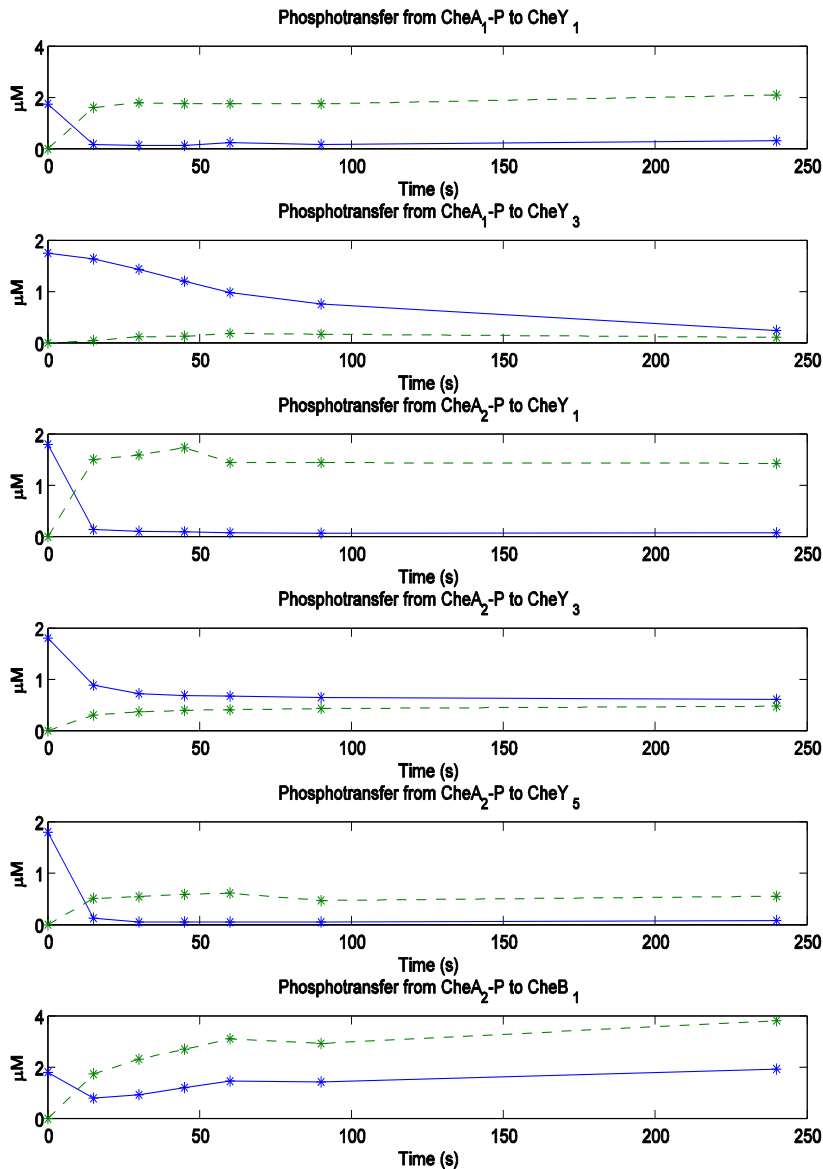


E. coli



R. sphaeroides

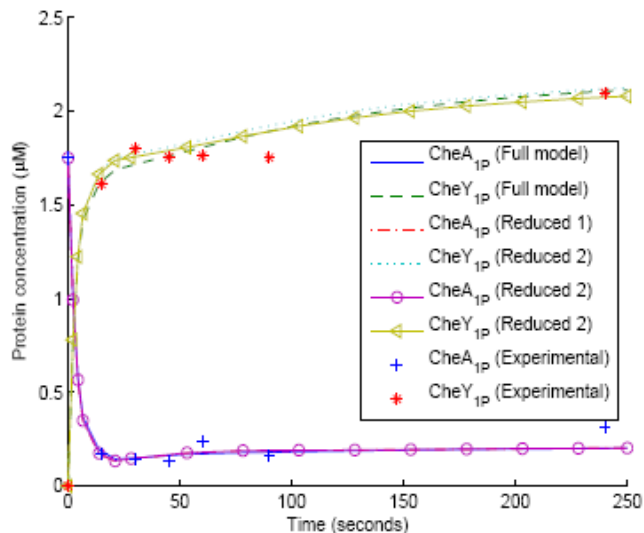
R. sphaeroides – Experimental Data



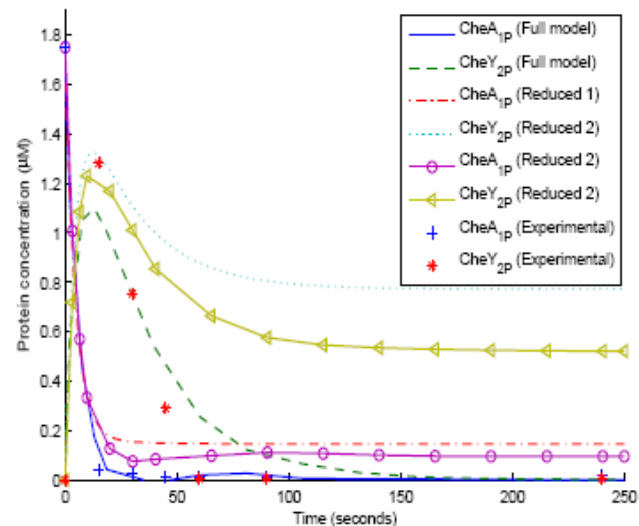
R. sphaeroides – Model parameterisation

Numerical optimisation (parameter fitting)

- Apply and compare a range of numerical optimisation methods for obtaining model parameters from experimental data.
- Allows us to compare various fits to the data to obtain the most robust set of parameters (as well as the method fits).

