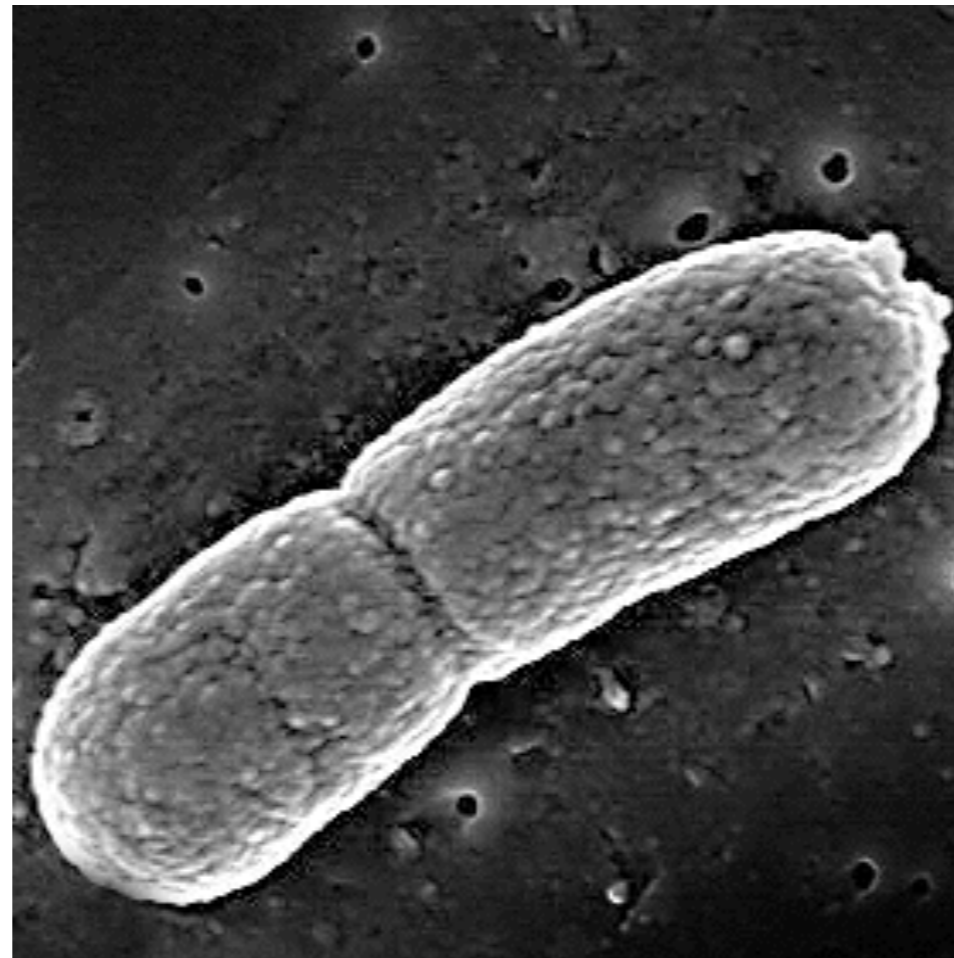
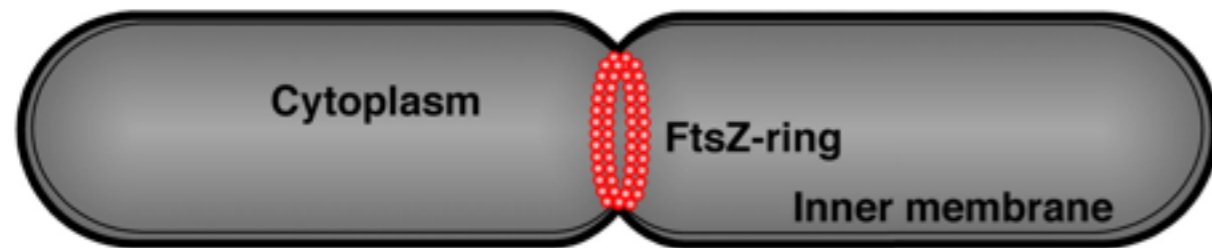


MAE 545: Lecture 10 (10/20)

How bacteria find midline for cell division?



Contraction of FtsZ-ring divides bacterial cell in two



**FtsZ is analogous to tubulin
(assembly by GTP hydrolysis)**

**Bacterial division is extremely
precise. FtsZ forms at**

$$(0.50 \pm 0.01) L$$



1 μm

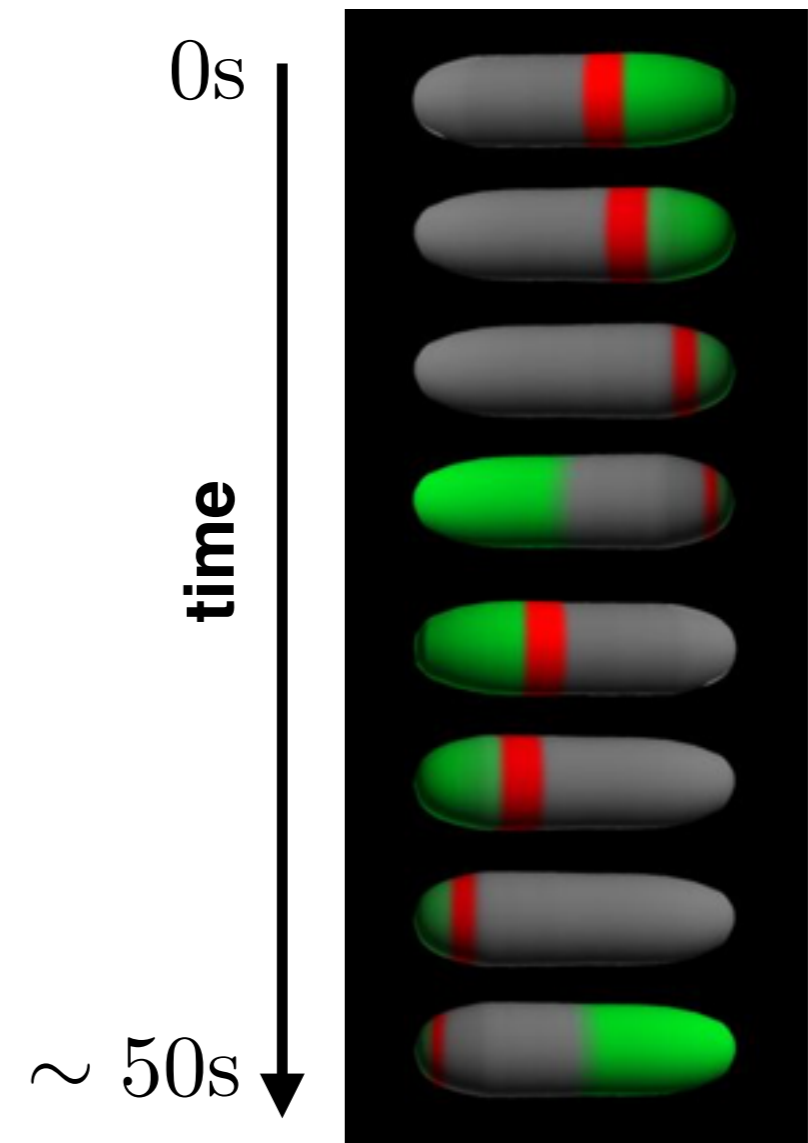
**How does bacteria know where
to place the contractile ring?**

Min system oscillations provide cues for the formation of FtsZ ring



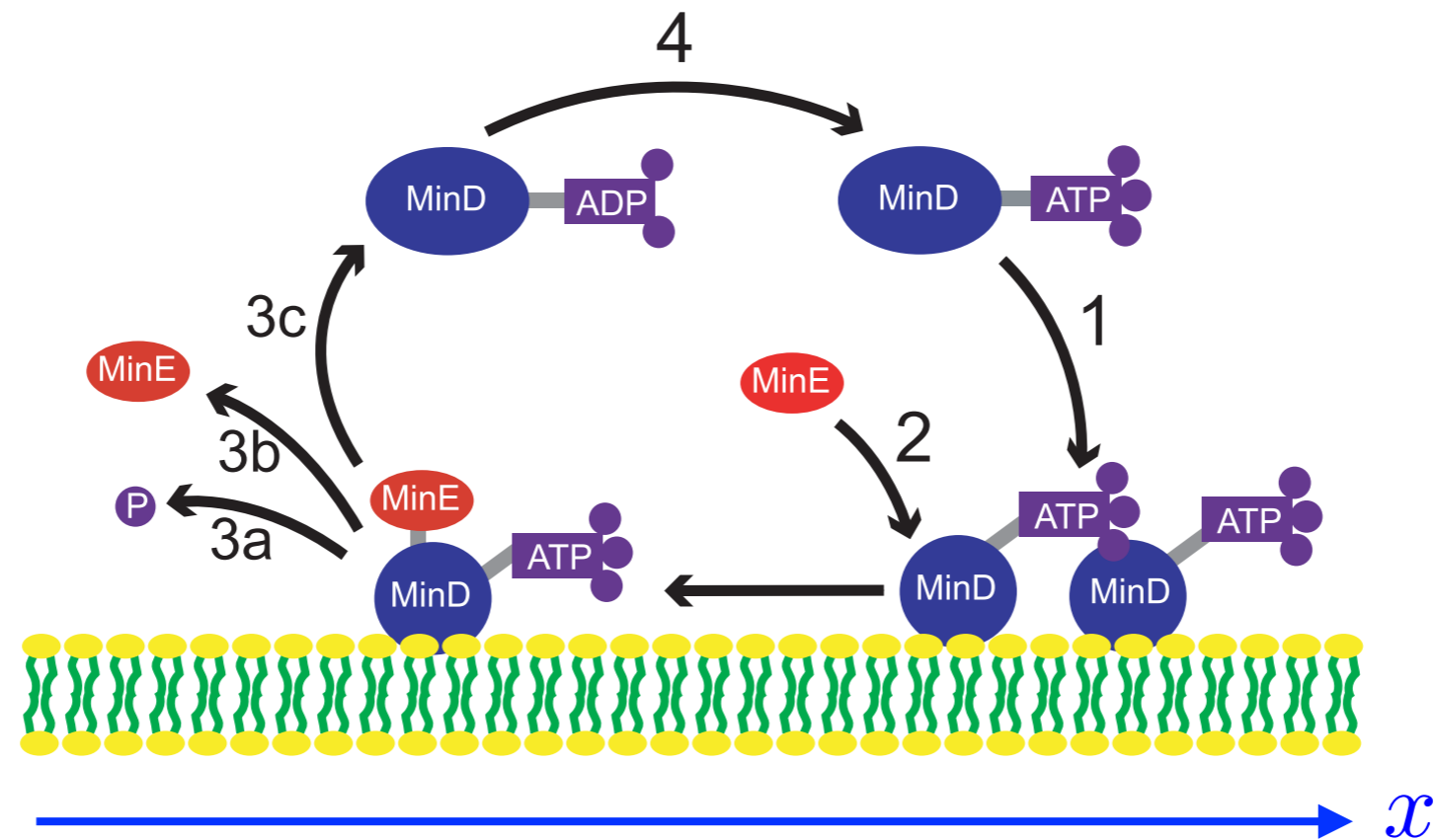
Predator-prey like dynamics between MinD and MinE proteins produce oscillations on a minute time scale, which is much shorter than typical division time (~20 min).

On average MinC/MinD proteins are depleted near the cell center, where FtsZ ring forms!



H. Meinhardt and P.A.J. de Boer,
PNAS 98, 14202 (2001)

1D model of Min system



Concentrations of proteins in the cytoplasm

$$C_{D:ADP}(x, t) \quad C_{D:ATP}(x, t) \quad C_E(x, t)$$

Concentrations of membrane bound proteins

$$C_d(x, t) \quad C_{de}(x, t)$$

Note: in paper they treat the full 3D model

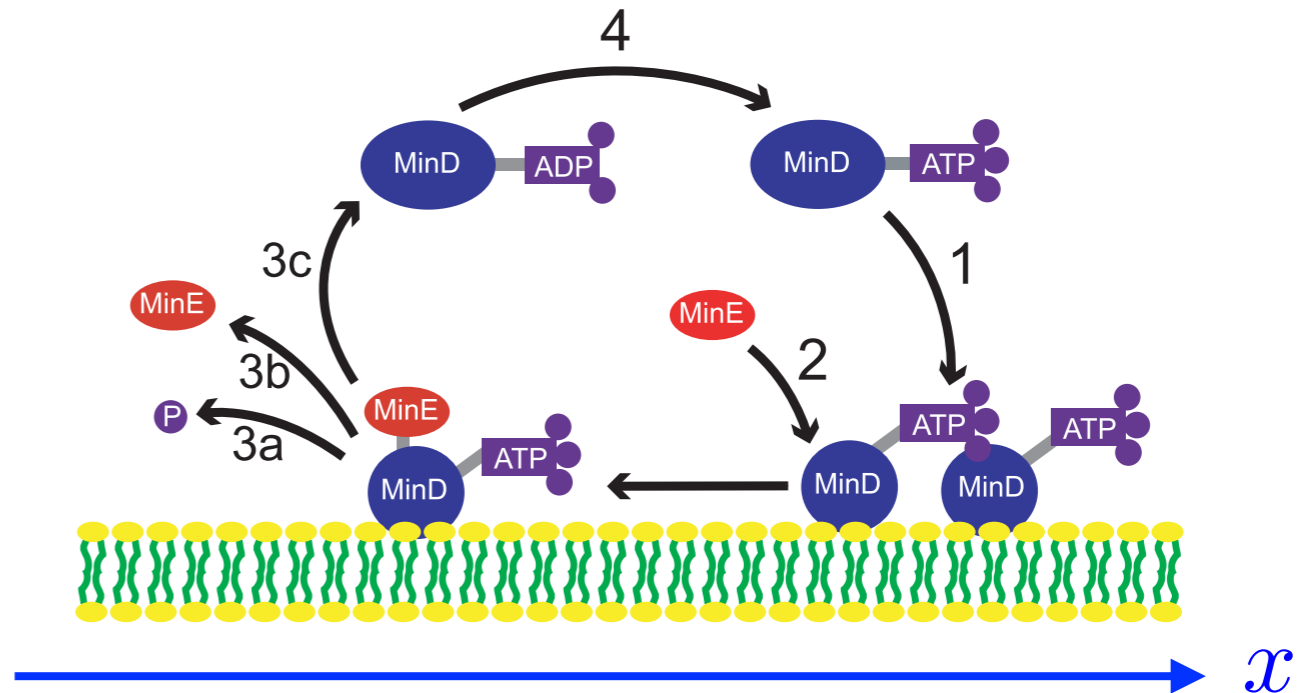
K.C. Huang *et al.*, PNAS 100, 12724 (2003)

Diffusion of Min proteins

Diffusion of Min proteins in cytoplasm

$$D_D \approx D_E \approx 2.5 \mu\text{m}^2/\text{s}$$

Diffusion of membrane bound Min proteins is negligible!



