

Outline

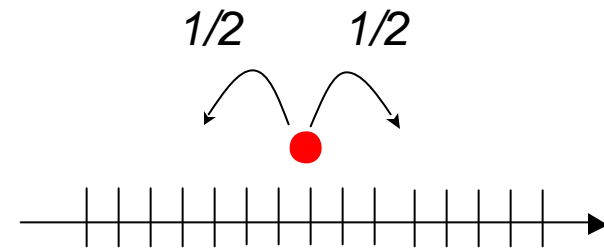
1. Position jump process
2. Velocity jump process
3. Macroscopic flux – dispersion, chemotaxis, chemokinesis
4. Example – biased random walk
5. Chemotactic wave paradox
6. Chemotaxis-driven linear instability

Position Jump Process (1)

N steps, net displacement - m : $-N \leq m \leq N$

(N odd/even \Rightarrow m odd/even)

$$P_N(m) = \frac{N!}{\left(\frac{N+m}{2}\right)! \left(\frac{N-m}{2}\right)!} \left(\frac{1}{2}\right)^N$$



$\langle m \rangle = 0$ $\langle (m - \langle m \rangle)^2 \rangle = N$ \longrightarrow *diffusive behavior*

Approximate solution: $P_N(m) \approx \sqrt{\frac{2}{\pi N}} e^{-\frac{m^2}{2N}}$

use $\log(n!) \approx (n + 1/2)\log n - n + 1/2 \log \pi + O(n^{-1})$ &

$$\log\left(1 \pm \frac{m}{N}\right) \approx \pm \frac{m}{N} - \frac{m^2}{2N^2} + O\left(\frac{m^3}{N^3}\right)$$

Position Jump Process (2)

step size: l frequency of jumps: l

Continuous space and time:

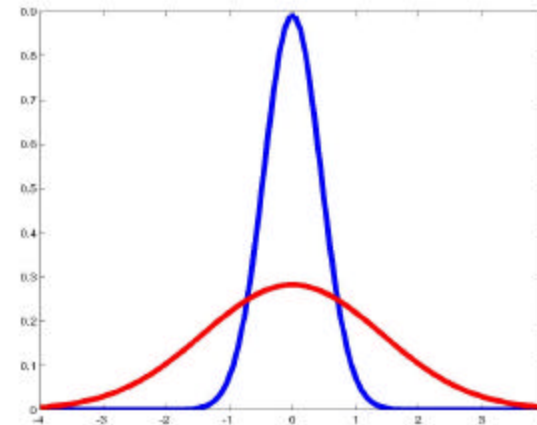
$$p(x, t)dx = P_N(m) \frac{dx}{2l} = \frac{1}{\sqrt{2pl^2t}} e^{-x^2/2l^2t} dx$$

$$D \equiv ll^2/2 \rightarrow p(x, t) = \frac{1}{\sqrt{2pDt}} e^{-x^2/4Dt}$$

$$\frac{\partial p}{\partial t} = \frac{\partial}{\partial x} \left(D \frac{\partial p}{\partial x} \right) \text{ with } p(x, t=0) = \mathbf{d}(x)$$

$$\text{Rewrite: } \frac{\partial p}{\partial t} = -\frac{\partial}{\partial x} J \leftarrow \text{flux } J \equiv -D \frac{\partial p}{\partial x}$$

$p(x, t) > 0$ for all t



Approximation for large times

*earlier times: correlations in motion
-> cell migration is like that*

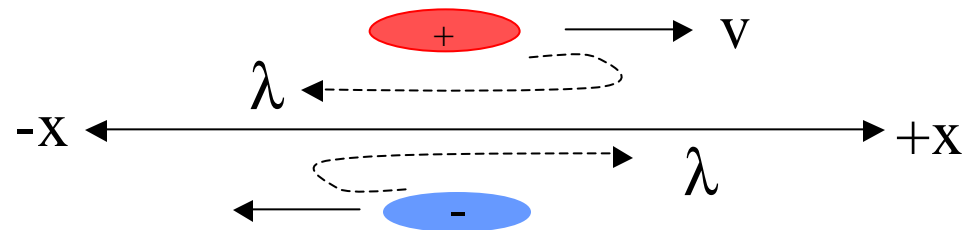
$$\langle x^2 \rangle = 2Dt$$

Velocity Jump Process (1)

