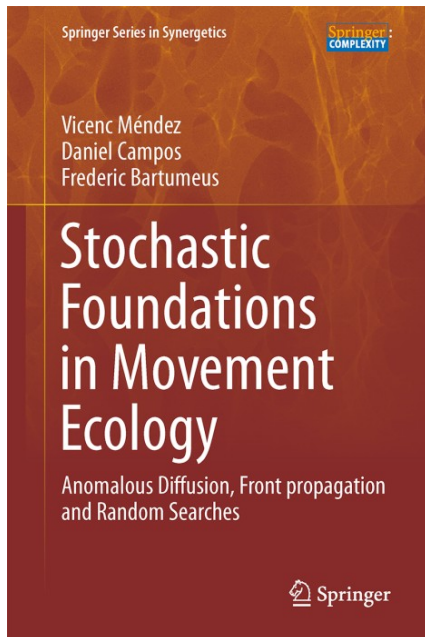


Advanced Study Group 2015: Statistical Physics and Anomalous Dynamics of Foraging



MAX-PLANCK-GESELLSCHAFT



STOCHASTIC FOUNDATIONS IN MOVEMENT ECOLOGY III:

*Mean first passage times and the exploitation-
exploration tradeoff*

Vicenç Méndez, Daniel Campos, Frederic Bartumeus

Universitat Autònoma de Barcelona

Centre d'Estudis Avançats de Blanes (CEAB-CSIC)



Stochastic Foundations in Movement Ecology III: MFPTs and the exploitation-exploration tradeoff

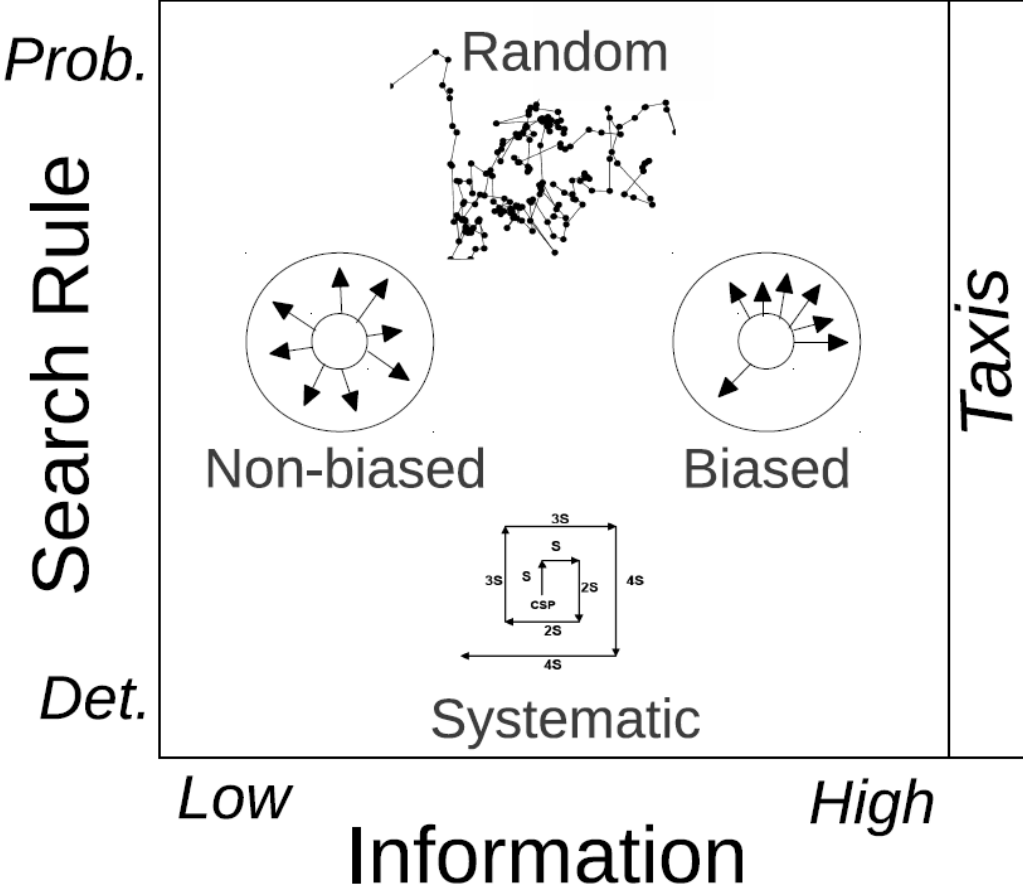
Sources:

- Raposo et al. PLoS Comput Bio (2011)
- Mendez et al. (2014) Chapter 6 and 9.
- Bartumeus et al. *PLoS ONE* (2014)
- Campos et al. *Phys. Rev. E* (2015)
- Bartumeus et al. *Ecology Letters* (submitted)
- Campos et al. (in preparation)

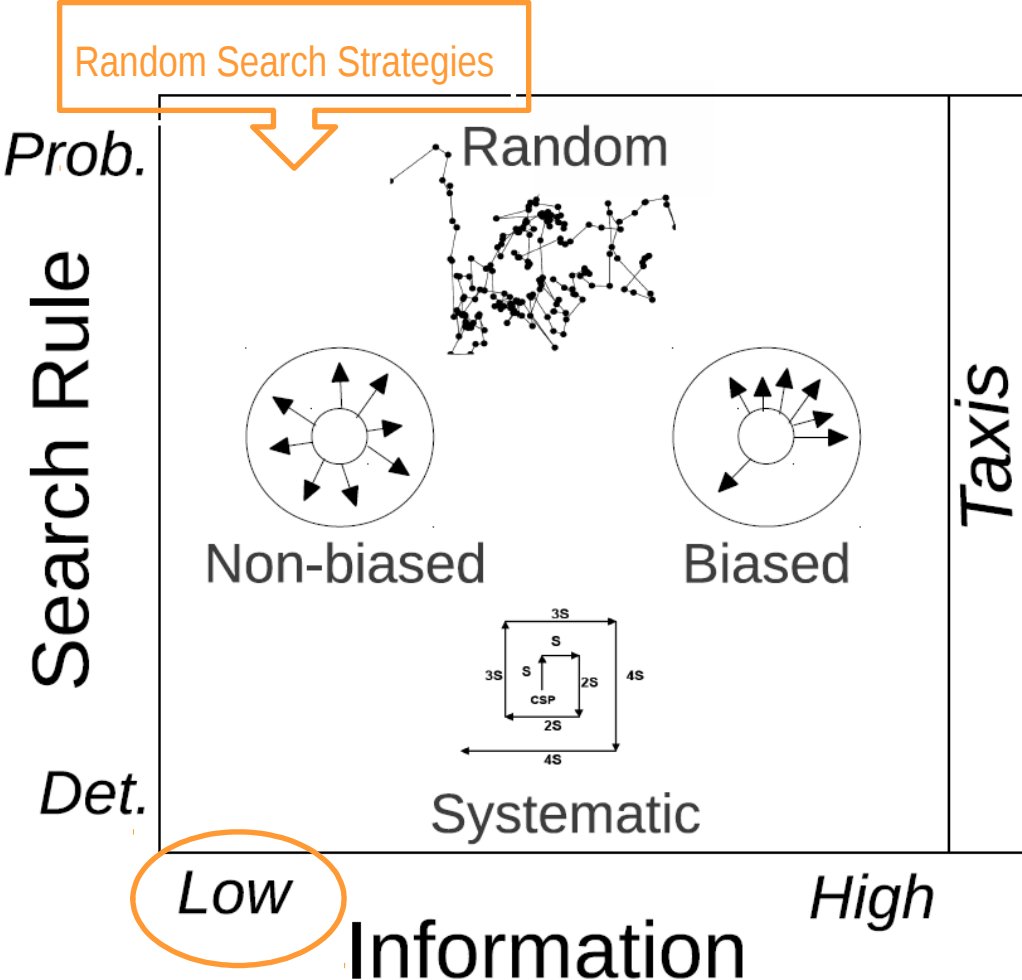
Collaborators:

- Raposo E., da Luz M.G.E., Viswanathan G.M. (Brazil)
- Campos D., Mendez V. (Spain)

Search Strategies



Search Strategies



Sensory Biology??

probabilistic

Search Rule

deterministic



Low

Information

High

probabilistic

Search Rule

deterministic



Low

Information

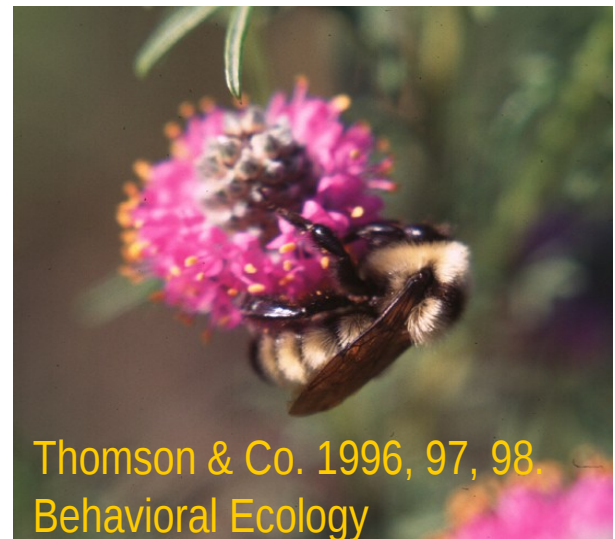
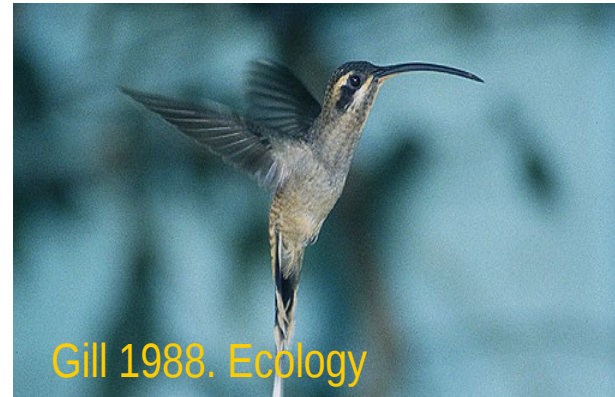
High

Animal Search Strategies: all types of rules

Relocation experiments



Learning, Trapline foraging

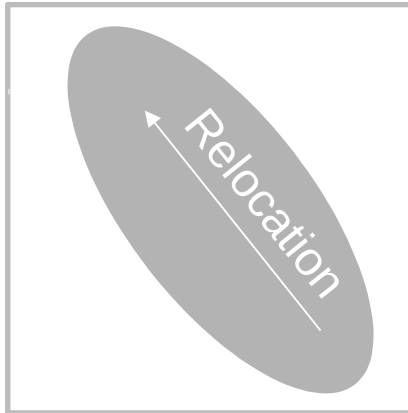


Relocation Experiments

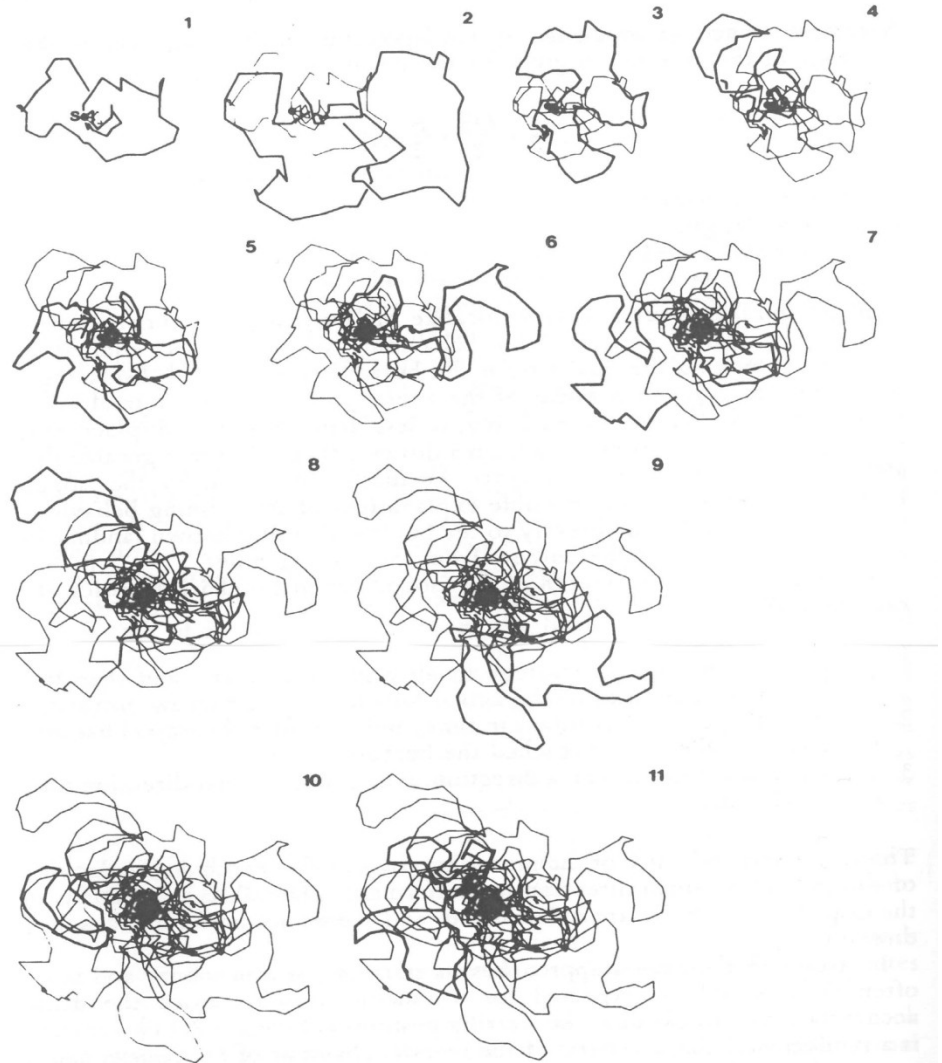


Hoffmann 1983a,b
Behav. Ecol. Sociobiol.