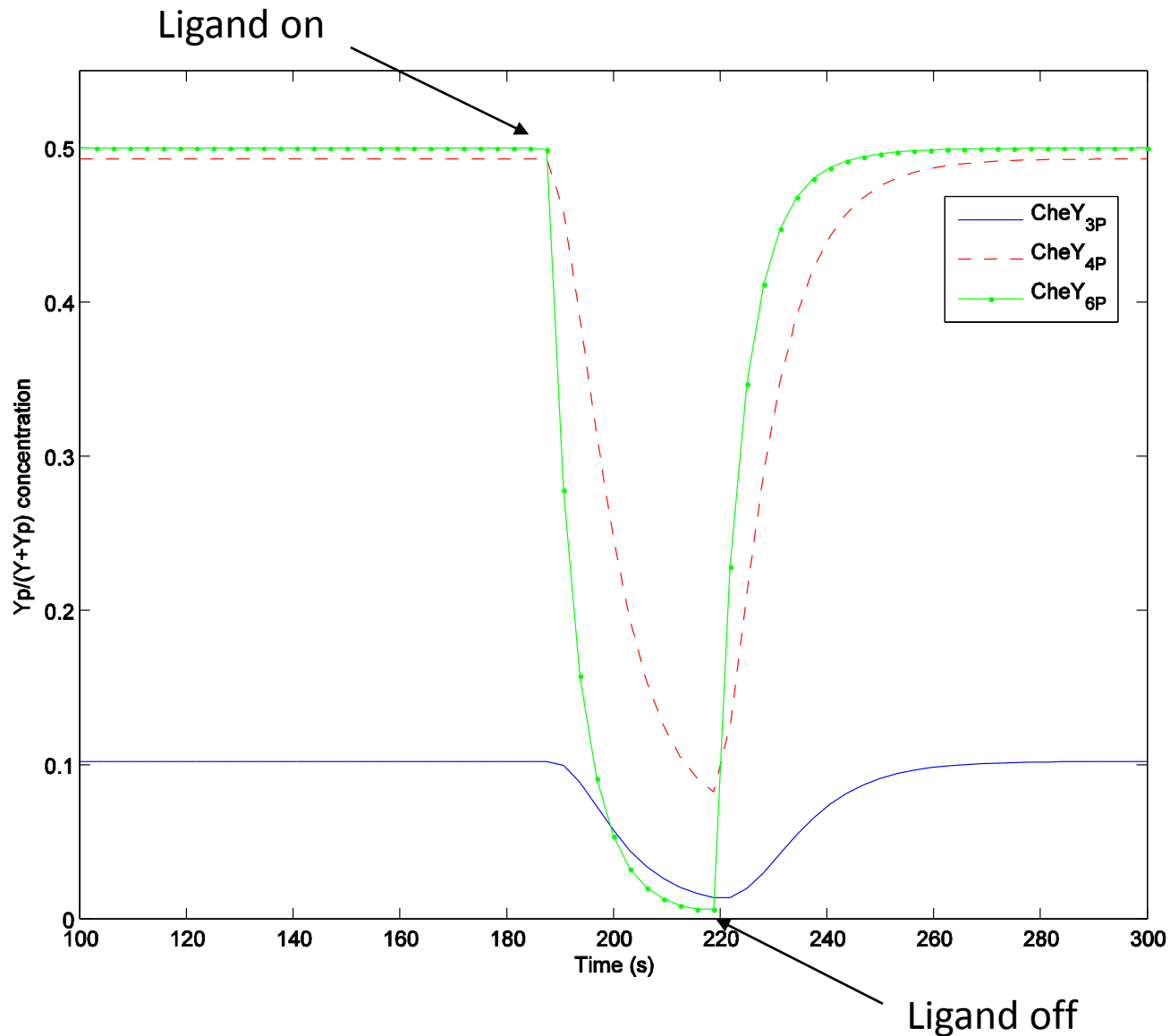


(a) CheA_{1P} to CheY_{1P}.