instantaneous changes in velocity

$$\frac{\partial p^+}{\partial t} + v \frac{\partial p^+}{\partial x} = -I p^+ + I p^-$$

$$\frac{\partial p^-}{\partial t} - v \frac{\partial p^-}{\partial x} = -I p^- + I p^+$$



$$p(x, t) \equiv p^+(x, t) + p^-(x, t) \quad \text{total probability}$$

$$j(x, t) \equiv v(p^+(x, t) - p^-(x, t)) \quad \text{flux}$$

$$\frac{\partial^2 p}{\partial t^2} + 2I \frac{\partial p}{\partial t} = v^2 \frac{\partial^2 p}{\partial x^2}$$

$$p(x, 0) = p_0(x), \quad \frac{\partial p(0, x)}{\partial t} = -\frac{\partial j(0, x)}{\partial x}$$

$$\langle x(t)^2 \rangle \equiv \int_{-\infty}^{\infty} x^2 p(x, t) dx = \frac{v^2}{I} \left\{ t - \frac{1}{2I} (1 - e^{-2It}) \right\} \longrightarrow$$

Velocity Jump Process (2)

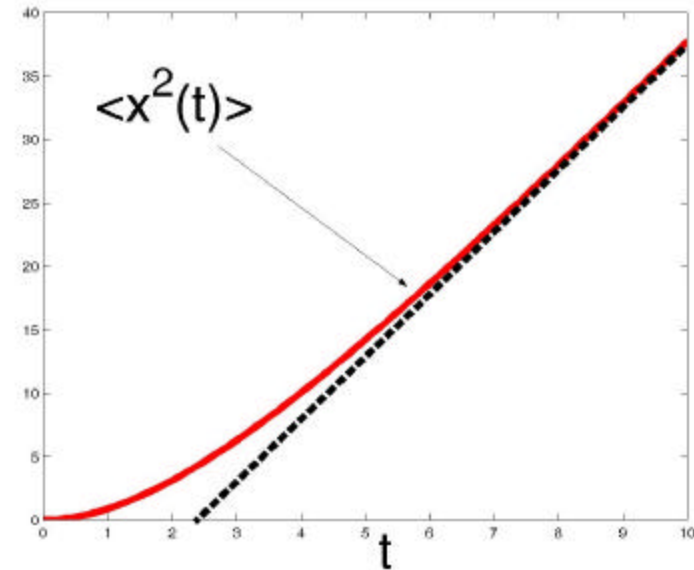
ballistic and diffusive regimes

$$\langle x(t)^2 \rangle = \begin{cases} t \ll 1: \langle x^2(t) \rangle \sim v^2 t^2 \\ t \gg 1: \langle x^2(t) \rangle \sim \frac{v^2 t}{l} \end{cases}$$

$$D \equiv \frac{n^2}{2l}$$

$T \equiv 2l$ – "persistence time"

