

Constructing a Stochastic Model of Bumblebee Flights from Experimental Data

Friedrich Lenz¹ Aleksei V. Chechkin² Rainer Klages¹

¹Queen Mary University of London, School of Mathematical Sciences

²Institute for Theoretical Physics NSC KIPT, Kharkov, Ukraine

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Motivation

bumblebee foraging:

find food (nectar, pollen) in complex landscapes

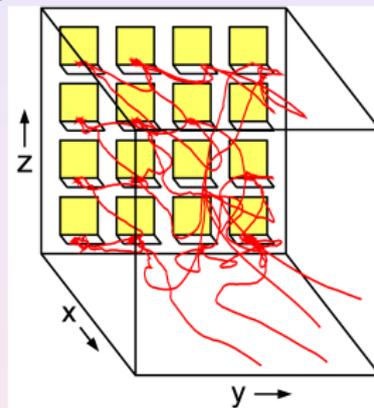


What type of motion?

Study bumblebee foraging in a *laboratory experiment*.

Bumblebee experiment

Ings, Chittka, *Current Biology* **18**, 1520 (2008):
bumblebee foraging in a cube of $\simeq 75\text{cm}$ side length

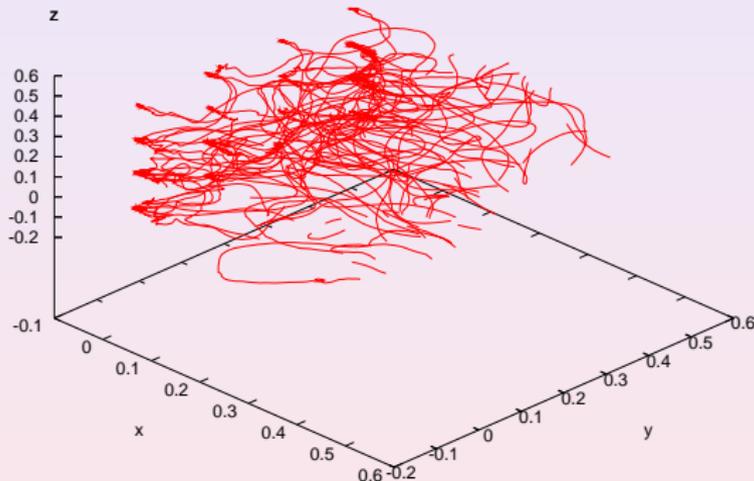


- artificial yellow flowers: 4x4 grid on one wall
- two cameras track the position (50fps) of a single bumblebee (*Bombus terrestris*)
- #bumblebees=30, #data points ≈ 49000

nb: Here we only focus on one aspect of the full experiment.

The main question

What **type of motion** do the bumblebees perform *away from the flower carpet* in terms of **stochastic dynamics**?



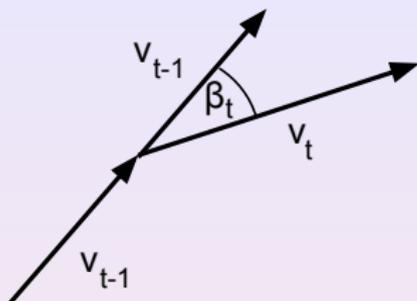
nb: The foraging dynamics *in interaction with the flowers* has been studied in [Lenz et al., PRL 108, 098103 \(2012\)](#).

Reorientation (or CRW) model

describe biological movements in a plane by speed $s(t) = |v(t)|$ and turning angle β in comoving frame:

Correlated Random Walk model

$$\beta(t) = \xi(t), \quad s(t) = \text{const.}$$



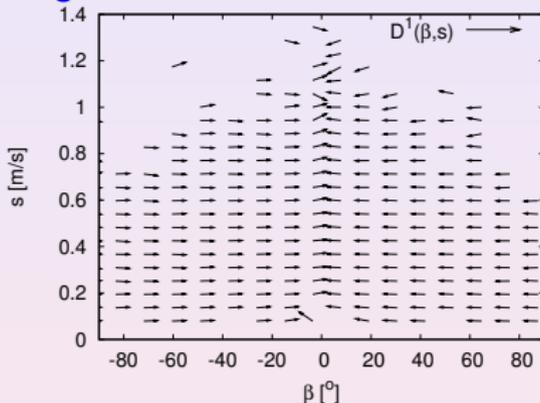
where $\xi(t)$ is typically drawn i.i.d. from a wrapped normal distribution; model captures directional **biological persistence**

goal: construct a **generalized CRW model from experimental data** for reproducing ‘free’ bumblebee flights by using **Langevin-type dynamics**: drift terms plus noise

$$\begin{aligned} \frac{d\beta(t)}{dt} &= h(\beta(t), s(t)) + \tilde{\xi}(t) \\ \frac{ds(t)}{dt} &= g(\beta(t), s(t)) + \psi(t) \end{aligned}$$

