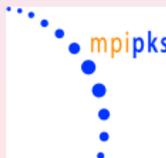


Search for Food of Birds, Fish and Insects

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Queen Mary University of London, School of Mathematical Sciences

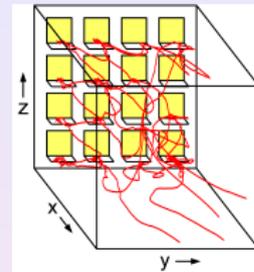
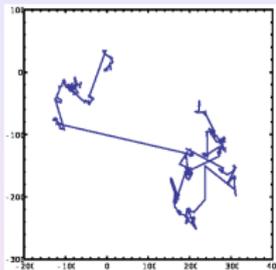
Diffusion Fundamentals VI, Technische Universität Dresden
26 August 2015



Queen Mary
University of London



Overview



Theme of this talk:

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

Three parts:

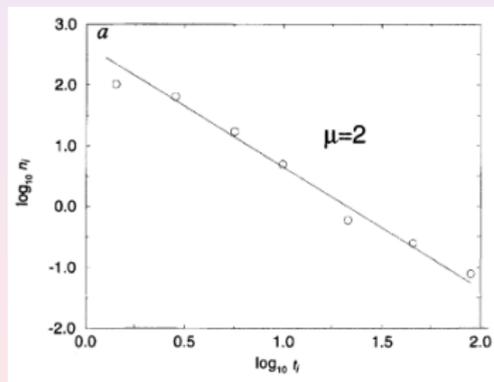
- 1 Lévy flight hypothesis: review
- 2 Biological data: analysis and interpretation
- 3 Foraging bumblebees: own research

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 $\sim t^{-\mu}$)



Lévy flights in a nutshell

Lévy flights have **well-defined mathematical properties**:

- a **Markovian** stochastic process (*no memory*)
- with probability distribution function of flight lengths exhibiting **power law tails**, $\rho(l) \sim l^{-1-\alpha}$, $0 < \alpha < 2$;
- it has **infinite variance**, $\langle l^2 \rangle = \infty$,
- satisfies a **generalized central limit theorem** (Gnedenko, Kolmogorov, 1949) and
- is **scale invariant**

for an outline see, e.g., Shlesinger et al., *Nature* **363**, 31 (1993)

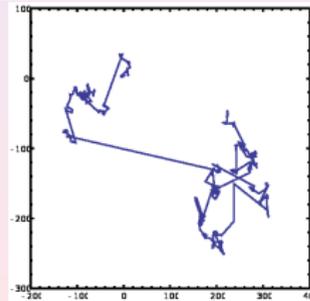
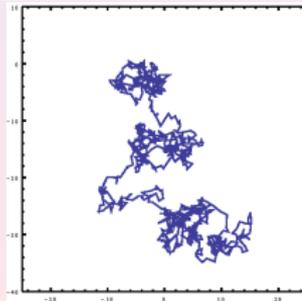
more realistic are *Lévy walks*; Zaburdaev et al., *RMP* **87**, 483 (2015)

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



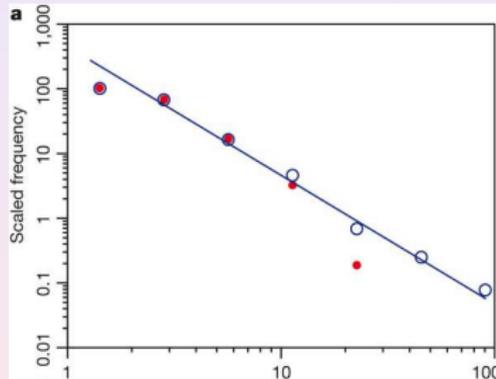
Brownian motion (left) vs. **Lévy flights** (right)

- Lévy flights also obtained for bumblebee and deer data

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):