v – velocity

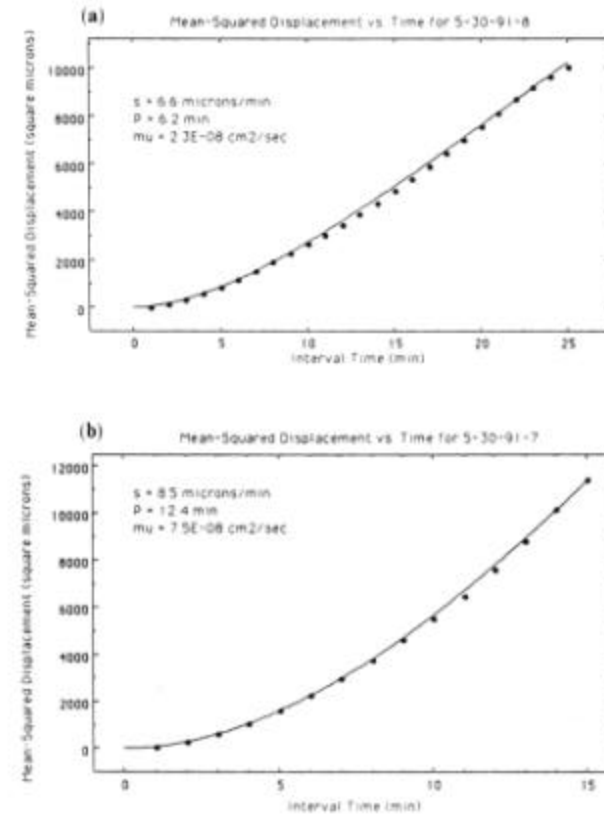
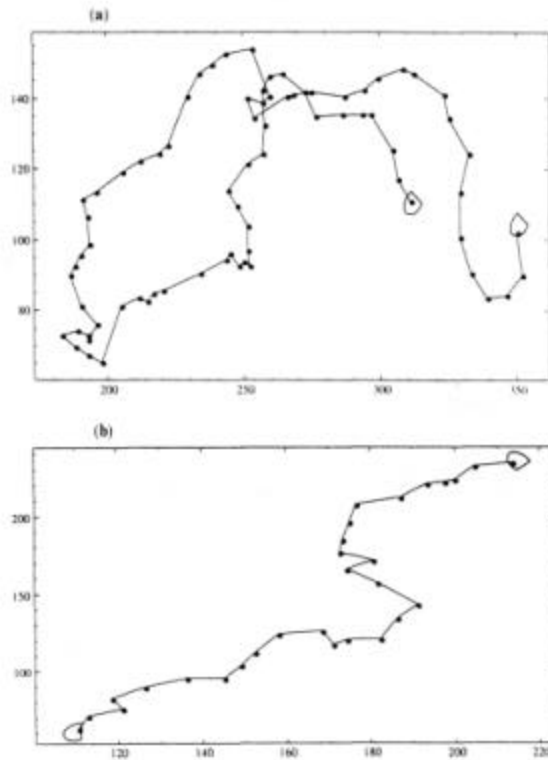


$$j(x, t) = -D \frac{\partial p(x, t)}{\partial x} \quad \text{only when } l \rightarrow \infty$$

expression for flux has memory:

$$j(x, t) = e^{-2lt} j(x, 0) - v^2 \int_0^t e^{-2l(t-t')} \frac{\partial p(x, t')}{\partial x} dt'$$

Velocity Jump Process (3) fitting data



1. **Microscopic** parameters can be extracted from data
2. Next step: expressions for **macroscopic** fluxes

BE Farrell et al, Cell Motil Cytoskeleton, 16:279-293, 1990 - example

RB Dickinson, RT Tranquillo, AICHE J, 39 (12): 1995, 1993 - estimation algorithms

Macroscopic Flux (1)

$$\frac{\partial n^+}{\partial t} + \frac{\partial}{\partial x}(vn^+) = I^- n^- - I^+ n^+$$
$$\frac{\partial n^-}{\partial t} - \frac{\partial}{\partial x}(vn^-) = I^+ n^+ - I^- n^-$$

total cell density: $n \equiv n^+ + n^-$

flux: $j \equiv v(n^+ - n^-)$

steady state for the flux ($t \gg (I^+ + I^-)^{-1}$)

$$j_{eq} = \frac{-v^2 \frac{\partial n}{\partial x} - nv \frac{\partial v}{\partial x} - vn(I^+ - I^-)}{(I^- + I^+) - v^{-1} \frac{\partial v}{\partial t}}$$

Macroscopic Flux (2)

$$T_p \equiv [(I^- + I^+) - v^{-1} \frac{\partial v}{\partial t}]^{-1} \quad \text{persistence time}$$

$$\mathbf{m} \equiv T_p v^2 \quad \text{random motility coefficient}$$

$$V_c \equiv T_p v (I^- - I^+) \quad \text{chemotactic velocity}$$



$$j_{eq} = -\mathbf{m} \frac{\partial n}{\partial x} + V_c n - T_p v n \frac{\partial v}{\partial x}$$

in phenomenological models

$$j_{eq} = -\mathbf{m} \frac{\partial n}{\partial x} + \underline{\mathbf{a}} n$$

Three contributions to flux:

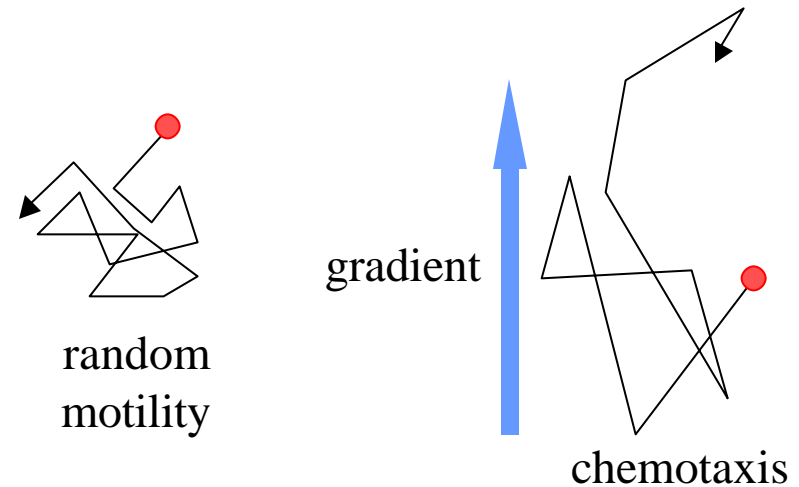
1. random motility
2. chemotaxis (right- and left- moving cells reverse differently)
3. chemokinesis (gradient in cell velocity)

To couple to external concentration field, combine with the experimentally determined dependencies of \mathbf{m} and T_p

Flux in a 1D Gradient (1)

Motivated by Berg & Brown 1972 Experiments

- runs & tumbles
- tumble duration is zero
- use velocity jump process in 1D
- motion in a gradient



$$T_p = (I^- + I^+)^{-1}$$

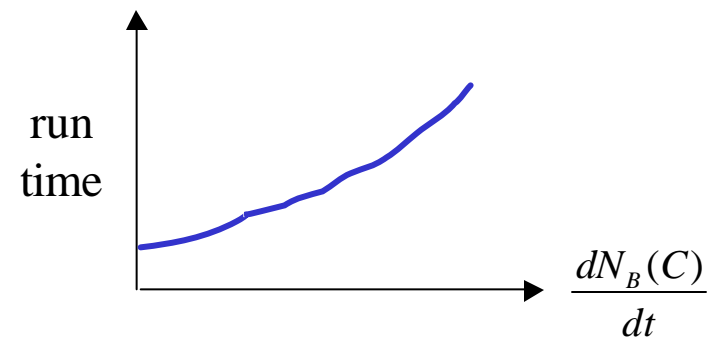
$$I^{+/-} = p_T^{+/-} (1 - y) / 2$$

p_T : is the tumbling probability

y : "directional persistence"

probability of reversing after tumbling

receptor-mediated mechanism:
 N_B – # of occupied receptors



Flux in a 1D Gradient (2)

$$t = t_0 e^{s \frac{dN_b}{dt}}$$

relate to the
frequency
of tumbles



$$p_T^{+/-} = 1/t^{+/-} = p_0 e^{-s \frac{dN_B}{dt}}$$

time derivative

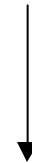
seen by the "bacterium"



$$\frac{dN_B}{dt} = \frac{\partial N_B}{\partial t} \pm v \frac{\partial N_B}{\partial x}$$

$$I^+ + I^- = (1-y) e^{-s \frac{\partial N_b}{\partial t}} \cosh\left(s v \frac{\partial N_b}{\partial x}\right)$$

$$I^+ - I^- = p_0 (1-y) e^{-s \frac{\partial N_b}{\partial t}} \sinh\left(s v \frac{\partial N_b}{\partial x}\right)$$



$$m = \frac{v^2}{p_0 (1-y)} e^{s \frac{\partial N_b}{\partial t}} \operatorname{sech}\left(s v \frac{\partial N_b}{\partial x}\right)$$

$$V_c = v \tanh\left(s v \frac{\partial N_b}{\partial x}\right)$$

Flux in a 1D Gradient (3)