$$\frac{\partial C_{D:ADP}}{\partial t} = D_D \frac{\partial^2 C_{D:ADP}}{\partial x^2}$$

$$\frac{\partial C_{D:ATP}}{\partial t} = D_D \frac{\partial^2 C_{D:ATP}}{\partial x^2}$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2}$$

$$\frac{\partial C_d}{\partial t} =$$

$$\frac{\partial C_{de}}{\partial t} =$$

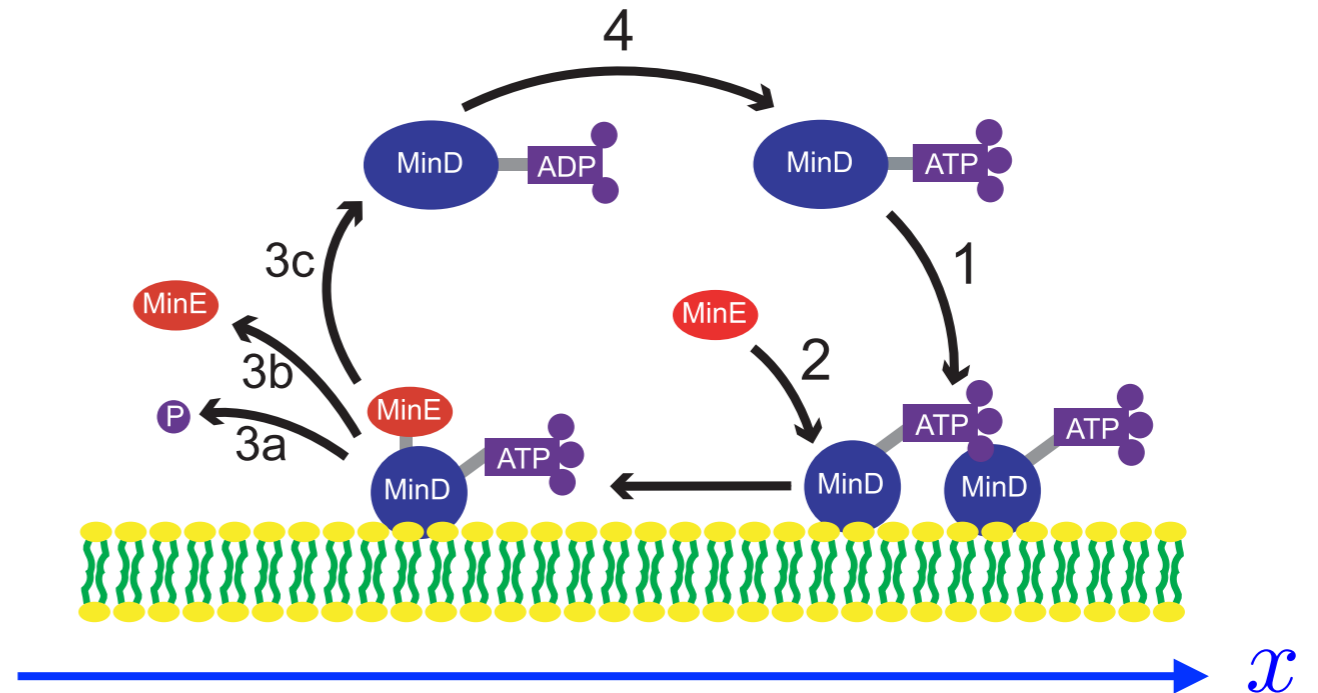
Reaction 1

MinD-ATP binds to the membrane

$$\sigma_D \approx 0.1 \text{s}^{-1}$$

MinD-ATP in the membrane recruits more MinD-ATP proteins

$$\sigma_{dD} \approx 0.002 \mu\text{m s}^{-1}$$



$$\frac{\partial C_{D:ADP}}{\partial t} = D_D \frac{\partial^2 C_{D:ADP}}{\partial x^2}$$

$$\frac{\partial C_{D:ATP}}{\partial t} = D_D \frac{\partial^2 C_{D:ATP}}{\partial x^2} - \sigma_D C_{D:ATP} - \sigma_{dD} C_{D:ATP} [C_d + C_{de}]$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2}$$

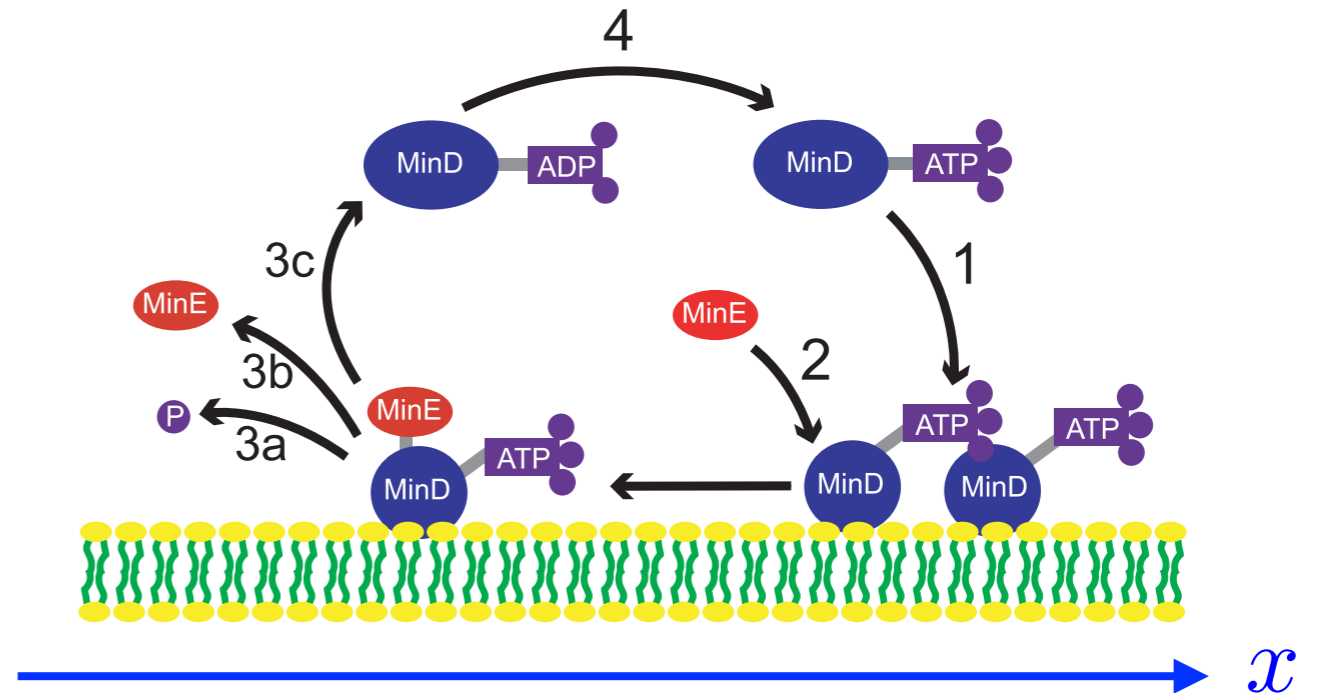
$$\frac{\partial C_d}{\partial t} = \sigma_D C_{D:ATP} + \sigma_{dD} C_{D:ATP} [C_d + C_{de}]$$

$$\frac{\partial C_{de}}{\partial t} =$$

Reaction 2

**MinE binds to MinD-ATP
in the membrane**

$$\sigma_E \approx 0.12 \mu\text{m s}^{-1}$$



$$\frac{\partial C_{D:ADP}}{\partial t} = D_D \frac{\partial^2 C_{D:ADP}}{\partial x^2}$$

$$\frac{\partial C_{D:ATP}}{\partial t} = D_D \frac{\partial^2 C_{D:ATP}}{\partial x^2} - \sigma_D C_{D:ATP} - \sigma_{dD} C_{D:ATP} [C_d + C_{de}]$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2} - \sigma_E C_d C_E$$

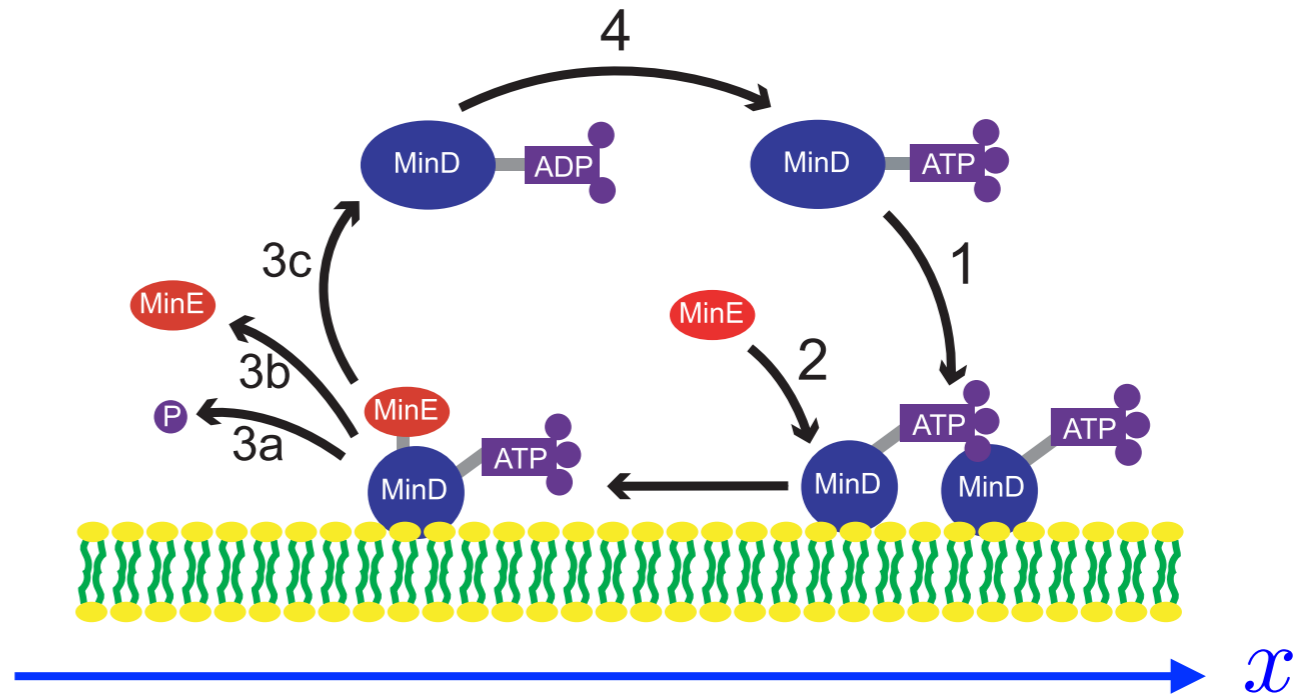
$$\frac{\partial C_d}{\partial t} = \sigma_D C_{D:ATP} + \sigma_{dD} C_{D:ATP} [C_d + C_{de}] - \sigma_E C_d C_E$$

$$\frac{\partial C_{de}}{\partial t} = \sigma_E C_d C_E$$

Reaction 3

MinE increases the rate of ATP hydrolysis and afterwards both MinE and MinD fall off the membrane

$$\sigma_{de} \approx 0.7s^{-1}$$



$$\frac{\partial C_{D:ADP}}{\partial t} = D_D \frac{\partial^2 C_{D:ADP}}{\partial x^2} + \sigma_{de} C_{de}$$

$$\frac{\partial C_{D:ATP}}{\partial t} = D_D \frac{\partial^2 C_{D:ATP}}{\partial x^2} - \sigma_D C_{D:ATP} - \sigma_{dD} C_{D:ATP} [C_d + C_{de}]$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2} - \sigma_E C_d C_E + \sigma_{de} C_{de}$$

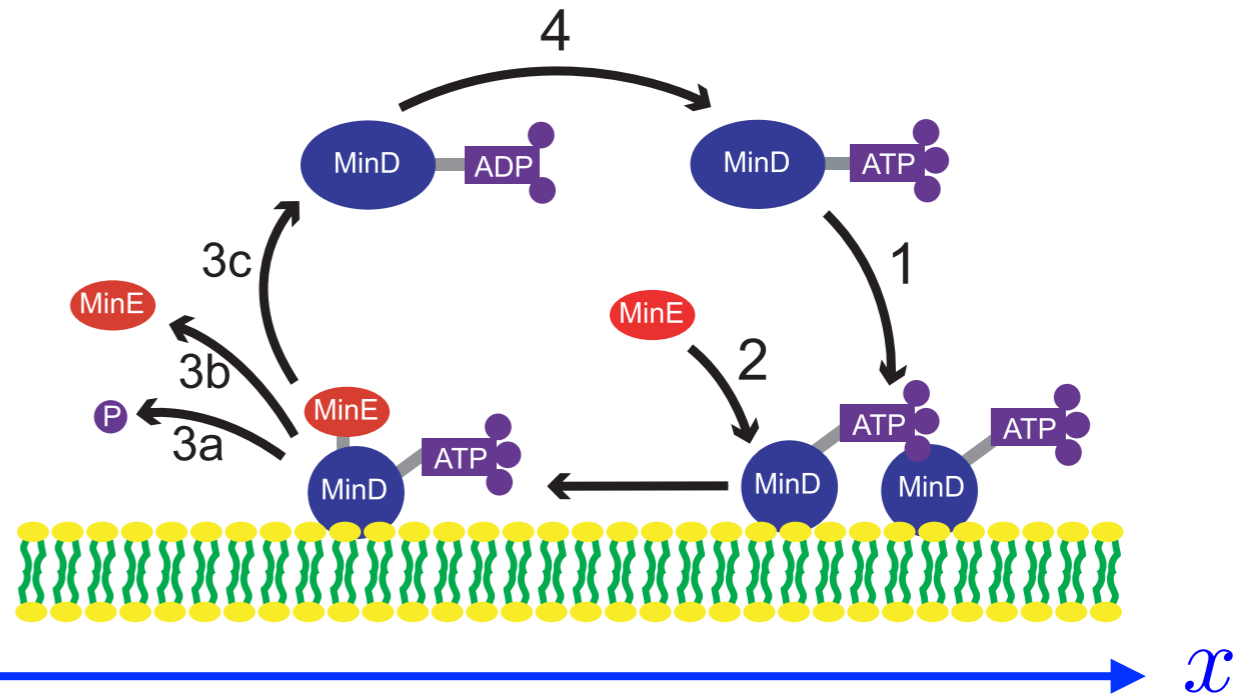
$$\frac{\partial C_d}{\partial t} = \sigma_D C_{D:ATP} + \sigma_{dD} C_{D:ATP} [C_d + C_{de}] - \sigma_E C_d C_E$$

$$\frac{\partial C_{de}}{\partial t} = \sigma_E C_d C_E - \sigma_{de} C_{de}$$

Reaction 4

Replacement of ADP with ATP for MinD proteins in the cytoplasm

$$\sigma_D^{\text{ADP} \rightarrow \text{ATP}} \sim 1\text{s}^{-1}$$



$$\frac{\partial C_{\text{D:ADP}}}{\partial t} = D_D \frac{\partial^2 C_{\text{D:ADP}}}{\partial x^2} + \sigma_{de} C_{de} - \sigma_D^{\text{ADP} \rightarrow \text{ATP}} C_{\text{D:ADP}}$$

$$\frac{\partial C_{\text{D:ATP}}}{\partial t} = D_D \frac{\partial^2 C_{\text{D:ATP}}}{\partial x^2} - \sigma_D C_{\text{D:ATP}} - \sigma_{dD} C_{\text{D:ATP}} [C_d + C_{de}] + \sigma_D^{\text{ADP} \rightarrow \text{ATP}} C_{\text{D:ADP}}$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2} - \sigma_E C_d C_E + \sigma_{de} C_{de}$$