Drift coefficients: phase space dynamics

assume Markovianity for estimating **Fokker-Planck drift coefficients** h and g ; normalized **drift vector field**:

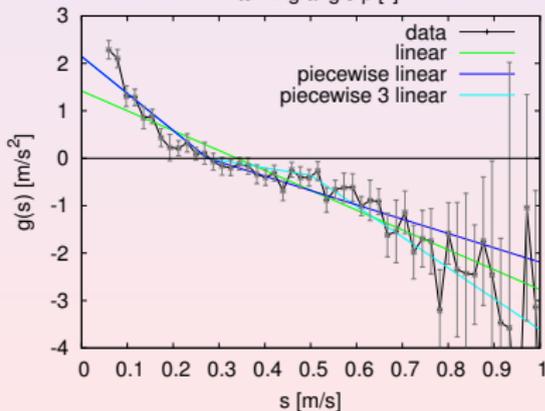
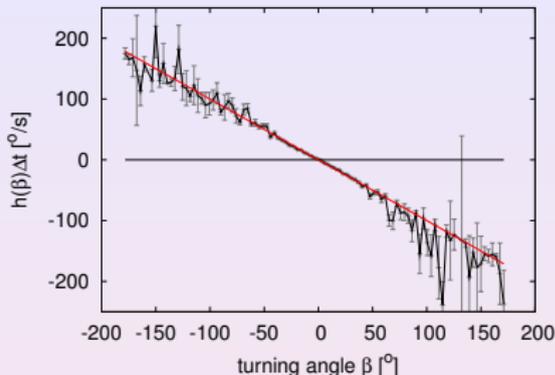


indicates that the **cross-dependencies** of $h(\beta(t), s(t))$ on s and of $g(\beta(t), s(t))$ on β are **weak**; vector field splits into

$$d\beta/dt = h(\beta(t)) + \tilde{\xi}(t)$$

$$ds/dt = g(s(t)) + \psi(t)$$

Estimation of drift terms from data



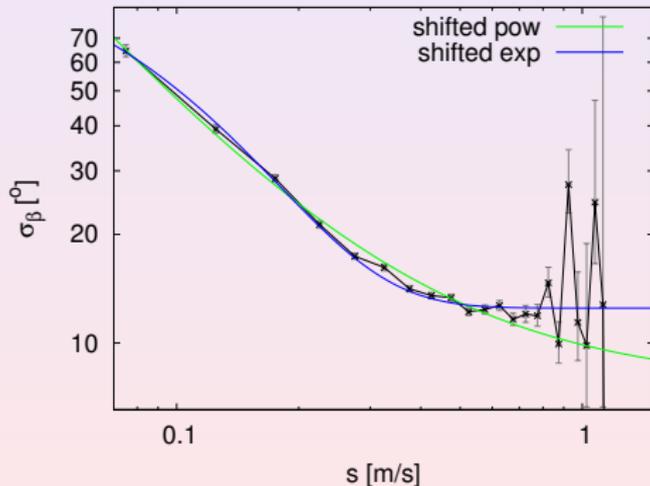
extract **projection** $h(\beta)$ from data:
 $h(\beta) \simeq -k\beta$ with $k \approx 1/\Delta t$
 integrating $d\beta/dt = h(\beta(t)) + \tilde{\xi}(t)$
 wrt Δt yields $\beta(t) = \xi(t)$

extract **projection** $g(s)$ from data:
 \exists preferred speed s_0 ; piecewise
 linear approximation for $g(s)$ in
 $ds/dt = g(s(t)) + \psi(t)$ yields
 $g(s) \approx (s - s_0) \cdot \begin{cases} -d_1, & s < s_0 \\ -d_2, & s \geq s_0 \end{cases}$
 with $d_1 > d_2 > 0$

Velocity-dependent angle noise

pdf for the **turning angles** β at each speed s is approximated by a **Gaussian**;

however, the **variance** σ_β is **s-dependent** (cf. naive reasoning):