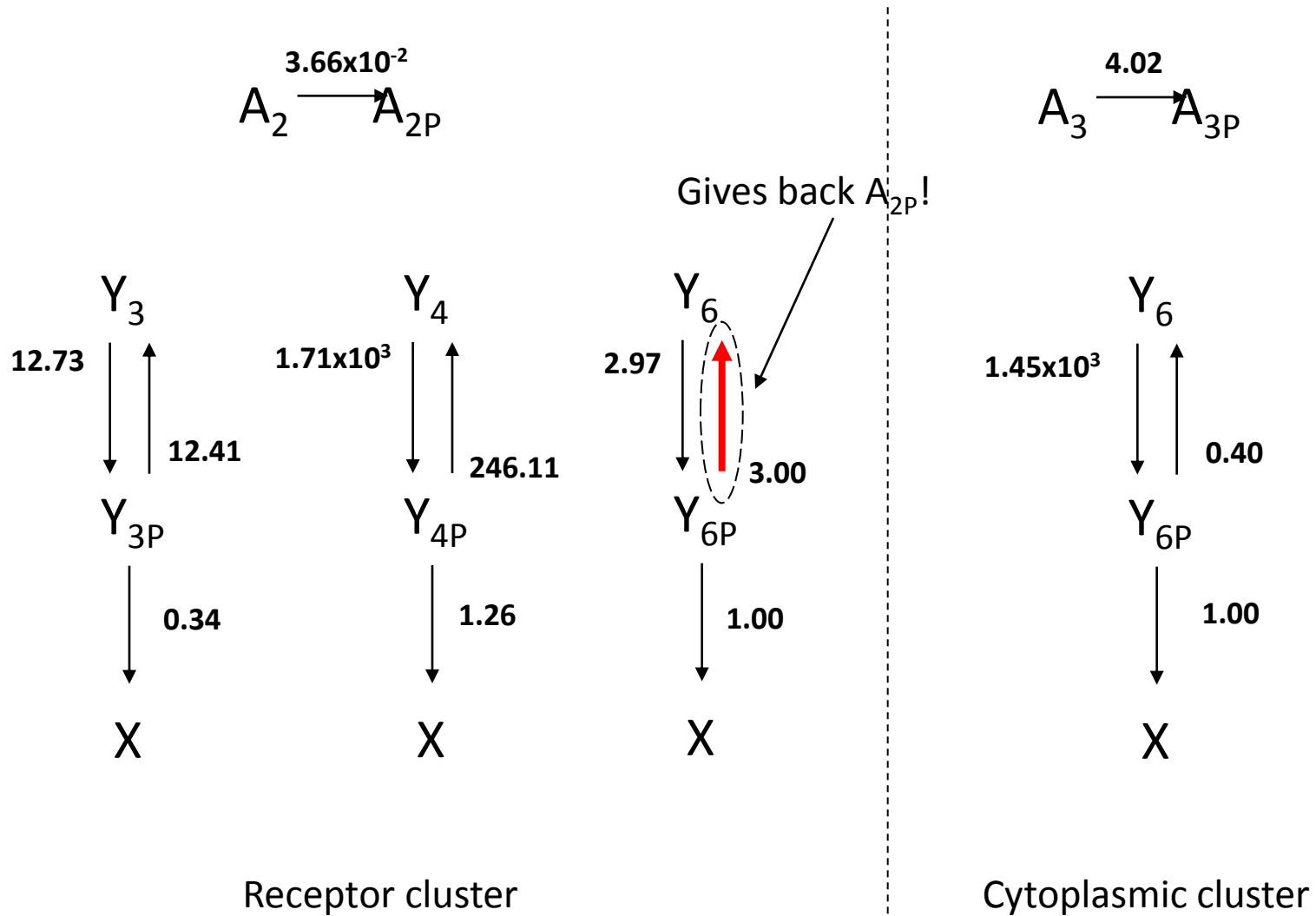


(b) CheA_{1P} to CheY_{2P}.