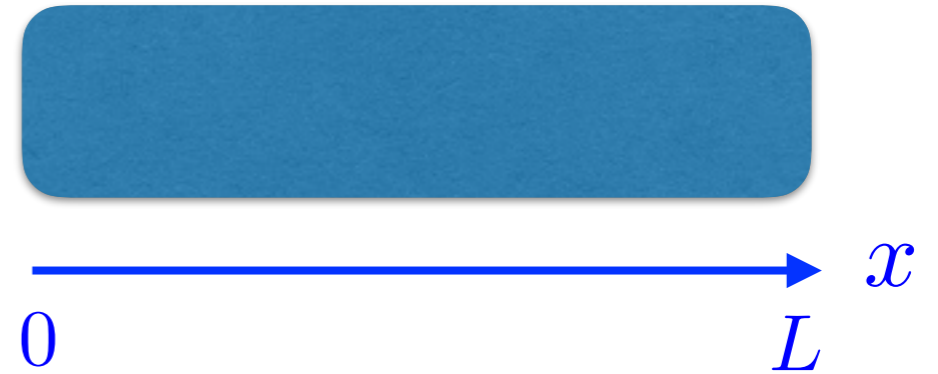
$$\frac{\partial C_d}{\partial t} = \sigma_D C_{\text{D:ATP}} + \sigma_{dD} C_{\text{D:ATP}} [C_d + C_{de}] - \sigma_E C_d C_E$$

$$\frac{\partial C_{de}}{\partial t} = \sigma_E C_d C_E - \sigma_{de} C_{de}$$

Boundary conditions

No flux of proteins through the edge

$$\begin{aligned}\frac{\partial C_{D:ADP}}{\partial x}(x=0, t) &= \frac{\partial C_{D:ADP}}{\partial x}(x=L, t) = 0 \\ \frac{\partial C_{D:ATP}}{\partial x}(x=0, t) &= \frac{\partial C_{D:ATP}}{\partial x}(x=L, t) = 0 \\ \frac{\partial C_E}{\partial x}(x=0, t) &= \frac{\partial C_E}{\partial x}(x=L, t) = 0\end{aligned}$$



$$\frac{\partial C_{D:ADP}}{\partial t} = D_D \frac{\partial^2 C_{D:ADP}}{\partial x^2} + \sigma_{de} C_{de} - \sigma_D^{\text{ADP} \rightarrow \text{ATP}} C_{D:ADP}$$

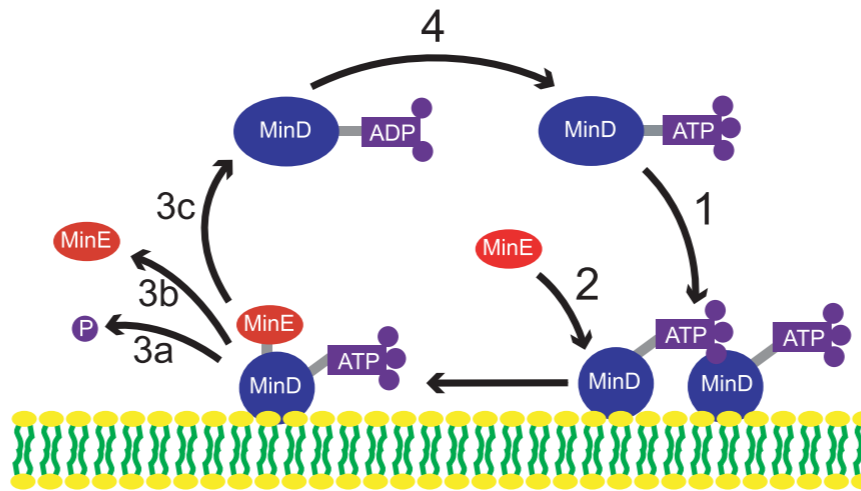
$$\frac{\partial C_{D:ATP}}{\partial t} = D_D \frac{\partial^2 C_{D:ATP}}{\partial x^2} - \sigma_D C_{D:ATP} - \sigma_{dD} C_{D:ATP} [C_d + C_{de}] + \sigma_D^{\text{ADP} \rightarrow \text{ATP}} C_{D:ADP}$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2} - \sigma_E C_d C_E + \sigma_{de} C_{de}$$

$$\frac{\partial C_d}{\partial t} = \sigma_D C_{D:ATP} + \sigma_{dD} C_{D:ATP} [C_d + C_{de}] - \sigma_E C_d C_E$$

$$\frac{\partial C_{de}}{\partial t} = \sigma_E C_d C_E - \sigma_{de} C_{de}$$

How do we analyze such system of PDEs?



$$\frac{\partial C_{D:ADP}}{\partial t} = D_D \frac{\partial^2 C_{D:ADP}}{\partial x^2} + \sigma_{de} C_{de} - \sigma_D^{\text{ADP} \rightarrow \text{ATP}} C_{D:ADP}$$

$$\frac{\partial C_{D:ATP}}{\partial t} = D_D \frac{\partial^2 C_{D:ATP}}{\partial x^2} - \sigma_D C_{D:ATP} - \sigma_{dD} C_{D:ATP} [C_d + C_{de}] + \sigma_D^{\text{ADP} \rightarrow \text{ATP}} C_{D:ADP}$$

$$\frac{\partial C_E}{\partial t} = D_E \frac{\partial^2 C_E}{\partial x^2} - \sigma_E C_d C_E + \sigma_{de} C_{de}$$

$$\frac{\partial C_d}{\partial t} = \sigma_D C_{D:ATP} + \sigma_{dD} C_{D:ATP} [C_d + C_{de}] - \sigma_E C_d C_E$$

$$\frac{\partial C_{de}}{\partial t} = \sigma_E C_d C_E - \sigma_{de} C_{de}$$

$$\frac{\partial C_i(x, t)}{\partial t} = F_i(\{C_j(x, t)\}) + D_i \frac{\partial^2 C_i(x, t)}{\partial x^2}$$

Stable fixed point for uniformly distributed concentrations

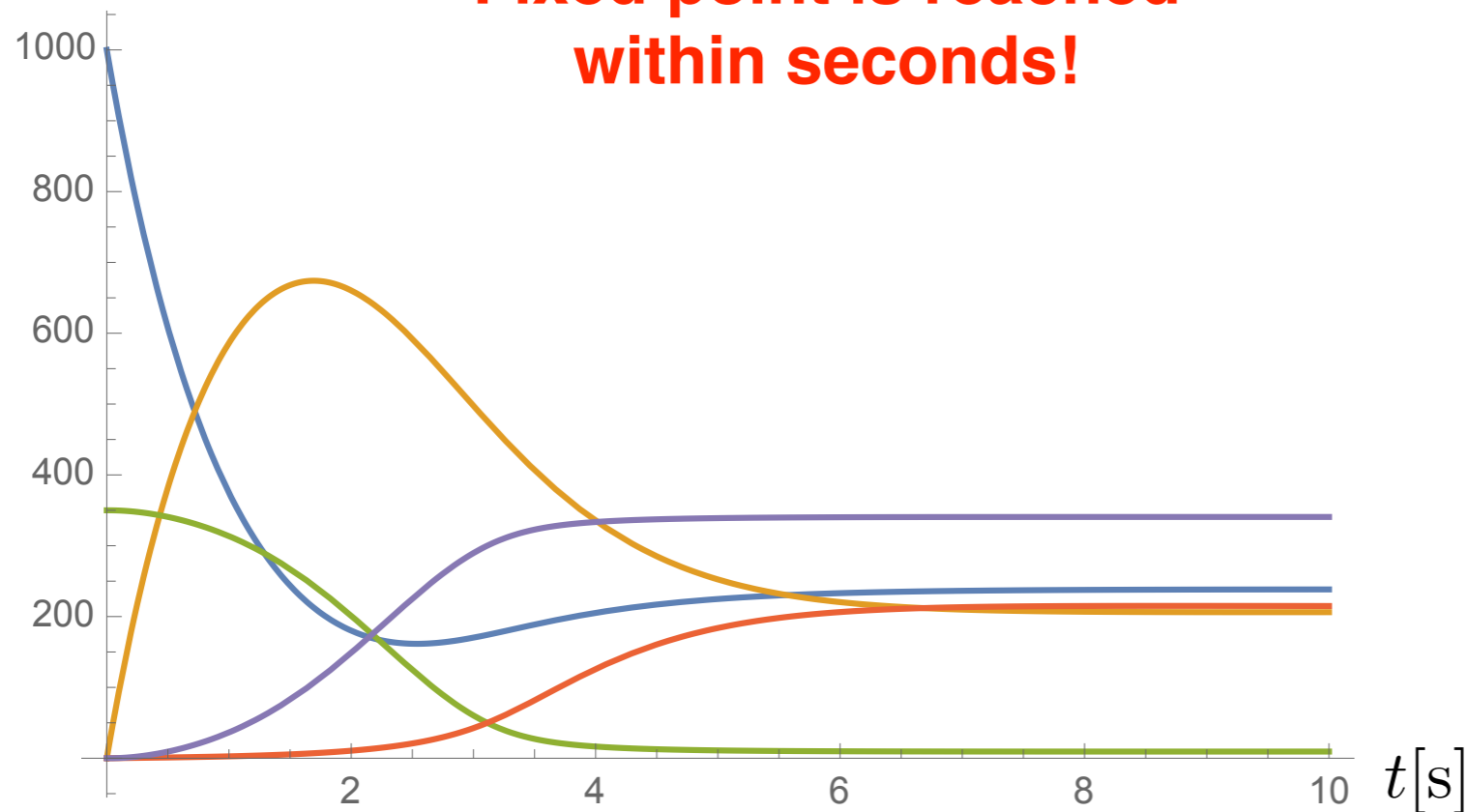
$$\frac{\partial C_i(x, t)}{\partial t} = F_i(\{C_j(x, t)\}) + D_i \frac{\partial^2 C_i(x, t)}{\partial x^2}$$

First let us assume that concentration profiles are independent of x and find stable fixed point concentrations.

$$F_i(\{C_j^*\}) = 0$$

$[\mu\text{m}^{-1}]$

Fixed point is reached within seconds!



$$C_{\text{D:ADP}}^* \approx 240 \mu\text{m}^{-1}$$

$$C_{\text{D:ATP}}^* \approx 210 \mu\text{m}^{-1}$$

$$C_{\text{E}}^* \approx 10 \mu\text{m}^{-1}$$

$$C_{\text{d}}^* \approx 210 \mu\text{m}^{-1}$$

$$C_{\text{de}}^* \approx 340 \mu\text{m}^{-1}$$

Linear stability analysis of fixed point

$$\frac{\partial C_i(x, t)}{\partial t} = F_i(\{C_j(x, t)\}) + D_i \frac{\partial^2 C_i(x, t)}{\partial x^2}$$

Let's assume small perturbations around the fixed point

$$c_i(x, t) = C_i(x, t) - C_i^*$$

and linearize the PDE

$$\frac{\partial c_i(x, t)}{\partial t} = \sum_j M_{ij}^0 c_j(x, t) + D_i \frac{\partial^2 c_i(x, t)}{\partial x^2}$$

$$M_{ij}^0 = \left. \frac{\partial F_i}{\partial C_j} \right|_{C^*} = \begin{pmatrix} -\sigma_D^{\text{ADP} \rightarrow \text{ATP}}, & 0, & 0, & 0, & \sigma_{de} \\ +\sigma_D^{\text{ADP} \rightarrow \text{ATP}}, & -\sigma_D - \sigma_{dD}(C_d^* + C_{de}^*), & 0, & -\sigma_{dD}C_{D:\text{ATP}}^*, & -\sigma_{dD}C_{D:\text{ATP}}^* \\ 0, & 0, & -\sigma_E C_d^*, & -\sigma_E C_E^*, & \sigma_{de} \\ 0, & \sigma_D + \sigma_{dD}(C_d^* + C_{de}^*), & -\sigma_E C_d^*, & \sigma_{dD}C_{D:\text{ATP}}^* - \sigma_E C_E^*, & \sigma_{dD}C_{D:\text{ATP}}^* \\ 0, & 0, & \sigma_E C_d^*, & \sigma_E C_E^*, & -\sigma_{de} \end{pmatrix}$$

Linear stability analysis of fixed point

$$\frac{\partial c_i(x, t)}{\partial t} = \sum_j M_{ij}^0 c_j(x, t) + D_i \frac{\partial^2 c_i(x, t)}{\partial x^2}$$

It is convenient to analyze this PDE in Fourier space, but note that only $\cos(kx)$ modes are consistent with boundary conditions.