Search Rule



Information

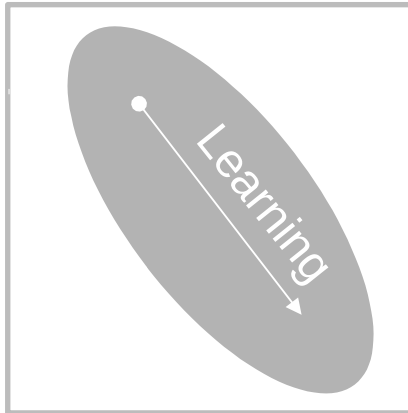


Learning Experiments

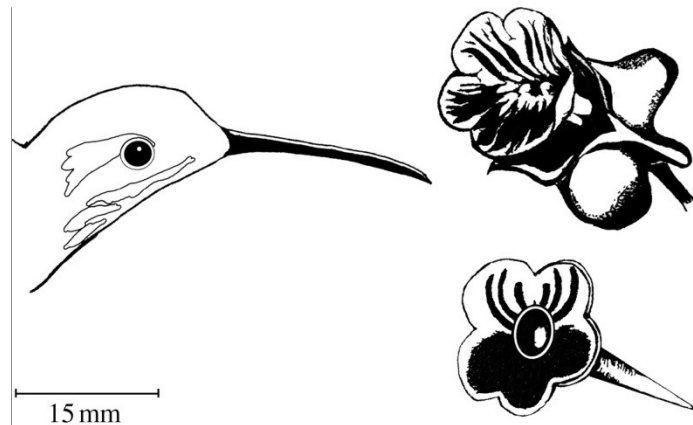
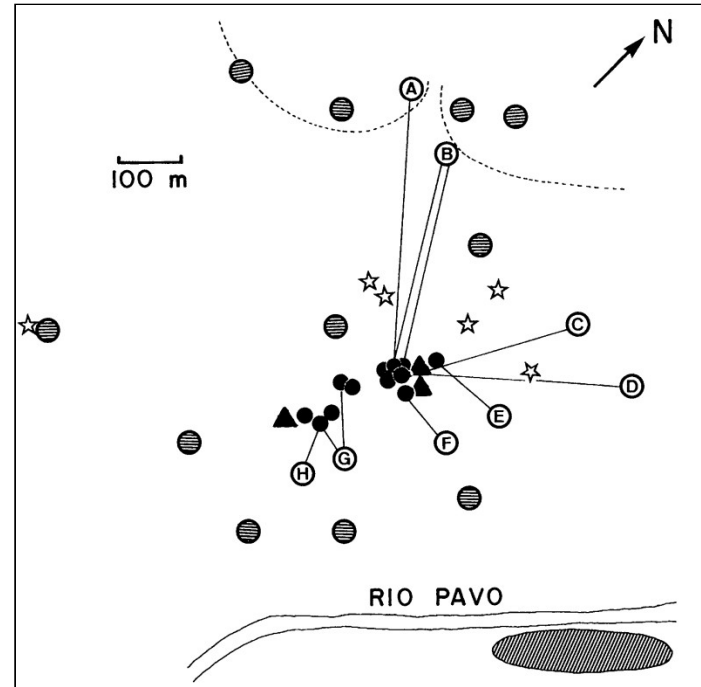


Gill 1988. Ecology.

