Anomalous dynamics of cell migration

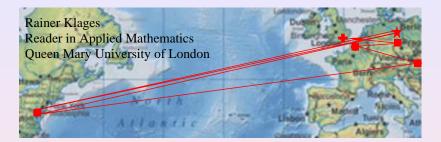
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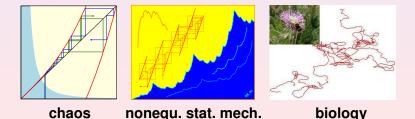
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Multiscale Analysis and Modeling of Collective Migration in Biological Systems, ZIF Bielefeld, 12 October 2017



My own scientific foraging





Cell migration	Experimental results	Theoretical modeling	Lévy motion	Fluctuation relations	Conclusions
Outline					

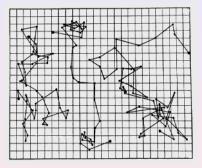
single cell migration:

- **Experimental results:** statistical data analysis
- Theoretical modeling: anomalous dynamics and its biophysical interpretation
- S Lévy motion: what is it, and search optimization for cells
- Fluctuation relations: experimental test of a theoretical model

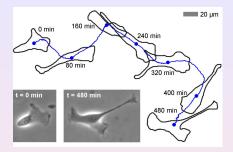
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Brownian motion of migrating cells?

Brownian motion



Perrin (1913) three colloidal particles, positions joined by straight lines



Dieterich et al. (2008) single biological cell crawling on a substrate

Brownian motion?

conflicting results: **yes:** Dunn, Brown (1987) **no:** Hartmann et al. (1994) Cell migration

Experimental results

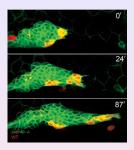
eoretical model

Lévy motior

Fluctuation relatio

Conclusions

Why and how do cells migrate?



example:

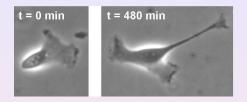
motion of the primordium in developing zebrafish; collective cell migration Lecaudey et al. (2008)



either via **membrane protrusions and retractions** or **blebbing** here: no microscopic details How does a cell migrate *as a whole* in terms of a **stochastic diffusion process**?

Our cell types and some typical scales

Experimental results



- renal epithelial MDCK-F (Madin-Darby canine kidney) cells; two types: wildtype (NHE⁺) and NHE-deficient (NHE⁻)
- observed up to 1000 minutes: here *no* limit $t \to \infty$!
- cell diameter 20-50μm; mean velocity ~ 1μm/min; lamellipodial dynamics ~ seconds

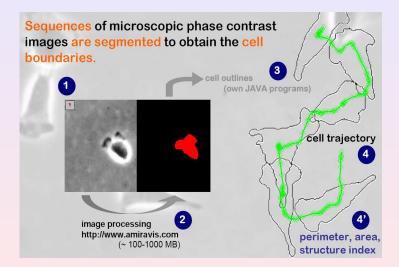
movies: NHE+: t=210min, dt=3min

NHE-: t=171min, dt=1min

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Measuring cell migration





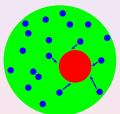
'Newton's law of stochastic physics':

 $\dot{\mathbf{v}} = -\kappa \mathbf{v} + \sqrt{\zeta} \boldsymbol{\xi}(t)$

Langevin equation (1908)

for a tracer particle of velocity **v** immersed in a fluid

force decomposed into viscous damping and random kicks of surrounding particles



Application to cell migration?

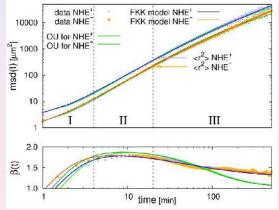
but: cell migration is active motion, not passively driven!

cf. active Brownian particles (e.g., Romanczuk et al., 2012)

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Mean square displacement

• $msd(t) := \langle [\mathbf{x}(t) - \mathbf{x}(0)]^2 \rangle \sim t^{\beta}$ with $\beta \to 2 \ (t \to 0)$ and $\beta \to 1 \ (t \to \infty)$ for Brownian motion; $\beta(t) = d \ln msd(t)/d \ln t$

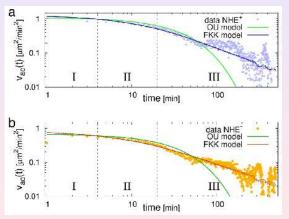


anomalous diffusion if $\beta \neq 1$ ($t \rightarrow \infty$); here: superdiffusion

Velocity autocorrelation function

Experimental results

- $v_{ac}(t) := \langle \mathbf{v}(t) \cdot \mathbf{v}(0) \rangle \sim \exp(-\kappa t)$ for Brownian motion
- fits with same parameter values as msd(t)



crossover from stretched exponential to power law

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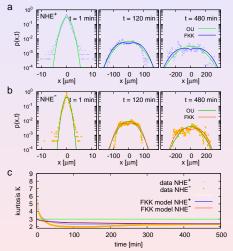
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Position distribution function