R. sphaeroides - Stimulation of the cytoplasmic cluster



R. sphaeroides - Phosphorelay pathway



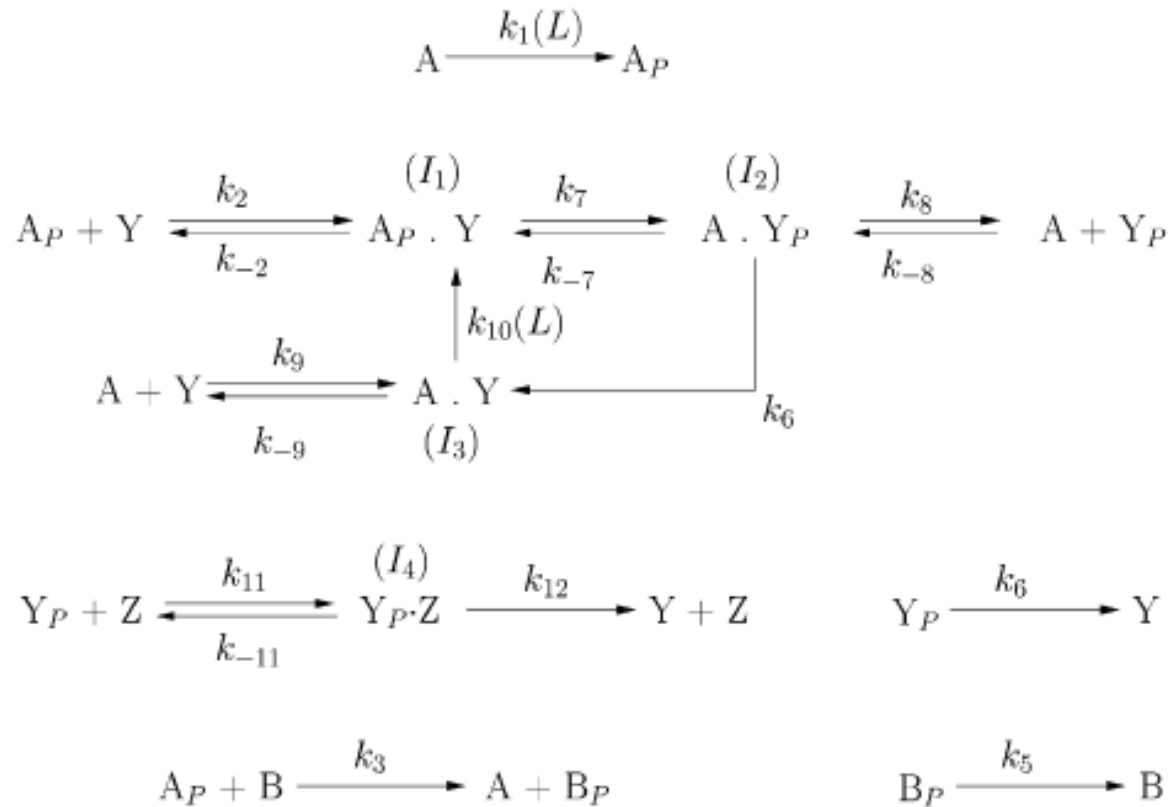
R. sphaeroides - Experimental Results

Time (s)	A3-P	B2-P	A2-P
0	21.21	0	0
15	20.58	0.97	0.03
30	21.55	0.92	0.03
60	20.55	0.92	0.05
120	19.70	0.94	0.07
240	19.29	1.00	0.11

E. coli – Phosphotransfer Reactions

Process	Reaction	Details
Autophosphorylation	$\text{CheA} \xrightarrow{k_1} \text{CheA}_p$	
Phosphotransfer	$\text{CheA}_p + \text{CheY} \xrightarrow{k_2} \text{CheA} + \text{CheY}_p$	CheA _p to CheY
	$\text{CheA}_p + \text{CheB} \xrightarrow{k_3} \text{CheA} + \text{CheB}_p$	CheA _p to CheB
Dephosphorylation	$\text{CheY}_p + \text{CheZ} \xrightarrow{k_4} \text{CheY} + \text{CheZ}$	Dephosphorylation by CheZ
	$\text{CheY}_p \xrightarrow{k_5} \text{CheY}$	Natural dephosphorylation
	$\text{CheB}_p \xrightarrow{k_5} \text{CheB}$	Natural dephosphorylation

E. coli - Complex formation



E. coli Complex formation - Parameterisation

Table 1

Parameter values used in our model as obtained from the indicated sources.

Parameter	Description	Value	Reference
k_1	Rate of CheA autophosphorylation	3.75 s^{-1}	Assumed
k_2	Rate of phosphotransfer from CheA _p + CheY to I ₁	$2.50 \times 10^6 (\text{Ms})^{-1\text{c}}$	Stewart (1997)
k_{-2}	Rate of I ₁ to CheA _p and CheY	$15 \text{ s}^{-1\text{c}}$	Stewart (1997)
k_3	Rate of phosphotransfer from CheA _p to CheB	$1.5 \times 10^7 (\text{Ms})^{-1}$	Bray website data ^b
k_5	Rate of CheB _p natural dephosphorylation	0.7 s^{-1}	Stewart (1993)
k_6	Rate of CheY _p natural dephosphorylation and A·Y _p to A·Y	$8.5 \times 10^{-2} \text{ s}^{-1}$	Sourjik and Berg (2002a)