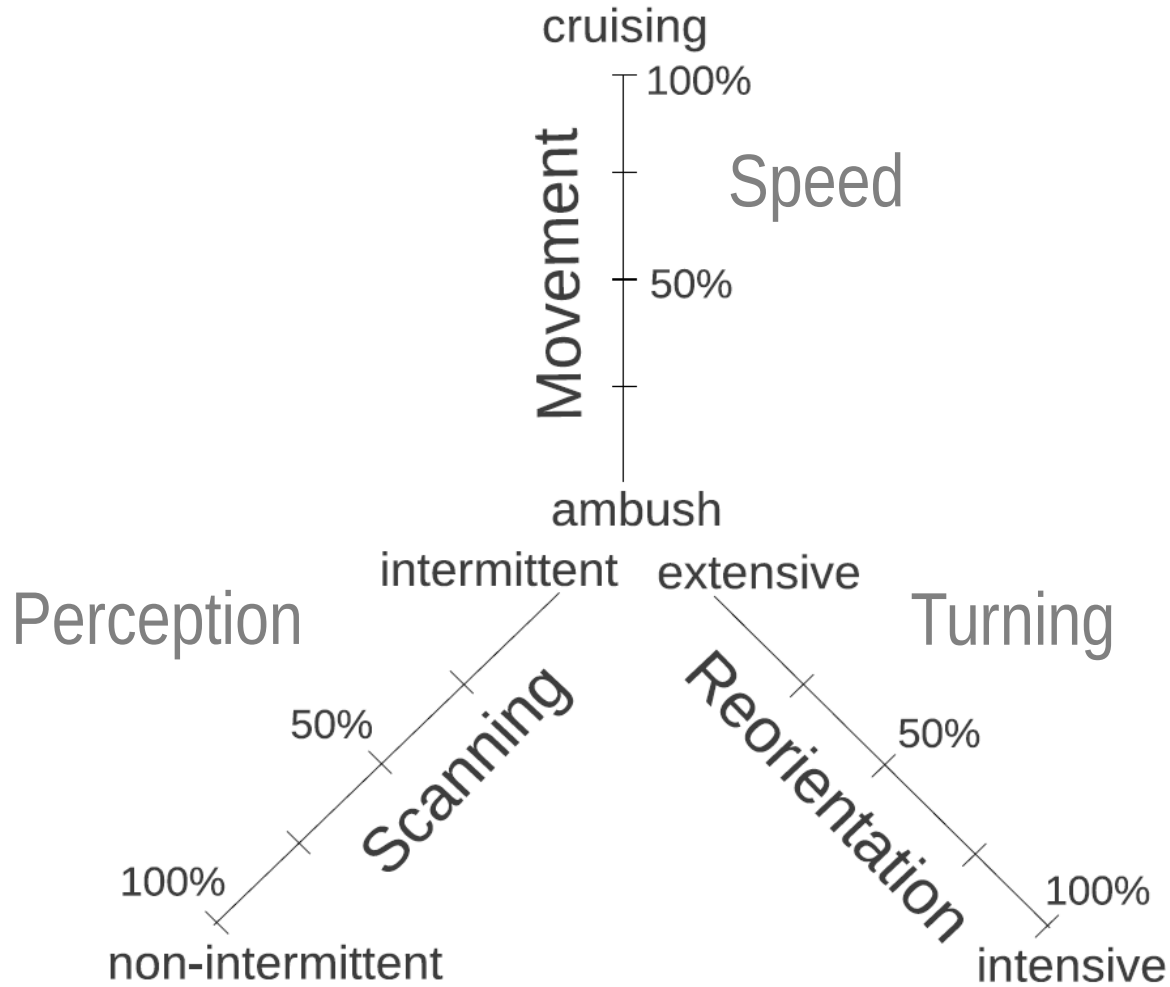
Search Rule



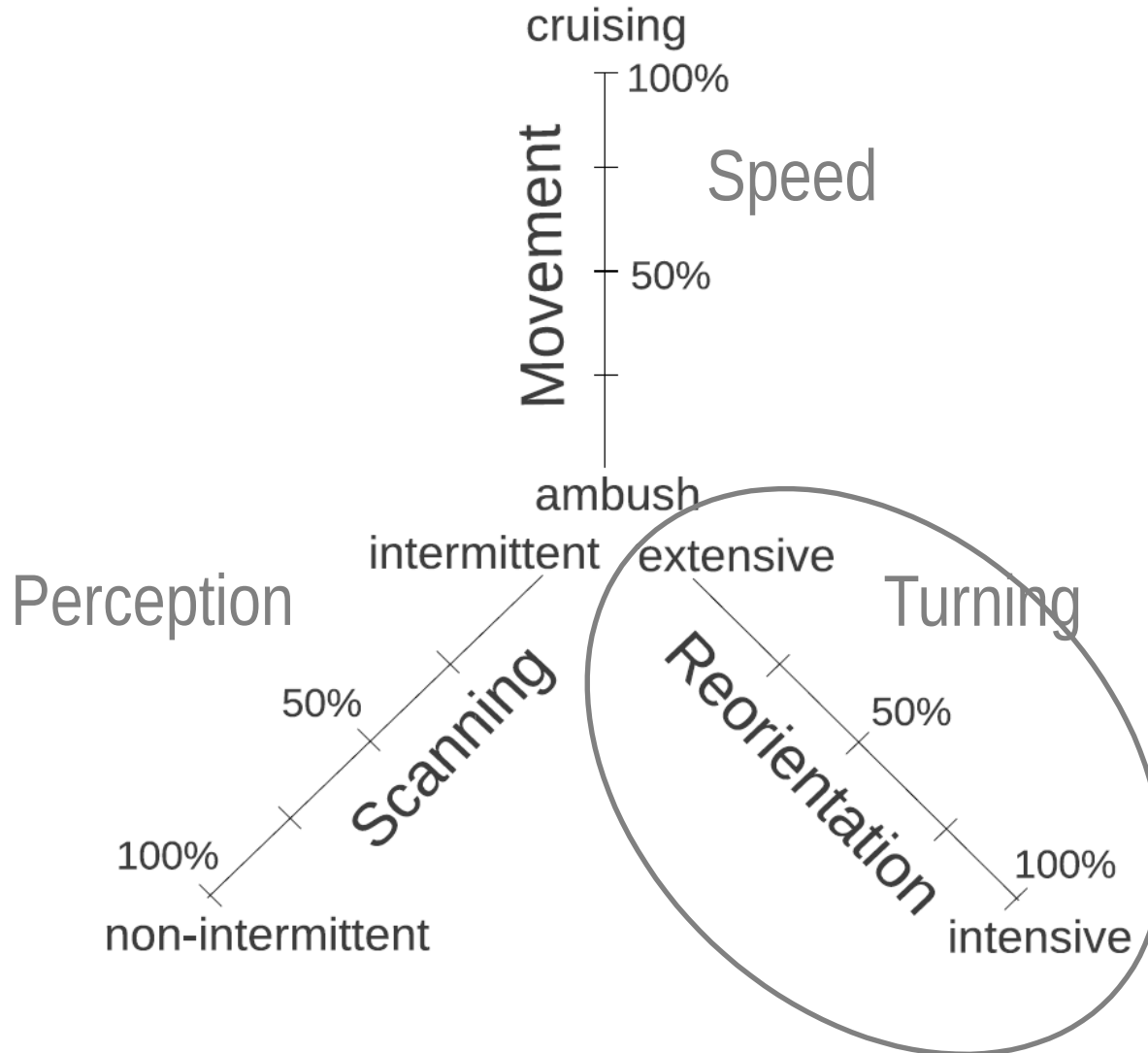
Information



Searcher perspective



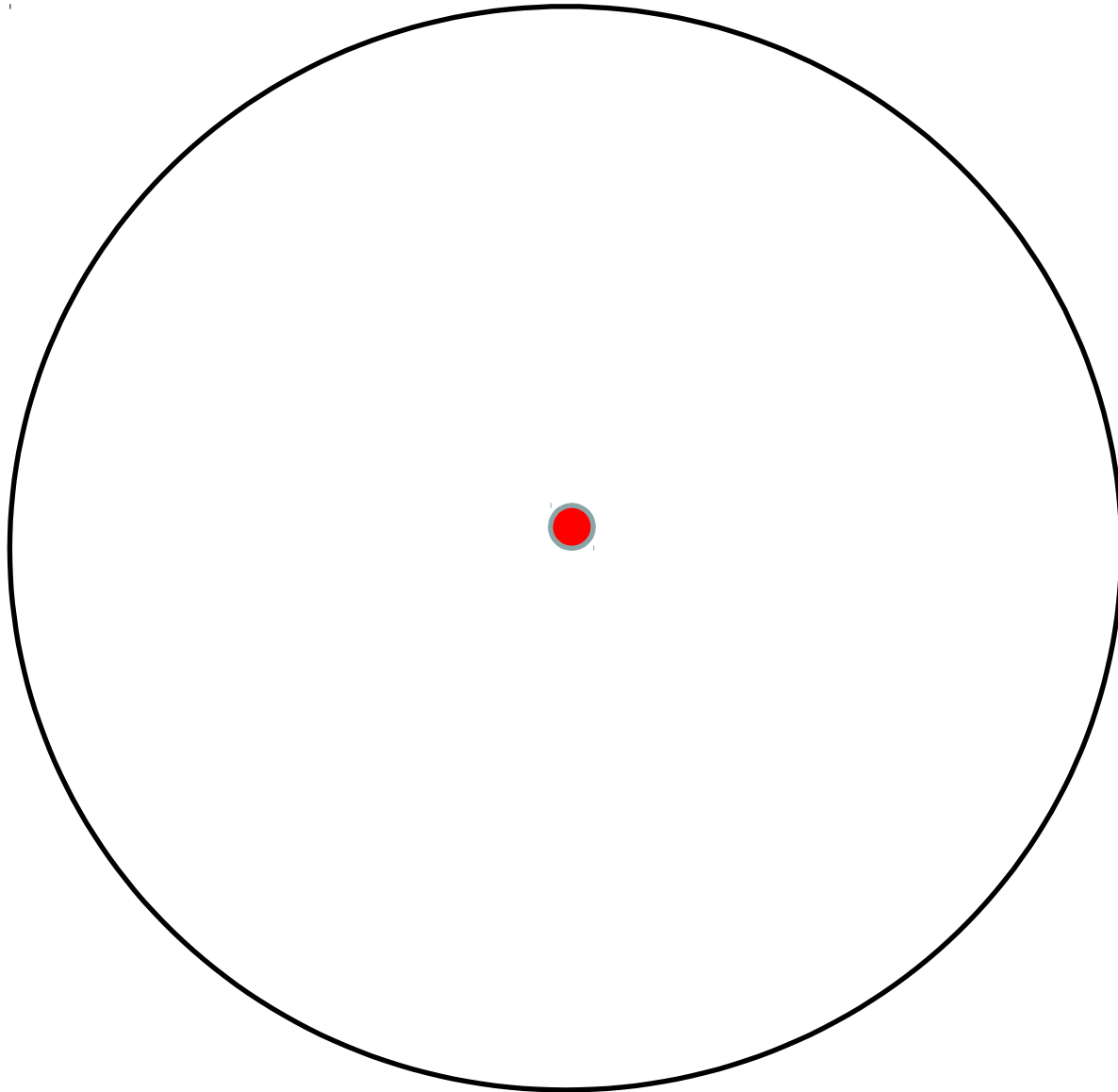
Searcher perspective



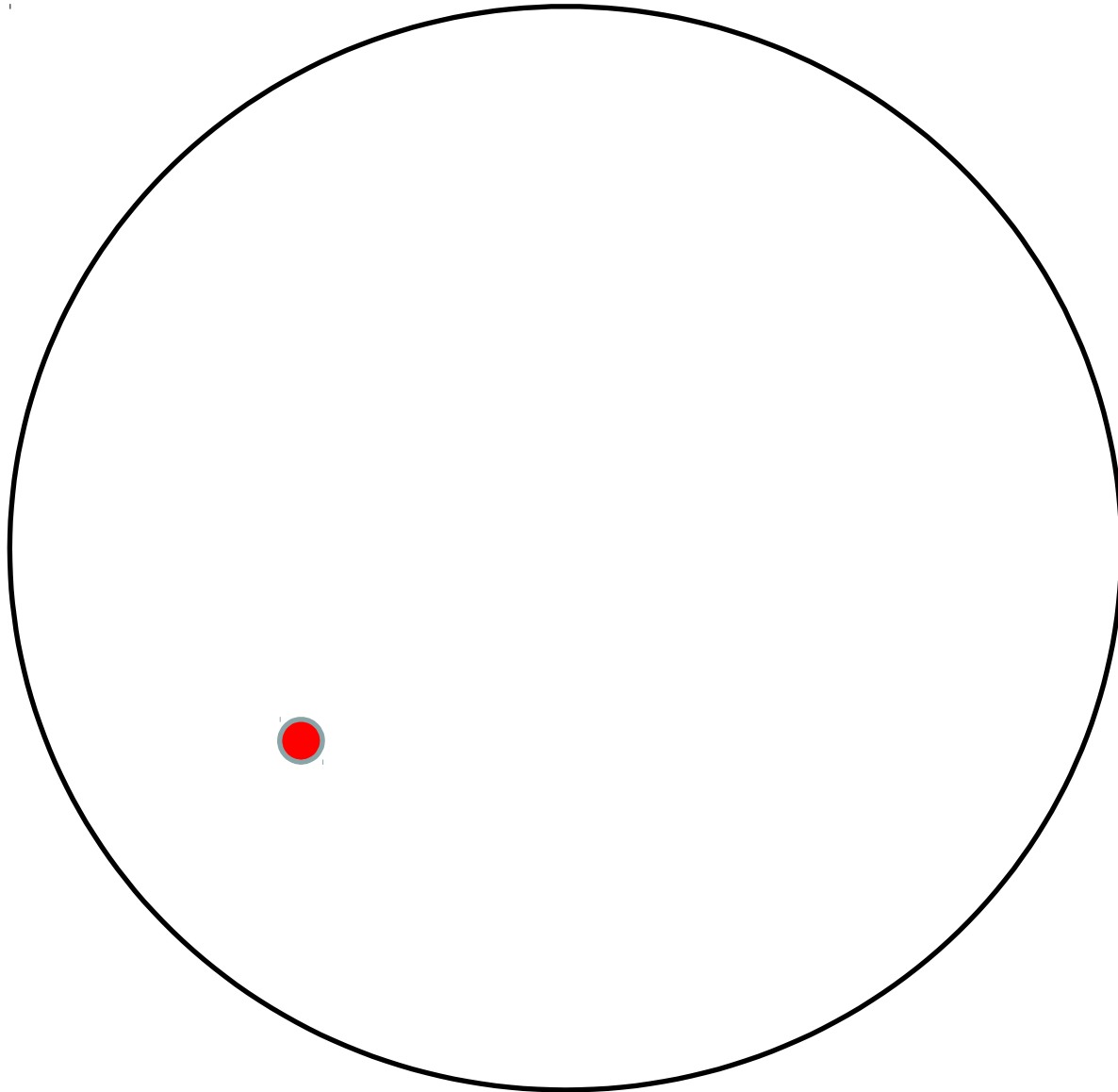
Why to turn at all?



Mean First Passage Times



Mean First Passage Times



Turning behaviour in a random search

The condition for turning

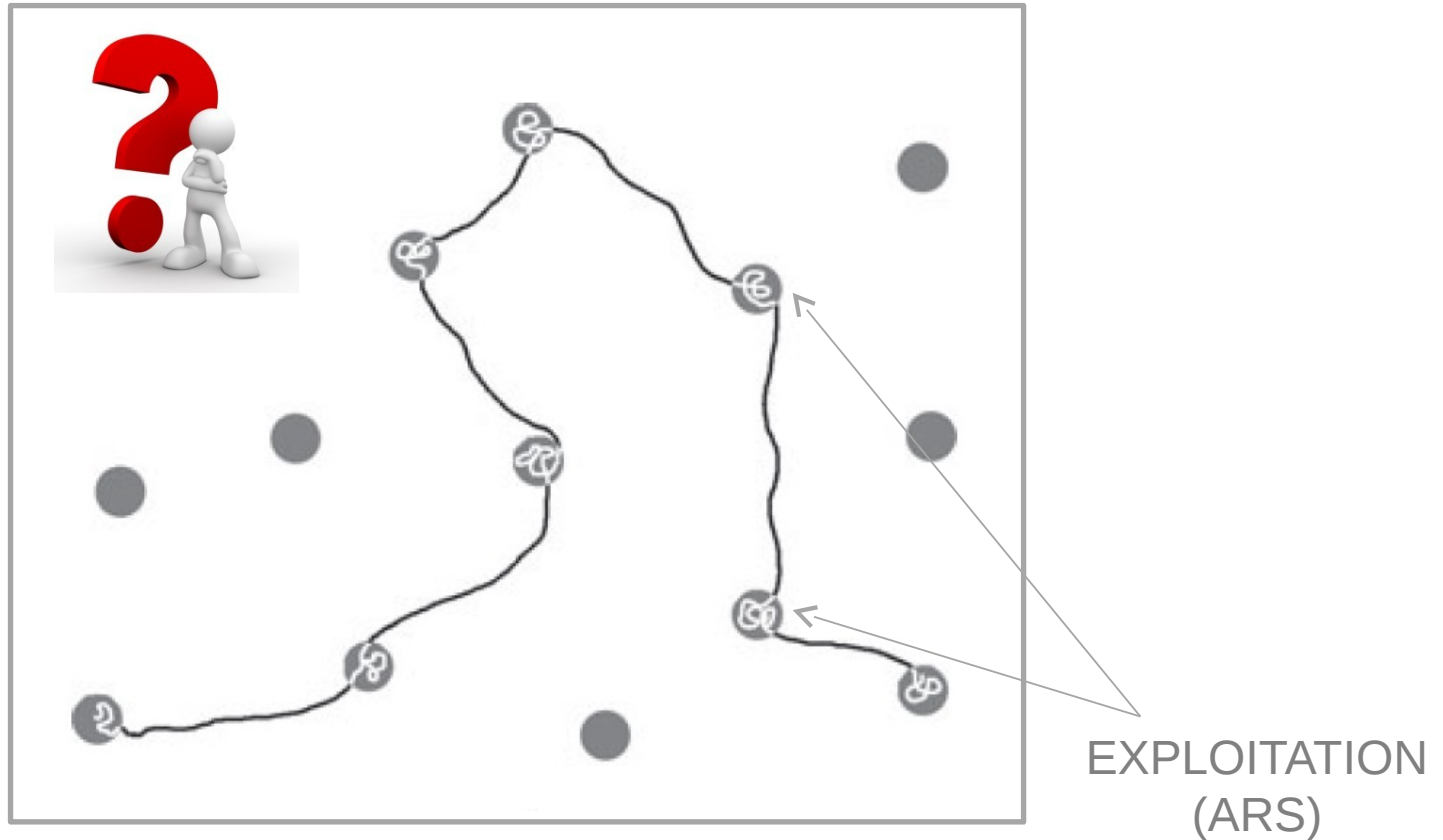
Assimetry

Heterogeneous searcher-to-target distances

Examples:

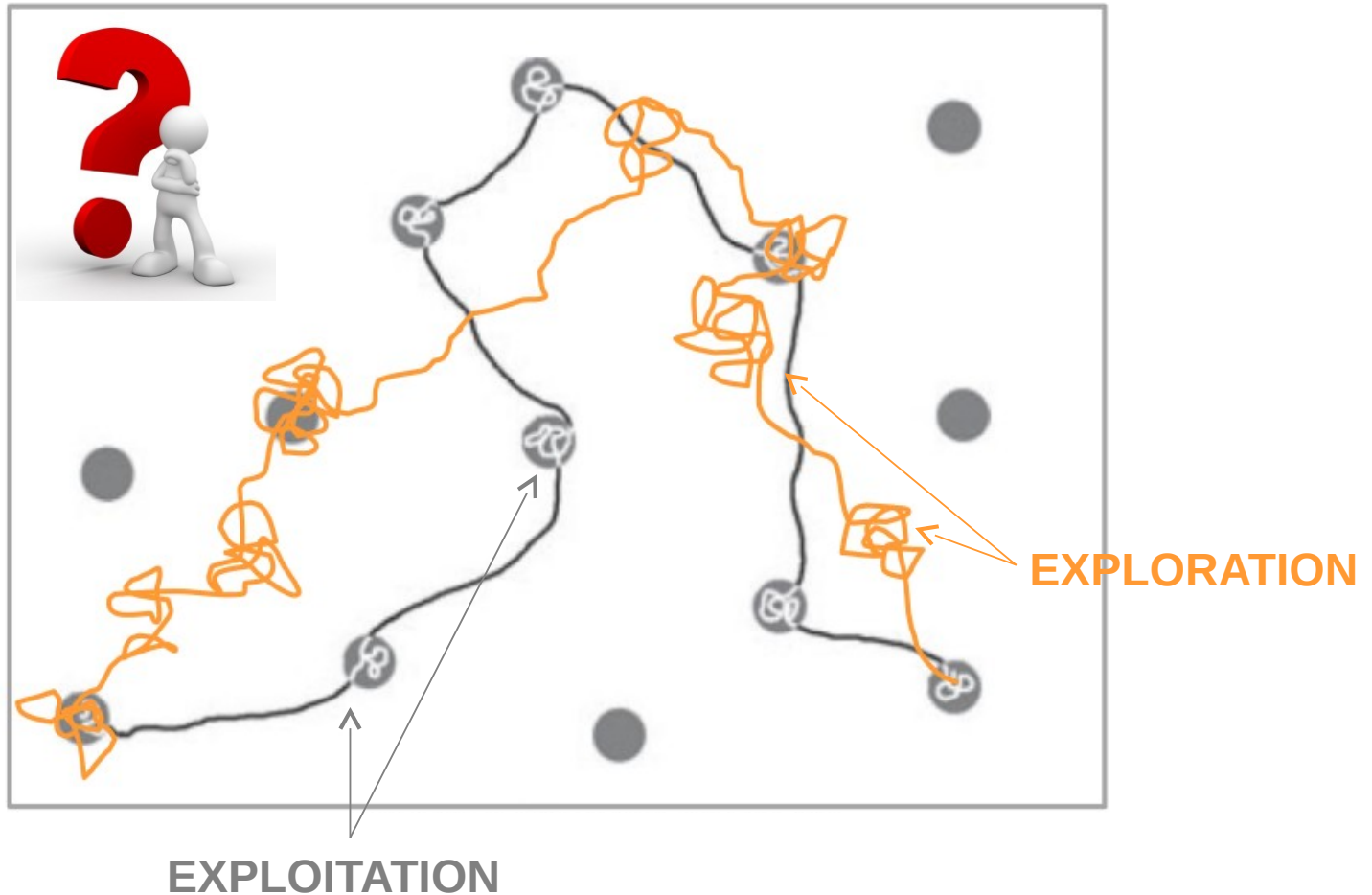
- Revisitable targets (non-destructive)
- Perception errors
- Patchy/highly heterogeneous landscapes

Turning behaviour as “cue”-driven



Simon Benhamou
Ecology Letters 2014 (*Ideas and Perspectives*)

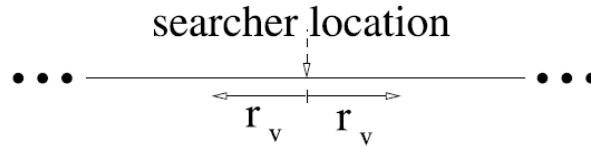
Turning behaviour as a “sampling” strategy



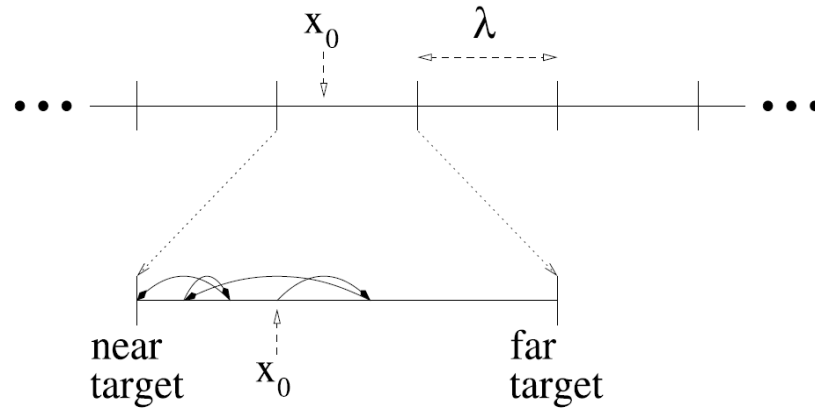
- To avoid missing nearby targets
- To improve 2D spatial coverage

Stochastic Optimal Foraging Theory

Perception:



Search process:



Back to initial conditions:



Stochastic Optimal Foraging Theory

We consider a 1D-random walker that can choose a direction (left/right) with equal probability and can take move lengths (ℓ) from a pdf $p(\ell)$

$$\int_{-\infty}^{+\infty} p(\ell) d\ell = 1, [p(-\ell) = p(\ell)]$$

We compute a statistical search efficiency as: $\eta = \frac{N_{\text{found}}}{L_{\text{tot}}}$

$$L_{\text{tot}} = N_{\text{found}} \langle L \rangle, \quad \eta = \frac{1}{\langle L \rangle}$$

$$N = N_{\text{found}} \langle n \rangle, \quad \langle |\ell| \rangle(x_0), \quad \eta \approx (\langle n \rangle \langle |\ell| \rangle)^{-1}$$

$\langle L \rangle$ average distance travelled between two targets found

$\langle n \rangle$ average number of steps between two targets found

$\langle |\ell| \rangle$ average step length between two targets found

Stochastic Optimal Foraging Theory

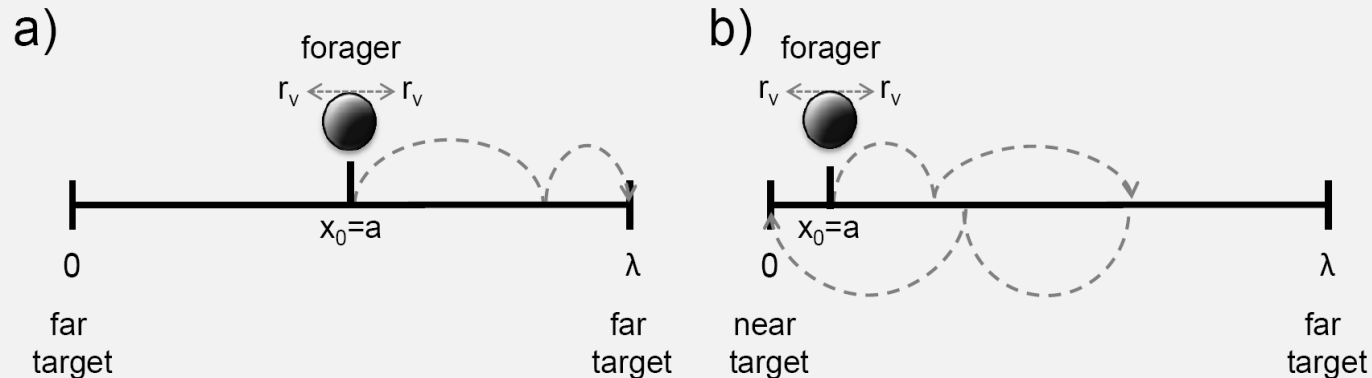
$\langle L \rangle$ average distance travelled between two targets found