• $P(x, t) \rightarrow \text{Gaussian}$ $(t \rightarrow \infty)$ and kurtosis $\kappa(t) := \frac{\langle x^4(t) \rangle}{\langle x^2(t) \rangle^2} \rightarrow 3 \ (t \rightarrow \infty)$ for Brownian motion (green lines, in 1d)

• other solid lines: fits from our model; parameter values as before

note: model needs to be amended to explain short-time distributions



crossover from peaked to broad non-Gaussian distributions

Cell migration	Experimental results	Theoretical modeling ●○○○	Lévy motion	Fluctuation relations	Conclusions
The mo	odel				

• Fractional Klein-Kramers equation (Barkai, Silbey, 2000):

$$\frac{\partial P}{\partial t} = -\frac{\partial}{\partial x} \left[vP \right] + \frac{\partial^{1-\alpha}}{\partial t^{1-\alpha}} \kappa \left[\frac{\partial}{\partial v} v + v_{th}^2 \frac{\partial^2}{\partial v^2} \right] P$$

with probability distribution P = P(x, v, t), damping term κ , thermal velocity $v_{th}^2 = kT/m$ and Riemann-Liouville fractional (generalized ordinary) derivative of order $1 - \alpha$ for $\alpha = 1$ Langevin's theory of Brownian motion recovered

• analytical solutions for msd(t) and P(x, t) can be obtained in terms of special functions (Barkai, Silbey, 2000; Schneider, Wyss, 1989)

• 4 fit parameters v_{th} , α , κ (plus another one for short-time dynamics)

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What is a fractional derivative?

letter from Leibniz to L'Hôpital (1695): $\frac{d^{1/2}}{dx^{1/2}} = ?$

one way to proceed: we know that for integer m, n

$$\frac{d^m}{dx^m}x^n = \frac{n!}{(n-m)!}x^{n-m} = \frac{\Gamma(n+1)}{\Gamma(n-m+1)}x^{n-m};$$

assume that this also holds for m = 1/2, n = 1

$$\Rightarrow \quad \frac{d^{1/2}}{dx^{1/2}}x = \frac{2}{\sqrt{\pi}}x^{1/2}$$

extension leads to the *Riemann-Liouville fractional derivative*, which yields power laws in Fourier (Laplace) space:

$$rac{d^{\gamma}}{dx^{\gamma}}F(x) \leftrightarrow (ik)^{\gamma}\tilde{F}(k)$$

∃ well-developed mathematical theory of fractional calculus, see Sokolov, Klafter, Blumen, Phys. Today 2002 for a short intro



Physical meaning of the fractional derivative?

• the generalized Langevin equation

$$\dot{\mathbf{v}} + \int_0^t dt' \,\kappa(t-t')\mathbf{v}(t') = \sqrt{\zeta}\,\xi(t)$$

e.g., Mori, Kubo (1965/66)

with time-dependent friction coefficient $\kappa(t) \sim t^{-\alpha}$ generates the same msd and v_{ac} as the fractional Klein-Kramers eq.

• fractional derivatives model power law correlations:

 $\frac{\partial^{\gamma} P}{\partial t^{\gamma}} := \frac{\partial^{m}}{\partial t^{m}} \left[\frac{1}{\Gamma(m-\gamma)} \int_{0}^{t} dt' \frac{P(t')}{(t-t')^{\gamma+1-m}} \right] , \ m-1 \leq \gamma \leq m$

- cell anomalies might originate from **glassy behavior** of the cytoskeleton gel, where power law exponents are conjectured to be universal (Fabry et al., 2003; Kroy et al., 2008)
- anomalous dynamics observed for **many different cell types** by at least 10 different groups



• results show *diffusion for short times slower* than Brownian motion while *long-time motion is faster*: **intermittent dynamics** can minimize search times



Bénichou et al. (2006)

• question about optimal search strategy related to the Lévy flight hypothesis; for cells: see Krummel et al. (2016)

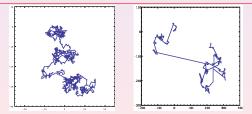


Optimizing the success of random searches

famous article by Viswanathan et al., Nature 401, 911 (1999):

- question about "best statistical strategy to adapt in order to search efficiently for randomly located objects"
- random walk model leads to Lévy flight hypothesis:

Lévy flights provide an *optimal search strategy* for *sparse, randomly distributed, immobile, revisitable targets in unbounded domains*



Brownian motion (left) vs. Lévy flights (right) big debate about the validity of this hypothesis!

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a random walk generating Lévy flights:

 $\begin{array}{l} x_{n+1} = x_n + \ell_n \text{ with steps of length } |\ell_n| = \ell \text{ to the left/right, sign} \\ \text{determined by coin tossing; } \ell_n \text{ drawn from a Lévy } \alpha \text{-stable} \\ \text{distribution} \\ \hline \rho(\ell_n) \sim |\ell_n|^{-1-\alpha} \left(|\ell_n| \gg 1 \right), \ 0 < \alpha < 2 \end{array}$

P. Lévy (1925ff)



• fat tails: larger probability for long jumps than for a Gaussian!