Boundary conditions also restrict the values for wavenumber k

$$\frac{\partial c_i}{\partial x}(x=0, t) = \frac{\partial c_i}{\partial x}(x=L, t) = 0 \longrightarrow k = \frac{n\pi}{L}, \quad n = 0, 1, 2, \dots$$

Let's rewrite the PDE in Fourier space

$$c_i(x, t) = \sum_k \tilde{c}_i(k, t) \cos(kx)$$

$$\frac{\partial \tilde{c}_i(k, t)}{\partial t} = \sum_j M_{ij}^0 \tilde{c}_j(k, t) - D_i k^2 \tilde{c}_i(k, t) = \sum_j M_{ij}(k) \tilde{c}_j(k, t)$$

Note: in higher dimensions use solutions of Helmholtz equation with appropriate boundary conditions $u(\vec{r}) = -k^2 \nabla^2 u(\vec{r})$

Linear stability analysis of fixed point

$$\frac{\partial \tilde{c}_i(k, t)}{\partial t} = \sum_j M_{ij}^0 \tilde{c}_j(k, t) - D_i k^2 \tilde{c}_i(k, t) = \sum_j M_{ij}(k) \tilde{c}_j(k, t)$$

From linear algebra we know that the solution of this equation can be expressed in terms of eigenvalues and eigenvectors of matrix $M_{ij}(k)$:

$$\tilde{c}_i(k, t) = \sum_{\alpha} A_{\alpha}(k) v_i^{\alpha}(k) e^{\lambda_{\alpha}(k)t} \quad \lambda_{\alpha} v_i^{\alpha} = \sum_j M_{ij}(k) v_j^{\alpha}$$

Thus small perturbations from fixed point evolve as

$$c_i(x, t) = \sum_{\alpha, k} A_{\alpha}(k) v_i^{\alpha}(k) e^{\lambda_{\alpha}(k)t} \cos(kx)$$

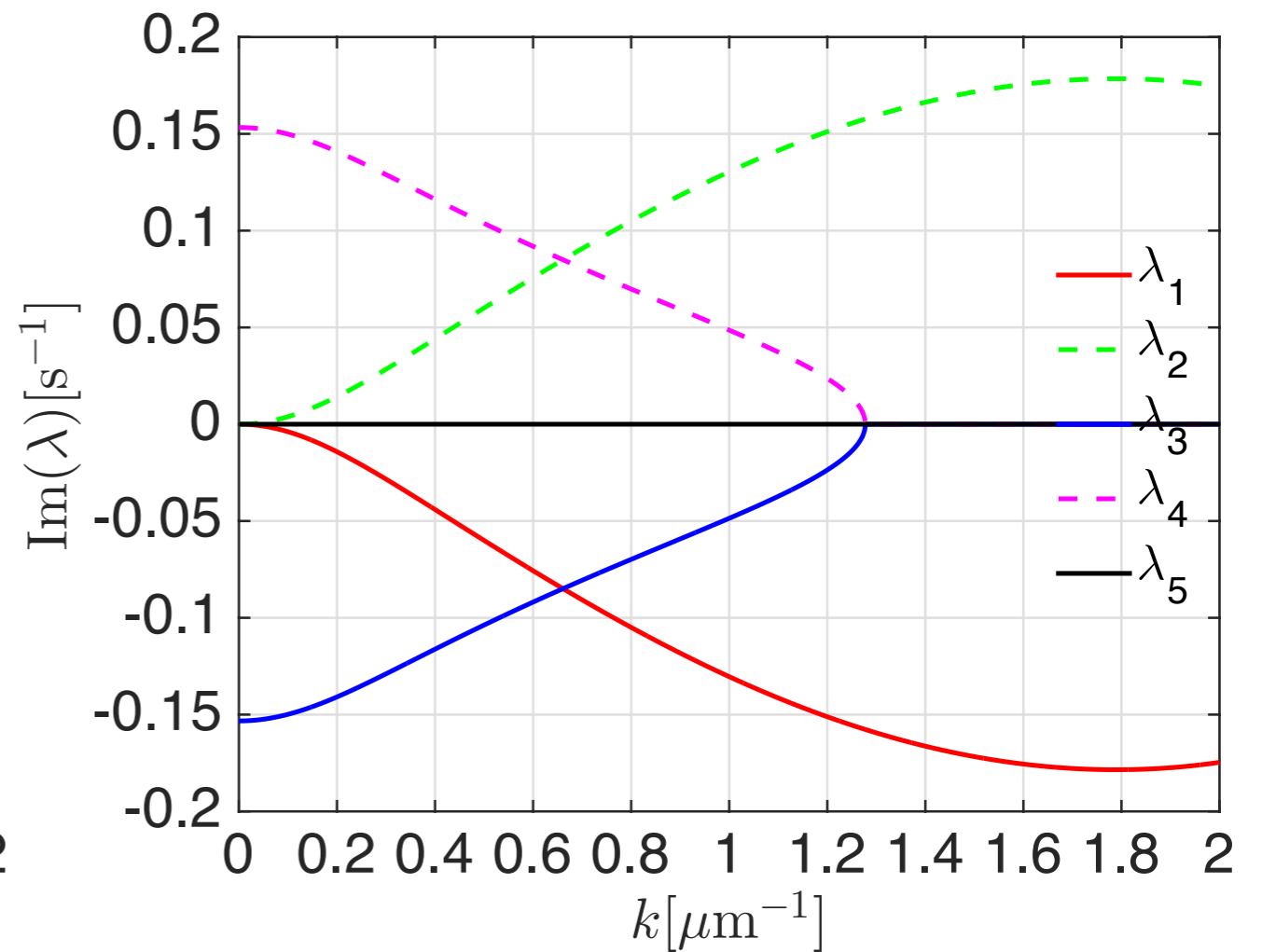
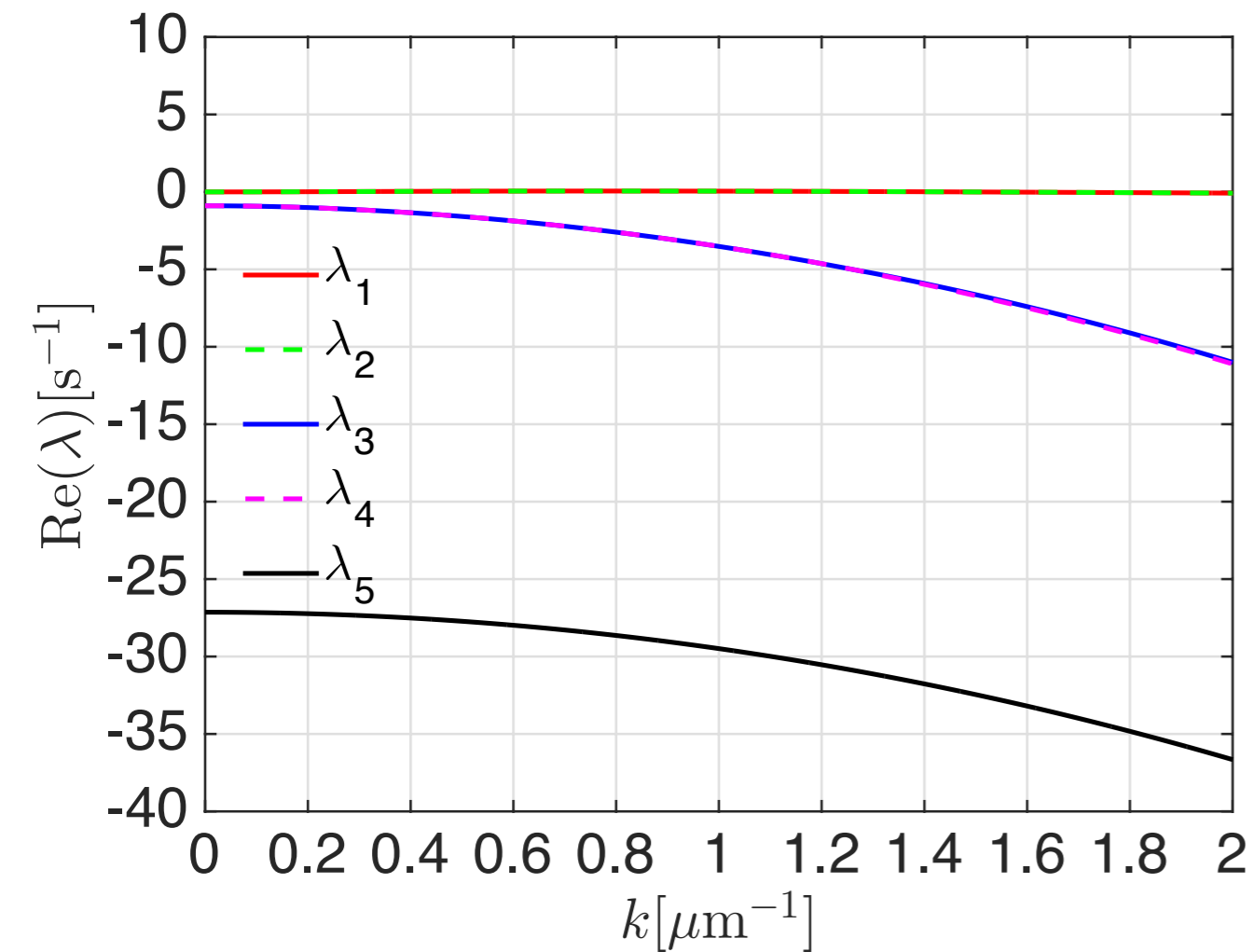
Fixed point is stable if and only if all eigenvalues have negative real parts for all allowed wavenumbers k !

$$\text{Re} [\lambda_{\alpha}(k)] < 0$$

For unstable fixed points the mode that corresponds to the eigenvalue with the largest real part dominates!

Eigenvalues in the model Min system

$$M_{ij}(k) = \begin{pmatrix} -\sigma_D^{\text{ADP} \rightarrow \text{ATP}} - D_D k^2, & 0, & 0, & 0, & \sigma_{de} \\ +\sigma_D^{\text{ADP} \rightarrow \text{ATP}}, & -\sigma_D - \sigma_{dD}(C_d^* + C_{de}^*) - D_D k^2, & 0, & -\sigma_{dD} C_{D:\text{ATP}}^*, & -\sigma_{dD} C_{D:\text{ATP}}^* \\ 0, & 0, & -\sigma_E C_d^* - D_E k^2, & -\sigma_E C_E^*, & \sigma_{de} \\ 0, & \sigma_D + \sigma_{dD}(C_d^* + C_{de}^*), & -\sigma_E C_d^*, & \sigma_{dD} C_{D:\text{ATP}}^* - \sigma_E C_E^*, & \sigma_{dD} C_{D:\text{ATP}}^* \\ 0, & 0, & \sigma_E C_d^*, & \sigma_E C_E^*, & -\sigma_{de} \end{pmatrix}$$

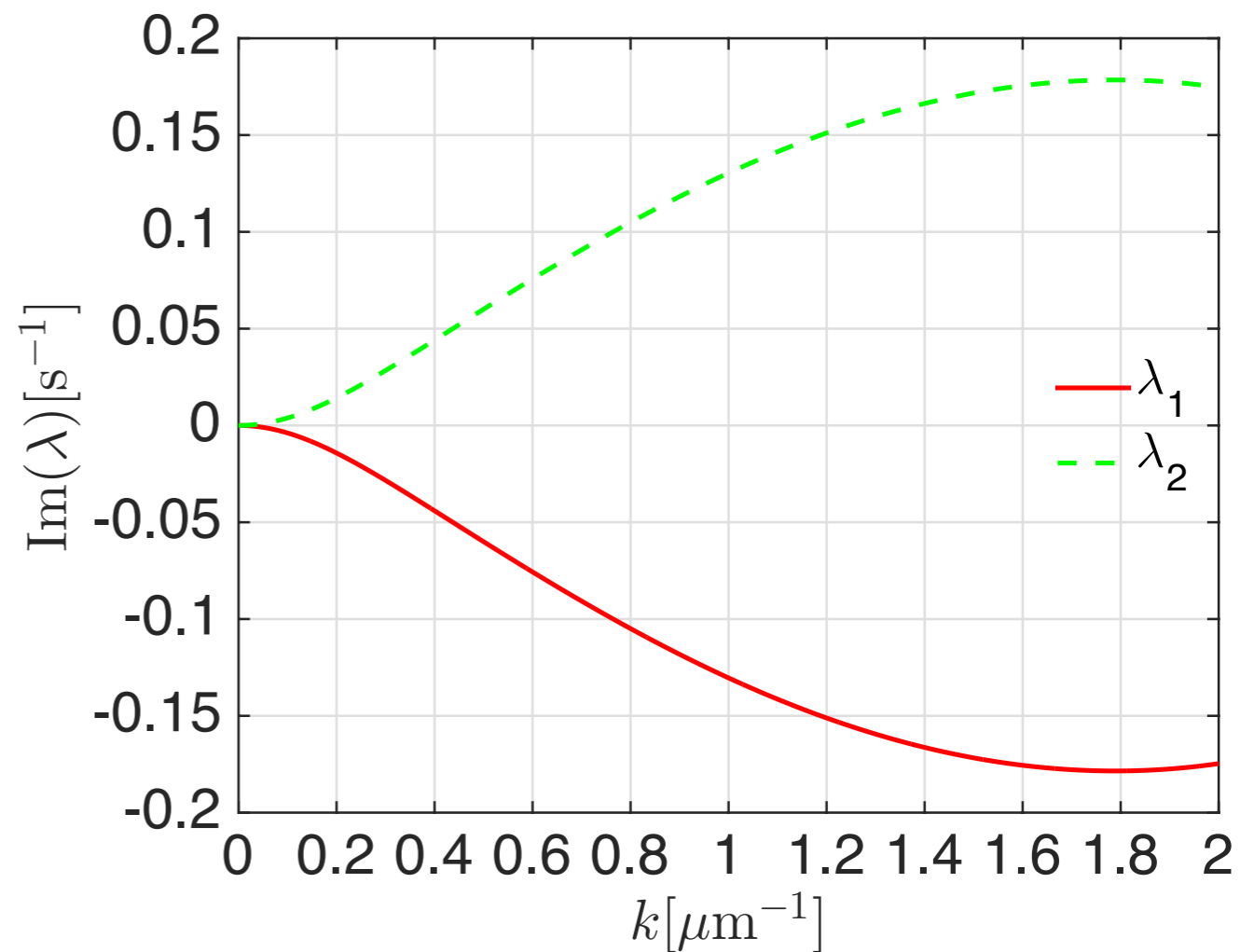
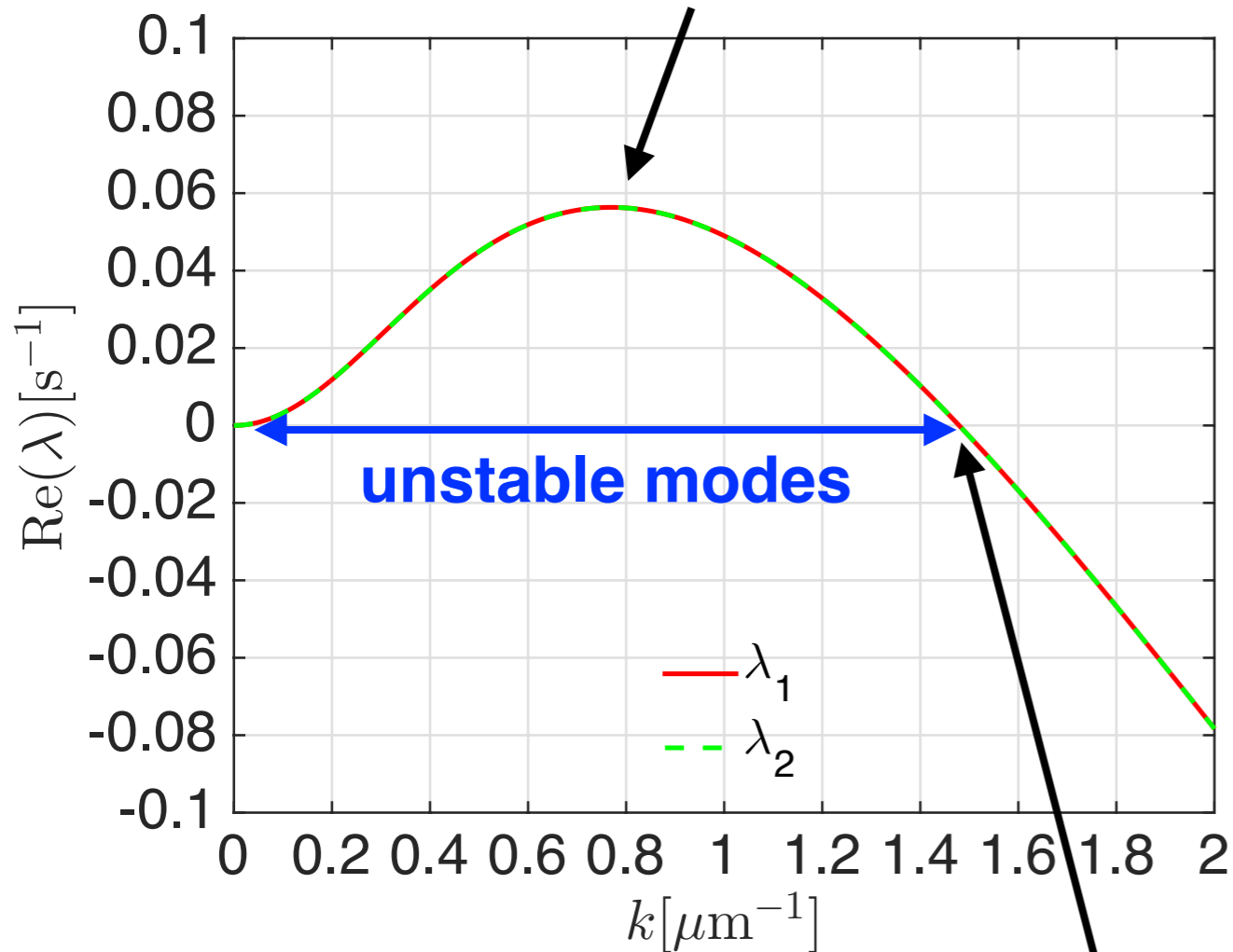