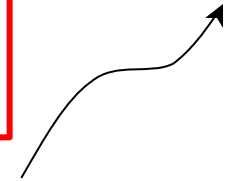
Simple Ligand/receptor Equilibrium

$$N_B = \frac{N_{total}c}{K_D + c} \Rightarrow \frac{dN_B}{dc} = \frac{N_T K_d}{(K_d + c)^2}$$

$$\mathbf{m} = \frac{v^2}{p_0(1-y)} e^{s \frac{\partial c}{\partial t} \frac{dN_B}{dc}} \left[\cosh\left(\mathbf{s} v \frac{\partial c}{\partial x} \frac{dN_B}{dc} \right) \right]^{-1}$$

$$V_c = v \tanh\left(\mathbf{s} v \frac{\partial c}{\partial x} \frac{dN_B}{dc} \right)$$

chemotactic coefficient, χ

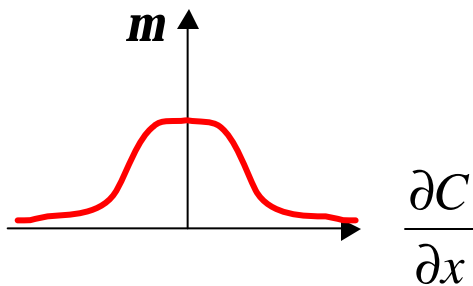
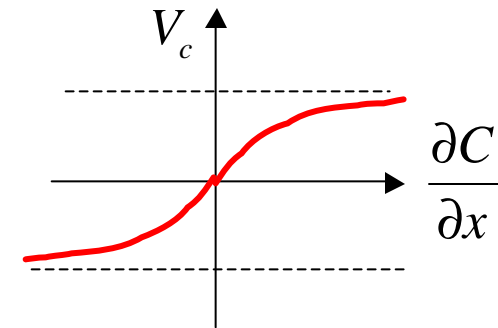
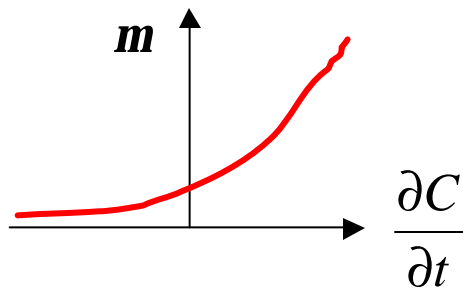


small gradients:

$$\mathbf{m} = \frac{v^2}{p_0(1-y)} \left[1 + \mathbf{s} \frac{\partial c}{\partial t} \frac{dN_B}{dc} \right], \quad V_c = \boxed{\mathbf{s} v^2 \frac{dN_B}{dc}} \frac{\partial c}{\partial x}$$

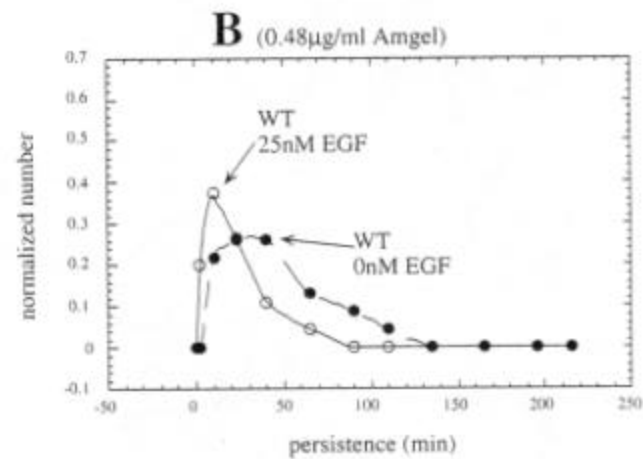
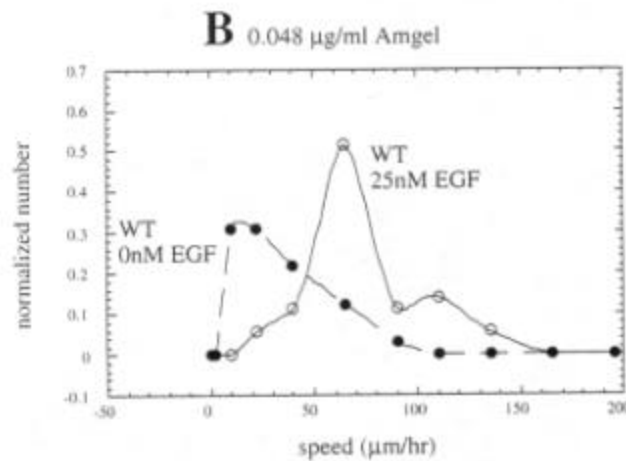
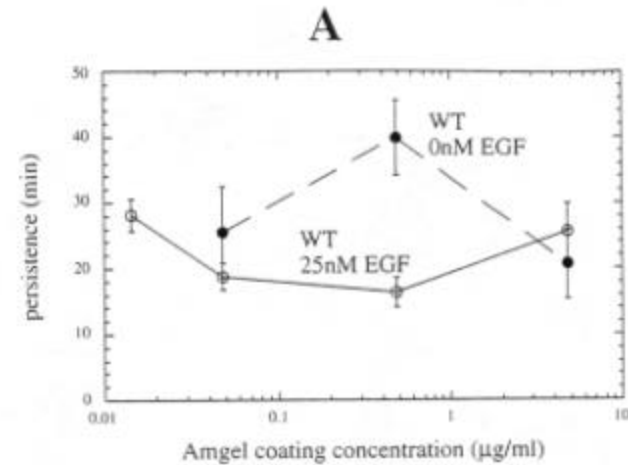
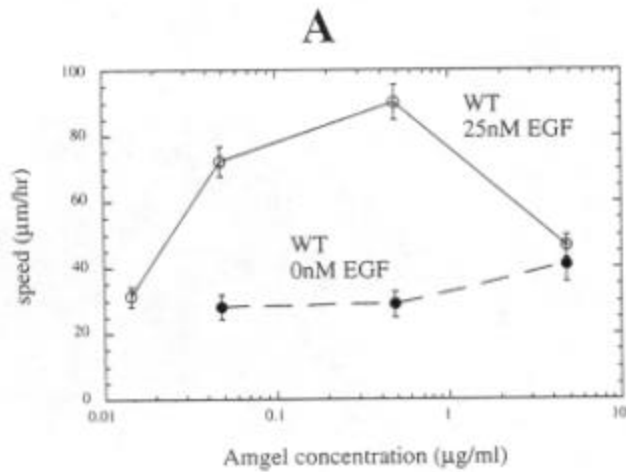
If the model is correct: macroscopic flux can be estimated from data on binding and microscopic parameters for cell migration

