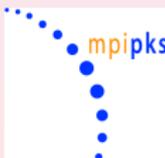


Statistical Physics and Anomalous Dynamics of Foraging

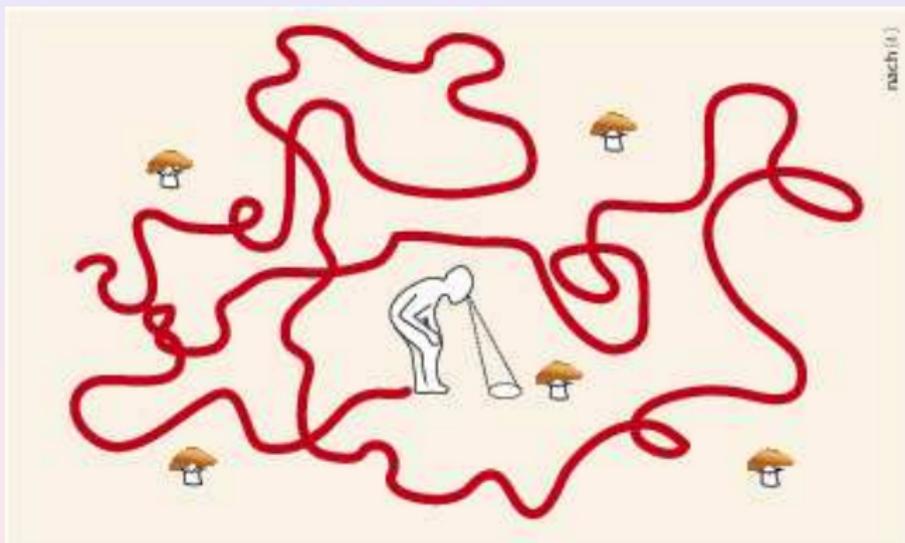
Rainer Klages

(Max Planck Institute for the Physics of Complex Systems, Dresden)
Queen Mary University of London, School of Mathematical Sciences

4th Workshop on Fractional Calculus, probability and
Non-Local Operators: Applications and Recent
Developments; BCAM, 24 November 2016

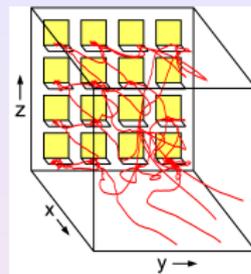
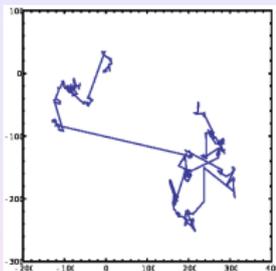


The problem



from: [Chupeau, Nature Physics \(2015\)](#)

Outline of my talk



Theme of this talk:

Can search for food by biological organisms be understood by mathematical modeling?

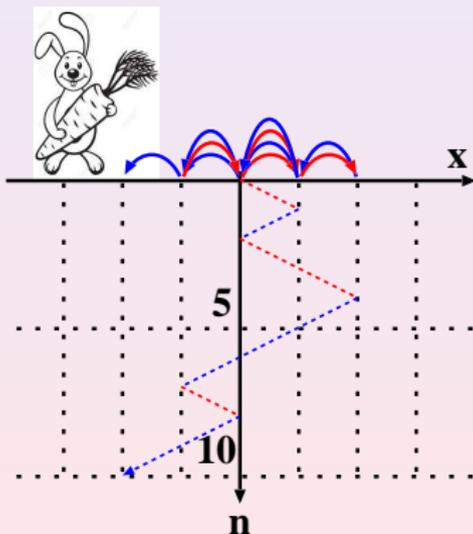
Three parts:

- 1 Lévy flight hypothesis: review
- 2 Biological data: analysis and interpretation
- 3 Stochastic modeling: fractional calculus and non-local operators

A mathematical theory of random migration

Karl Pearson (1906):

model movements of biological organisms by a **random walk** in one dimension: position x_n at discrete time step n



$$x_{n+1} = x_n + \ell_n$$

- here: steps of length $|\ell_n| = \ell$ to the **left/right**; sign determined by **coin tossing**
- **Markov process**: the steps are *uncorrelated*
- generates **Gaussian distributions** for x_n (central limit theorem)