k_7	Rate of I ₁ to I ₂	650 s^{-1}	Stewart (1997)
k_{-7}	Rate of I ₂ to I ₁	50 s^{-1}	Stewart (1997)
k_8	Rate of I ₂ to CheA + CheY _p	$250 \text{ s}^{-1\text{d}}$	Li et al. (1995)
k_{-8}	Rate of CheA + CheY _p to I ₂	$2.08 \times 10^7 (\text{Ms})^{-1 \text{d}}$	Li et al. (1995)
k_9	Rate of CheA + CheY to I ₃	$7.50 \times 10^6 (\text{Ms})^{-1}$	Stewart and van Bruggen (2004)
k_{-9}	Rate of I ₃ to CheA + CheY	15 s^{-1}	Stewart and van Bruggen (2004)
k_{10}	Rate of I ₃ to I ₁	3.75 s^{-1}	Assumed equivalent to k_1
k_{11}	Rate of CheY _p + CheZ to I ₄	$5.60 \times 10^6 (\text{Ms})^{-1}$	Silversmith et al. (2008)
k_{-11}	Rate of I ₄ to CheY _p + CheZ	0.04 s^{-1}	Silversmith et al. (2008)
k_{12}	Rate of I ₄ to CheY + CheZ	4.90 s^{-1}	Silversmith et al. (2008)
D_Y	CheY diffusion coefficient	$10 \mu\text{m}^2 \text{ s}^{-1}$	Elowitz et al. (1999)
D_{Y_p}	CheY _p diffusion coefficient	$10 \mu\text{m}^2 \text{ s}^{-1}$	Segall et al. (1985) Segall et al. (1985) Elowitz et al. (1999)
D_B	CheB diffusion coefficient	$7 \mu\text{m}^2 \text{ s}^{-1}$	Assumed ^a
D_{B_p}	CheB _p diffusion coefficient	$7 \mu\text{m}^2 \text{ s}^{-1}$	Assumed ^a
A_T	Total CheA concentration	$7.9 \mu\text{M}^{\text{e}}$	Bray website data ^b
Y_T	Total CheY concentration	$9.70 \mu\text{M}$	Bray website data ^b
B_T	Total CheB concentration	$0.28 \mu\text{M}$	Bray website data ^b
Z_T	Total CheZ concentration	$3.8 \mu\text{M}^{\text{e}}$	Bray website data ^b
L_x	Average length of a cell	$3 \mu\text{m}$	Darnton et al. (2007)
L_y	Average width of a cell	$1 \mu\text{m}$	Darnton et al. (2007)