$\langle n \rangle$ average number of steps between two targets found

$\langle |\ell| \rangle$ average step length between two targets found

These 3 quantities depend on:

- An **initial position** ($x_0 = a$)
- A **boundary condition**: the average distance between targets ($x_0 = \lambda$)



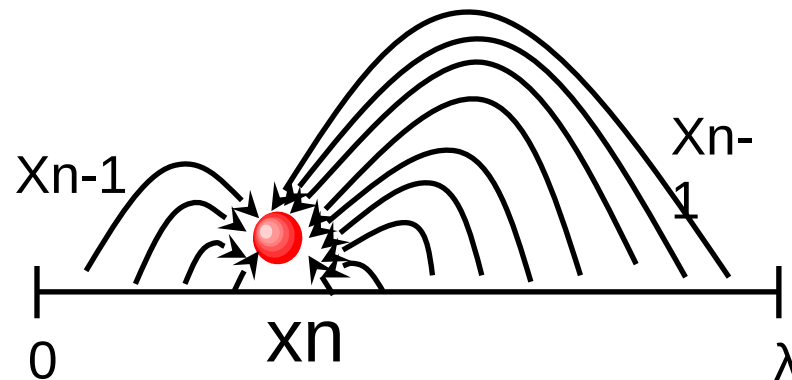
Stochastic Optimal Foraging Theory

Key concept of the calculations: a renewal approach

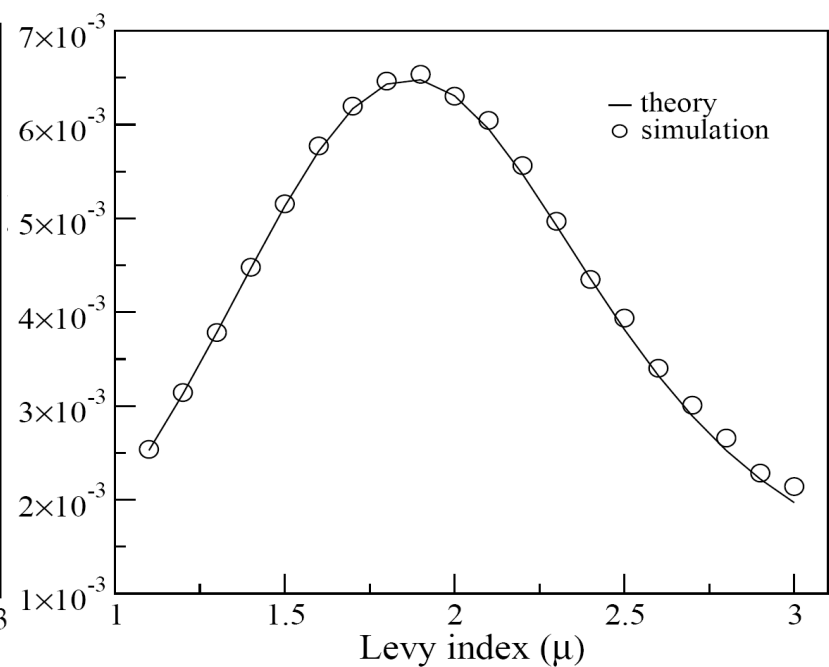
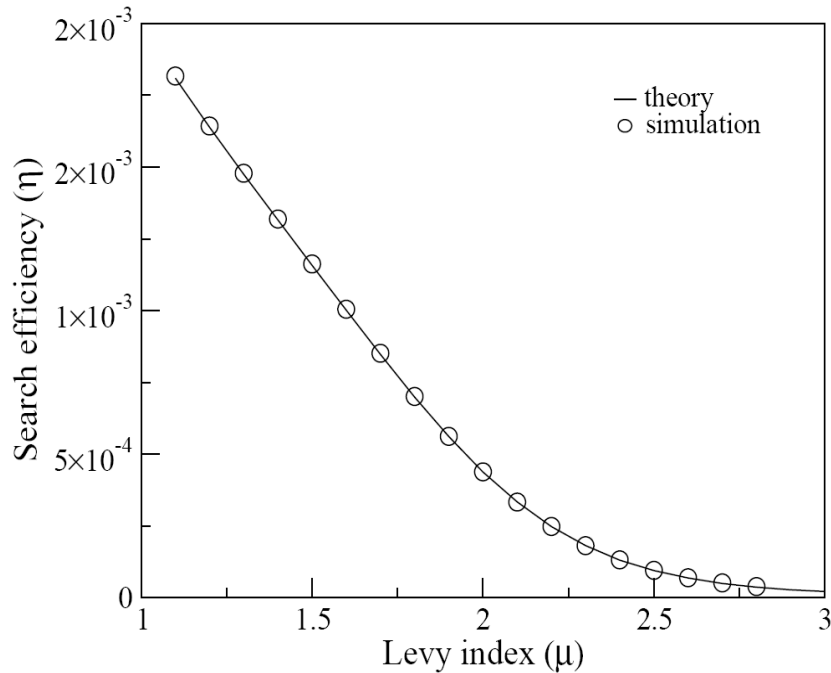
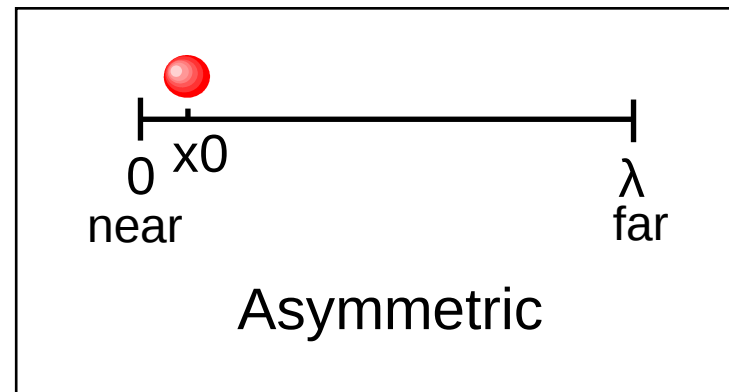
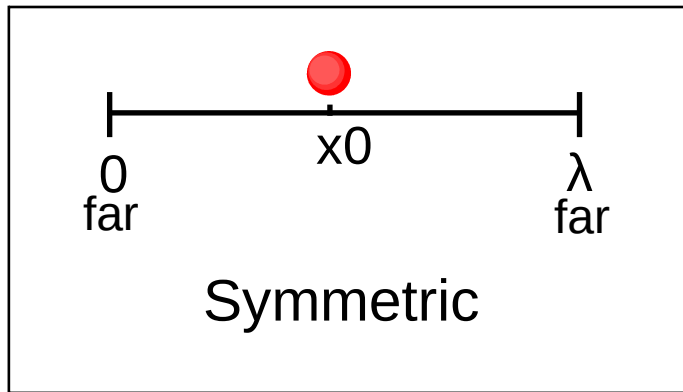
We note that:

$$\rho_n(x_n) = \int_{\ell_0}^{\lambda - \ell_0} \rho_{n-1}(x_{n-1})p(x_n - x_{n-1})dx_{n-1}, \quad \ell_0 \leq x_n \leq \lambda - \ell_0$$

The probability density for the walker to be at position x_n in the interval $[x_n, x_n+dx]$ after n steps can be defined as the probability density of being at the previous position x_{n-1} times the probability density of performing a step of length $x_n - x_{n-1}$. The integration accounts for all the possible previous positions x_{n-1} that lead to x_n .



Stochastic Optimal Foraging Theory



Ballistic \longleftrightarrow Brownian

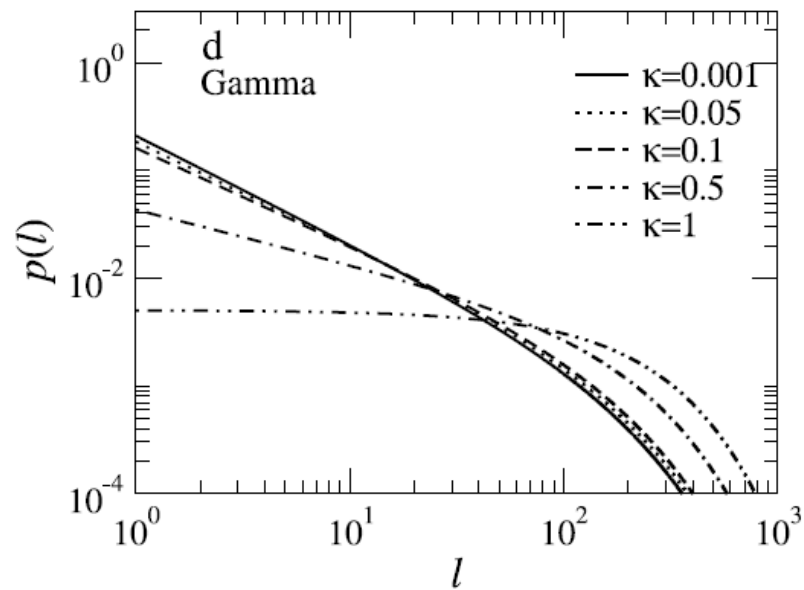
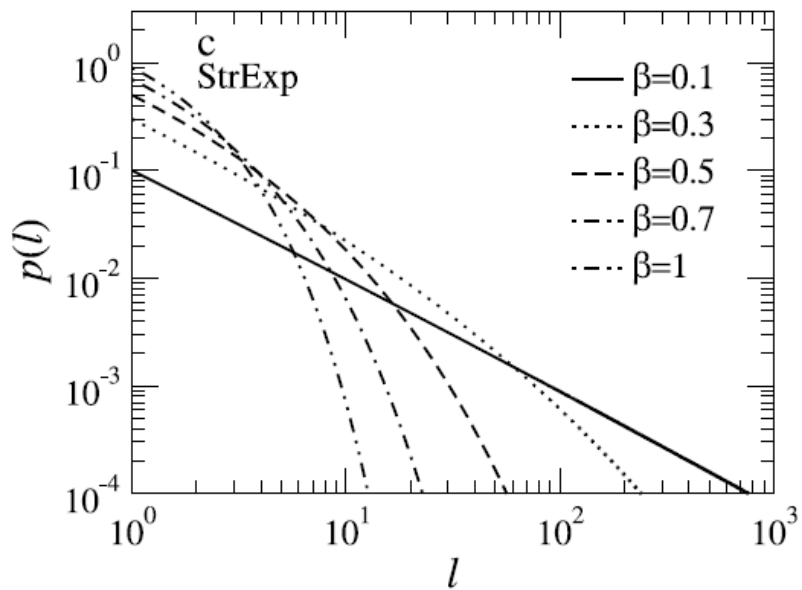
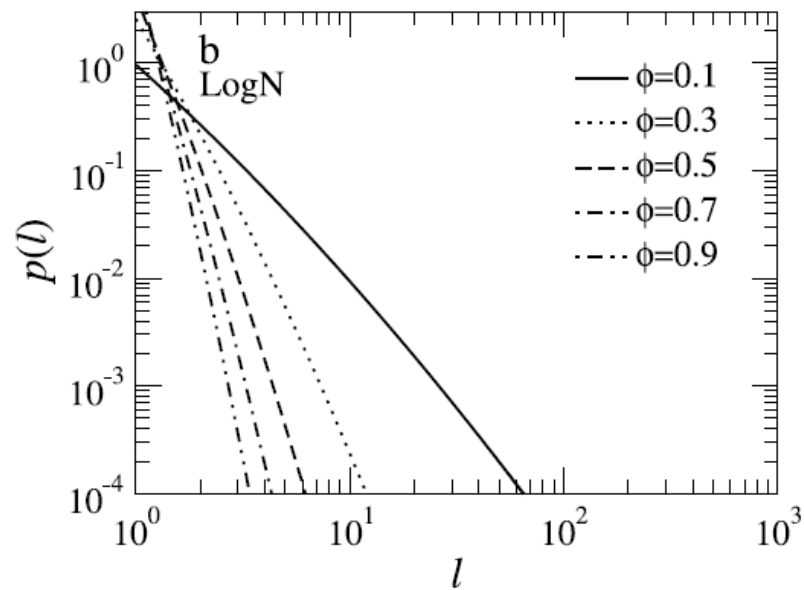
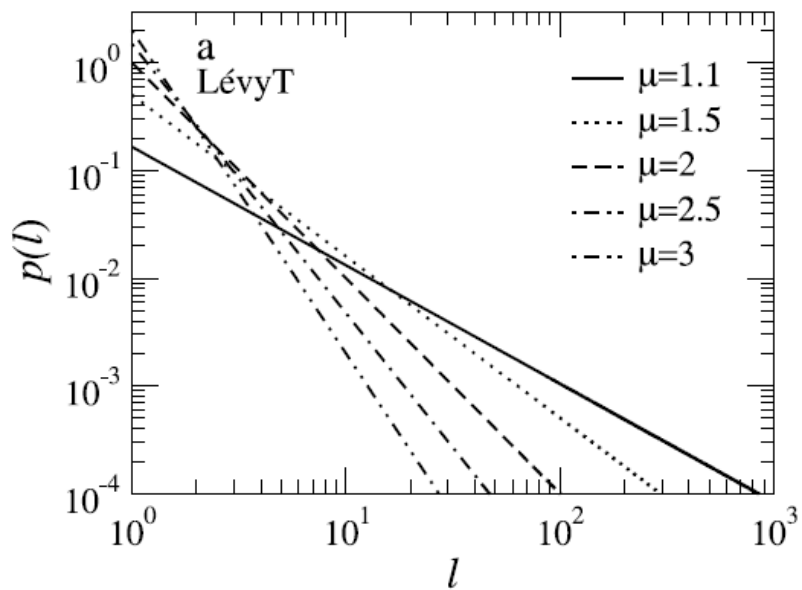
Ballistic \longleftrightarrow Brownian