Eigenvalues in the model Min system

wavelength of the
most unstable mode

$$k_{\max} \approx 0.8 \mu\text{m}^{-1}$$

$$2\pi/k_{\max} \approx 8.2 \mu\text{m}^{-1}$$



Note that only discrete set
of wavenumber is allowed!

$$k = \frac{n\pi}{L}, \quad n = 0, 1, 2, \dots$$

For bacteria that is shorter than

$$L < (\pi/1.5)\mu\text{m} \approx 2.1\mu\text{m}$$

fixed point is stable and there
are no oscillations!

For E.coli with $L \approx 4\mu\text{m}$

$$k = \pi/L \approx 0.8\mu\text{m}^{-1}$$

$$\lambda_{1,2} \approx (0.06 \pm i0.10)\text{s}^{-1}$$

Period of oscillations

$$(2\pi/0.010)\text{s} \approx 60\text{s}$$

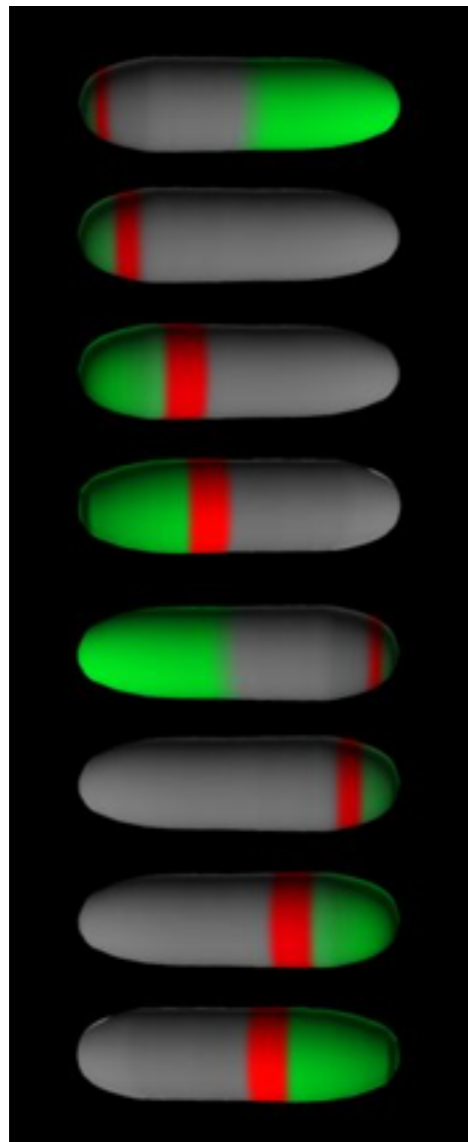
Oscillations in the model Min system

Tot MinD

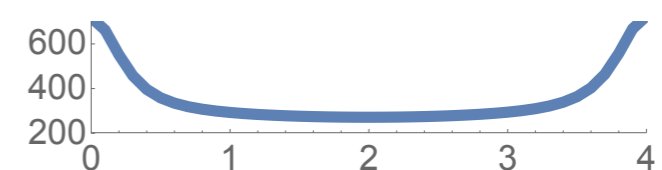
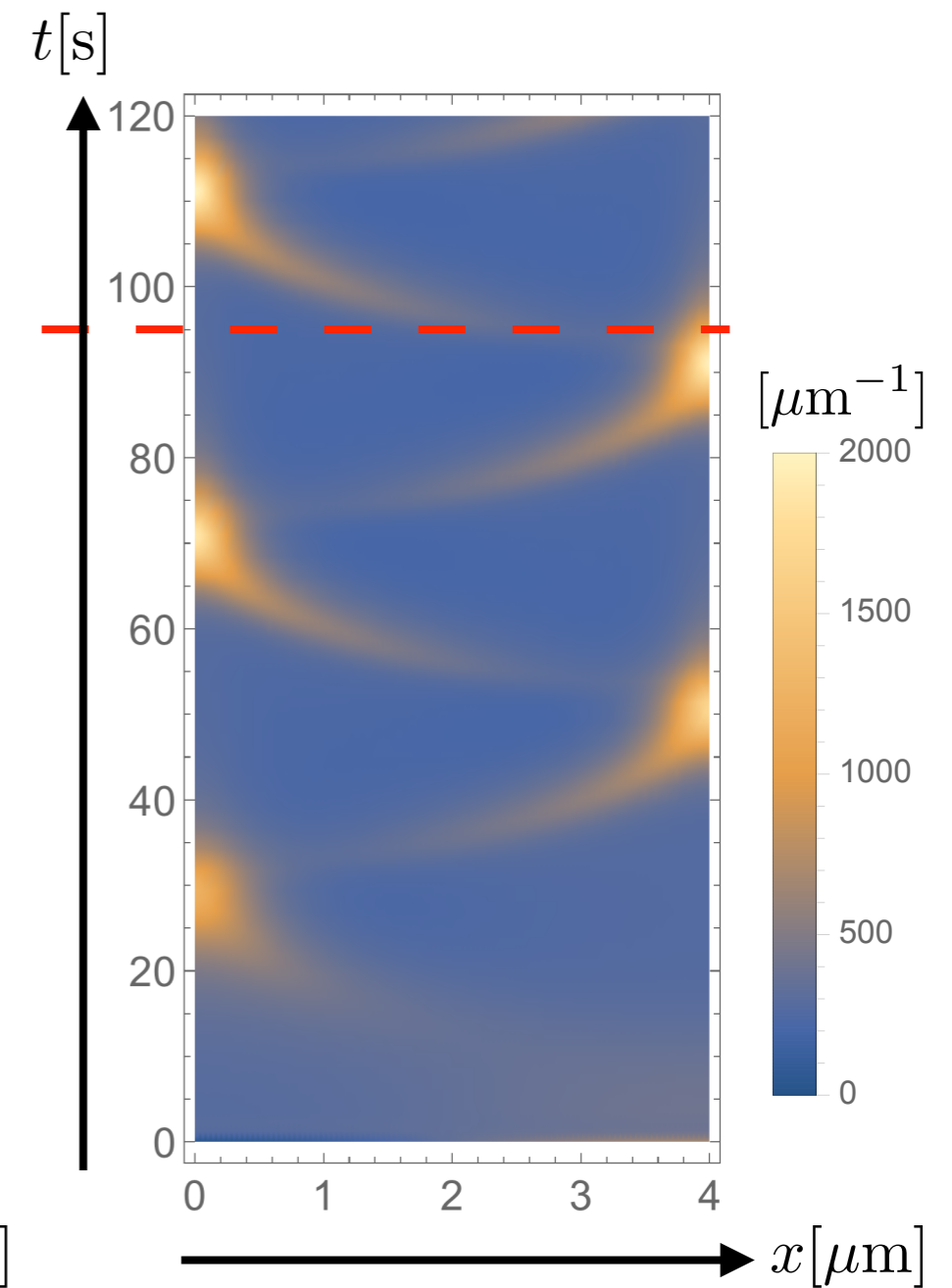
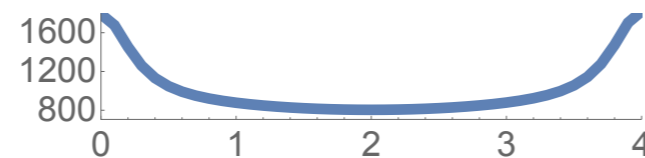
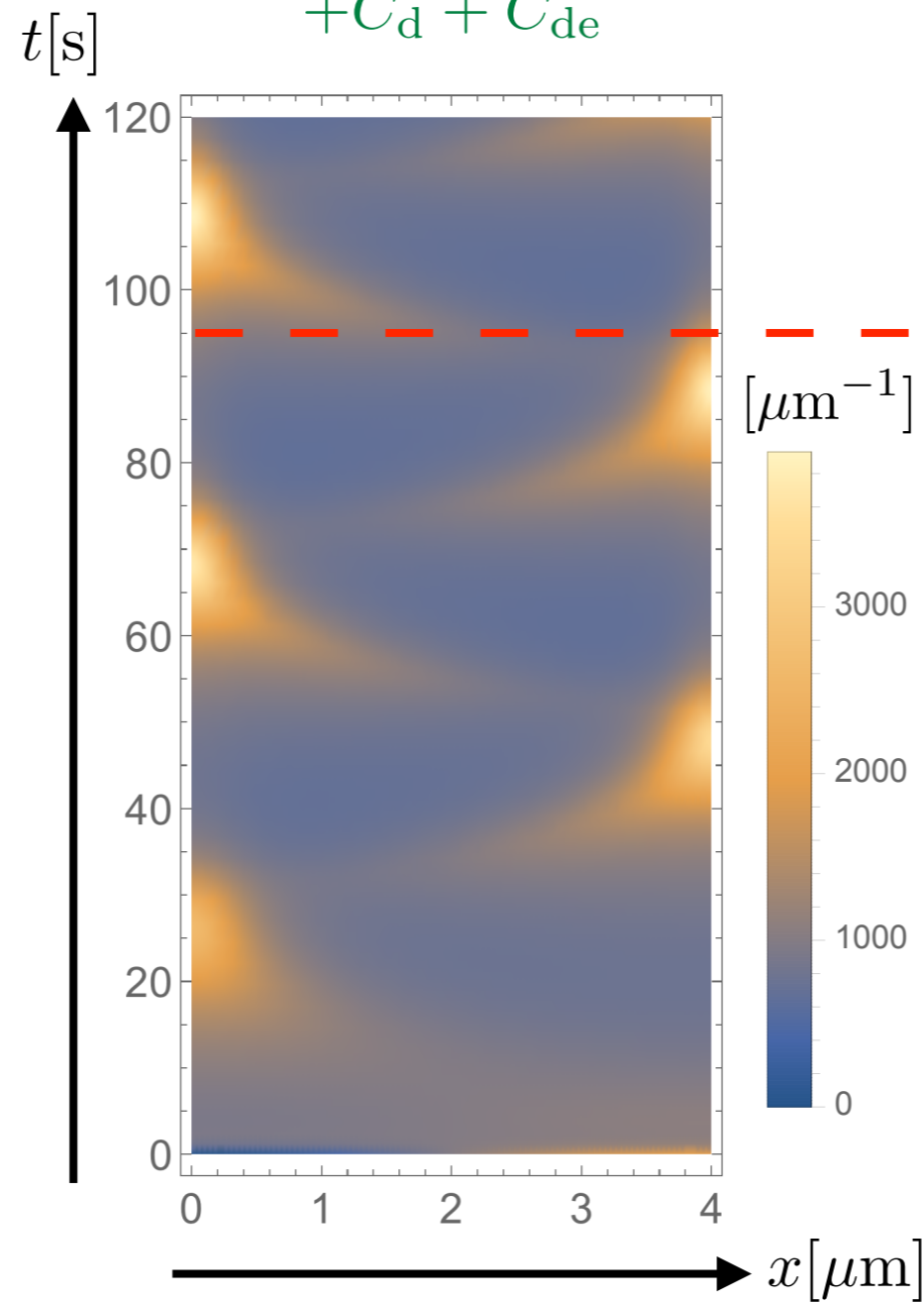
$$C_{D:ADP} + C_{D:ATP} + C_d + C_{de}$$