^a Calculated using the Einstein diffusion approximation.

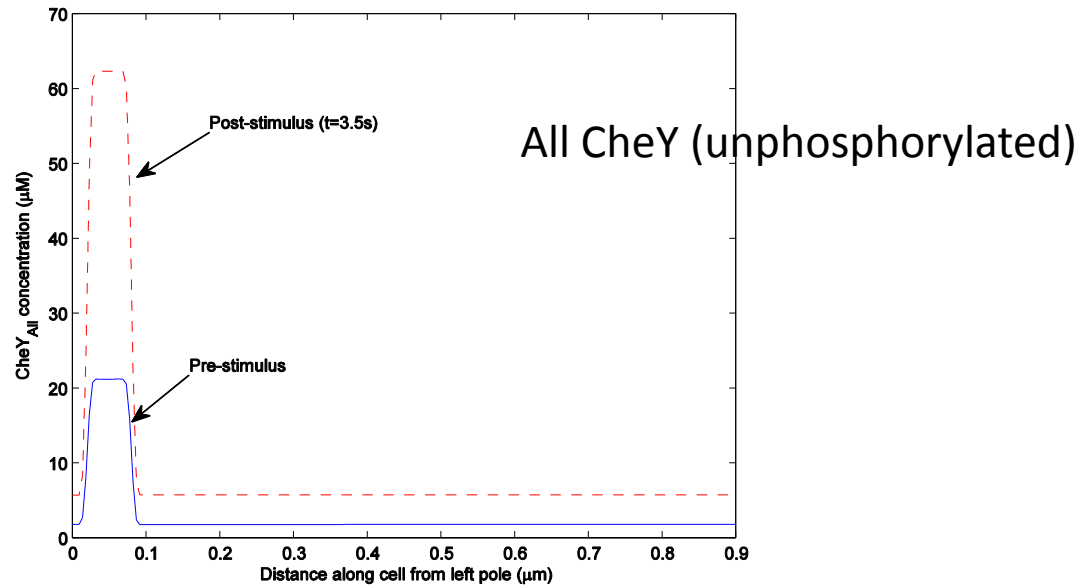
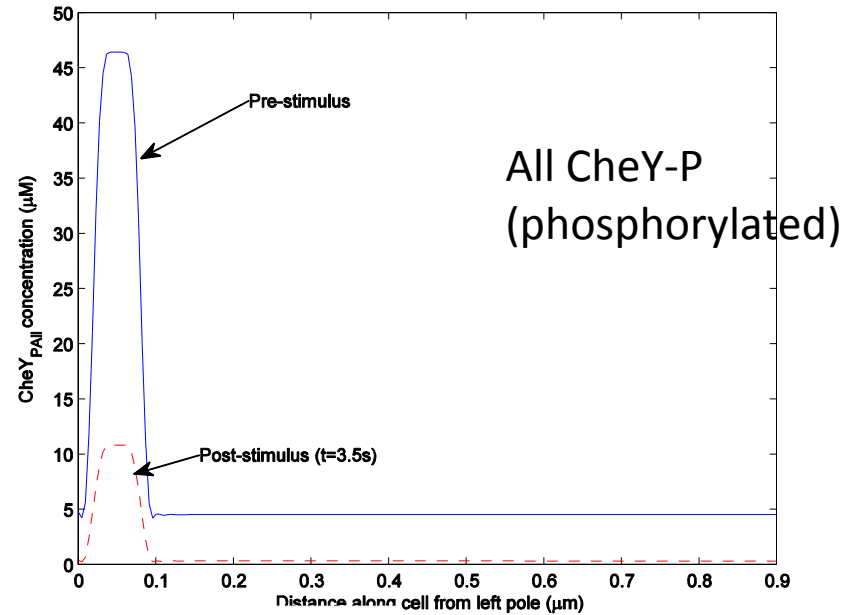
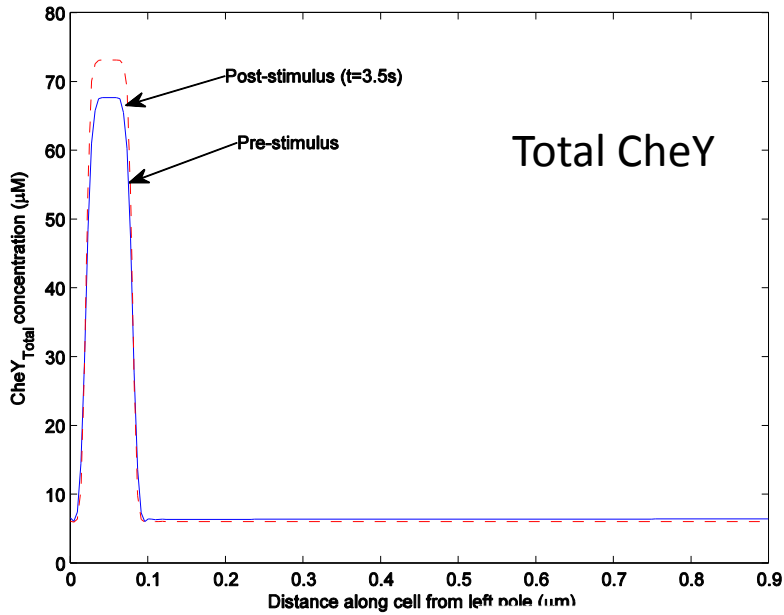
^b www.pdn.cam.ac.uk/groups/comp-cell/Rates.html.

