## Cellular Motility



David Rogers at Vanderbilt University.

## Course 6: Chemotaxis 2 - Eukaryotes

Thomas Lecuit
chaire: Dynamiques du vivant


COLLE GE DE FRANCE
$1530-$

## Two general classes of Mechanisms

- Temporal mechanism: comparison of chemoattractant at different positions and memory.
- Spatial mechanism: comparison of chemoattractant concentration along cell length


## Conclusions - Bacterial chemotaxis



- Biased random walk in a spatial gradient
- Temporal gradient sensing
- Memory (short term)


Key properties of chemotactic network

- Sensitivity - Gain : output/input ratio
- Adaptation: reset after input
- High amplitude range


## Chemical guidance - Chemotaxis

## 1. Bacterial chemotaxis

2. Eukaryote chemotaxis

- Directional sensing
- Polarity



David Rogers at Vanderbilt University.
Determinism

P. Maiuri, JF. Rupprecht, et al. Cell 161, 374-386 (2015)

Stochasticity - Persistence

## Chemical guidance of cell motility

Key features of chemotaxis:

- Specificity
- Cell surface sensing (receptors)
- Sensitivity to ratio (gradient) but not difference in concentration of attractant

How can cells respond to a chemoattractant gradient?

## Problems:

Bacteria can go up an an exponential gradient, over 20 mm at mM range of concentration. For a $2 \mu \mathrm{~m}$ cell to detect such a gradient, they would need to detect $0.0001 \%$ difference on both ends

Eukaryotic cell, e.g. Dictyostelium, maximal chemotactic response between $\sim 10$ and 100 nM , corresponding to relative spatial gradients of $\sim 15-30 \%$ across a $10 \mu \mathrm{~m}$ cell (see later).

## Chemotaxis of immune cells



Ilyia Metchnikoff (1845-1905)
Metchnikoff observed in Sicily (Messina) starfish larvae, noticed motile cells and hypothesized that this might underly response to external agents.

Used rose thorns under larval skin and observed leukocyte chemotaxis and phagocytosis


Metchnikoff's drawing of phagocytes at a site of inflammation (induced by silver nitrate) in the caudal fin of a Triton embryo.

Metchnikoff, E. Lectures on the Comparative Pathology of Inflammation. (Reprinted by Dover, New York, 1968).
Cited in: G. A. Dunn and G. E. Jones Nature Reviews Mol. Cell Biol. 5:667-672 (2004)

## Chemotaxis of immune cells

## Pifysiological Reviews <br> YOL. 211 <br> M:LT. J Hd <br> No. 3 <br> WENUTGXIF IN LETEOCYTE <br>  <br> 

- Leukocytes chemotaxis: attracted by bacteria, phagocytose bacteria
- Chemotaxis index: net path/total path 0.61-0.97, mean 0.85, s.e.m. 0.01.
- Leukocytes paths alternate runs and turns
- Chemotactic and non-chemotactic cells do not have different velocities
- Cells that do not perform chemotaxis have a higher probability of stopping and turning


McCutcheon, M., Physiol. Rev., 20, 319 (1946).

Leukocyte chasing a bacterium (Staphilococcus aureus)
David Rogers at Vanderbilt University.
https://www.youtube.com/watch?v=I_xh-bkiv_c

## Spatial sensing of chemical gradient

## Mechanisms of sensing chemical gradients by polymorphonuclear leukocytes

As the cells tend to move in straight lines for short periods of time, the traced paths could be broken into segments of straight movement (that is movements having less than a $10^{\circ}$ change in direction which were separated by relatively abrupt turns of varying magnitude).

- Leukocytes (PMN - granulocyte) alternate runs and turns

$65 \%$ of segments are oriented within $30^{\circ}$ of most direct path to the source
- Run length statistics: no evidence of bias (unlike Bacteria such as E. coli)
-Distance between turns (length of runs) is independent of direction with respect to the source of attractant (coeff corr. $-0.169, \mathrm{~N}=68 \mathrm{P}>0.1$ ).
- Turns angle statistics: evidence of directional bias (unlike E. coli)
- angle of turn was correlated in magnitude with angle of run prior to run with respect to direct path to
attractant (coeff corr. 0.35, N=68 P $<0.0 \mathrm{I}$ )
- when angle prior to turn was $>30^{\circ}$ with respect to direct path to the source of attractant, the next direction was towards source ( $\mathrm{P}<0.00 \mathrm{I}$ )
The angle bias is not compatible with a simple temporal comparison at two time points (unless with memory of directions)
Consistent with a spatial model of chemotaxis
Consistent with this, cells extend pseudopode in direction of source with higher probabilities before they move