Tot MinE

$$C_E + C_{de}$$

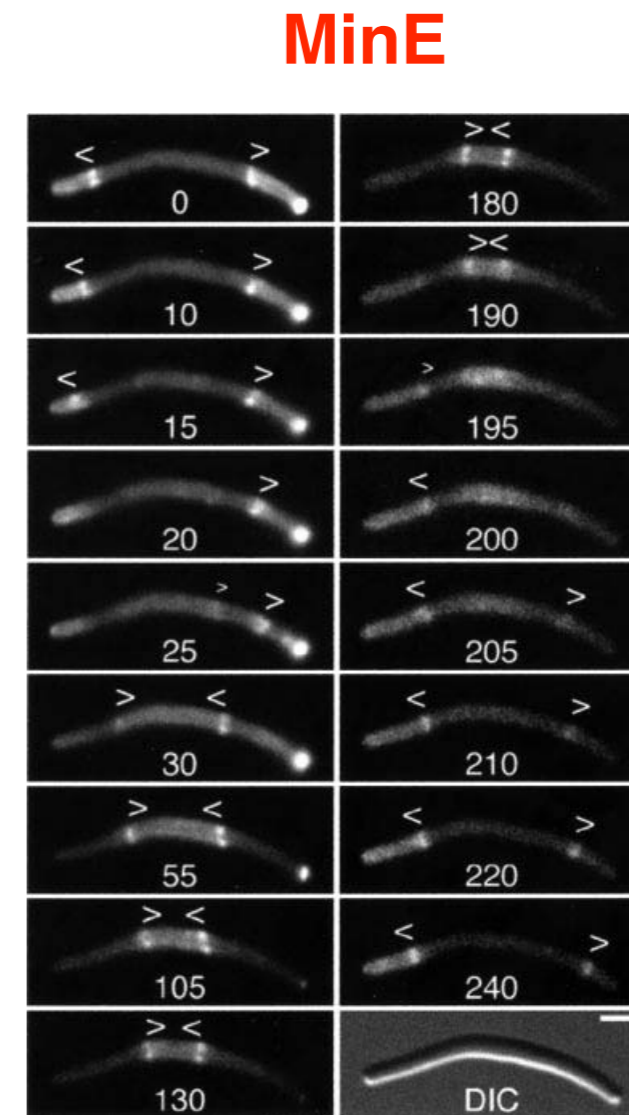
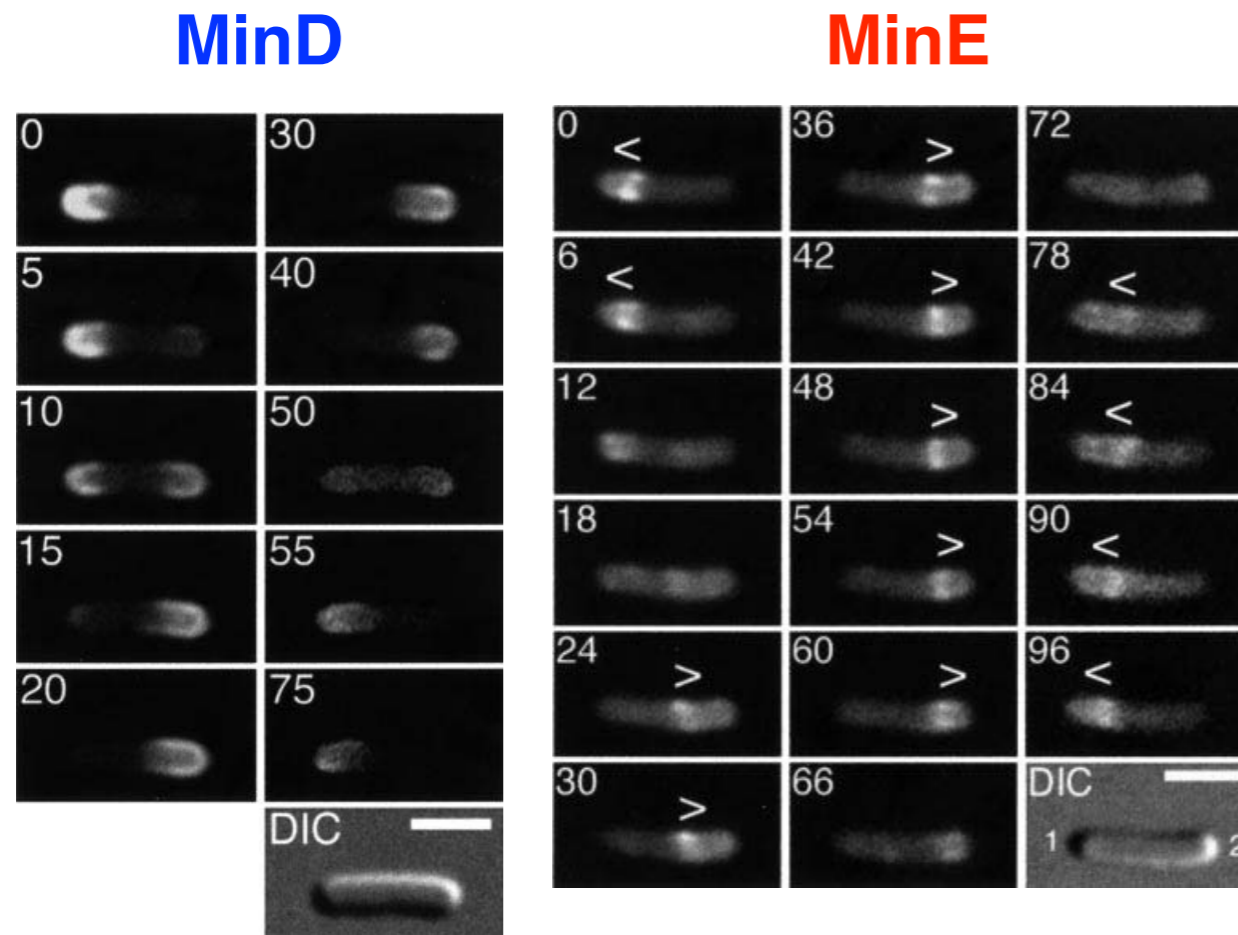


time averaged concentrations

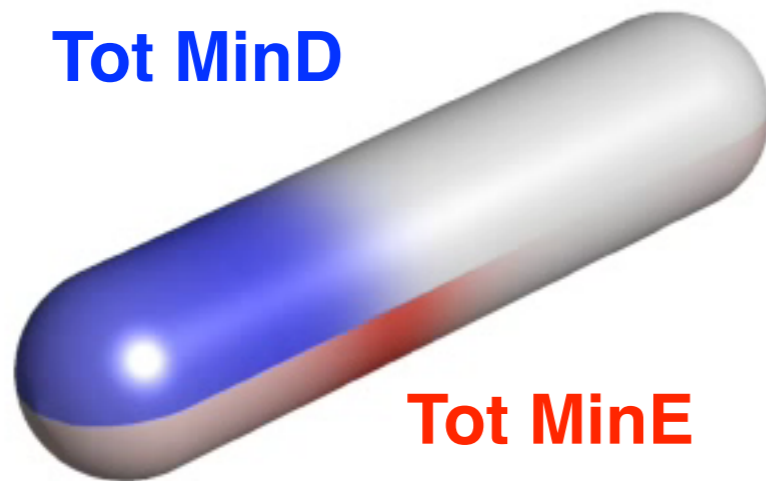


Min system oscillations in larger cells

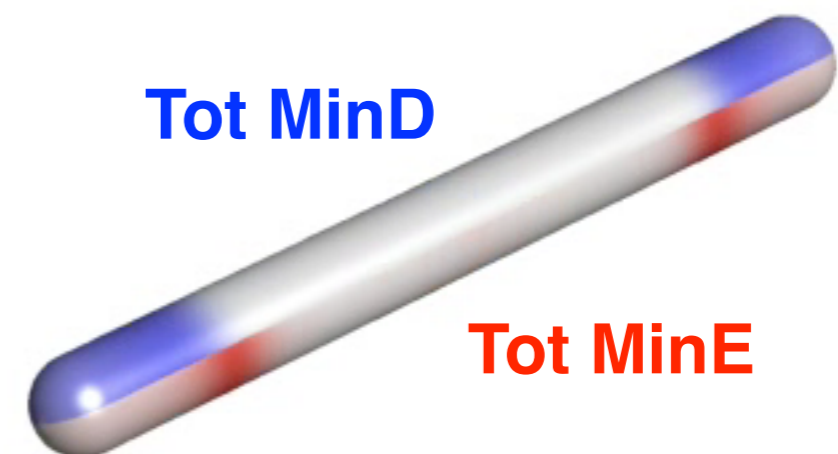
C.A. Hale *et al.*, EMBO 20, 1563 (2001)



<http://mesord.sourceforge.net>



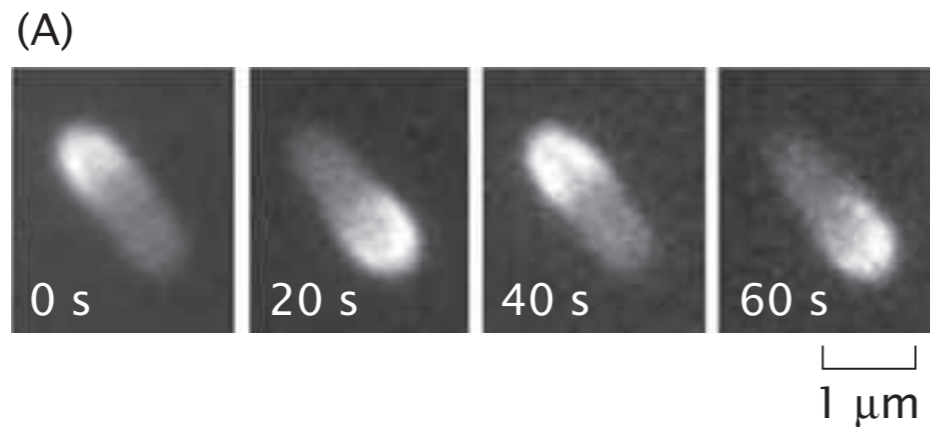
$$L = 4\mu\text{m}$$



$$L = 10\mu\text{m}$$

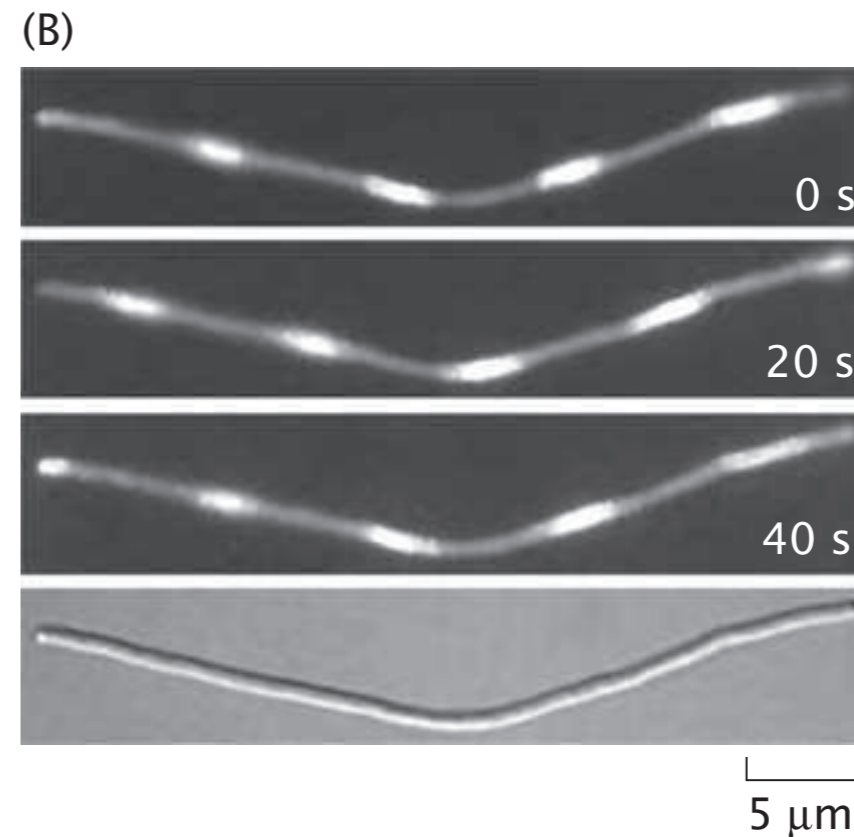
Min system oscillations in large cells

MinD oscillations in normal E. Coli



R. Phillips et al., Physical
Biology of the Cell

MinD oscillations in E. Coli, where division is prevented by removing FtsZ

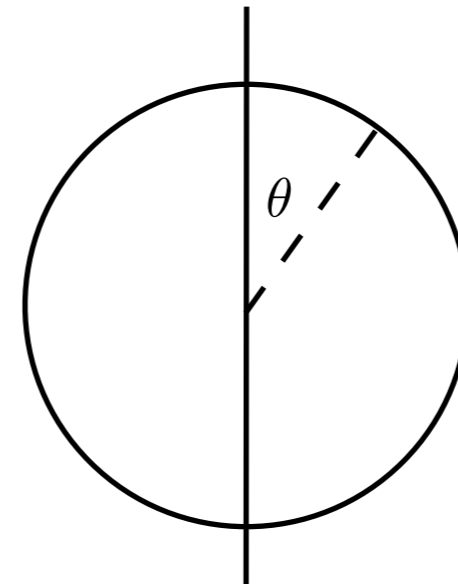
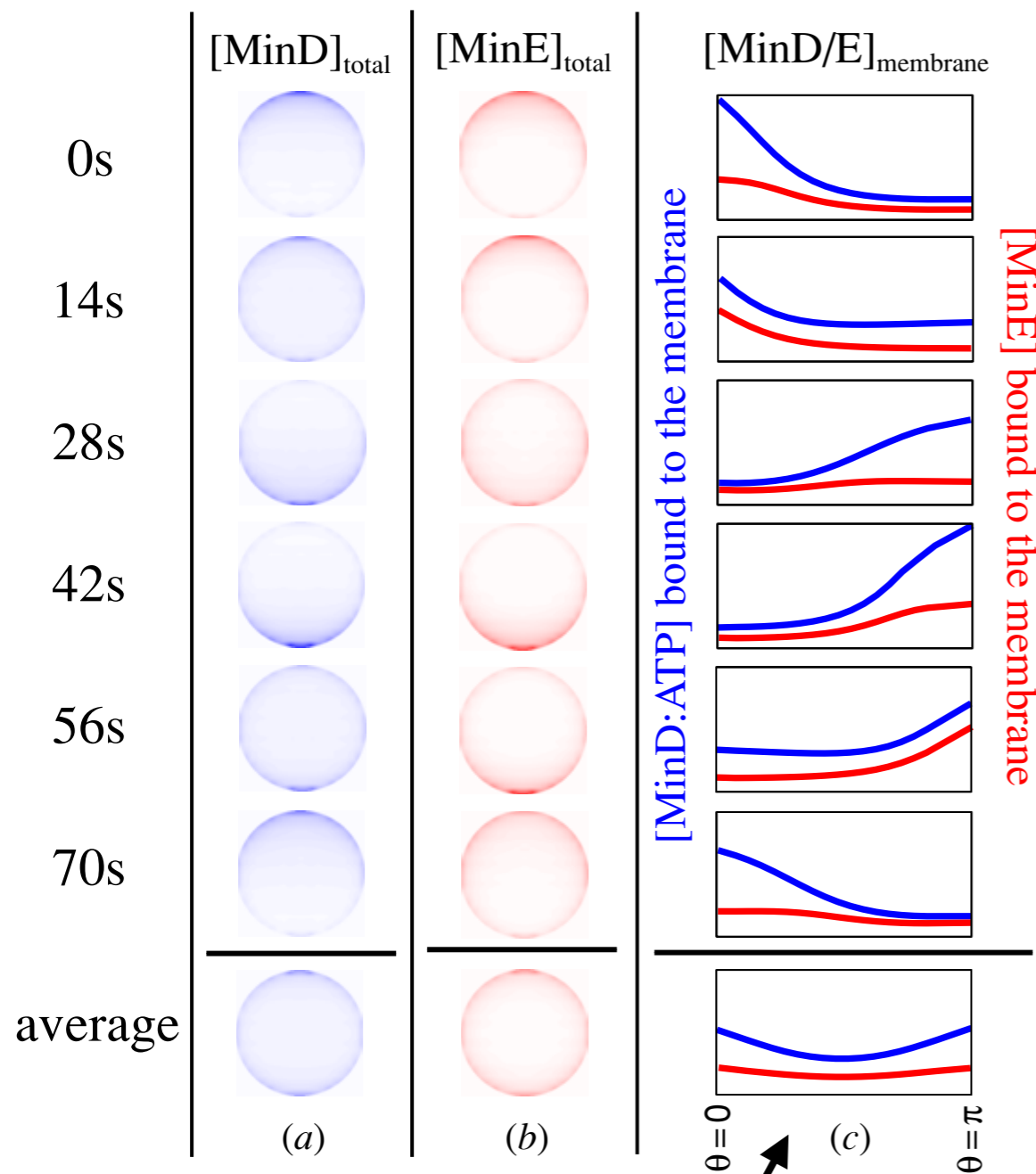


Can the same mechanism work for spherical bacteria?

