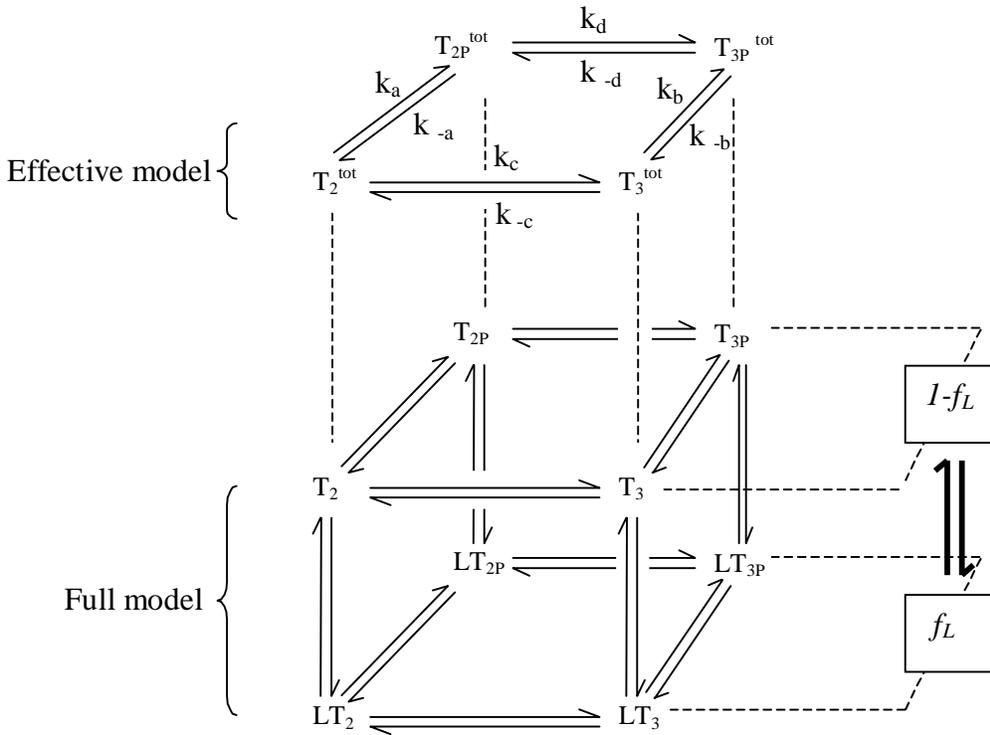


**Problem Set 2**  
**Due in class**

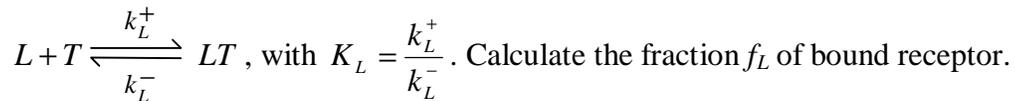
**Assigned: 10.06.04**  
**Due: 10.19.04**

1. Biochemistry of the chemotaxis network.



The *E. coli* chemotaxis network is represented here in simplified form. T represents the Tar receptor, and L the ligand or attractant. The receptor can be modified by phosphorylation (subscript P) or methylation (subscripts 2 or 3).

(10) a. The Tar receptor binds an extracellular ligand L according to



(10) b. We now assume that the ligand binding reaction is in rapid equilibrium, and only consider total amounts of each modified form of receptor. For example,

$T_{3P}^{tot} = T_{3P} + LT_{3P}$ , etc. How would you calculate the effective rate constants  $k_a, \dots, k_d$  and  $k_{-a}, \dots, k_{-d}$  between these total concentration pools in terms of the rate constants of the original methylation /demethylation and phosphorylation/dephosphorylation reactions?

- (10) c. The assumed rates in the models of Spiro *et al.* and Barkai *et al.* are shown in the table below. Write down explicitly the effective rate constants for each model, in terms of the symbols listed in the table.