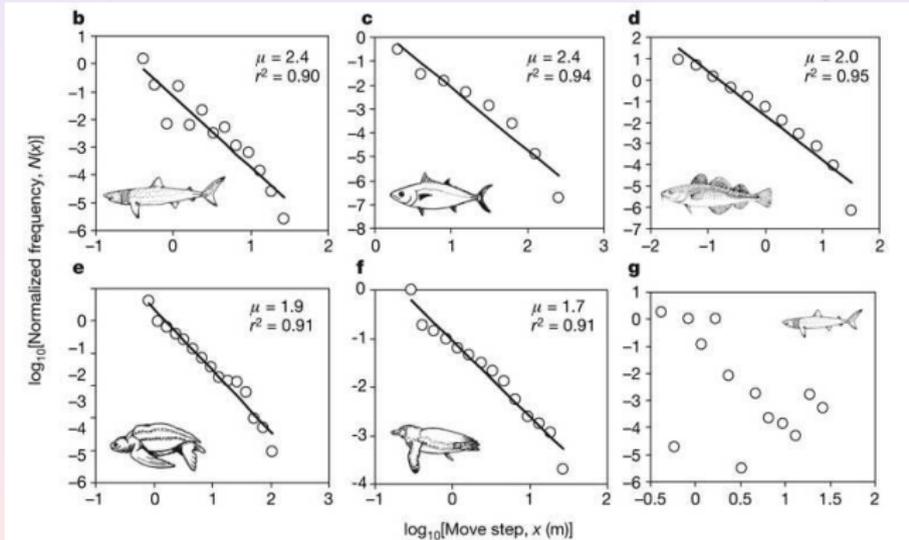


- **no Lévy flights:** new, more extensive data suggests (gamma distributed) stochastic process
- but claim that **truncated Lévy flights** fit yet new data Humphries et al., PNAS **109**, 7169 (2012) (and reply..)

Lévy or not Lévy?

Lévy paradigm: Look for power law tails in pdfs!

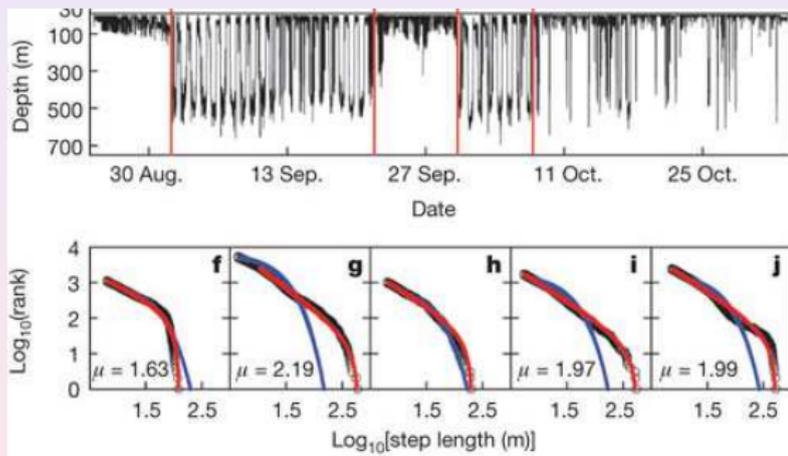
- **Sims et al., Nature 451, 1098 (2008): scaling laws of marine predator search behaviour; > 10⁶ data points!**



- prey distributions also display Lévy-like patterns...

Lévy flights induced by the environment?

- **Humphries et al., Nature 465, 1066 (2010): environmental context** explains Lévy and Brownian movement patterns of marine predators; $> 10^7$ data points!; for blue shark:



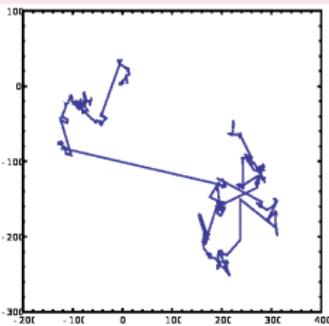
blue: exponential; red: truncated power law

- **note:** \exists day-night cycle, cf. oscillations; suggests to fit with two different pdfs (not done)

Optimal searches: adaptive or emergent?

strictly speaking **two different Lévy flight hypotheses:**

- ① Lévy flights represent an (evolutionary) **adaptive optimal search strategy**
Viswanathan et al. (1999)
the 'conventional' Lévy flight hypothesis



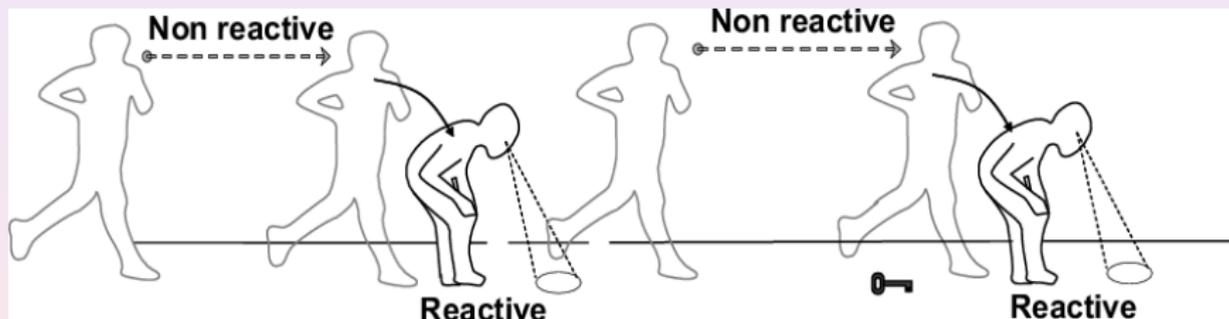
- ② Lévy flights **emerge** from the **interaction with a scale-free food source distribution**
Viswanathan et al. (1996)
more recent reasoning