Flight distributions

Stretched exponential	$p(\ell) = \alpha \Theta(\ell - \ell_0) \ell ^{\beta-1} \exp(-\phi \ell ^\beta)$
Gamma	$p(\ell) = \alpha \Theta(\ell - \ell_0) \ell ^{k-1} \exp(-\beta \ell)$
Lognormal	$p(\ell) = \frac{\alpha}{ \ell } \Theta(\ell - \ell_0) \exp\{-\phi [\ln(\beta \ell)]^2\}$
Power law	$p(\ell) = \alpha \frac{\Theta(\ell - \ell_0)}{ \ell ^\mu}$

where the theta function is such that $\Theta(|\ell| - \ell_0) = 0$ if $|\ell| < \ell_0$, and $\Theta(|\ell| - \ell_0) = 1$ otherwise

Flight distributions: two key parameters

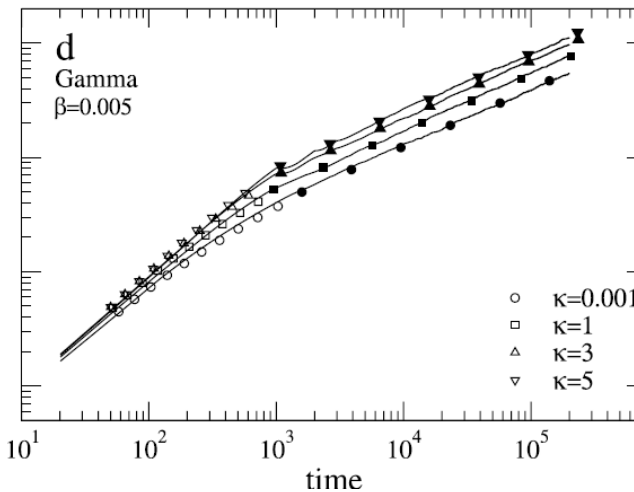
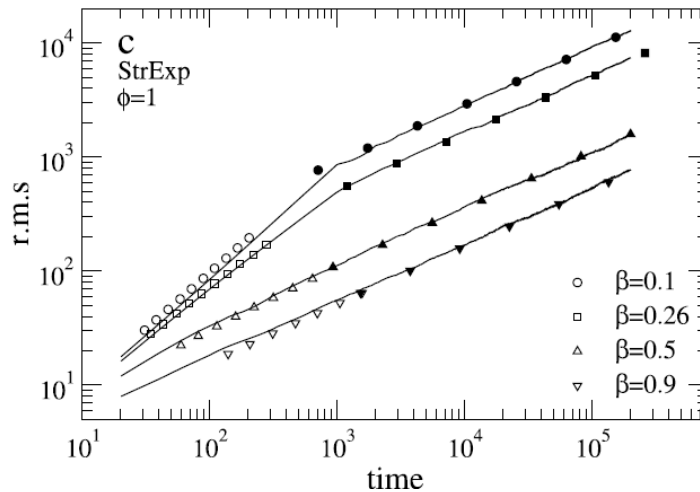
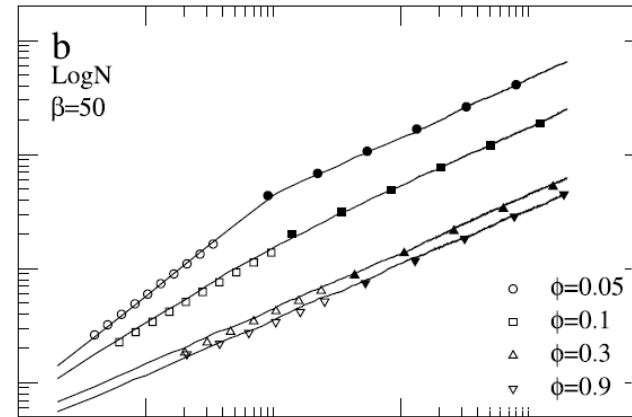
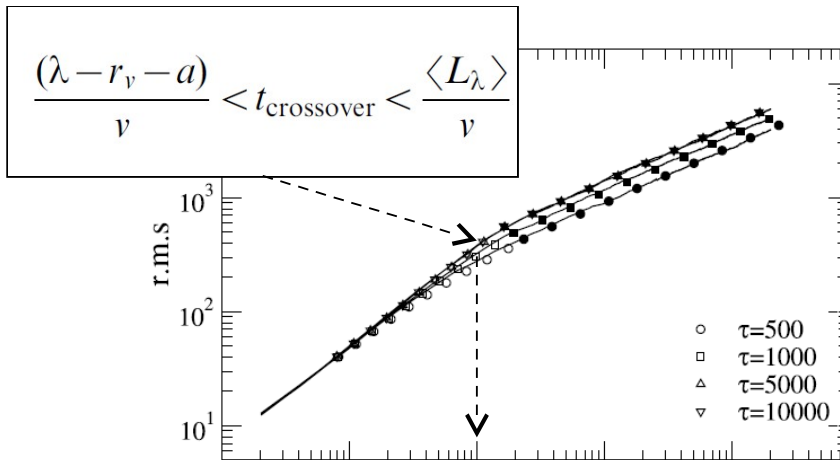


Root mean square displacement

Asymmetric condition

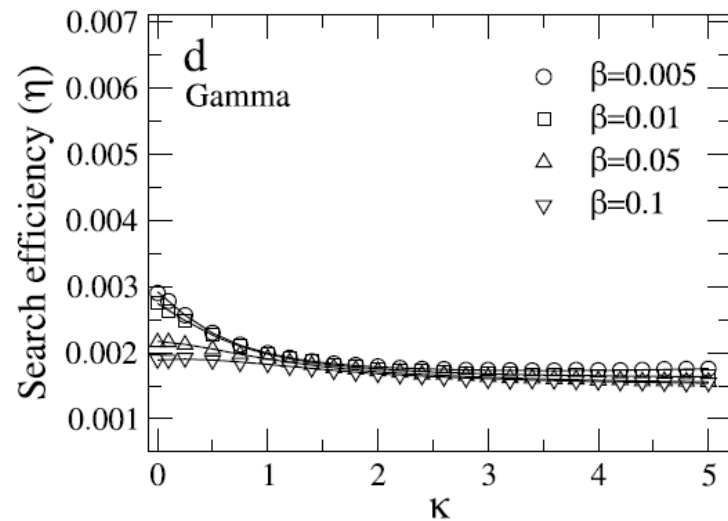
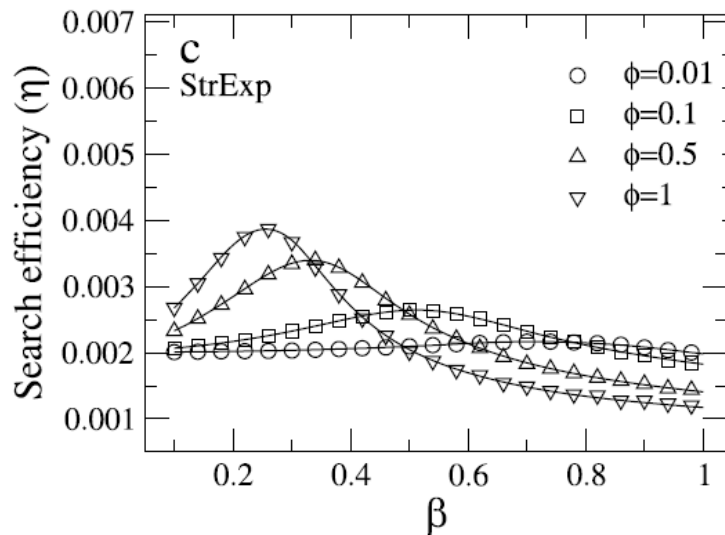
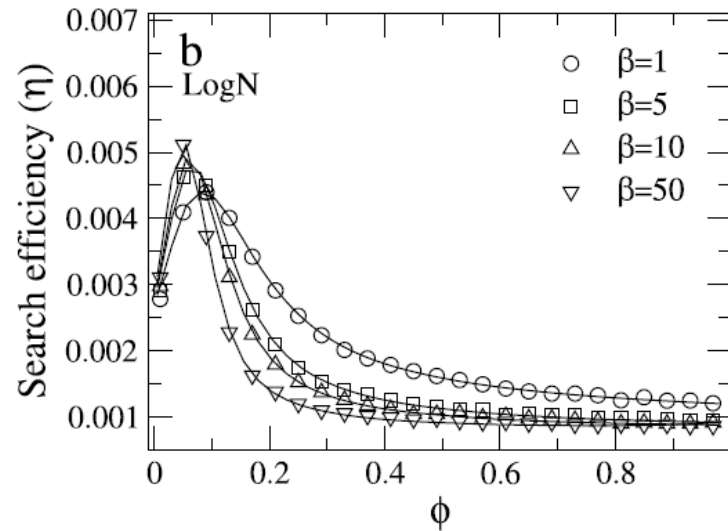
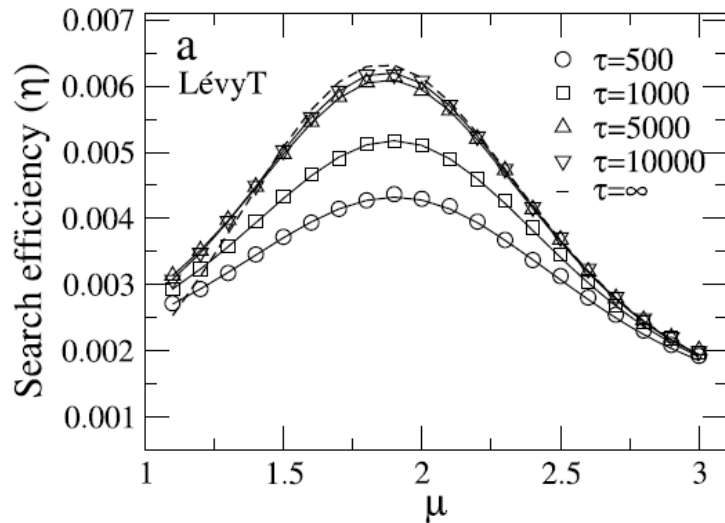
$$R \equiv [\langle (\Delta x)^2 \rangle]^{1/2} = [\langle x^2 \rangle - \langle x \rangle^2]^{1/2}$$

$$\left\{ \begin{array}{l} R_{\text{fpt}} = (\lambda - 2r_v)(p_0 p_\lambda)^{1/2} \\ R_{\text{Brownian}} = \left(\frac{\lambda^2 v p_\lambda}{\langle L \rangle} \right)^{1/2} t^{1/2} \end{array} \right.$$



The Search Efficiency (1/MFPT)

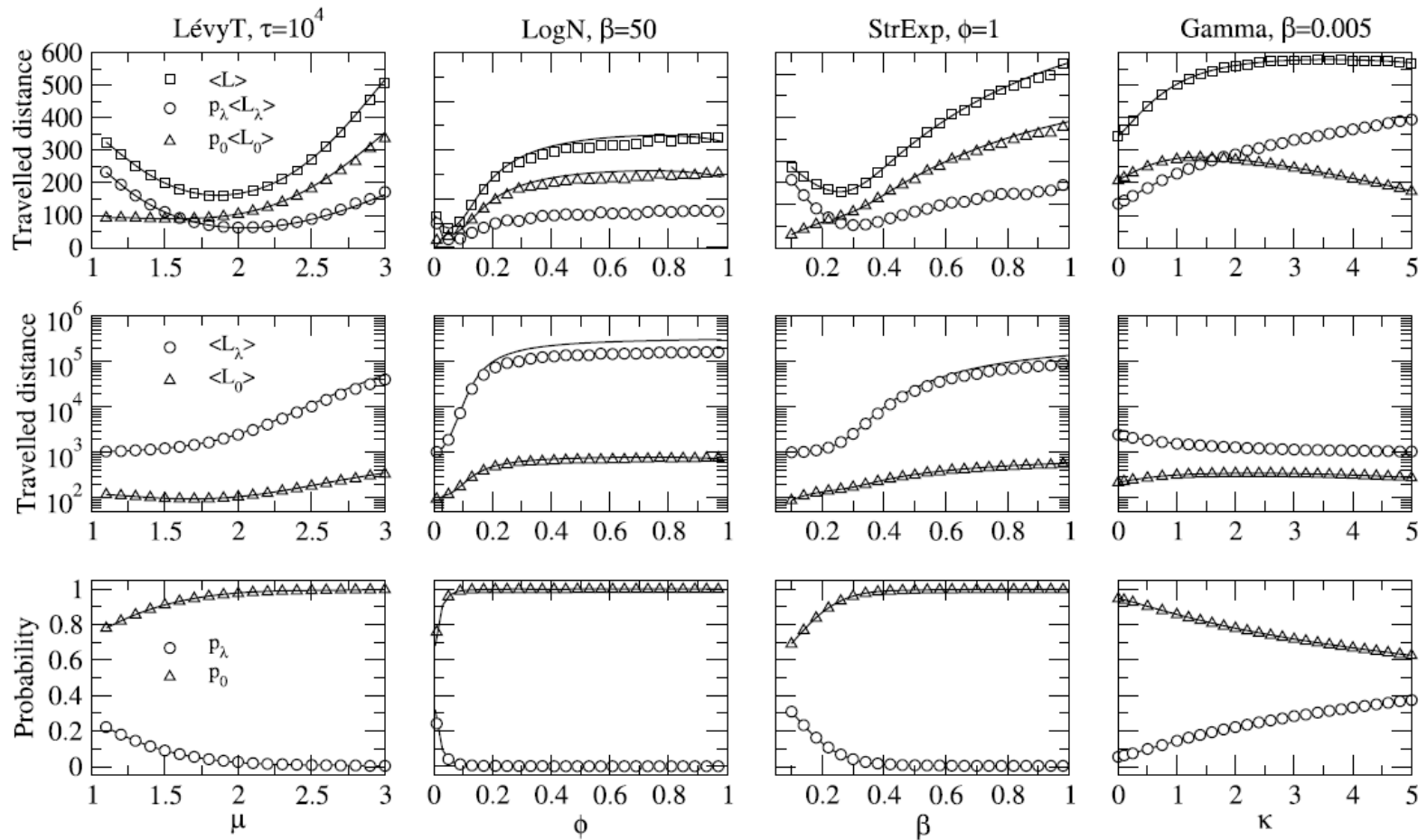
Asymmetric condition



The Search Efficiency: factorization

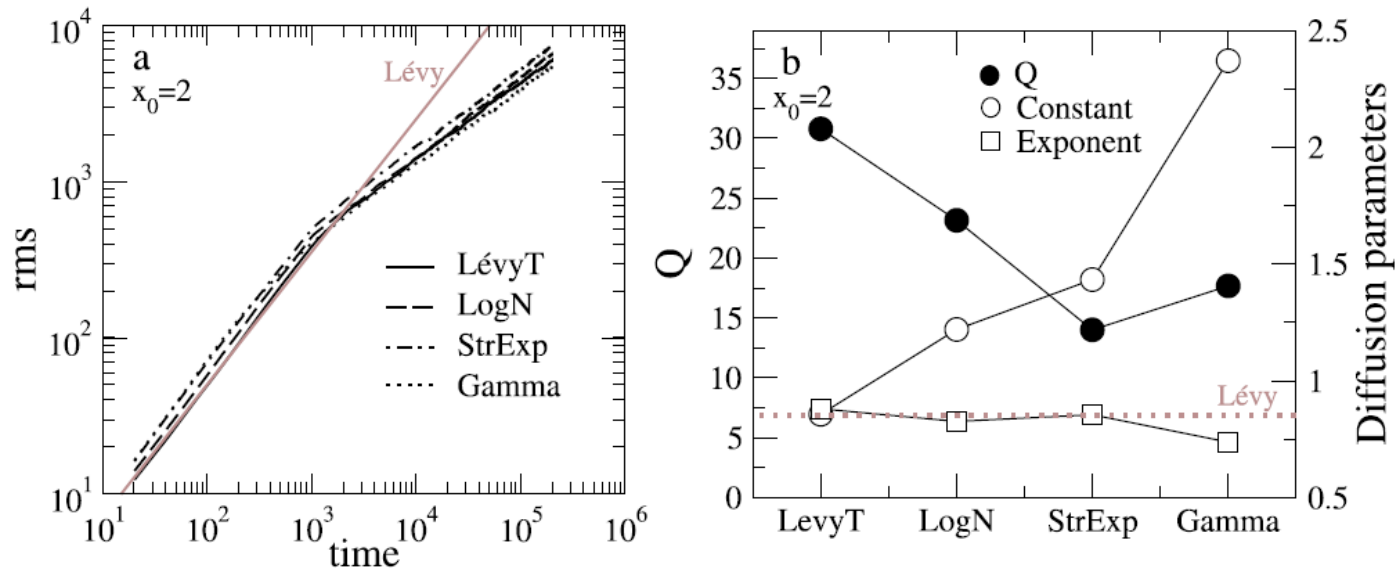
Asymmetric condition

$$\langle L \rangle = p_0 \langle L_0 \rangle + p_\lambda \langle L_\lambda \rangle$$