Flux in a 1D Gradient (4): Analysis



1. Random motility coefficient increases with temporal gradient
2. Random motility coefficient is a decreasing function of spatial gradient: at large gradients all cells swim in one direction
3. Chemotactic velocity has a limiting value: the population can not move faster than the maximal cell speed

Example: Growth Factor Mediated Cell Motility



“Chemotactic Wave Paradox”

Observation

aggregation to the source of chemical wave
pulse of cAMP is nearly symmetric

Devreotes & Tomchik, *Science* 212, 443-6, 1981

Simple-model:

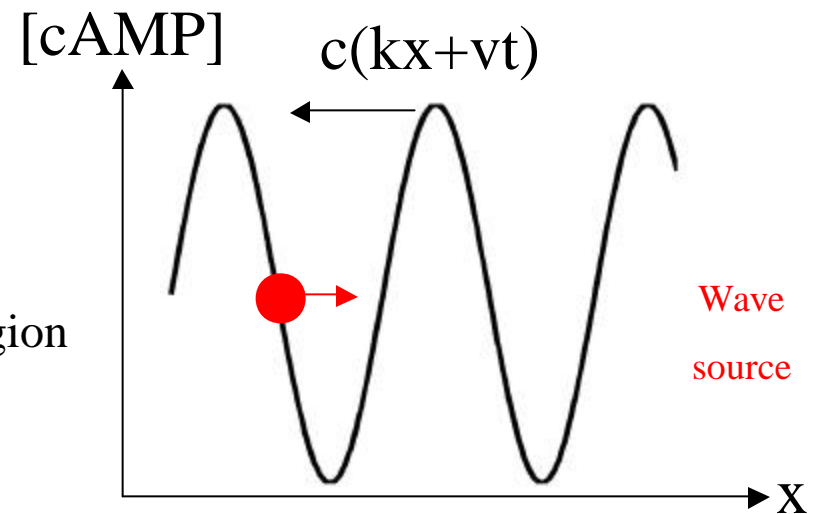
symmetric chemotactic velocity
no net directed motion

Worse: cells stay longer in the negative gradient region

Prediction: cells move away from the wave source

What is the problem?