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Properties of Lévy flights in a nutshell

- a Markov process (no memory)
- which obeys a generalized central limit theorem if the Lévy distributions are α-stable (for 0 < α < 2) Gnedenko, Kolmogorov, 1949
- implying that they are scale invariant and thus self-similar
- $\rho(\ell_n)$ has infinite variance

$$\langle \ell_n^2 \rangle = \int_{-\infty}^{\infty} d\ell_n \, \rho(\ell_n) \ell_n^2 = \infty$$

- Lévy flights have arbitrarily large velocities, as $v_n = \ell_n/1$
- position pdf given by the fractional diffusion equation

$$\frac{\partial f(x,t)}{\partial t} = \mathcal{K}_{\alpha} \frac{\partial^{\alpha} f(x,t)}{\partial |x|^{\alpha}}$$

with Riesz fract. derivative $\sim -|k|^{\alpha} f(k, t)$ in Fourier space



cure the problem of infinite moments and velocities by introducing an additional constraint:

a Lévy walker spends a time

 $t_n = \ell_n / v$, |v| = const.

to complete a step; yields finite moments and finite velocities in contrast to Lévy flights

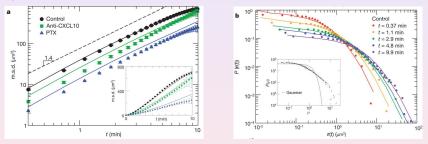
• Lévy walks generate anomalous (super) diffusion:

$$\langle x^2
angle \sim t^\gamma \ (t
ightarrow \infty)$$
 with $\gamma > 1$

see Shlesinger at al., Nature **363**, 31 (1993) for an outline; Zaburdaev et al., RMP **87**, 483 (2015) for details

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- T.H. Harris et al., Nature 486, 545 (2012):
- mean square displacement (for 3 different cell types) and position distribution function for T cells in vivo:



- T cell motility described by a generalized Lévy walk (Zumofen, Klafter, 1995)
- search more efficient than Brownian motion
- pdf not Lévy: how does the result fit to the Lévy hypothesis?

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Motivation: Fluctuation relations

Consider a (classical) particle system evolving from some initial state into a nonequilibrium steady state.

Measure the probability distribution $\rho(\xi_t)$ of *entropy production*

 ξ_t during time t:

$$\ln \frac{\rho(\xi_t)}{\rho(-\xi_t)} = \xi_t$$

Transient Fluctuation Relation (TFR)

Evans, Cohen, Morriss (1993); Gallavotti, Cohen (1995)

why important? of very general validity and

- generalizes the Second Law to small systems in nonequ.
- connection with fluctuation dissipation relations (FDRs)
- can be checked in experiments (Wang et al., 2002)

Anomalous TFR for Gaussian stochastic processes

theory: consider overdamped generalized Langevin equation $\dot{x} = F + \zeta(t)$ with force *F* and Gaussian power-law correlated noise $< \zeta(t)\zeta(t') >_{\tau=t-t'} \sim (\Delta/\tau)^{\beta}$ for $\tau > \Delta$, $\beta > 0$

that is external (i.e., no FDR):

- dynamics can generate **anomalous diffusion**, $\sigma_x^2 \sim t^{2-\beta}$ with $2 - \beta \neq 1 \ (t \to \infty)$
- yields an **anomalous work fluctuation relation**, $\ln \frac{\rho(W_t)}{\rho(-W_t)} = \mathbf{f}_{\beta}(\mathbf{t}) W_t$

A.V.Chechkin, R.K. et al., J.Stat.Mech. L11001 (2012); L03002 (2009) experiments: test this theory for murine neutrophil chemotaxis

Cell migration Experimental results Theoretical modeling Lévy motion Fluctuation relations occorrelations control occorrelation Summary: Anomalous cell migration

- anomalous dynamics: superdiffusion with power law velocity correlations and non-Gaussian position pdfs for long times
- **theoretical model:** coherent mathematical description of experimental data by an anomalous stochastic process
- temporal complexity: different cell dynamics on different time scales
- interpretation: possible biophysical significance of anomalous dynamics for optimizing search; cf. *Lévy flight foraging hypothesis*
- second law-like relation for cell chemotaxis

Conclusions

Cell migration	Experimental results	Theoretical modeling	Lévy motion	Fluctuation relations	Conclusions ○●○
Outlook	Κ				

• single vs. collective cell migration?

single cell motility controls glass and jamming transition Bi et al. (2016), or not Giavazzi et al. (2017)

• significance of *anomalous* diffusion for collective phenomena?

superdiffusion enhances colony formation of stem cells Barbaric et al. (2014);

non-trivial phase transitions in models of *active* Brownian particles Fodor et al. (2016)

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References						

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- R. Klages, Search for food of birds, fish and insects, in: A.Bunde et al. (Eds.), Diffusive Spreading in Nature, Technology and Society (Springer, 2017)
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