<i>Spiro model</i>			<i>Barkai model</i>		
	<i>L-unbound</i>	<i>L-bound</i>		<i>L-unbound</i>	<i>L-bound</i>
$k_a$	$k_8$	0	$k_a$	0	0
$k_{-a}$	$k_y$	$k_y$	$k_{-a}$	$k_0$	$k_0$
$k_b$	$3 k_8$	$1.1 k_8$	$k_b$	$k_{p1}$	$k_{p2}$
$k_{-b}$	$k_y$	$k_y$	$k_{-b}$	$k_{-p1}$	$k_{-p2}$
$k_c$	$k_1$	$k_3$	$k_c$	$k$	$k$
$k_{-c}$	$k_{-1}$	$k_{-1}$	$k_{-c}$	0	0
$k_d$	$k_1$	$k_3$	$k_d$	0	0
$k_{-d}$	$k_{-1}$	$k_{-1}$	$k_{-d}$	$k_m$	$k_m$

Note that some effective rate constants are now functions of L. This is appropriate, since we know for example that the receptors should become less phosphorylated as L increases.

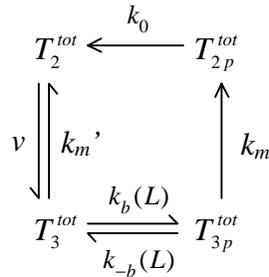
- i) In the Spiro model, do  $k_a/k_{-a}$  and  $k_b/k_{-b}$  increase or decrease with L?
- ii) In the Barkai model, we would like  $k_b$  to decrease and  $k_{-b}$  to increase with L. What does this imply about  $k_{p1}$ ,  $k_{p2}$ ,  $k_{-p1}$  and  $k_{-p2}$ ?
- (10) d. Spiro model. In steady state, after the slow methylation reactions have had time to equilibrate, let  $\alpha(L)$  represent the fraction of receptors that are methylated. Consider now the total concentration of phosphorylated and unphosphorylated receptors. Write out explicitly, in terms of  $\alpha(L)$ , the effective rates of phosphorylation ( $k_p$ ) and dephosphorylation ( $k_{-p}$ ) using

$$k_p = (1 - \alpha(L))k_a + \alpha(L)k_b, \quad k_{-p} = (1 - \alpha(L))k_{-a} + \alpha(L)k_{-b}$$

For perfect adaptation to be achieved, the phosphorylated fraction of receptor must be independent of L in steady state. You should have found above that  $k_p = k_y$ ; it is therefore sufficient for perfect adaptation that  $k_p = k_p^*$  is a constant.

Set  $k_8 = 15 \text{ s}^{-1}$ ;  $K_L = 1 \times 10^6 \text{ M}^{-1}$ ; and  $k_p^* = 15 \text{ s}^{-1}$ . Plot  $k_a$  and  $k_b$  for  $L = 0, \dots, 2 K_L$ . On the same graph, draw a horizontal line showing the desired  $k_p^*$ . Finally, set  $k_p = k_p^*$  in the equation above, solve for  $\alpha(L)$ , and plot this function. This is the magical form of  $\alpha(L)$  required for perfect adaptation. The model of Spiro *et al.* is carefully “tuned” in order to achieve this result. We can contrast this situation with the Barkai model in part f, which is perfectly adapting but requires no fine tuning.

- (10) e. Barkai model. Biochemical evidence suggests that the methylation reaction (whose rate constant was written as  $k$  in part c) operates at saturation with rate  $v$ . Show that under this assumption, the entire model reduces to the following reaction scheme:



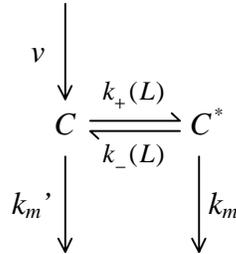
Note that  $v$  is a constant *rate* (measured in  $\text{M s}^{-1}$ ) while  $k_m$  is a rate constant (measured in  $\text{s}^{-1}$ ). What is the value of  $k_m'$  ?

Write down the equation for  $\frac{d(T_3^{tot} + T_{3p}^{tot})}{dt}$  and solve for  $T_{3p}^{tot}$  in steady state. Show

that this value is independent of  $L$  if and only if  $k_m' = 0$ . This is the essence of the Barkai model: perfect adaptation is easy to achieve, as long as only the *phosphorylated* receptors are demethylated by the CheB protein.

## 2. Adaptation and frequency response of the chemotaxis network.

With a slight change of notation, the Barkai model of the chemotaxis network (see Problem 1f) can be represented as



Here,  $v$  represents the rate of creation of  $C$ , the unphosphorylated receptor;  $C^*$  is the phosphorylated or active form of the receptor, the actual signal which induces bacterial tumbling;  $k_m$  and  $k_m'$  are the rate constants of demethylation reactions; and finally,  $k_+$  and  $k_-$  are rate constants that represent the effect of ligand binding on the phosphorylation state of the receptor.