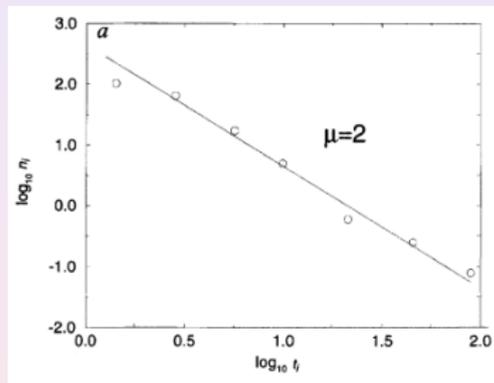
Lévy flight search patterns of wandering albatrosses

famous paper by **Viswanathan et al.**, *Nature* **381**, 413 (1996):

for **albatrosses** foraging in the South Atlantic the flight times were recorded



the histogram of flight times



was fitted by a **Lévy distribution** (power law $\sim t^{-\mu}$)

- may be due to the **food distribution on the ocean surface being scale invariant: Lévy Environmental Hypothesis**

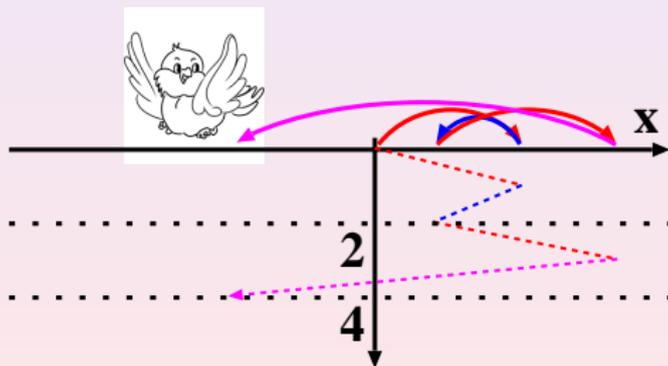
What are Lévy flights?

a random walk generating **Lévy flights**:

$x_{n+1} = x_n + l_n$ with l_n drawn from a **Lévy α -stable distribution**

$$\rho(l_n) \sim |l_n|^{-1-\alpha} (|l_n| \gg 1), \quad 0 < \alpha < 2$$

P. Lévy (1925ff)



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

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 $0 < \alpha < 2$)
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**

$$\frac{\partial f(\mathbf{x}, t)}{\partial t} = K_\alpha \frac{\partial^\alpha f(\mathbf{x}, t)}{\partial |\mathbf{x}|^\alpha}$$

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

Lévy walks

cure the problem of infinite moments and velocities:

- a **Lévy walker** spends a time

$$t_n = \ell_n/v, \quad |v| = \text{const.}$$

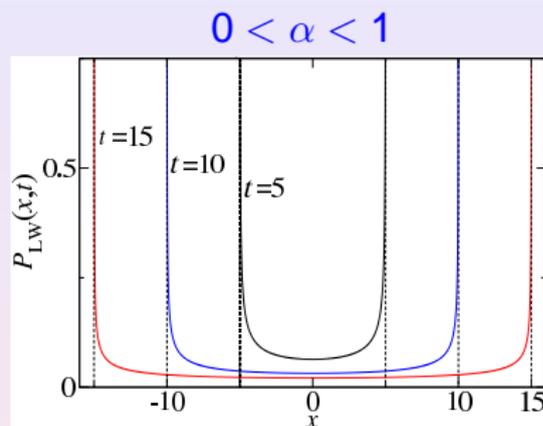
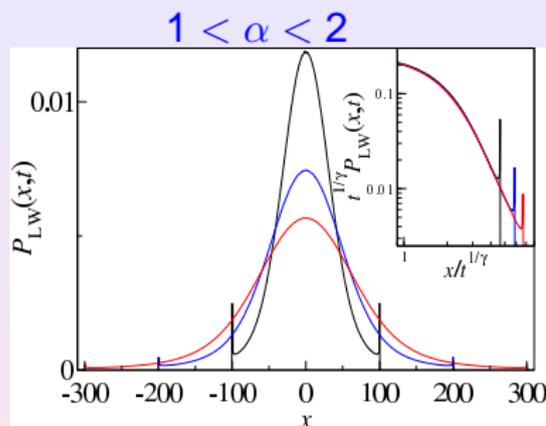
to complete a step; yields **finite moments** and **finite velocities** in contrast to Lévy flights

- Lévy walks generate **anomalous (super) diffusion**:

$$\langle x^2 \rangle \sim t^\gamma \quad (t \rightarrow \infty) \quad \text{with } \gamma > 1$$

see Shlesinger et al., *Nature* **363**, 31 (1993) for an outline;
RK, Radons, Sokolov (Eds.), *Anomalous transport* (Wiley, 2008)