The Search Efficiency (1/MFPT)

Asymmetric condition

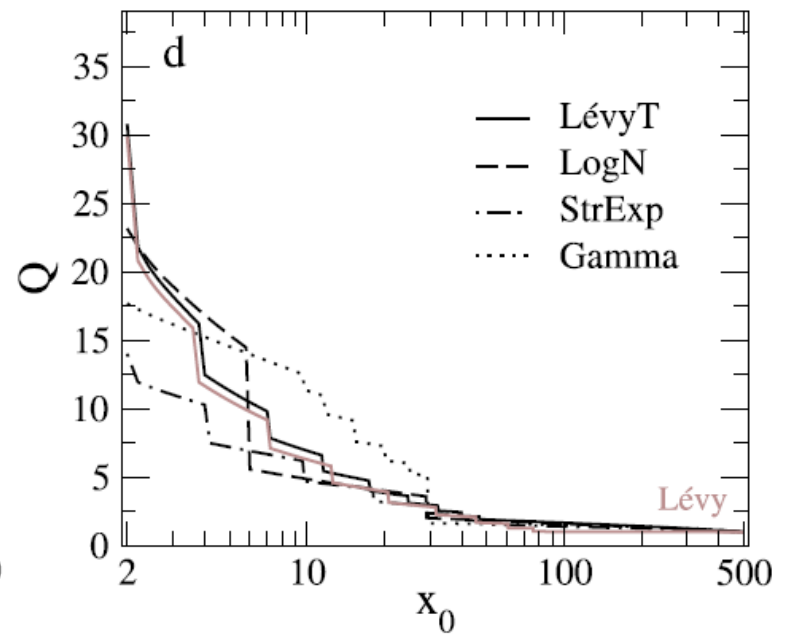
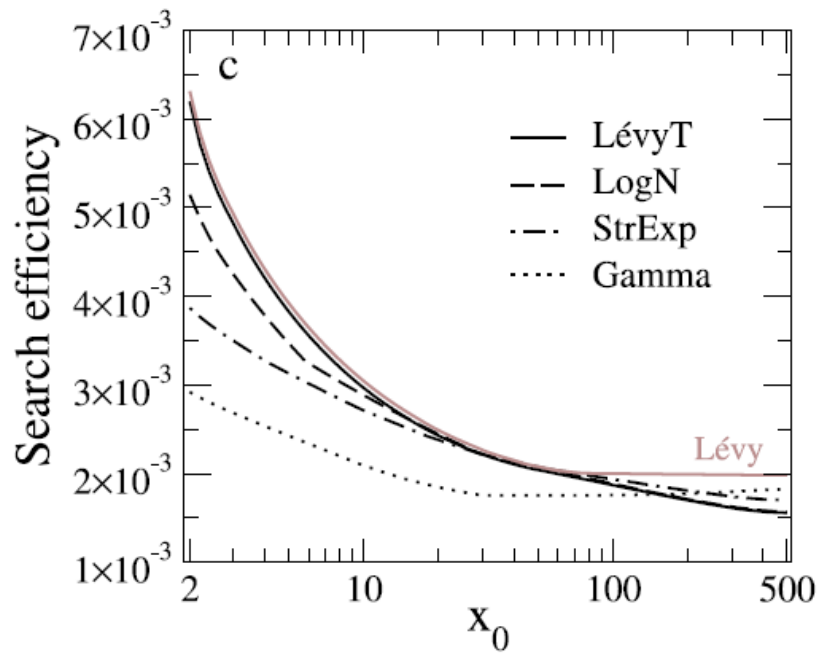


Ratio between the average numbers of encounters of the closest (first order revisit) and farthest (new and high order revisits) targets

$$Q = \frac{\langle N_0 \rangle}{\langle N_N \rangle + \langle N_H \rangle}$$

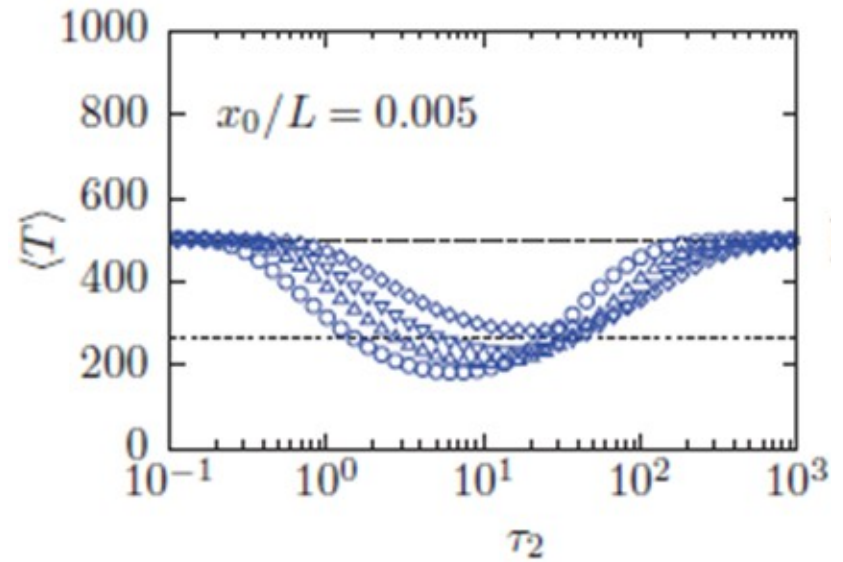
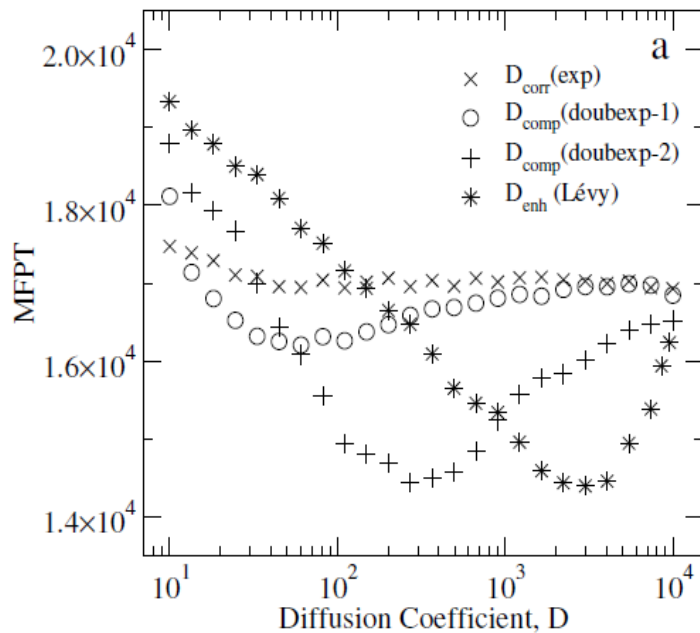
The Search Efficiency (1/MFPT)

From asymmetric to symmetric condition...



The Search Efficiency (1/MFPT)

A mixture of scales...with the right scales....beats Levy



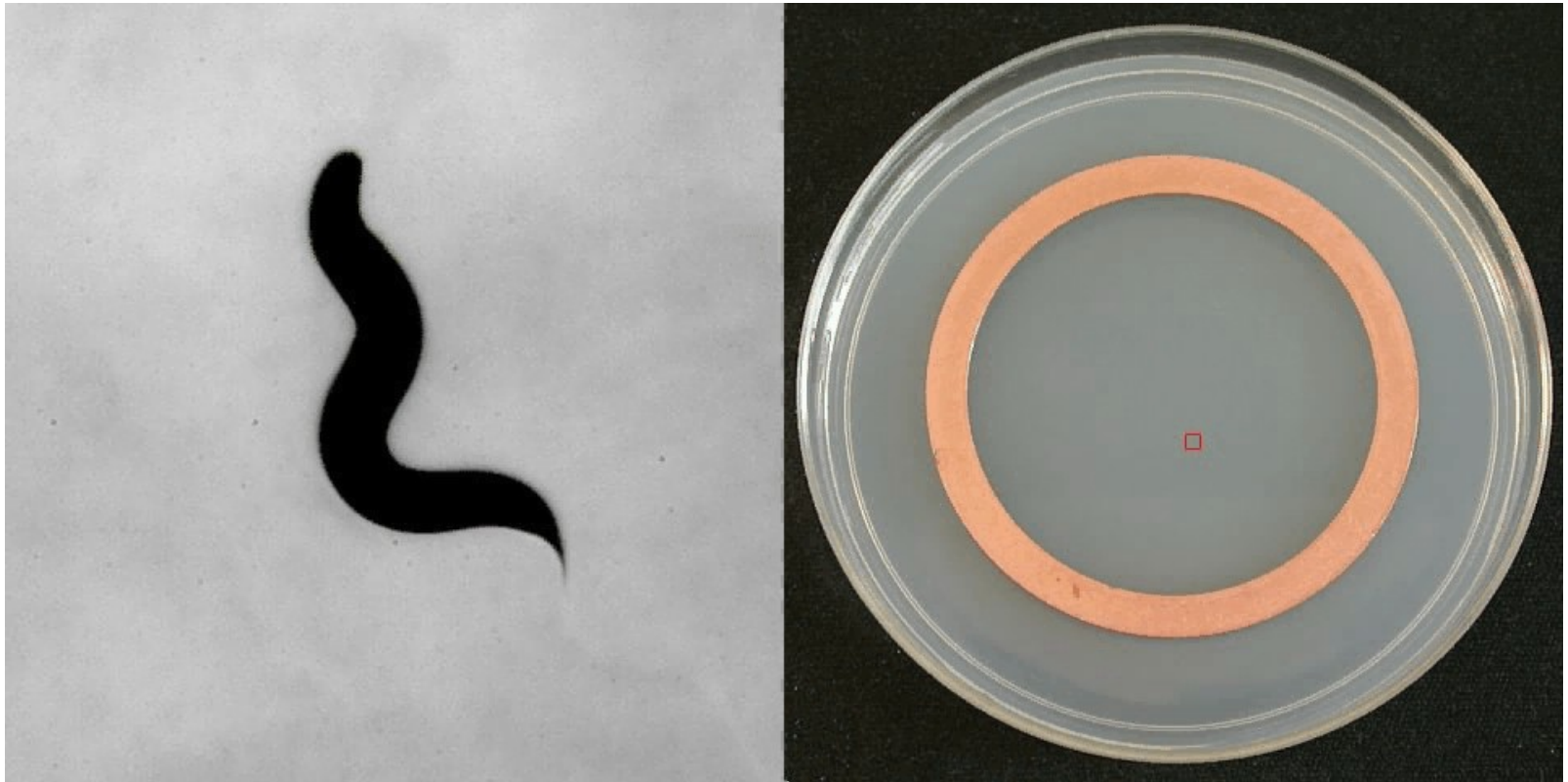
Why to turn at all?



To solve exploitation-exploration tradeoffs

- A) *Reorientation behaviour (time-to-reorientation) can control movement scales.*
- B) *If there is information about relevant landscape scales one should match reorientations to those particular scales.*
- C) *The larger the uncertainty the larger is the number of scales needed to solve the exploitation-exploration tradeoff*

Chaenorabditis elegans



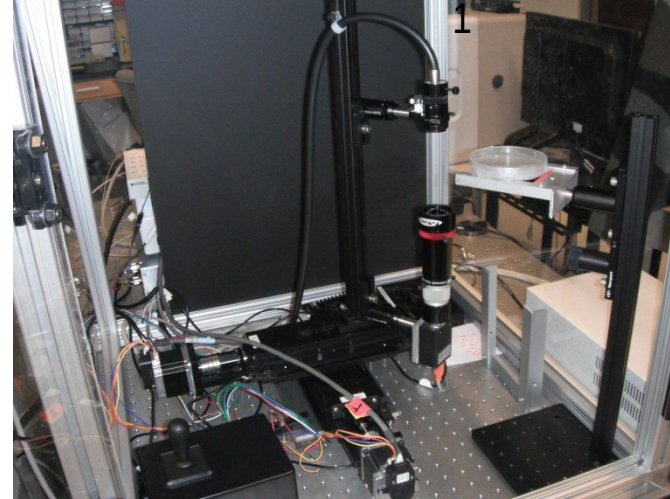
- > Locomotion includes crawling or swimming and they perform stereotyped turns
Omega / Reversals / Pirouettes / Pauses
- > Evidence of random movements and chemotaxis
- > Mutants (sensorial and motor) and engineering genetic techniques

Tracking system and behavioral annotation

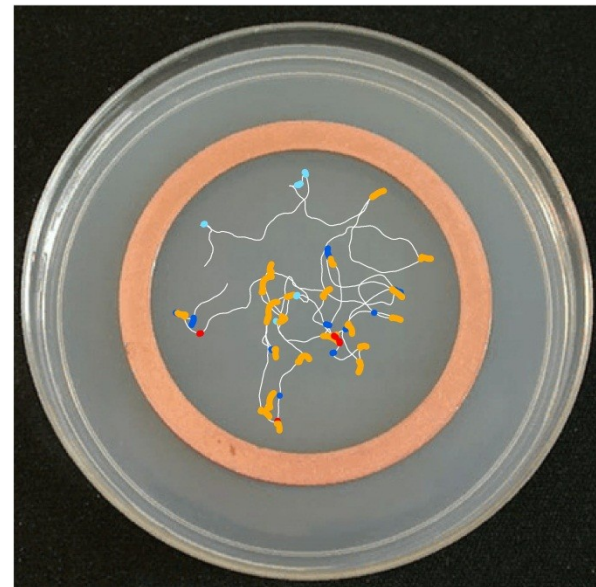
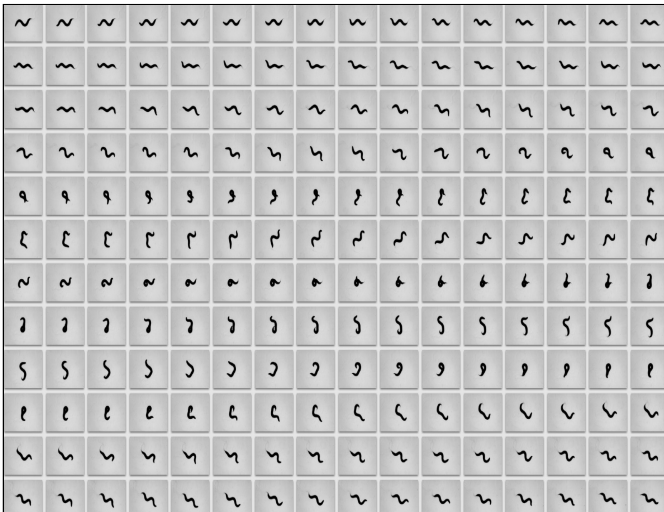
W.Ryu Lab. U.Toronto, Canada



4 frames sec-



Behaviour



Turning behaviour in a random search

How to turn?