## Chemotaxis of immune cells

- Neutrophiles are attracted to wound sites by a gradient of peroxyde:


Niethammer, P., Grabher, C., Look, A. T. \& Mitchison, T. J. Nature 459, 996-999 (2009)

## Chemotaxis of immune cells

- Neutrophiles are attracted to wound sites by a gradient of peroxyde:

$\mathrm{H}_{2} \mathrm{O}_{2}$ sensor


## Average arrival of firs




9 min after $\quad 35$ min after wounding wounding wounding

Niethammer, P., Grabher, C., Look, A. T. \& Mitchison, T. J. Nature 459, 996-999 (2009)


Kinetics of neutrophile (magenta)-macrophage (green) interactions at wounds in zebrafish
S. Tauzin et al. J Cell Biol (2014) 207 (5): 589-598.

## Chemotaxis of Neutrophils

## Cells respond to a gradient of bacterial peptides

$N$-formyl oligopeptides: produced and released by bacteria
formyl-methionyl-leucine-phenylalanine (fMLP) induces polarization and motility of Neutrophils

fMLP in micropipette
O. Weiner, J. Sedat and H. Bourne
D.A. Bloes et al. Nature Reviews Microbiology (2014) doi:10.1038/nrmicro3390

## Chemotaxis induces cell polarization

## Neutrophils


D.A. Bloes et al. Nature Reviews Microbiology (2014) doi:10.1038/nrmicro3390

## Chemotaxis of the slime mold Dictyostelium

Individual Dictyostelium discoideum (Mycetozoa) cells are attracted by bacteria (folic acid)
During swarming induced by starvation, cAMP is released by cells that attract neighboring cells
Waves (period T $\sim 6 \mathrm{~min}$ ) of cAMP induce wave of cell polarization and motility


Life cycle


https://www.youtube.com/watch?v=oYRF7BaaaJY
MPI Dynamics and Self-Organisation (Goettingen) - Bodenschatz and Westendorf labs


Thomas Gregor lab - Institut Pasteur/ Princeton Univ 13

## Chemotaxis of the slime mold Dictyostelium

## Cells respond to a gradient of cAMP



Carole Parent lab at the University of Michigan Life Sciences Institute.
Dictyostelium discoideum cells are attracted by cAMP released in a gradient from a pipette

## Chemotaxis induces cell polarization



## Chemotaxis induces cell polarization

## Dictyostelium



AR. Kimmel and CA. Parent Science 300:1525-1527 (2003)

## The chemotactic response pathways



## Cell polarization



## Cell polarity affects sensitivity to chemical gradient

- Cells exhibit a behavioral polarity, i.e. their capacity to respond to a chemical cue, which reflects/depends on their morphological polarity/locomotion
I. Cell persistence in moving cells with uniform chemoattractant:
-Cell protrusions form near preexisting ones


2. Increasing the concentration of chemoattractant increases the production of protrusion at the front (but the frequency of turn does not change)


Resumption of polarity after cell pauses:


## Cell polarity affects cells sensitivity to chemical gradient

- Cell exhibit a behavioral polarity, their capacity to respond to a chemical cue, which reflects/depends on their morphological polarity/locomotion

3. Cells U-turn when an opposite gradient is formed, instead of formation of pseudopod at the back

The trailing edge (uropod) is insensitive to chemoattractant


Migrating neutrophils (PMNs) exhibit persistent motility, and their morphological polarity (eg. front-back cytoskeletal polarity for instance, contractile at the back and protrusive at the front) biases their capacity to repolarise in response to external chemical cues.


## Cell polarity affects cells sensitivity to chemical gradient

- Insensitivity of the Neutrophils' trailing edge to chemoattractant is prevented by inhibition of Myosinll activation



## Self-organizing cell polarity

- Actin cytoskeleton assemblies (contractile and protrusive networks) are not simple readouts of cell polarity, but feedback on cell polarity and sensitivity to chemoattractant
- Amplification of signal detection and processing


Mechanochemical Turing like instability

Local excitation (positive feedback loop)

Backness
Long range inhibition

## Self-organizing cell polarity

## Oscillatory actin dynamics at the leading edge

- Actin polymerization at the leading edge is dynamic in space and time (waves and oscillations)
- This is a fast oscillator ( $5-10$ s period)



## Self-organizing cell polarity

## Cytoskeletal Oscillatory Network - Fast



Fast oscillator underlies small undulations to the cell boundaries and behaves as an idling motor system

## Excitability: Bistability, Excitability and Oscillations.



$D_{A}<D_{B}$

x

x

Temporal instability

[A]

t

x

Self-organization in nonequilibrium chemical systems:

- activator auto-activation
- inhibitor induction


## Decay rate:

the decay rate of inhibitor must be lower than that other activator ( $r_{b}<r_{a}$ )

Existence of refractory period: time needed to clear inhibitor (or to re-synthesise depleted substrate)
(Note: the shells are a formal analogy, patterns are not driven by chemical instability, but by neural network Turing instabilities)

Hans Meinhardt. The algorithmic beauty of sea shells. Ed. Springer-Verlag (2009)
Boettiger A, Ermentrout B, Oster G. The neural origins of shell structure and pattern in aquatic mollusks. PNAS. 106(16):6837-42 (2009)

## Excitability: Bistability, Excitability and Oscillations.



- FitzHugh-Nagumo Model :
(heuristic model adapted for study of action potential)

$$
\begin{array}{l|l}
\frac{d A}{d t}=A-A^{3}-B & \text { Fast reaction } \\
\frac{d B}{d t}=\varepsilon(A-x B+y) & \text { Slow reaction } \\
(\varepsilon \ll 1)
\end{array}
$$


(nullclines in AB plane: steady state response of $A$ to $B$ and of $B$ to $A$ )

## Excitability: Bistability, Excitability and Oscillations.

## Spatial temporal patterns of activity: Trigger waves

- Spatial coupling by diffusion - Synchronization

$$
\begin{aligned}
& \frac{\partial A}{\partial t}=D \frac{\partial^{2} A}{\partial x^{2}}+A-A^{3}-B \\
& \frac{\partial B}{\partial t}=D \frac{\partial^{2} B}{\partial x^{2}}+\varepsilon(A-x B+y) \\
& \text { diffusion } \\
& \text { reaction }
\end{aligned}
$$



## Excitability: Bistability, Excitability and Oscillations.

## Spatial temporal patterns of activity: Trigger waves

- Diffusion is a mechanism for crossing the threshold in space

- a low $A$ value state will increase its $A$ concentration due to diffusion of nearby high A state but will relax back
- Higher diffusion increases rate of local propagation but makes it more transient

- The presence of a reaction captures suprathreshold state and move to high A state
- This depends on the relative time scale of diffusion and reaction
- Speed: $s=2 \sqrt{\frac{D}{\tau}}$ doubling time of + feedback
L. Gelens, G.A. Anderson and James Ferrel. MBoC 25:3486-3493. 2014


## Self-organizing cell polarity

## Cytoskeletal Oscillatory Network - Fast



## Motility

Fast oscillatory


Stochastic



Fast oscillator underlies small undulations to the cell boundaries and behaves as an idling motor system

## Self-organizing cell polarity

## Slow excitable transduction network of polarisation

- Signatures of excitable system:
- Wave dynamics of directional sensing network (in absence of actin)


$$
(n=7)
$$




Pi3K sensor

- Low intensity stimulation gives maximal response



## Self-organizing cell polarity

## Coupling between two excitable networks




STEN and CON are correlated in cell protrusions but not In small fluctuations of the plasma membrane

## Self-organizing cell polarity

## Coupling between two excitable networks



Chemical, mechanical, electrical, noise


- Integrated inputs, external chemical signals and internal noise, bias components of STEN and cause STEN to pass the threshold of activation

Directional sensing
Motility

- CON is entrained to a larger more stable state associated with actin polymerization and cell protrusion

- Activation of the STEN by stochastic noise underlies random cell motility
- A cue that affects the threshold for excitation differently on opposite sides of a cell would be expected to guide cell migration. (e.g., inhibitor of $\mathrm{PI}(3) \mathrm{K}$ can act as chemorepellent)


## Excitability, memory and persistence of motility

Cells are exposed to a dynamic gradient of cAMP

- Why don't cells move backward in the back of the wave?
- This is not by temporal gradient measurement but by a mechanism of cell memory

- When exposed to a dynamic gradient in microfluidic chamber, Dictyostelium cells show 9 min persistence of motility and polarity: the chemotactic index is lower but positive in the back of the wave up to a wave period T~10 min. This persistence of migration is indicative of a memory with a timescale similar to the Dictyostelium wave period ( $\sim 6 \mathrm{~min}$ )
- Cell memory and persistence can be explained by refractory period intrinsic to excitable dynamics of polarity network: time scale of degradation of inhibitor.
- Cells are persistent when exposed to rapid and abrupt changes in chemoattractant, but depolarize in presence of slow, shallow gradient changes

mean concentration of 50 nM with a relative spatial gradient of $17 \%$ across $10 \mu \mathrm{~m}$ chemotactic index $\quad \mathrm{CI}=V_{x} / V$,