Set  $\alpha_+ = \frac{\partial k_+}{\partial L} < 0$ , and  $\alpha_- = \frac{\partial k_-}{\partial L} > 0$ . This ensures that a sudden increase of ligand concentration causes a drop in the phosphorylated fraction of the receptor.

- (10) a. Write down the equations for  $dC/dt$  and  $dC^*/dt$ . Solve for the steady state concentrations  $C_{ss}$  and  $C^*_{ss}$ . Under what conditions will  $C^*_{ss}$  be independent of  $L$ ?
- (10) b. Set  $\delta C = C - C_{ss}$ ,  $\delta C^* = C^* - C^*_{ss}$ . Derive the linearized equations representing fluctuations from steady state, driven by fluctuations  $\delta L(t)$  of the ligand concentration. You should obtain

$$\frac{d}{dt} \begin{bmatrix} \delta C \\ \delta C^* \end{bmatrix} = \begin{bmatrix} -(k_+ + k_m') & +k_- \\ +k_+ & -(k_- + k_m) \end{bmatrix} \begin{bmatrix} \delta C \\ \delta C^* \end{bmatrix} + (\alpha_- C^*_{ss} - \alpha_+ C_{ss}) \delta L \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

- (10) c. Assume for now that  $\delta L = 0$ ,  $k_m' = 0$ , and  $k_m = 0$ . Calculate the eigenvectors and eigenvalues of the above matrix. You will find that one of the eigenvalues is zero. Recalculate this eigenvalue to first order in  $k_m$ .

On a graph of  $\delta C$  vs.  $\delta C^*$ , plot the eigenvectors and note the slow and fast eigenvalues. Sketch a few typical timecourses for various initial values of  $\{\delta C, \delta C^*\}$ . This initial perturbation might arise if the system had first reached steady state for one value of  $L$ , but that value was abruptly changed. Sketch out such an event, showing a step increase in  $L$  at time  $t = 0$ , and the subsequent evolution of  $C$ ,  $C^*$ , and  $C_T$  as functions of time.

- (10) d. Now assume that  $\delta L, \delta C, \delta C^* \sim e^{i\omega t}$ . This corresponds to Fourier transforming the equation above.

Calculate the *transfer function*  $T(\omega) = \left| \frac{\delta C^*(\omega)}{\delta L(\omega)} \right|$ .

Claiming that perfect adaptation holds corresponds to claiming that  $T(\omega)$  has no dc component ( $T(\omega=0) = 0$ ). Show that this is true only if  $k_m' = 0$ . Assume from now on that  $k_m' = 0$ .

- (i) What is the behavior of  $T(\omega)$  as  $\omega \rightarrow 0$ ?
  - (ii) What is the behavior of  $T(\omega)$  as  $\omega \rightarrow \infty$ ?
  - (iii) Calculate the value  $\omega^*$  at which  $T(\omega)$  is maximized.
  - (iv) Make a sketch of  $T(\omega)$ , indicating all the important regimes.
- (10) e. From this sketch, it should be clear that the chemotaxis network serves as a bandpass filter: variations of L slower than the demethylation rate  $k_m$  are suppressed by the adaptation property of the network; fast fluctuations of L are suppressed because  $C^*$  cannot respond any faster than the phosphorylation rate.
- (i) Suppose  $f_{out}(t) = df_{in}(t)/dt$ . Calculate  $T_{diff}(\omega) = |f_{out}(\omega)/f_{in}(\omega)|$ . This is the transfer function of a differentiator. For what values of  $\omega$  does the chemotaxis network serve as a differentiator?
  - (ii) The network most efficiently transmits signals at the frequency  $\omega^*$  calculated in part d(iii). What is the value of  $\omega^*$ , assuming  $k_m \sim 0.01 \text{ s}^{-1}$  and  $k_+ \sim 10 \text{ s}^{-1}$ ?
  - (iii) It is said that “a cell compares the attractant concentration at any given time to that 4 seconds ago”, generating a tumble if it registers a decrease or a run if it registers an increase. That is, only by *differentiating* the input does the cell manage to swim up an attractant gradient. Is the timescale of 4 seconds consistent with your answer from the part (ii)?