- Abruptly (reorientation)
- Smoothly (persistent curvature)



When to turn?

- Time between turns
- Curvature control (loops)



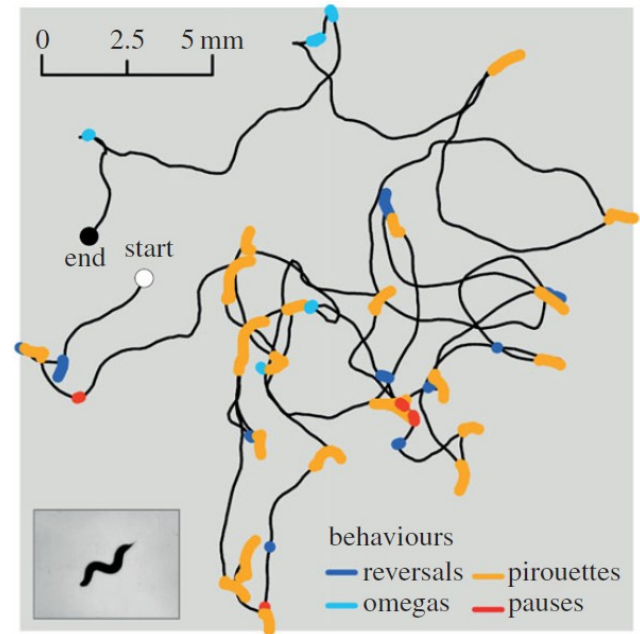
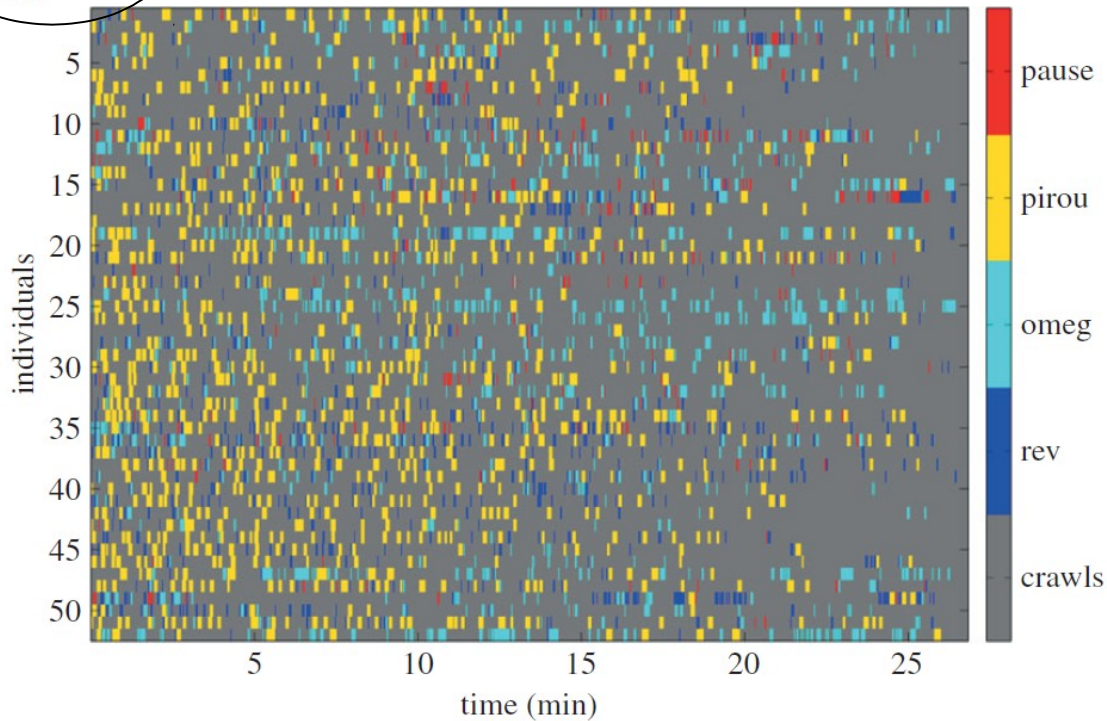
Search efficiency, space use, revisitability...

Simple relocation experiment: from food to no-food

food

(c)

No food (but no starvation). Minimal cues.

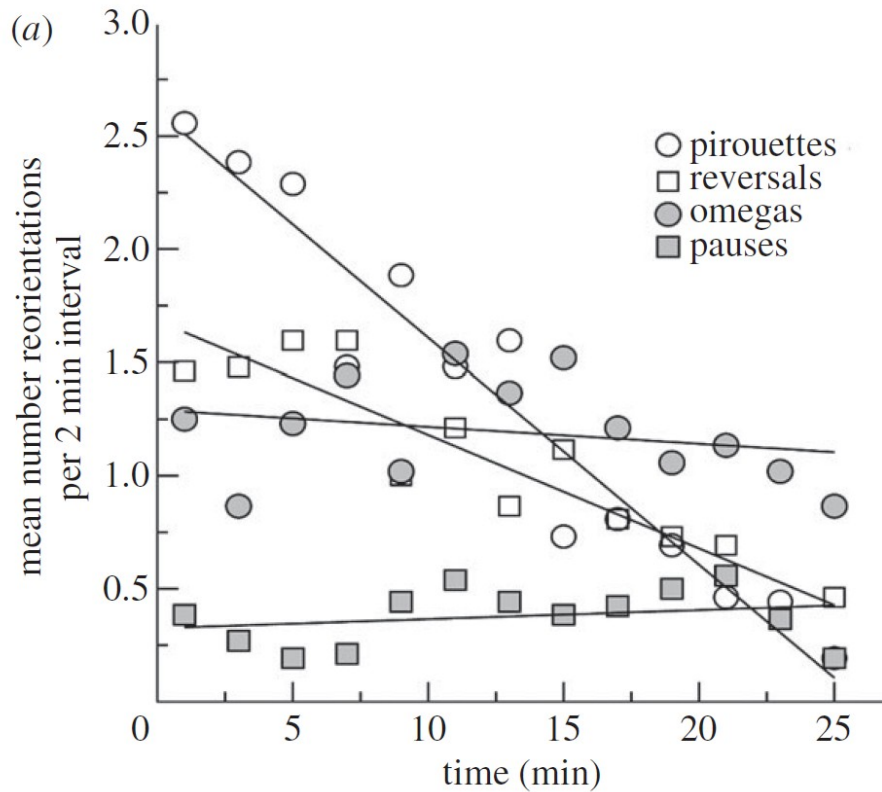


Salvador et al. (2014)
J Royal Soc Interface

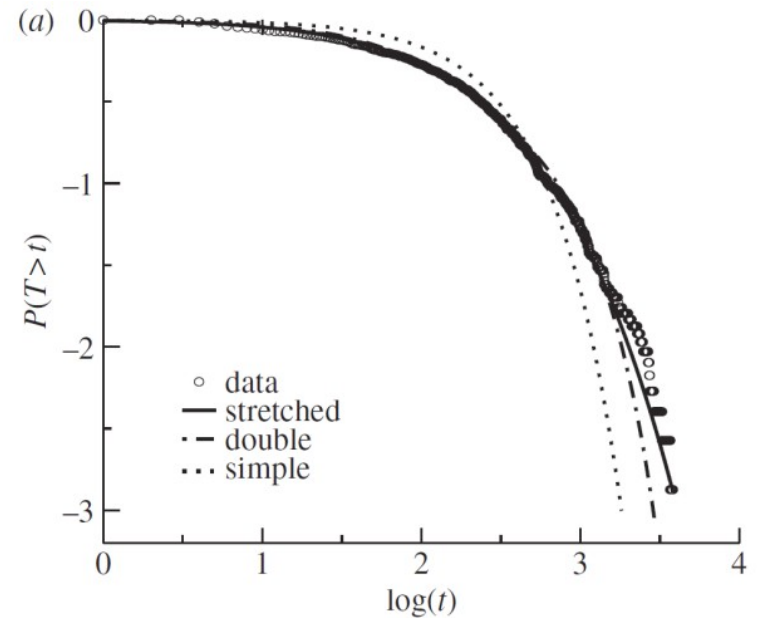
Different temporal dynamics for different turn types

food

No food (but no starvation). Minimal cues.

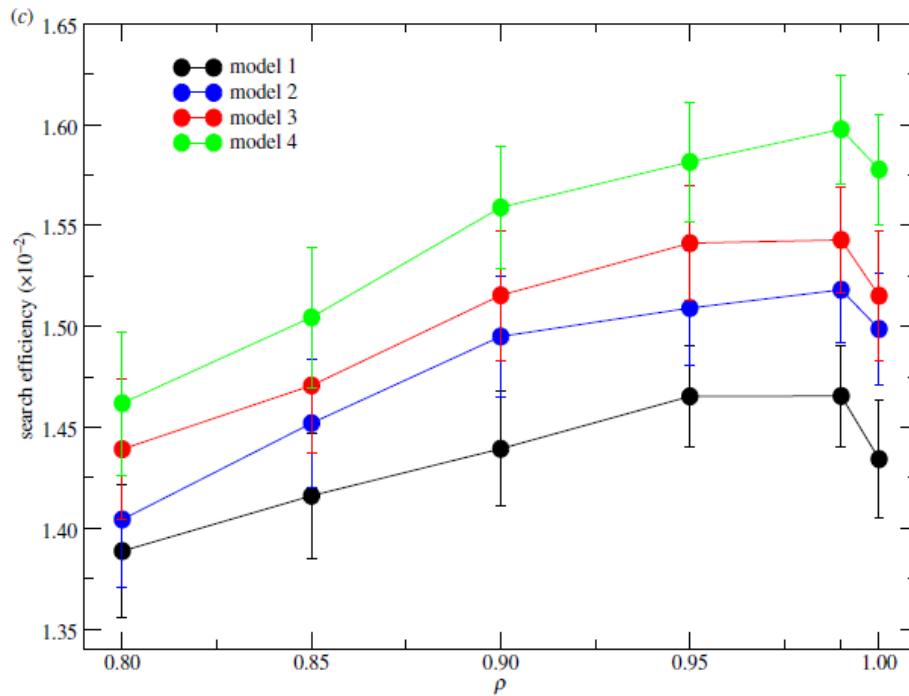
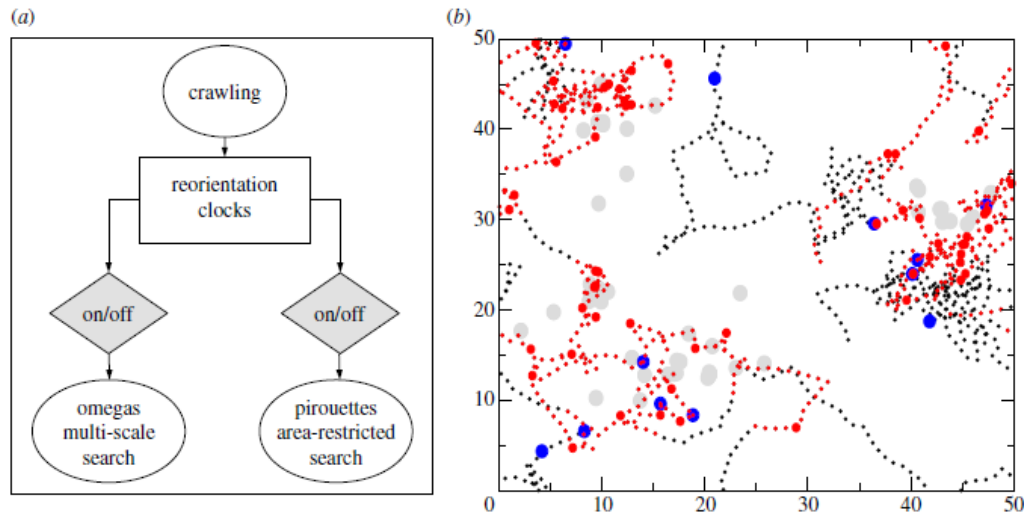


Pirouettes "food memory"
Omegas "template"