> Raf-GFP polarized recruitment is maintained when cells in a $0-100 \mathrm{nM}$ gradient are exposed to a uniform concentration of cAMP equal or greater than the concentration they had in the gradient


## Persistent random walk underlies eukaryotic chemotaxis

- Chemotaxis in Eukaryotic cells change the lifetime of persistent motility
- The default state is a random walk
- Chemotaxis causes a biased, persistent random walk
—Dictyostelium: In uniform chemoattractant, cells undergo a random walk with a persistence time of $\sim 3-$ 10 min
-In absence of chemoattractant, bone marrow derived dendritic cells (BMDCs) have variable persistence time
- What underlies the variable persistence time in motility?



## What underlies variable persistence?

## Persistence is coupled to cell speed

In all conditions, ID, 2D, 3D, on adhesive or non-adhesive substrates, cells consistently show the same relation between persistence time and speed. It applies to mesenchymal and amoeboid motility




Persistence time $=\Delta t$ Mean instantaneous speed $=\Delta x / \Delta t$



Bone Marrow derived Dendritic Cells (BMDC) Retinal pigment epithelial cells (Rpe)

## Cell persistence is coupled to actin flow speed

- The most commun feature of adhesive and non-adhesive motility is the existence of an actin retrograde flow driven by actin polymerization at the front and depolymerization at the back mediated by Myosinll contractility.

- Frictional or adhesion based coupling to the substrate enables forward cell movement
- Experimental perturbations lead to different retrograde actin flow at constant/similar cell motility
- Across conditions, the persistence time of motility scales exponentially with actin flow speed.



$$
\tau=A^{\prime} e^{\lambda^{\prime} V}
$$

- Cell motility $v$ and actin flow speed $V$ are linearly proportional, but across conditions the constant of proportionality is different.


## Cell persistence is coupled to actin flow speed

## Advection of flow regulators

- actin flow reinforces cell polarity by enhancing the asymmetry of polarity cues by advection

$$
\partial_{t} c(x, t)-\partial_{x}[\tilde{V} c(x, t)]=\tilde{D} \partial_{x}^{2} c(x, t)+\partial_{x} \zeta_{c},
$$

$$
\begin{aligned}
& \tilde{V}=V k_{\text {on }} /\left(k_{\text {on }}+k_{\text {off }}\right) \\
& \tilde{D}=D k_{\text {off }} /\left(k_{\text {on }}+k_{\text {off }}\right)
\end{aligned}
$$



At steady state:

$$
\bar{C}_{V}(x)=C e^{-\tilde{V} x / \tilde{D}}
$$




- Polarization lifetime is enhanced by retrograde flow And requires Myosin2 activity




## A model of flow dynamics predicts 3 regimes of motility

- The model which predicts the probability distribution of velocity, yields the exponential relation between velocity and polarity persistence time
- Prediction of 3 regimes based on 2 key parameters:
—the coupling strength $\beta_{c}$ between asymmetry of regulator $c(t)$ and flow speed $V . \quad V^{*}=\beta\left(c^{*}(0, t)-c^{*}(L, t)\right)$,
-the concentration of cues $C_{s}$ above which activity is saturated

$$
c^{*}(x, t)=\frac{C^{n}(x, t)}{C_{s}^{n}+C^{n}(x, t)} .
$$

- Experimental data are well fit by the model using these 2 free parameters:



## Deterministic vs Self-organised Guidance

How do cells navigate over long range in situ (development, immune system, cancer)?
>> Interaction between cells and environment:
cells generate/modify their own guidance cue through such interactions
The structure of the environment matters (eg. confinement)

## Decoding the chemical environment

## Precision

## Deterministic vs Stochastic decoding

## Deterministic vs Stochastic Guidance

## Informed choice model


compass choice model


Dictyostelium discoideum

- Cells make many pseudopods are regular intervals and select the better ones up the gradient
- The pseudopod that are better aligned with gradient have an increased survival (bias for survival)


Informed choice model
Reinforcement of most up gradient protrusion



## Why do cells often breakdown chemoattractants?

Chemotactic cells have a variety of mechanisms for depleting attractants:

- Receptor-ligand endocytosis
- Decoy receptors: the CXCR7 receptor competes with CXCR4 for binding to the ligand SDFI in the zebrafish lateral line. CXCR7 does not signal and traps the ligand.

Dona E, et al. and D. Gilmour. Nature 2013, 503:285-289.