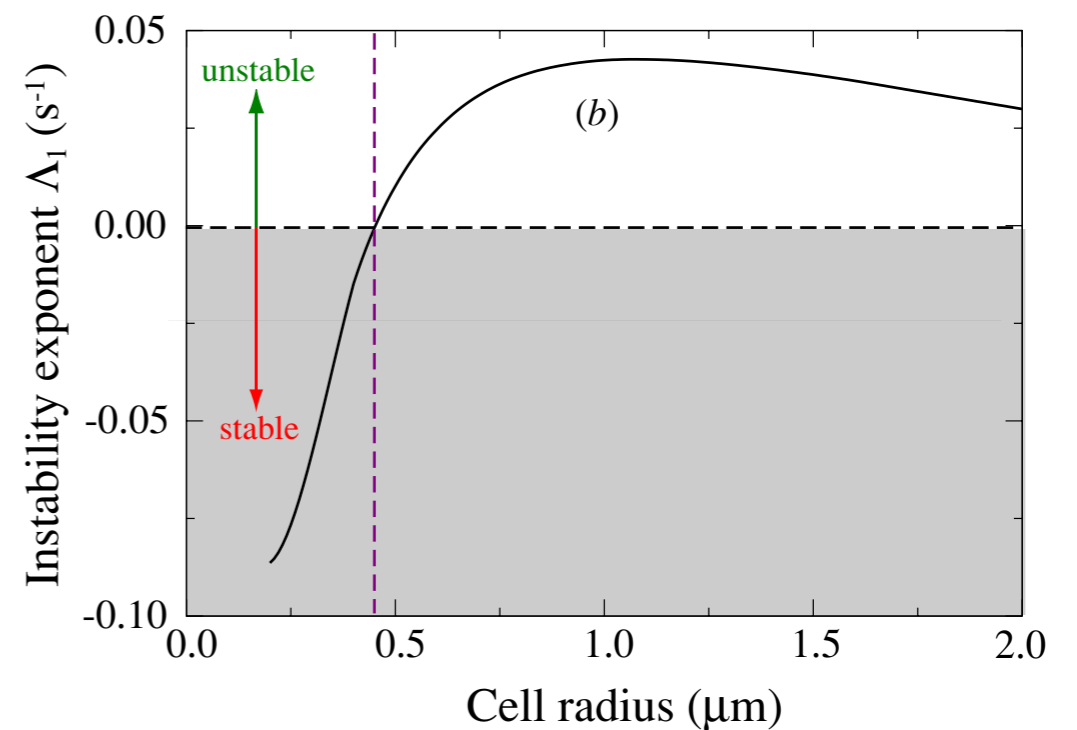


Min system oscillations in spherical bacteria

no MinE ring



Oscillations possible only for sufficiently large cells

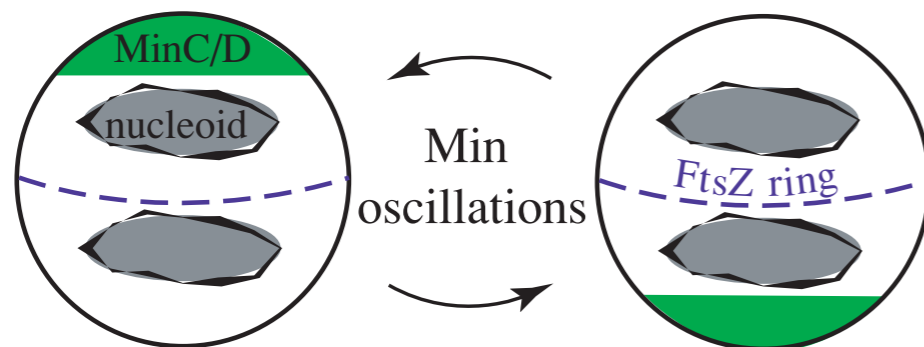


Min oscillations can identify midline

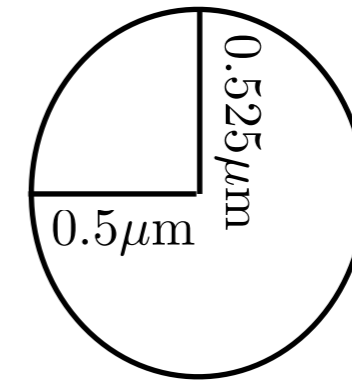
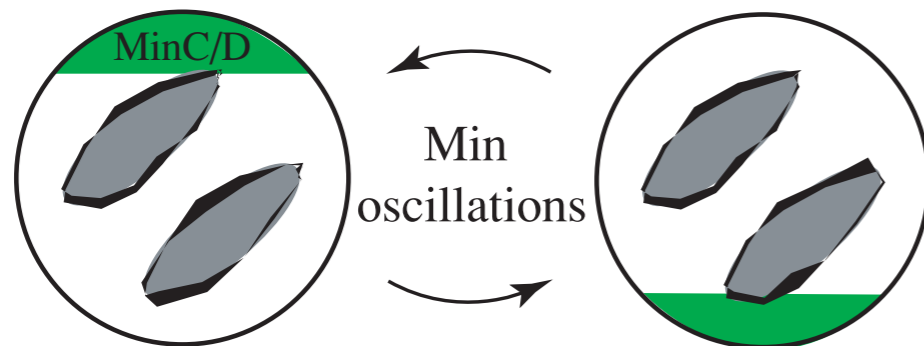
K.C. Huang and N.S. Wingreen, Phys. Biol. 1, 229 (2004)

Finding midline is not enough to divide properly in spherical bacteria

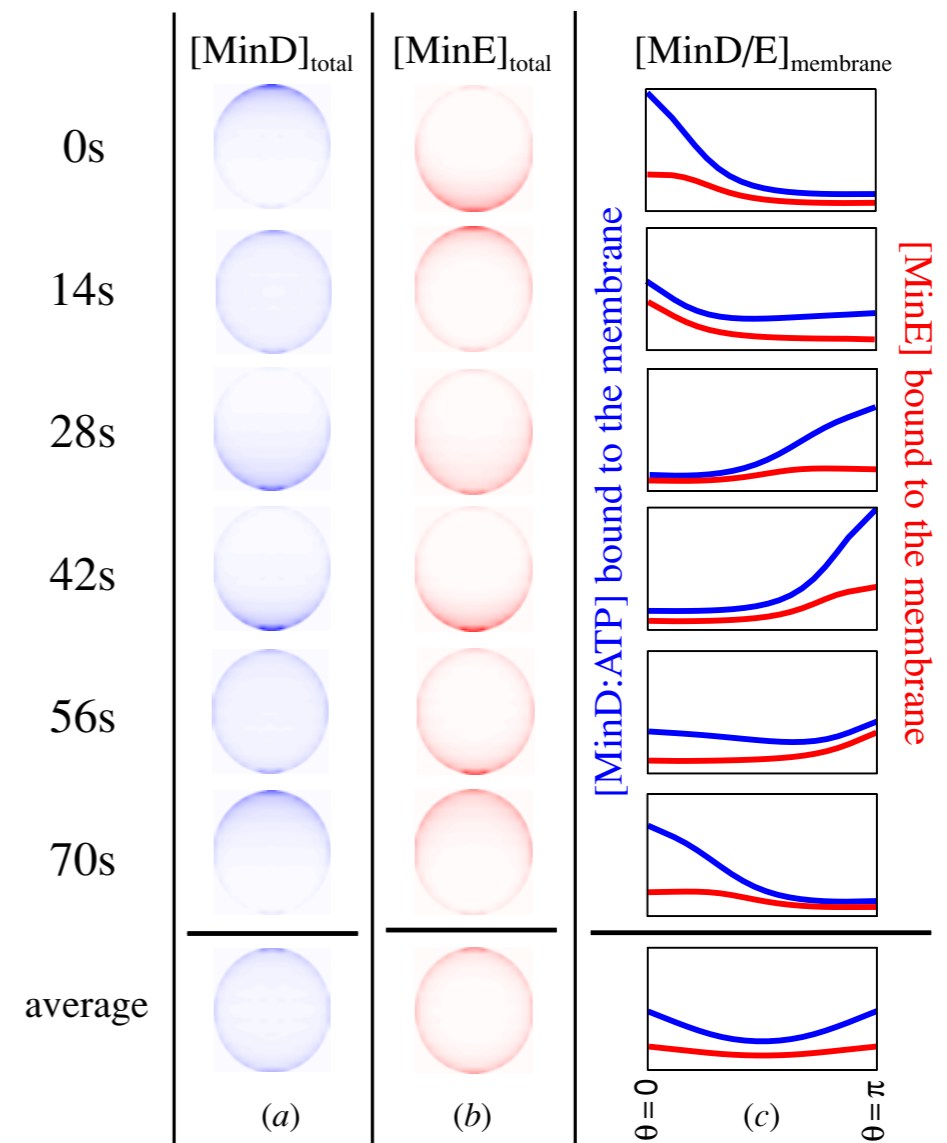
proper alignment



misalignment



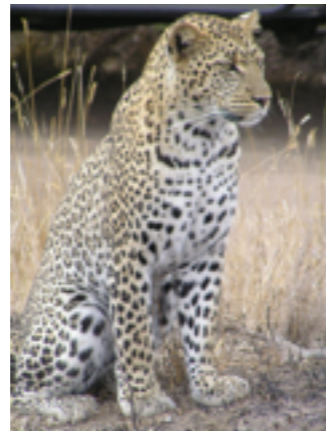
Slight deformation of the cell due to chromosomes could be enough to properly align Min oscillations



Patterns in nature



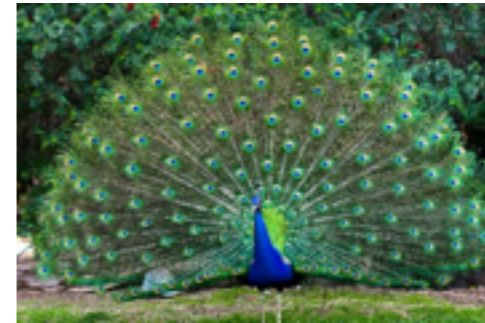
zebra



leopard



royal
angelfish



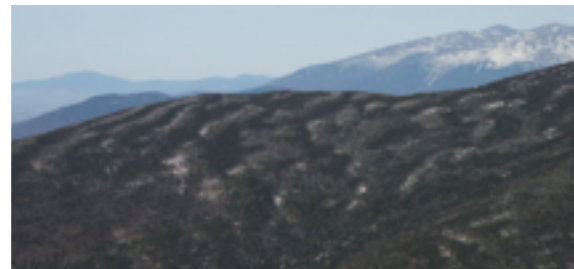
peacock



giant
pufferfish



tiger bush



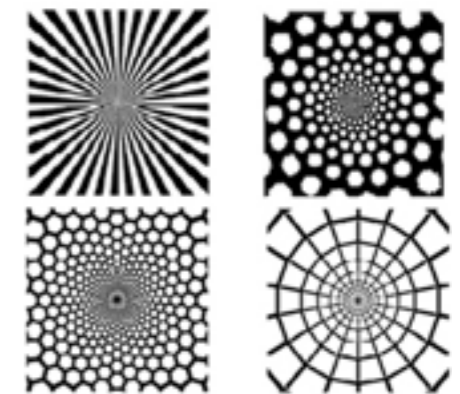
fir waves



mussels



clouds



hallucination
patterns

1952: Alan Turing wrote “The Chemical Basis of Morphogenesis”

Many of these patterns can be constructed with reaction-diffusion models.

What are the minimal requirements that produce such patterns?

Reaction-diffusion equations

$$\frac{\partial C_i(\vec{r}, t)}{\partial t} = F_i(\{C_j(\vec{r}, t)\}) + D_i \nabla^2 C_i(\vec{r}, t)$$

$i = 1, 2, \dots, N$ **N interacting components**

In the absence of diffusion find stable fixed points

$$F_i(\{C_j^*\}) = 0$$

Can diffusion destabilize such fixed points?

Linearize the PDE around the fixed point

$$c_i(\vec{r}, t) = C_i(\vec{r}, t) - C_i^*$$

$$\frac{\partial c_i(\vec{r}, t)}{\partial t} = \sum_{j=1}^N M_{ij}^0 c_j(\vec{r}, t) + D_i \nabla^2 c_i(\vec{r}, t) \quad M_{ij}^0 = \left. \frac{\partial F_i}{\partial C_j} \right|_{C^*}$$

Fourier transform

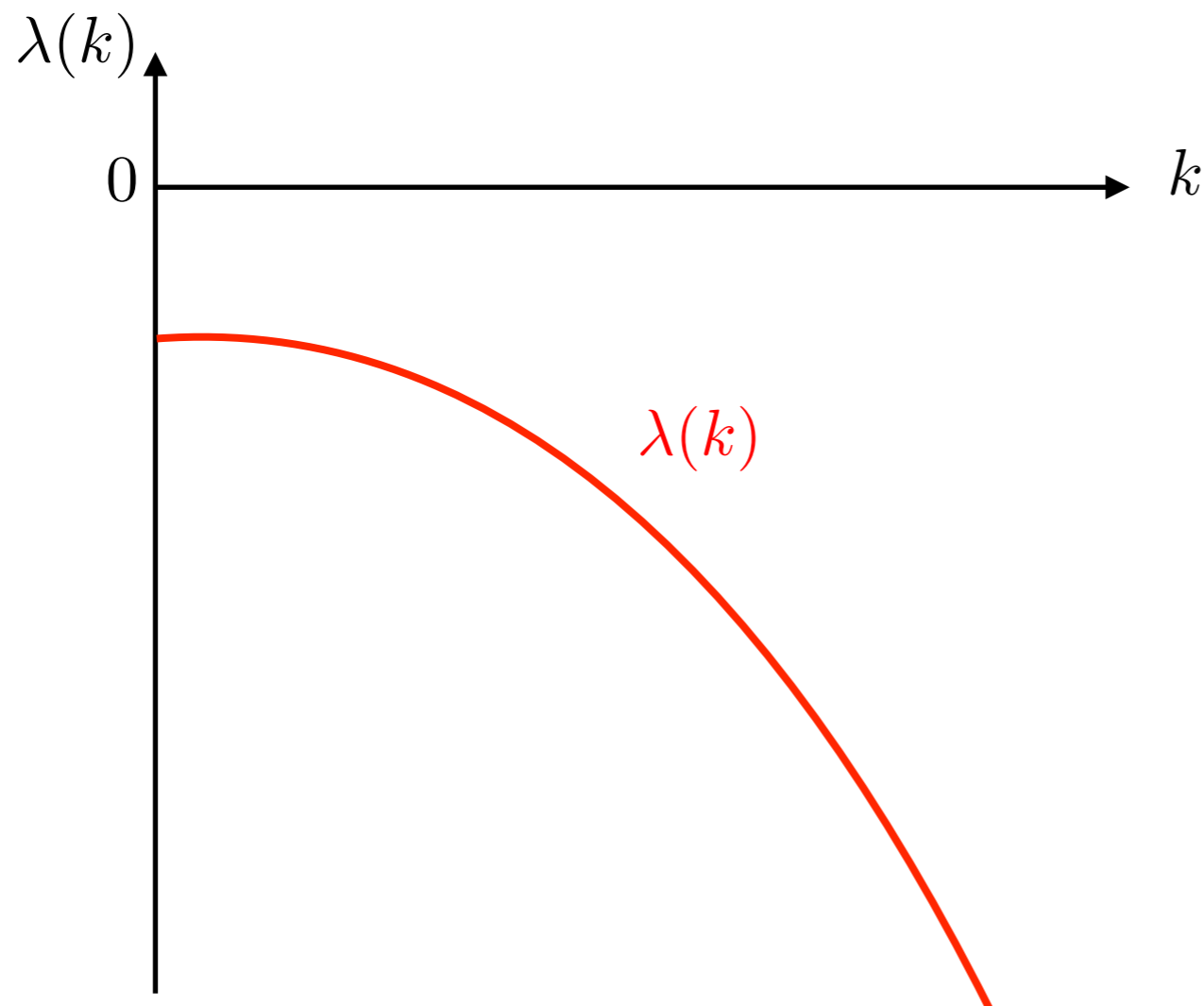
$$\frac{\partial \tilde{c}_i(\vec{k}, t)}{\partial t} = \sum_{j=1}^N (M_{ij}^0 - \delta_{ij} k^2 D_i) \tilde{c}_j(\vec{k}, t)$$

What are the eigenvalues for such system?