Position distribution functions for Lévy walks



Zaburdaev et al., RMP **87**, 483 (2015)

topic of very recent research:

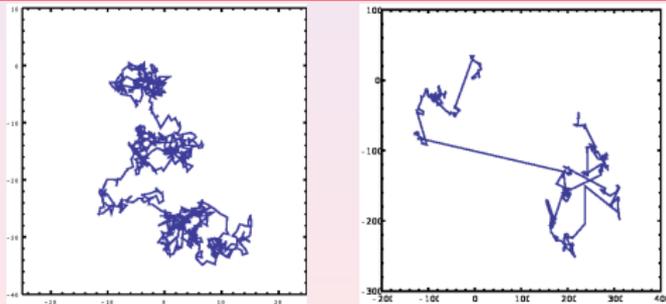
- derivation of an **integrodifferential wave equation** for a Lévy walk: Fedotov, PRE (2016)
- **analytical formulas for densities** of multidimensional Lévy walks: Magdziarz, Zorawik, PRE (2016)

Optimizing the success of random searches

another paper by **Viswanathan et al., Nature 401, 911 (1999)**:

- question posed 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



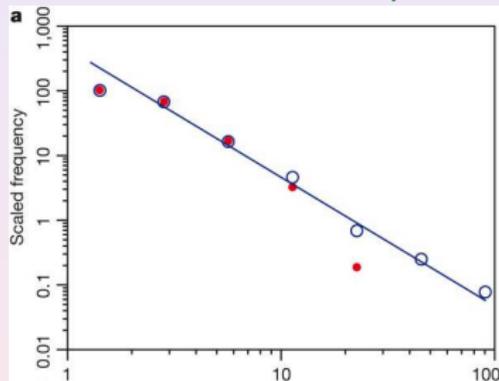
Brownian motion (left) vs. **Lévy flights** (right)

- yields the *second* **Lévy Foraging Hypothesis**

Revisiting Lévy flight search patterns

Edwards et al., Nature **449**, 1044 (2007):

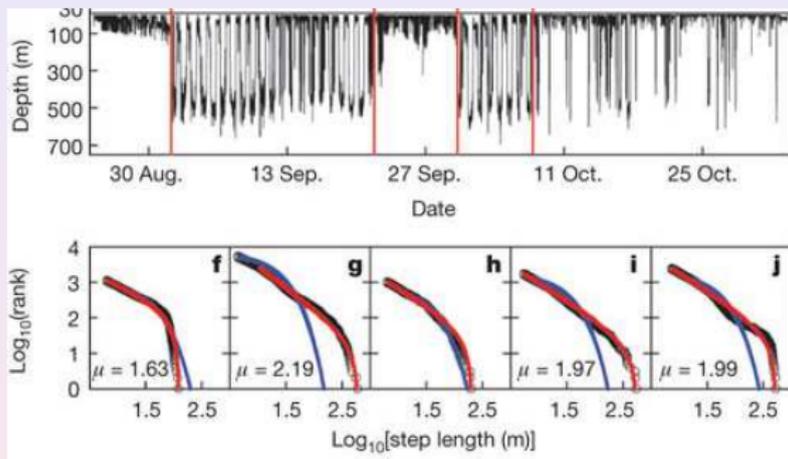
- Viswanathan et al. results revisited by **correcting old data** (Buchanan, Nature **453**, 714, 2008):



- **no Lévy flights:** new, more extensive data suggests (gamma distributed) stochastic process
- but claim that **truncated Lévy flights** fit yet new data
Humphries et al., PNAS **109**, 7169 (2012)

Lévy Paradigm: Look for power law tails in pdfs

Humphries et al., *Nature* **465**, 1066 (2010): blue shark data



blue: exponential; red: truncated power law

- ⊖ velocity pdfs extracted, *not* the jump pdfs of Lévy walks
- ⊕ environment explains Lévy vs. Brownian movement
- ⊖ data averaged over day-night cycle, cf. oscillations