- Cell surface enzymes that degrade attractants

Melanoma metastasis: lysophosphatidic acid (LPA) degradation in melanoma cells by Lipid Phosphatases
O. Susanto et al., J. Cell Sci. 130, 3455-3466 (2017)

Dictyostelium: cAMP chemoattractant is degraded by the extracellular phosphodiesteras PdsA
G. L. Garcia, et al and C. A. Parent, Mol. Biol. Cell 20, 3295-3304|(2009)

- Attractant breakdown leads to steep local, self-generated gradients


Chemotaxis occurs most efficiently when the ligand is present at close to the receptor's dissociation constant ( $\mathrm{K}_{\mathrm{d}}$ ). At substantially higher concentrations, the difference in receptor occupancy between the front and back of responding cells drops as the receptors become saturated, and chemotaxis becomes inefficient. Thus high attractant concentrations are incompatible with imposed gradients.

Self-generated gradients work best when the chemoattractant concentration is saturating, and cells break it down to a such a level that their can resolve the steepness locally.
in Dictyostelium, strong chemotactic responses to cAMP occur with $10 \mu \mathrm{M}$ of attractant, and the receptor $K_{d}$ is in the $n M$ range
L. Tweedy, O. Susanto and R. Insall. Current Opinion in Cell Biology 42:46-51 (2016)

## Self-generated gradients of chemotaxis

- Cells produce an activity that degrades the chemoattractant
- A gradient of attractant is formed at the edge of cell cluster, that steers cells forward leaving behind no attractant where cells have random motility.
- A front wave emerges that self-propagates
- Self-reinforcing process: if a few cells go pass the front, they will adopt random motility because they can't produce a new gradient of chemoattractant which is also saturating. If attractant diffuses behind the front, it will attract more cells


C


A


## Self-generated versus passive chemotactic gradients

## More robust migration in self-generated gradients

- Simulations

Imposed gradient:

- chemotaxis is not efficient because it is too shallow

Self-generated gradient:

- chemotaxis works best at near saturation of receptor binding, allowing formation of a steep local gradient

Persistent, biased random walk of cells in a bath with diffusion of attractant


- Experiments

Dictyostelium cells in a non degradable attractant Sp-cAMPS
cAMP gradient

$\dot{2} .5 \mu \mathrm{McAMP}$

## Pathfinding in a maze using self-generated gradients

## Cells detect and avoid dead ends

- When cells split between two free ends at a Tjunction, they partition between the two channels with greater precision than expected by chance.
If a cell enters a channel, the attractant will be degraded and this will feedback on the cells behind. So any imbalance in cell distribution will be corrected.
In other words cells influence their followers

- Cells enter T-junctions and are given a choice between entering a Free or a Dead end.
- In a dead end, the chemoattractant is rapidly degraded and follower cells avoid the dead end.
This results in a statistical bias in the
 distribution.
This bias increases at the dead end is shorter



# Pathfinding in a maze using self-generated gradients 

## Cells detect and avoid dead ends

Simulations


Dictyostelium cells


Pancreatic cancer cells


Pancreatic cancer cells

## Pathfinding in a maze using self-generated gradients

## Limits to the detection of dead ends

- More distant dead ends increase the error rate because the clearance of attractant is less efficient



## Pathfinding in a maze using self-generated gradients

## Limits to the detection of dead ends: cell speed and diffusion

- Longer dead ends leads to less accurate decision
- Speed: lower speed of cells leads to more accurate decision: cells have more time to clear attractant from a dead end
- Diffusion:At higher diffusion constant of chemoattractant, cells can more efficiently lower/ deplete attractant from a dead end
- A zero diffusion, there is no information to decode (cells cannot decode what is ahead of them)
- A low diffusion, short, parallel dead ends lure cells in the wrong direction.



## Self-organised guidance

## Reinforcement of guidance landscape by cells: spatial memory akin to stigmergia



- Conditioned substrate enhances cell motility
- Oscillatory migration in ID channels

Conditioned substrate



## General Conclusion

I. Biased random walk characterizes chemotaxis across scales
2. Two different mechanisms of gradient sensing:

- Spatial mechanism: comparison of chemoattractant concentration along cell length. This requires often (always?) self-generated gradients by depletion of activity. More robust and long range.
- Temporal mechanism: comparison of chemoattractant at different positions and requires memory.
3.Adaptation and memory manifest in different ways across scales
- Temporal gradient sensing in prokaryotes
- Persistence of motility in eukaryotes


## Mechanical guidance

I. Substrate interactions in 2D and 3D are inherently mechanical
2. The stiffness, topography, etc of the environment can affect motility
3. Cells also decode the mechanical properties of their environment