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

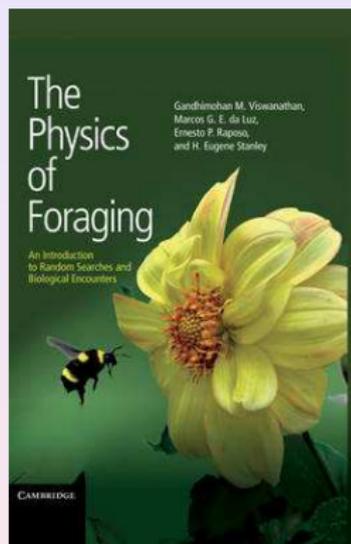
In search of a mathematical foraging theory

Summary:

- two different Lévy flight **hypothesis**: **adaptive** and **emergent**
- scale-free Lévy flight **paradigm**
- problems with the **data analysis**
- **intermittent** search strategies as alternatives

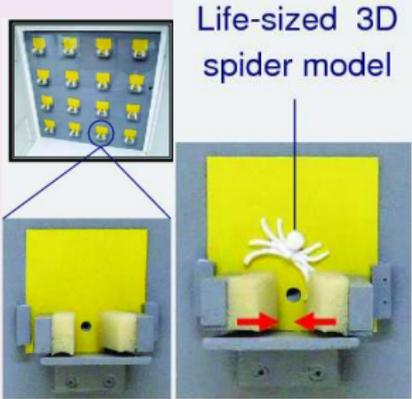
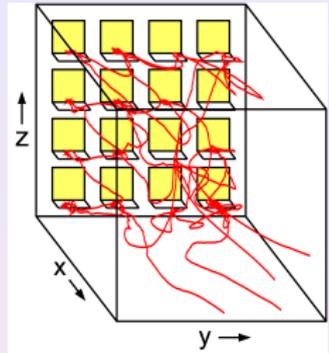
⇒ **ongoing discussions**:

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



Foraging bumblebees: the experiment

- tracking of **bumblebee flights** in the lab: foraging in an artificial carpet of **flowers with or without spiders**
- **no test** of the Lévy hypothesis but work inspired by the *paradigm*



safe and **dangerous** flowers

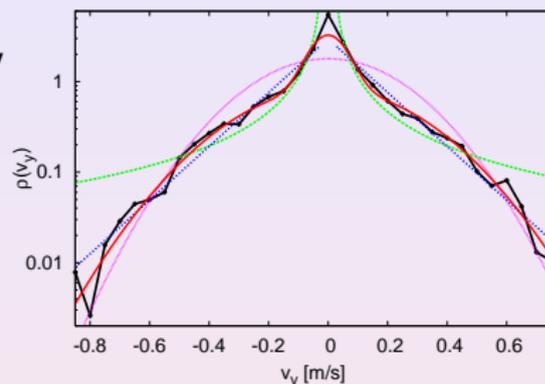
three experimental stages:

- 1 spider-free foraging
- 2 foraging under predation risk
- 3 memory test 1 day later

Flight velocity distributions

experimental **probability density** (pdf) of bumblebee v_y -velocities without spiders (bold black)

best fit: mixture of 2 Gaussians,
cp. to exponential, power law,
single Gaussian

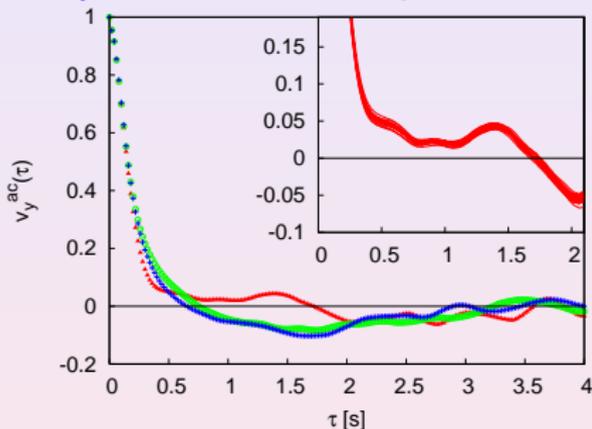


biological explanation: models spatially different flight modes near the flower vs. far away, cf. intermittent dynamics

big surprise: no difference in pdf's between different stages under variation of environmental conditions!