^c Approximated using the stated dissociation value of $K_D = 6 \mu\text{M}$ as a guide.

^d Calculated from a dissociation value of $K_D = 4.0 \mu\text{M}$.

^e In our model setup the total size of the polar region is approximately 1/10th the volume of the cell and thus these concentrations are increased tenfold.

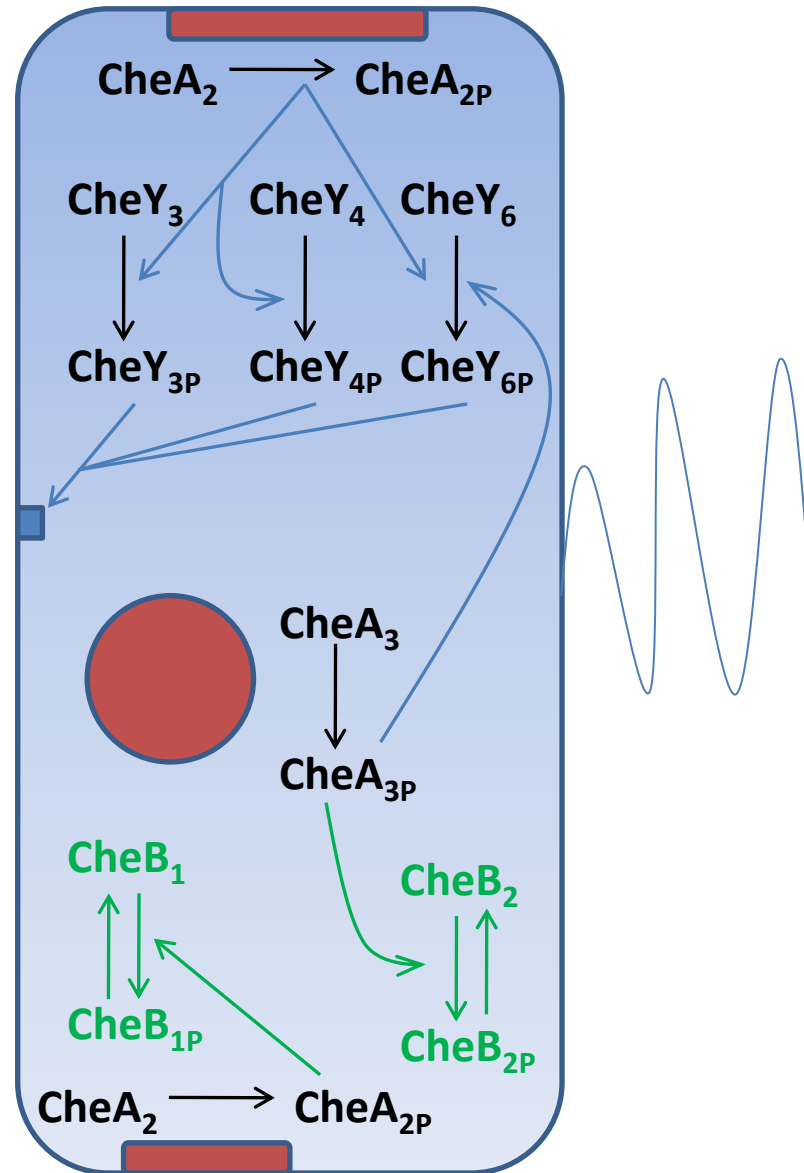
E. coli Complex formation – Wild-type cell



E. coli Complex formation

	CheY free	CheY _p free	CheA·CheY	CheA·CheY _p	CheA _p ·CheY	CheY _p ·CheZ
Wild-type ($Z_T = 3.8 \mu\text{M}$)						
Pre-stimulus	18.56	46.97	10.20	5.83	0.59	17.91
Post-stimulus	58.62	3.15	32.22	0.41	0.07	5.51
Low CheZ ($Z_T = 0.1 \mu\text{M}$)						
Pre-stimulus	1.74	95.23	0.16	2.17	0.18	0.51
Post-stimulus	4.7	81.98	2.85	9.22	0.73	0.51
High CheZ ($Z_T = 12 \mu\text{M}$)						
Pre-stimulus	40.64	6.55	22.77	1.33	0.32	28.39
Post-stimulus	60.81	0.81	32.56	0.18	0.06	5.57

R. sphaeroides - Adaptation



Current Work

- Elucidating adaptation in *R. sphaeroides*.
- Reducing model complexity in respect of protein complex formation.
- Developing spatiotemporal models (incorporating diffusion) of intracellular signalling in *R. sphaeroides*.

Acknowledgements