Summary: two different Lévy Flight Hypotheses

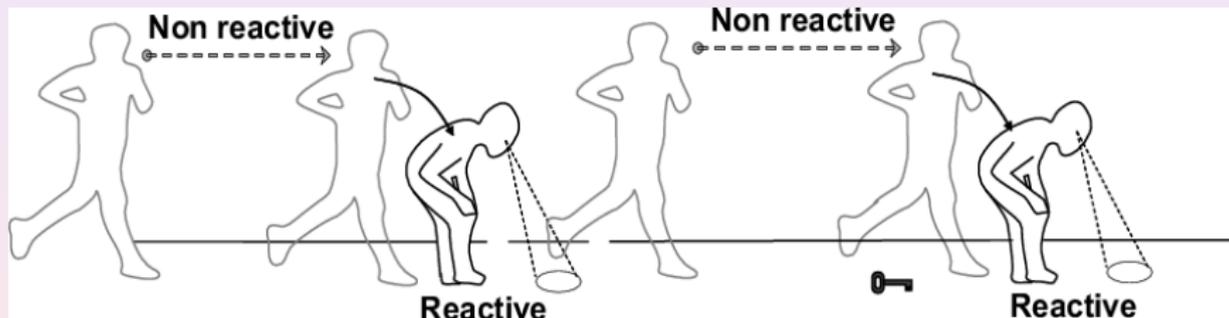
to be published

Bartumeus, Boyer, Chechkin, Giuggioli, RK, Pitchford, Watkins (2015)

An alternative to Lévy flight search strategies

Bénichou et al., Rev. Mod. Phys. **83**, 81 (2011):

- for *non-revisitable targets* **intermittent search strategies** minimize the search time



- popular account of this work in Shlesinger, Nature **443**, 281 (2006): “How to hunt a submarine?”; cf. also protein binding on DNA

Beyond the Lévy Flight Hypothesis

to be published

Bartumeus, Boyer, Chechkin, Giuggioli, RK, Pitchford, Watkins (2015)

In search of a mathematical foraging theory

Summary:

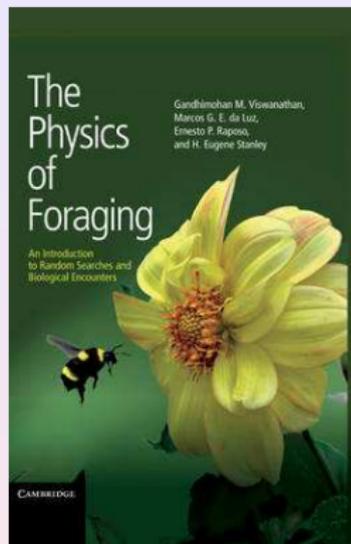
- scale-free Lévy flight **paradigm**
- problems with the **data analysis**
- two **Lévy Flight Hypotheses**:
adaptive and **emergent**
- **intermittent search** as an alternative
- need to go **beyond the Lévy Flight Hypotheses**

Ongoing discussions:

- mussels: **de Jager et al., Science (2011)**
- cells perform Lévy walks: **Harris et al., Nature (2012)** or not:
Dieterich, RK et al., PNAS (2008)

Applications:

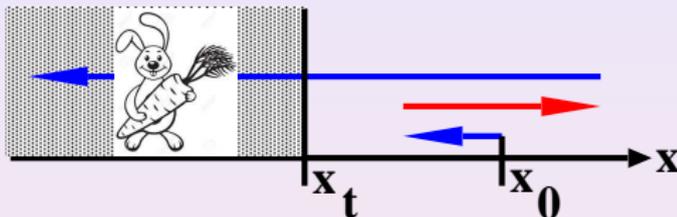
- search algorithms for robots? **Nurzaman et al. (2010)**



Searching for a single target

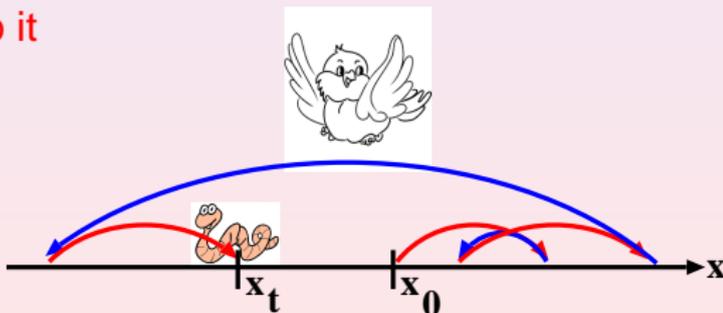
two basic types of foraging (James et al., 2010):

- 1 **cruise forager**: detects a target **while moving**



first passage problem

- 2 **saltatory forager**: only detects a target **when landing on it / next to it**



first arrival problem

First passage and first arrival: solutions

1 Brownian motion:

$$\varrho_{FP}(t) \sim t^{-3/2} \sim \varrho_{FA}(t)$$

Sparre-Andersen Theorem (1954)