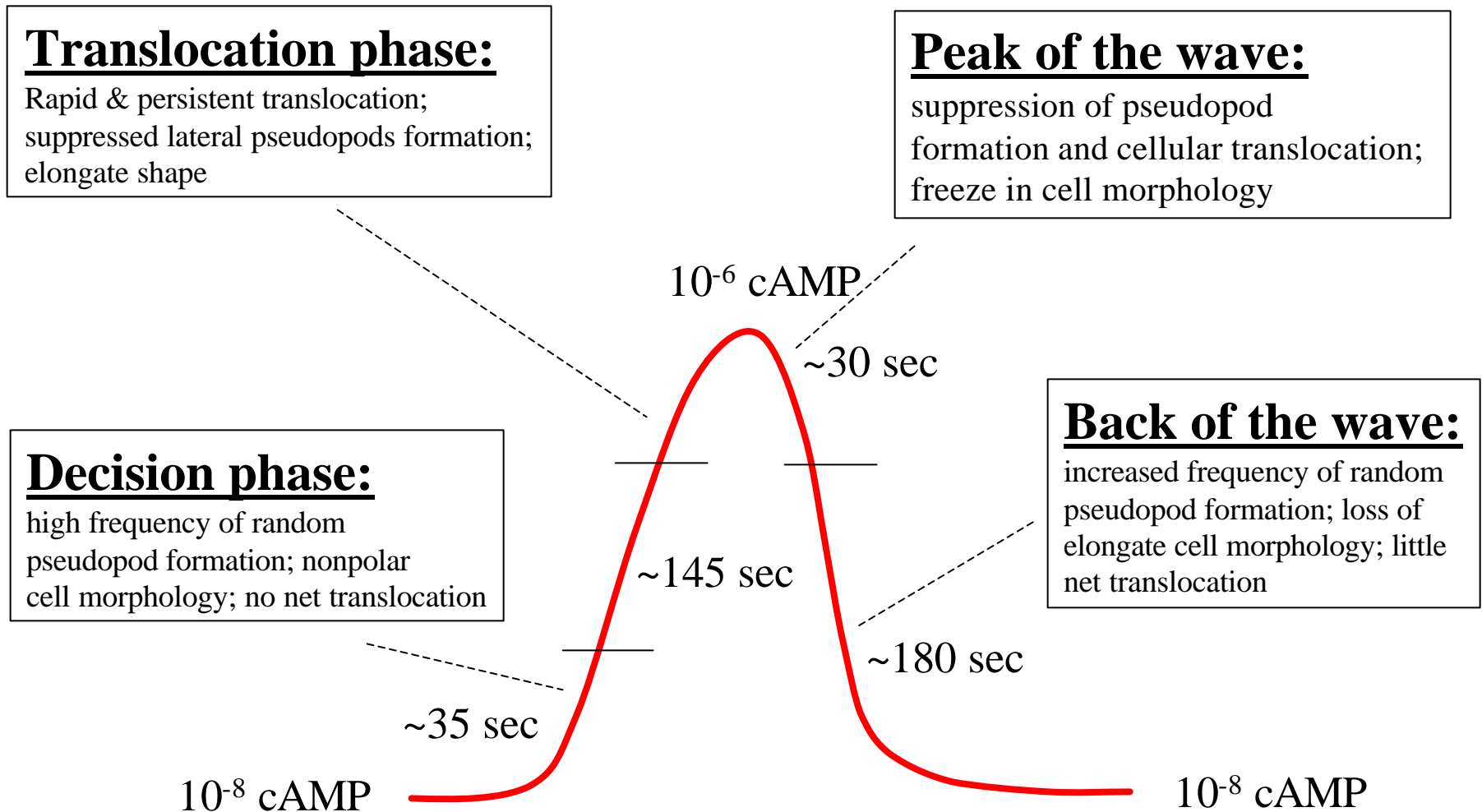
Experiment: Cells move only in the wave front and not in the back => chemotactic response can not be determined by the concentration gradient alone



$$c = c(a)$$

chemotactic
sensitivity

Model: Soll, Wessels, Sylwester, 1993

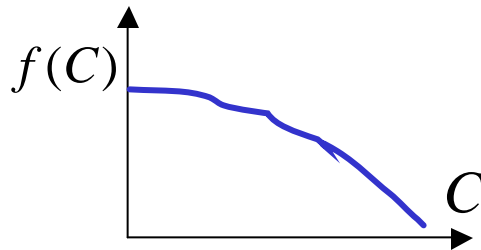


Chemotactic Sensitivity Is Dynamic

$$\frac{dx_{cell}}{dt} = c(C) \frac{dC(kx + vt)}{dx}$$
$$t \frac{dc}{dt} = f(C) - c$$

$$t / T \sim 1$$

Wave period



- Cells are sensitive in the wave front
- Refractory in the wave back
- Recovered by the next pulse