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Publications

Tindall, M.J., Determining model parameters from experimental data: Model reduction, numerical optimisation and asymptotic methods. In preparation.

Tindall, M.J., Gaffney, E., Maini, P.K. and Armitage, J.P. Theoretical insights into bacterial chemotaxis (Invited review).

Tindall, M.J., Porter, S.L., Maini, P.K. and Armitage, J.P. Modelling chemotaxis reveals the role of reversed phosphotransfer and a bi-functional phosphatase. *PLoS Comput. Biol.*, 6(8), e1000896, 2010.

Tindall, M.J., Porter, S.L., Wadhams, G., Maini, P.K. and Armitage, J.P. Spatiotemporal modelling of CheY complexes in *Escherichia coli* chemotaxis. *Prog. Biophys. Mol. Biol.*, E-print PMID: 19540260, 2009.

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Tindall, M.J., Porter, S.L., Maini, P.K., Gaglia, G. and Armitage, J.P. Overview of mathematical approaches used to model bacterial chemotaxis I: The single cell. *Bull. Math. Biol.*, 70(6), 1525-69, 2008.

Tindall, M.J., Maini, P.K., Porter, S.L. and Armitage, J.P. Overview of mathematical approaches used to model bacterial chemotaxis II: Bacterial populations. *Bull. Math. Biol.*, 70(6), 1570-607, 2008.