One component system ($N=1$)

$$\frac{\partial \tilde{c}_1(\vec{k}, t)}{\partial t} = (M_{11}^0 - k^2 D_1) \tilde{c}_1(\vec{k}, t) \equiv \lambda(k) \tilde{c}_1(\vec{k}, t)$$

Because fixed point is stable in the absence of diffusion, we must have $M_{11} < 0$.



There are no diffusion induced instabilities for one component system!

Two component system ($N=2$)

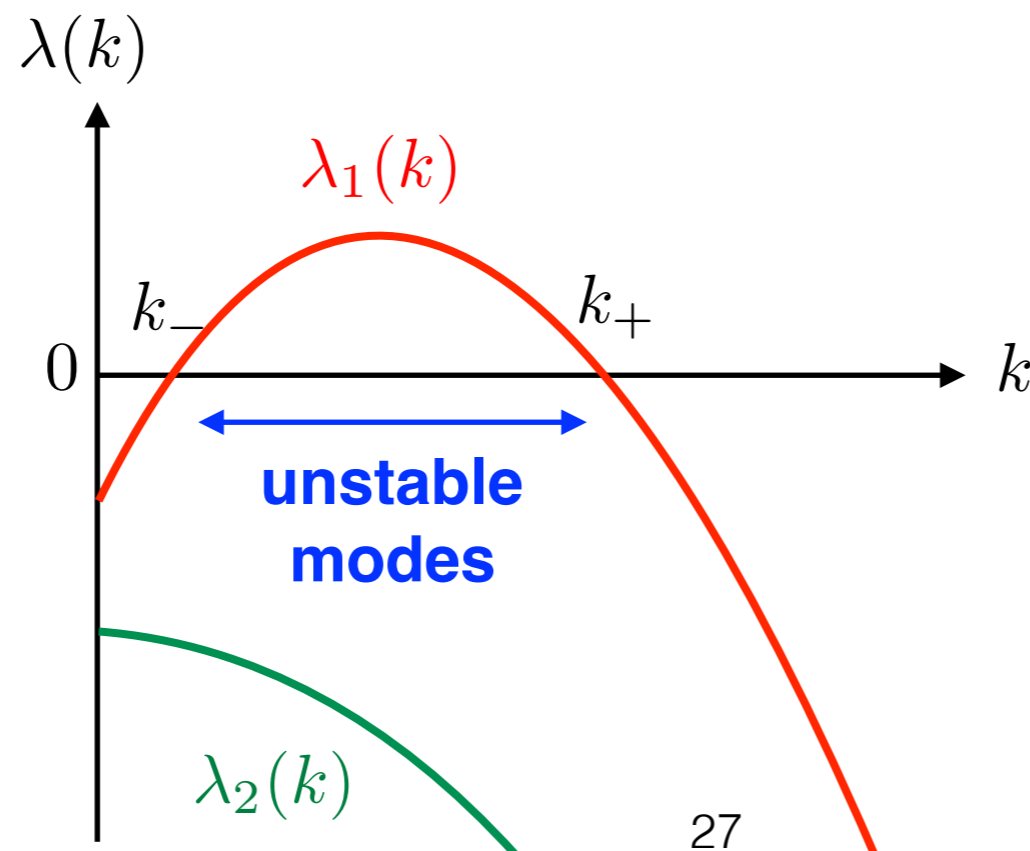
$$\frac{\partial}{\partial t} \begin{pmatrix} \tilde{c}_1(\vec{k}, t) \\ \tilde{c}_2(\vec{k}, t) \end{pmatrix} = \begin{pmatrix} M_{11}^0 - k^2 D_1, & M_{12}^0 \\ M_{21}^0, & M_{22}^0 - k^2 D_2 \end{pmatrix} \begin{pmatrix} \tilde{c}_1(\vec{k}, t) \\ \tilde{c}_2(\vec{k}, t) \end{pmatrix}$$

Relation between eigenvalues and trace of the matrix

$$\lambda_1(0) + \lambda_2(0) = M_{11}^0 + M_{22}^0 < 0$$

$$\lambda_1(k) + \lambda_2(k) = M_{11}^0 + M_{22}^0 - k^2(D_1 + D_2) < 0$$

**Therefore we must have one positive and one negative eigenvalue for Turing instability!
No temporal oscillations are possible!**



What are the conditions for matrix M_{ij}^0 that lead to Turing instability?

Two component system ($N=2$)

$$\frac{\partial}{\partial t} \begin{pmatrix} \tilde{c}_1(\vec{k}, t) \\ \tilde{c}_2(\vec{k}, t) \end{pmatrix} = \begin{pmatrix} M_{11}^0 - k^2 D_1, & M_{12}^0 \\ M_{21}^0, & M_{22}^0 - k^2 D_2 \end{pmatrix} \begin{pmatrix} \tilde{c}_1(\vec{k}, t) \\ \tilde{c}_2(\vec{k}, t) \end{pmatrix}$$

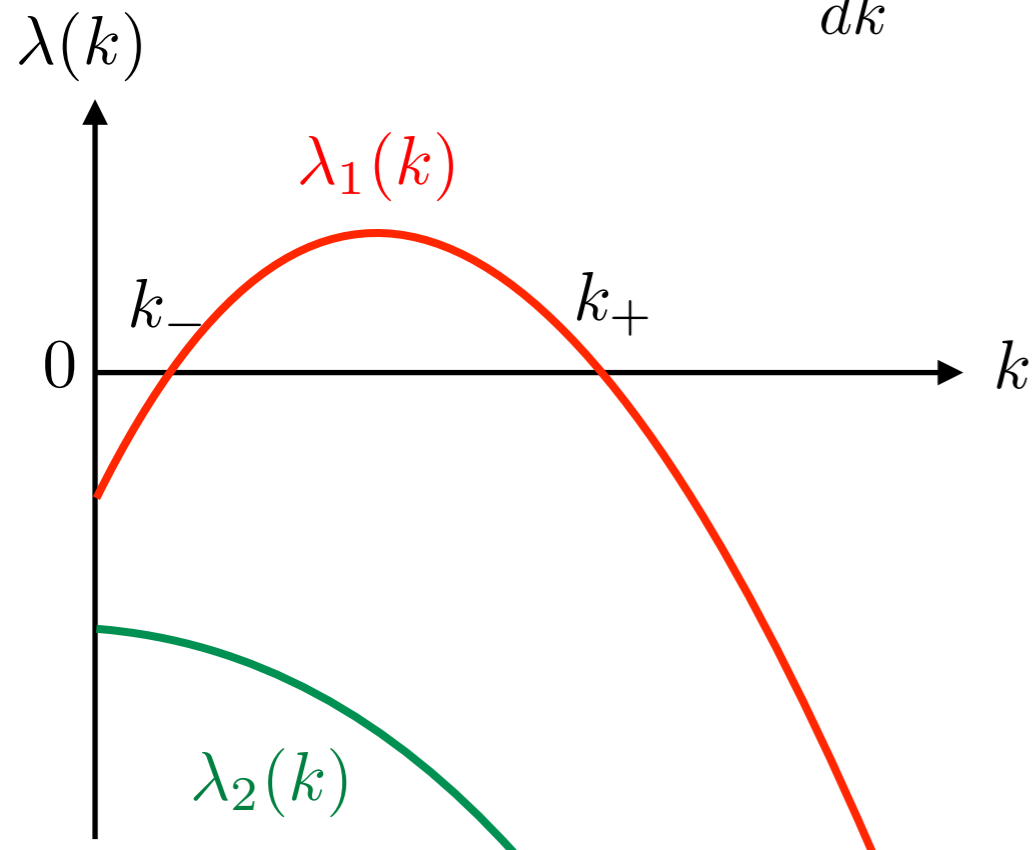
Relation between eigenvalues and determinant of the matrix

$$\lambda_1(k)\lambda_2(k) = (M_{11}^0 - k^2 D_1)(M_{22}^0 - k^2 D_2) - M_{12}^0 M_{21}^0$$

$$\lambda_1(k)\lambda_2(k) = M_{11}^0 M_{22}^0 - M_{12}^0 M_{21}^0 - k^2 (M_{11}^0 D_2 + M_{22}^0 D_1) + k^4 D_1 D_2$$

Determinant becomes negative and reaches minimal value at $k^* \in (k_-, k_+)$.

$$\frac{d(\lambda_1(k)\lambda_2(k))}{dk} = 0 \longrightarrow k^{*2} = \frac{M_{11}^0 D_2 + M_{22}^0 D_1}{2D_1 D_2} > 0$$



$$\lambda_1(0) + \lambda_2(0) < 0$$

$$M_{11}^0 M_{22}^0 < 0$$

$$\lambda_1(0)\lambda_2(0) > 0$$

$$M_{12}^0 M_{21}^0 < 0$$

Two component system ($N=2$)

$$\frac{\partial}{\partial t} \begin{pmatrix} \tilde{c}_1(\vec{k}, t) \\ \tilde{c}_2(\vec{k}, t) \end{pmatrix} = \begin{pmatrix} M_{11}^0 - k^2 D_1, & M_{12}^0 \\ M_{21}^0, & M_{22}^0 - k^2 D_2 \end{pmatrix} \begin{pmatrix} \tilde{c}_1(\vec{k}, t) \\ \tilde{c}_2(\vec{k}, t) \end{pmatrix}$$

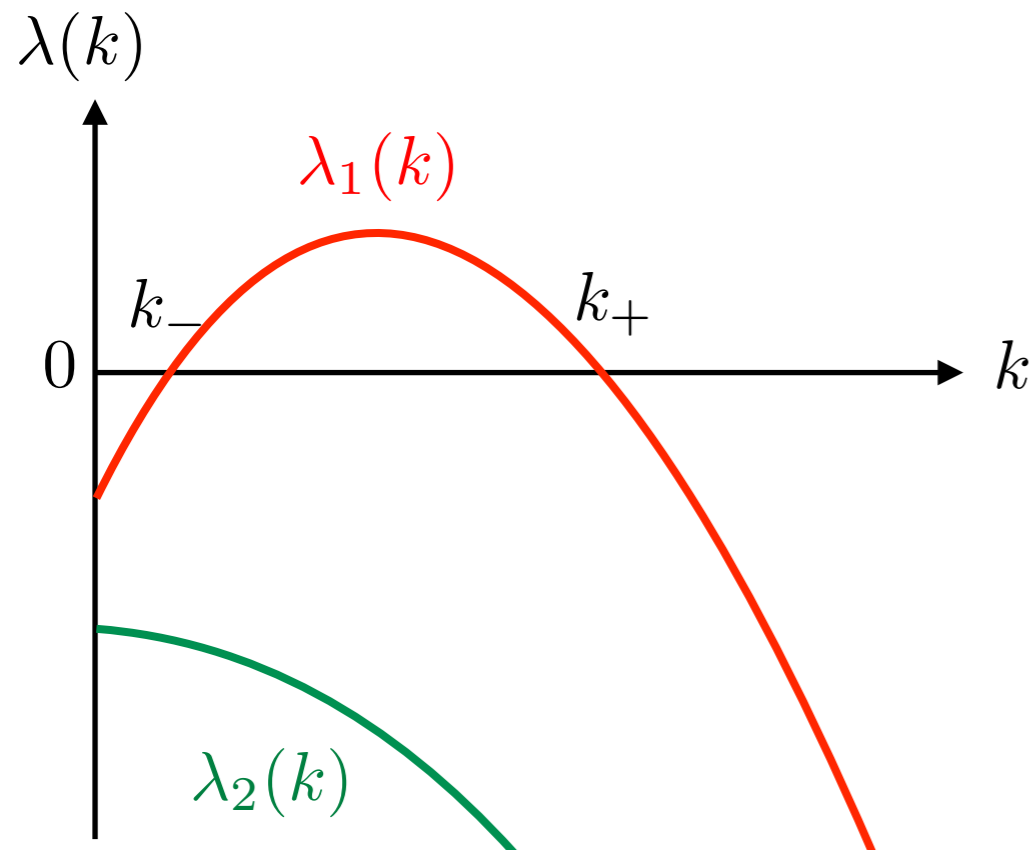
Without loss of generality we can assume $M_{11}^0 < 0$, $M_{22}^0 > 0$

$$M_{11}^0 + M_{22}^0 < 0 \longrightarrow$$

$$|M_{11}^0| > |M_{22}^0|$$

$$\frac{M_{11}^0 D_2 + M_{22}^0 D_1}{2D_1 D_2} > 0 \longrightarrow$$

$$D_1 > \frac{|M_{11}^0|}{|M_{22}^0|} D_2 > D_2$$



Finite wavelength Turing instabilities arise by long-ranged inhibition and short-range excitation. The resulting patterns are fixed in time.

In the system with 3 or more components oscillating patterns in time are also possible.