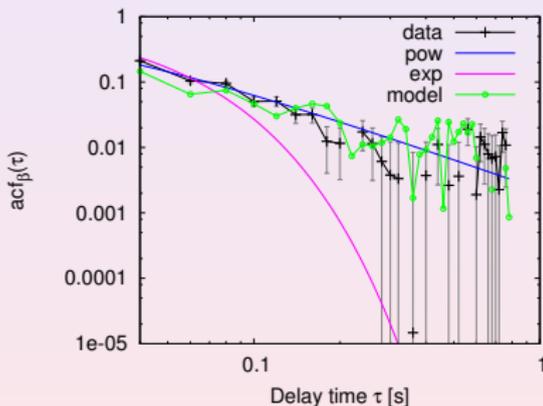
$$\beta(t) = \xi_s(t)$$

$$\xi_s(t) \sim \mathcal{N}(0, f(s(t)))$$

$$f(s) = c_1 e^{-c_2 s} + c_3$$

Noise autocorrelation functions

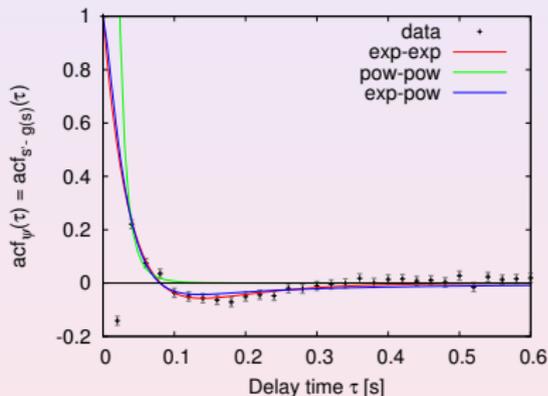
noise $\xi_s(t)$ of turning angles
 β is a steep power law:



noise of speed changes

$\psi(t) = ds/dt - g(s(t))$ is

Gaussian with **anti-correlations**:



best approximated by

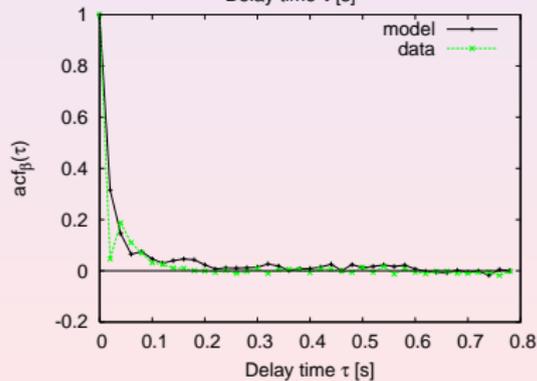
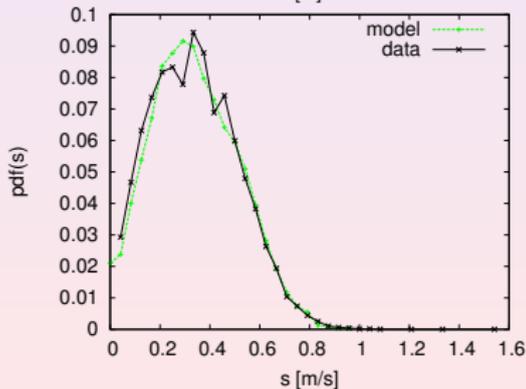
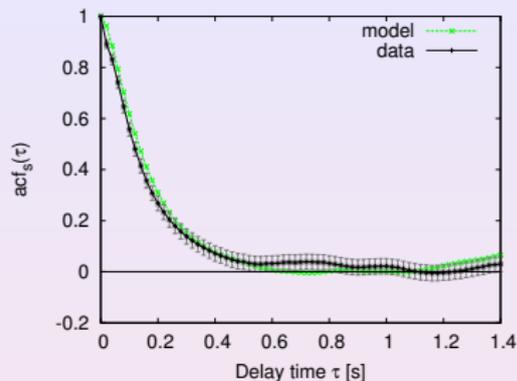
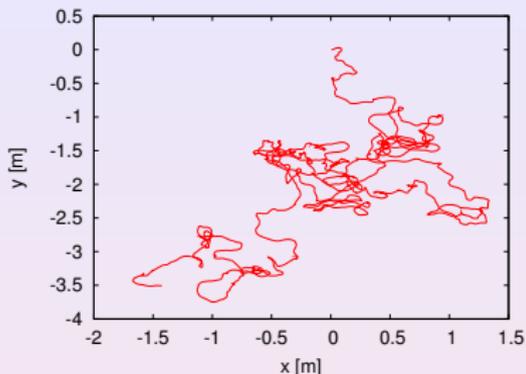
$$\text{acf}_\psi(t) \approx ae^{-\lambda_1 t} + (1 - a)e^{-\lambda_2 t}$$

Summary: The complete model

$$\begin{aligned}\beta(t) &= \xi_s(t) \\ \frac{ds}{dt} &= g(s(t)) + \psi(t)\end{aligned}$$

- **turning angles** β given by **power law-correlated Gaussian noise** $\xi_s(t) \sim \mathcal{N}(0, \sigma_\xi(s))$ with $\sigma_\xi(s) = c_1 e^{-c_2 s} + c_3$
- **piecewise linear drift** $g(s)$ for **speed** s
- ψ **Gaussian noise** and **anti-correlated** via sum of exponentials

Simulation and comparison to real data



very good agreement given the number of approximations

Summary and Reference

- We have constructed a generalized **Langevin-type correlated random walk model** that well reproduces bumblebee flights in a small cube.
- **Question 1:** Does it also work for **bee flights in the wild?**
- **Question 2:** Can we model **other animal flights** with it?

Reference:

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