References:

1. T. Hofer et. al, Appl. Math. Lett. (7), 1-5, 1994
2. R.E. Goldstein, Phys. Rev. Lett.(77), 775-778, 1996

Chemotaxis-driven Linear Instability (1)

Keller & Segel, 1971: cells migrate in a self-imposed field of chemoattractant

$$\frac{\partial n}{\partial t} = -\frac{\partial}{\partial x} \left(-\mathbf{m} \frac{\partial n}{\partial x} + \mathbf{c} n \frac{\partial c}{\partial x} \right), \quad \frac{\partial n}{\partial x} \Big|_{0,L} = 0$$

$$\frac{\partial c}{\partial t} = -\frac{\partial}{\partial x} \left(-D \frac{\partial c}{\partial x} \right) + fn - kc, \quad \frac{\partial c}{\partial x} \Big|_{0,L} = 0$$

$$s.s.: \bar{n} = N/L, \bar{c} = \bar{n}f/k$$

$$n(x,t) = \bar{n} + n'(x,t)$$

$$c(x,t) = \bar{c} + c'(x,t)$$

Linearized equations:

Solution:

$$\frac{\partial n'}{\partial t} = \mathbf{m} \frac{\partial^2 n'}{\partial x^2} - \mathbf{c} \bar{n} \frac{\partial^2 c'}{\partial x^2}$$

$$\frac{\partial c'}{\partial t} = D \frac{\partial^2 c'}{\partial x^2} + f n' - kc'$$

$$\begin{pmatrix} n'(x,t) \\ c'(x,t) \end{pmatrix} = \sum_{i=1}^{\infty} \begin{pmatrix} A_i \\ B_i \end{pmatrix} \cos(q_i x) \exp(\mathbf{I}_i t)$$

← why $i \neq 0$?

Linear instability of uniform state: $\mathbf{I}_i > 0$

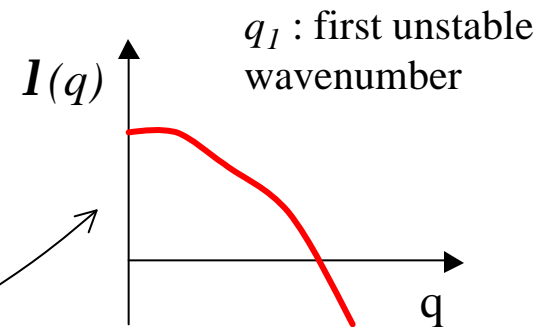
Keller-Segel (2)

For every wavenumber q :

$$\begin{bmatrix} 1 + \mathbf{m}q^2 & -\mathbf{c}\bar{n}q^2 \\ -f & 1 + Dq^2 + k \end{bmatrix} \begin{pmatrix} A \\ B \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Nontrivial solutions ($A \neq 0, B \neq 0$) when $\det(M) \neq 0$

$I_{1,2}$ satisfy $(1 + \mathbf{m}q^2)(1 + Dq^2 + k) - \mathbf{c}\bar{n}q^2 f = 0$



Condition for instability y : $\mathbf{m}(Dq^2 + k) < \mathbf{c}\bar{n}f$

Using the B.C. :

$$\mathbf{m} \left[\frac{D(\pi i)^2}{L^2} + k \right] < \mathbf{c}\bar{n}f$$

Interpretation :

- 1) small \mathbf{m}, D, k, i
- 2) large L
- 3) large \mathbf{c}, \bar{n}, f

This is just linear analysis ...

Keller-Segel (3)

- Instability is promoted by

low random motility & chemoattractant degradation
high chemotactic sensitivity, secretion rate, cell density

- Problems

no saturating effect: $\lim_{t \rightarrow \infty} n(x, t) = \mathbf{d}(x)$

instability does not appear to involve linear mechanism

mechanism is more complicated

References:

1. E.F. Keller and L.A. Segel, J. theor. Biol. (26), 399-415, 1970
2. T. Hillen and K. Painter, Adv. Appl. Math. (26), 280-315, 2001