Velocity autocorrelation function || to the wall

$$V_y^{AC}(\tau) = \frac{\langle (v_y(t) - \mu)(v_y(t+\tau) - \mu) \rangle}{\sigma^2}$$



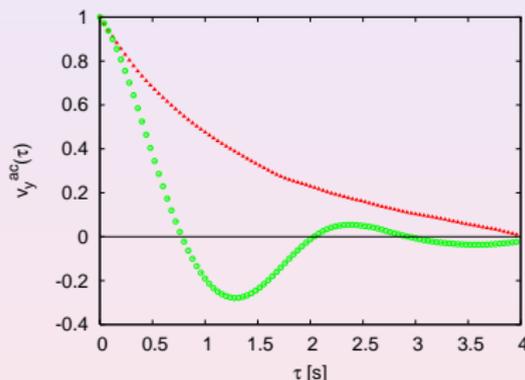
3 stages: spider-free, predation thread, memory test

all changes are in the flight correlations, not in the pdfs

model: Langevin equation

$$\frac{dv_y}{dt}(t) = -\eta v_y(t) - \frac{\partial U}{\partial y}(y(t)) + \xi(t)$$

η : friction, ξ : Gauss. white noise



result: velocity correlations with repulsive interaction U
bumblebee - spider off / on

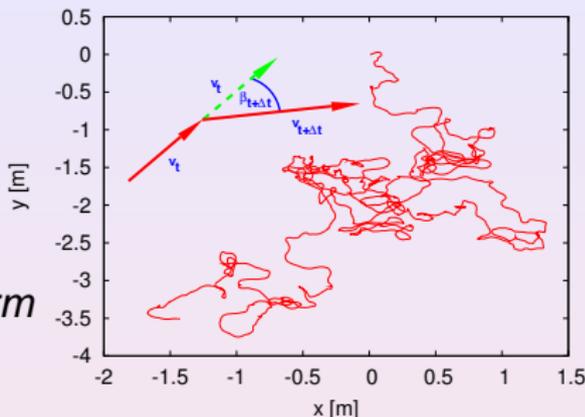
Lenz et al., PRL **108**, 098103 (2012)

Modeling free bumblebee flights

reorientation model:

describe 2d movement in comoving frame by

- speed $v(t) = \text{const.}$
- turning angle $\beta(t) = \xi(t)$ as random variable from *non-uniform pdf* modeling **persistence**

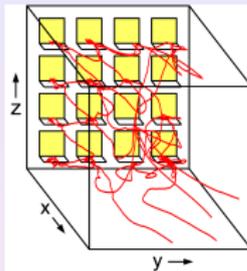
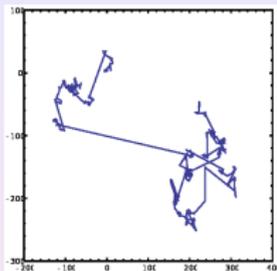


generalized model for bumblebee flights far away from flowers constructed from experimental data:

- $\beta(t) = \xi_v(t)$: power law **correlated Gaussian noise**
- $\frac{dv}{dt} = g(v(t)) + \psi(t)$: **generalized Langevin equation** with anti-correlated Gaussian noise

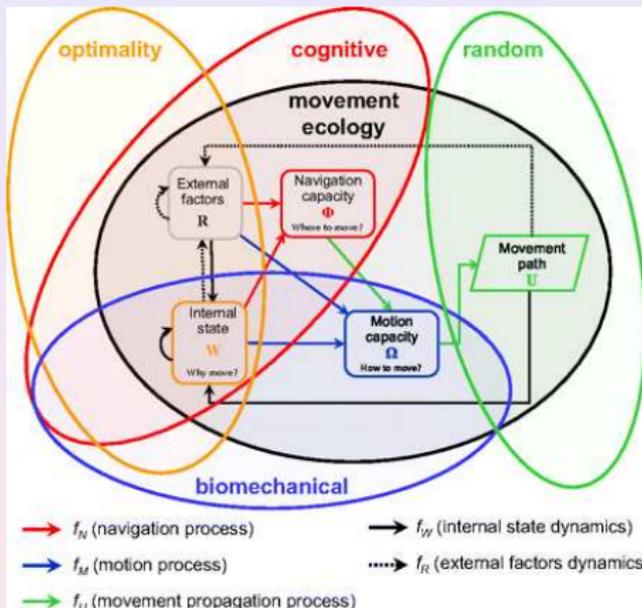
F.Lenz, A.V.Chechkin, R.K., PLoS ONE 8, e59036 (2013)

Summary



- Be careful with (power law) paradigms for data analysis.
- Other quantities may contain crucial information about foraging; **example**: bumblebee flights under predation threat.
- **Conclusion**: A more general biological embedding is needed!

Perspective: Movement Ecology Paradigm



Nathan et al., PNAS **105**, 19052 (2008)

state space approach $\mathbf{u}_{t+1} = F(\Omega, \Phi, \mathbf{r}_t, \mathbf{w}_t, \mathbf{u}_t)$ for the location \mathbf{u}_t of an organism at time t

Outlook: more to come...

Advanced Study Group on Statistical physics and anomalous dynamics of foraging MPIPKS Dresden, July - Dec. 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

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