2 Lévy flights:

$$\varrho_{FP}(t) \sim t^{-3/2} \text{ (Chechkin et al., 2003; numerics only)}$$

$$\varrho_{FA}(t) = 0 \text{ (} 0 < \alpha \leq 1 \text{)}; \varrho_{FA}(t) \sim t^{-2+1/\alpha} \text{ (} 1 < \alpha < 2 \text{)}$$

(Palyulin et al., 2014)

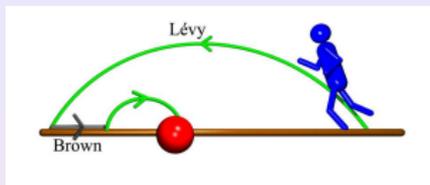
3 Lévy walks:

$$\varrho_{FP}(t) \sim t^{-1-\alpha/2} \text{ (} 0 < \alpha \leq 1 \text{)}; \varrho_{FP}(t) \sim t^{-3/2} \text{ (} 1 < \alpha < 2 \text{)}$$

(numerics: Korabel, Barkai (2011); analytically: Artuso et al., 2014)

$\varrho_{FA}(t)$: **the same as for Lévy flights**, cf. simulations
(Blackburn et al., 2016)

Combined Lévy-Brownian motion search



- intermittency modeled by the **fractional diffusion equation**

$$\frac{\partial f(x, t)}{\partial t} = K_{\alpha} \frac{\partial^{\alpha} f(x, t)}{\partial |x|^{\alpha}} + K_B \frac{\partial^2 f(x, t)}{\partial x^2}$$

with Riesz fractional derivative (see before)

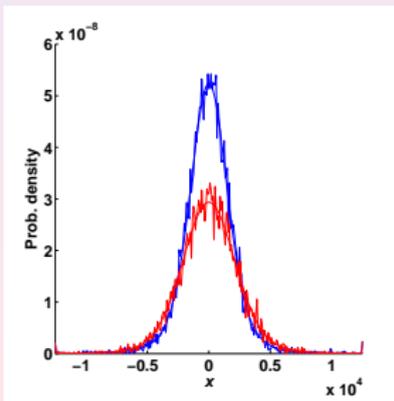
- define **search reliability** by cumulative probability P of reaching a target: $P = \lim_{s \rightarrow 0} \int_0^{\infty} \varrho_{FA}(t) \exp(-st) dt$
- **result: Brownian motion regularizes Lévy search**,
 $0 < P < 1$ for $0 < \alpha < 1$
- define and calculate **search efficiency** by

$$\varepsilon = \langle \text{visited \# targets} / \text{\# steps} \rangle \simeq \langle 1/t \rangle = \int_0^{\infty} \varrho_{FA}(s) ds$$

Palyulin et al., JPA, 2016

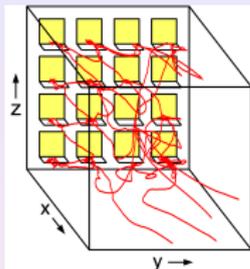
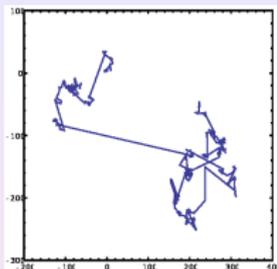
Deriving Lévy-Brownian motion from a Lévy walk

- model **short-range n -dim correlated Lévy walks** by a **fractional Klein-Kramers equation** (Friedrich et al., 2006)
- for $1 < \alpha < 2$ derive system of moment equations combined with a Cattaneo truncation scheme
- leads to the **same fractional diffusion equation** in the long time limit as seen before
- *however:*



Taylor-King et al., PRE, 2016

Summary



- Be careful with **(power law) paradigms** for data analysis.
- A **more general biological embedding** is needed to better understand foraging.
- Much work to be done to apply **other types of anomalous stochastic processes** for modeling foraging problems.

Advanced Study Group

Statistical physics and anomalous dynamics of foraging

MPIPKS Dresden, July - December 2015



F.Bartumeus (Blanes, Spain), D.Boyer (UNAM, Mexico),
A.V.Chechkin (Kharkov, Ukraine), L.Giuggioli (Bristol, UK),
convenor: RK (London, UK), J.Pitchford (York, UK)

ASG webpage: http://www.mpipks-dresden.mpg.de/~asg_2015

Literature:

RK, *Search for food of birds, fish and insects*, book chapter
(preprint, 2016)