Omega turns:
heavy-tail distribution

Toy model



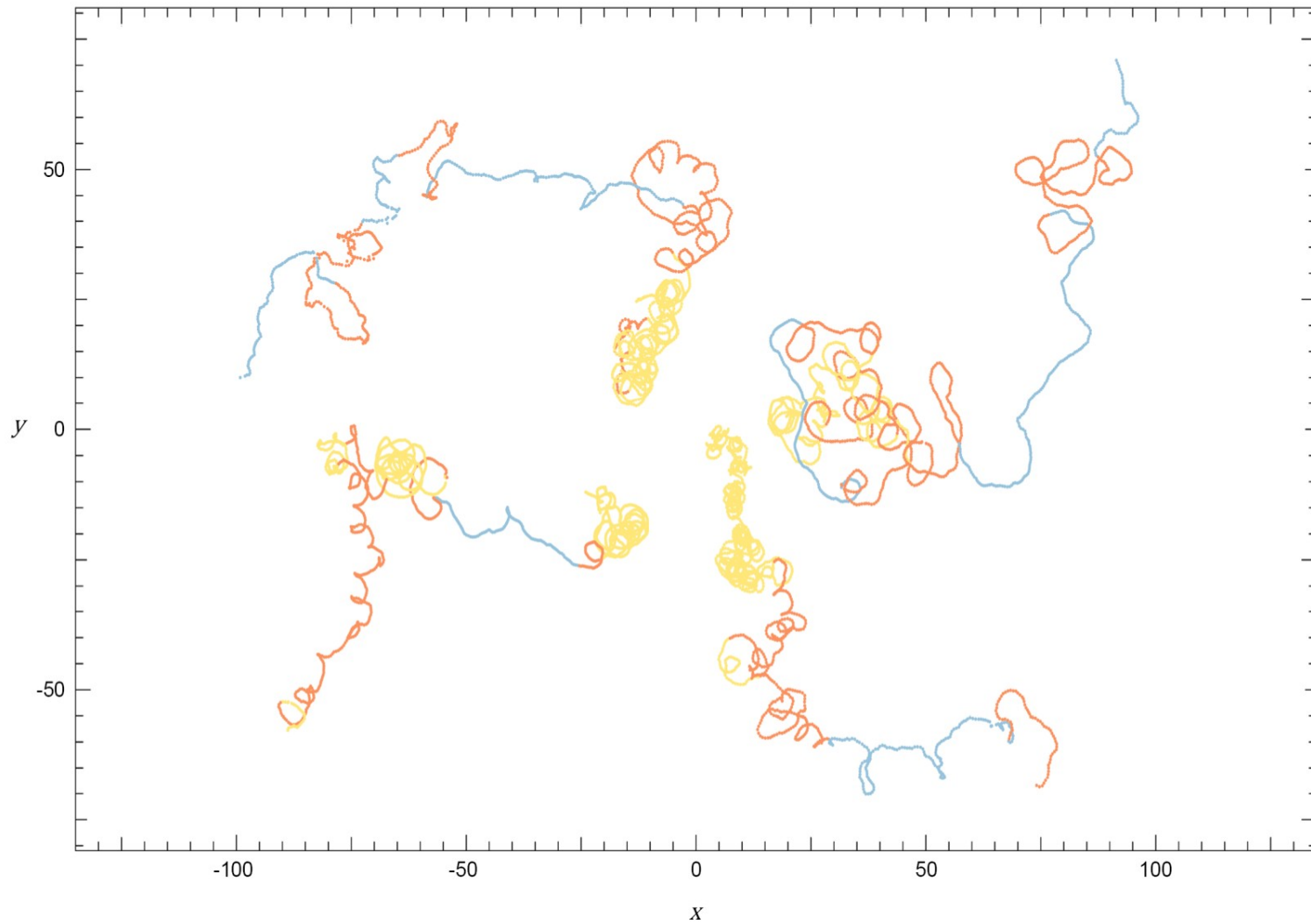
CRW + omegas + pirouettes

CRW + Exp (pirouettes)

CRW + strExp (omegas)

CRW

Behavioural Annotation: *Curvature*



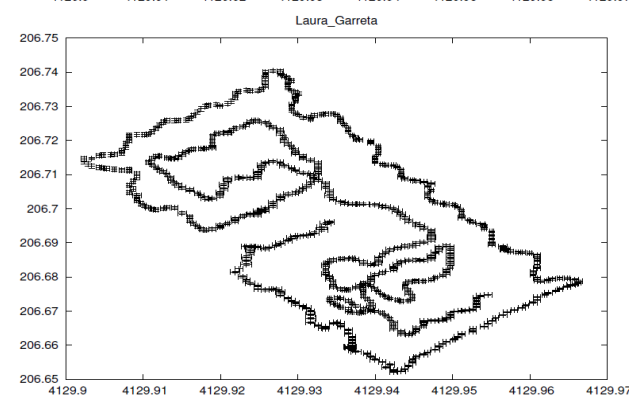
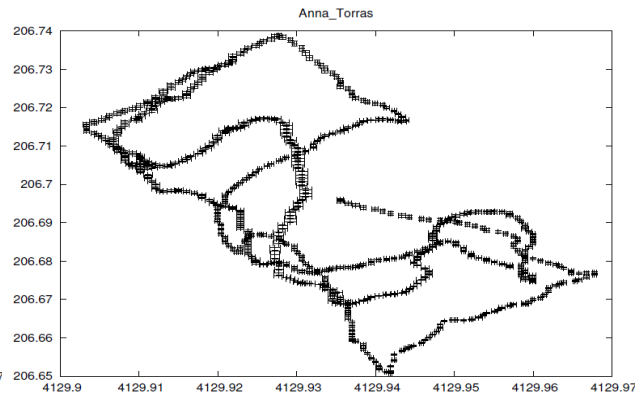
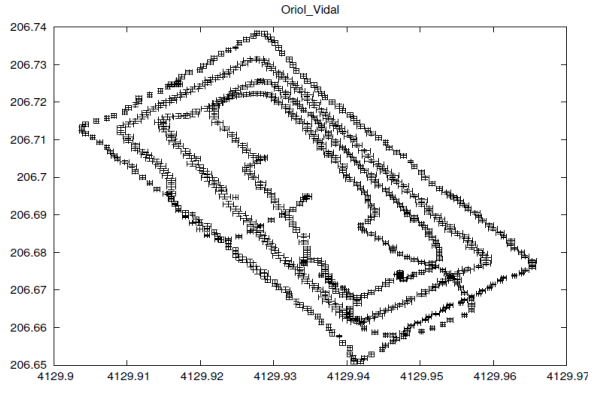
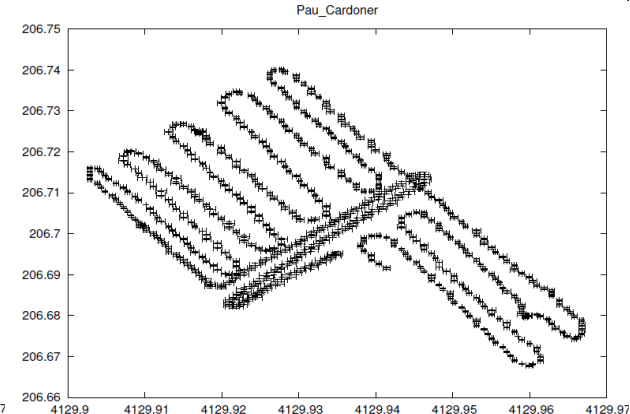
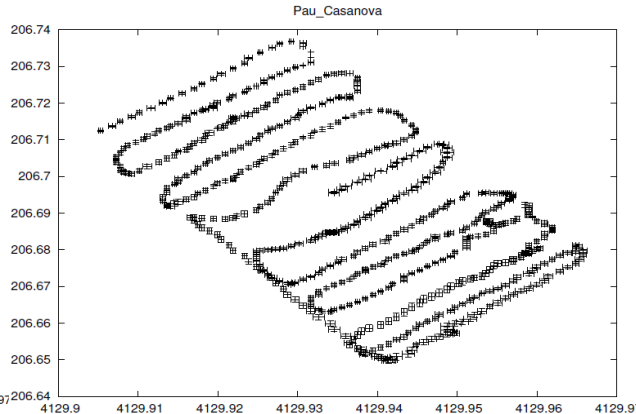
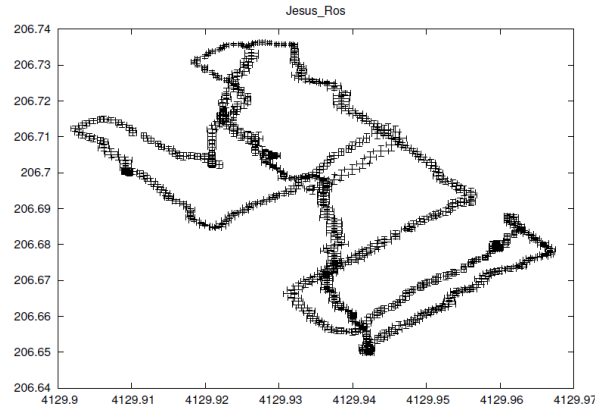
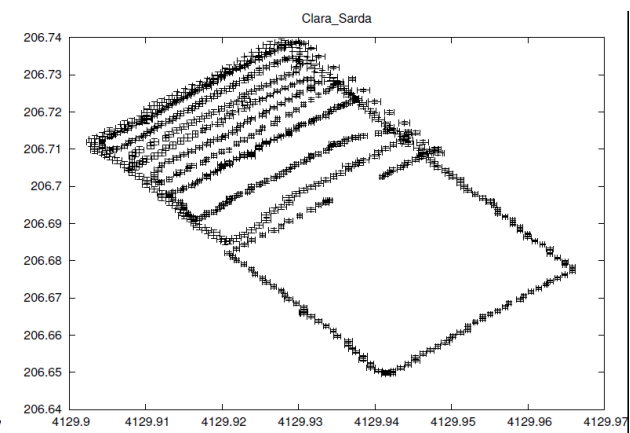
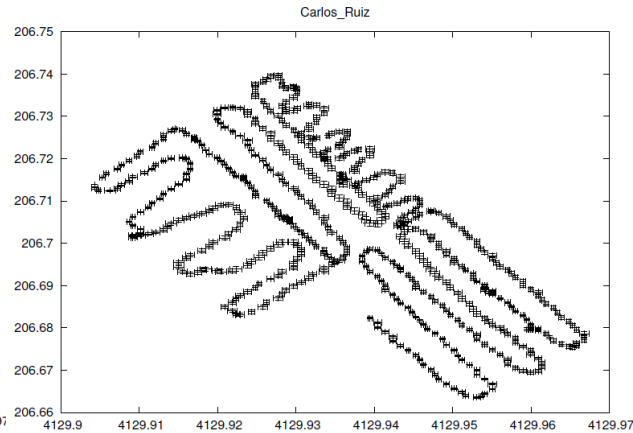
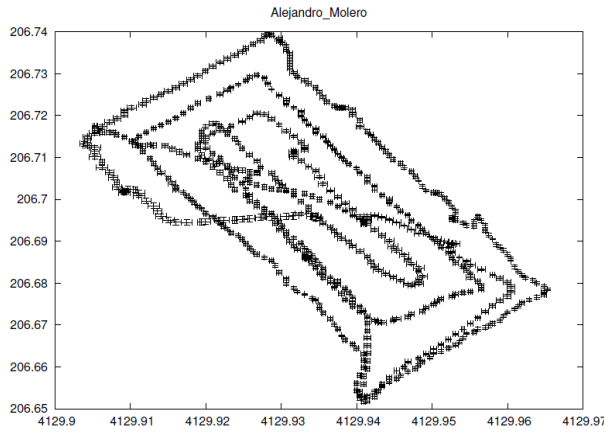
Human Search Strategies



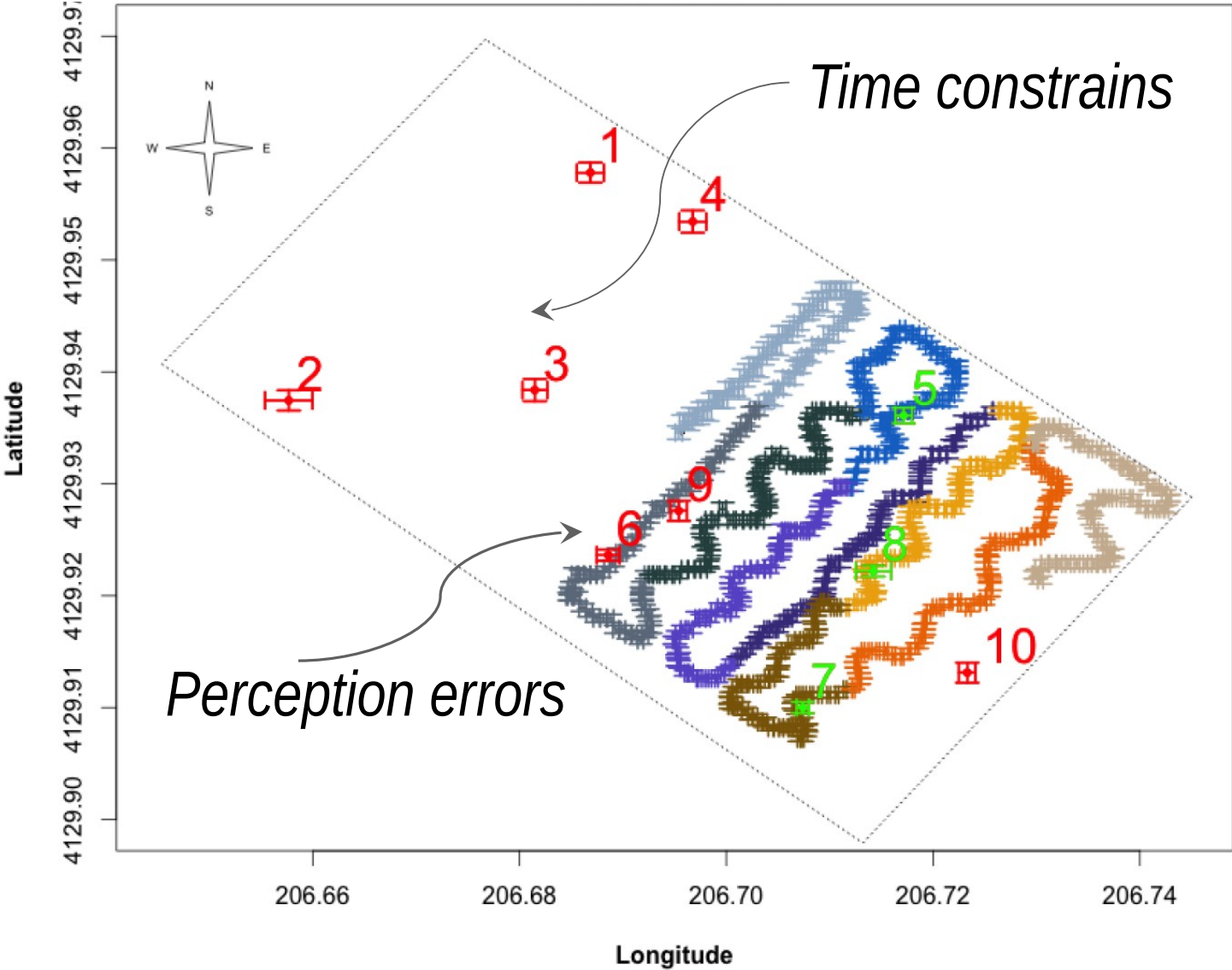
10 minutes...10 coins...of 10 cents



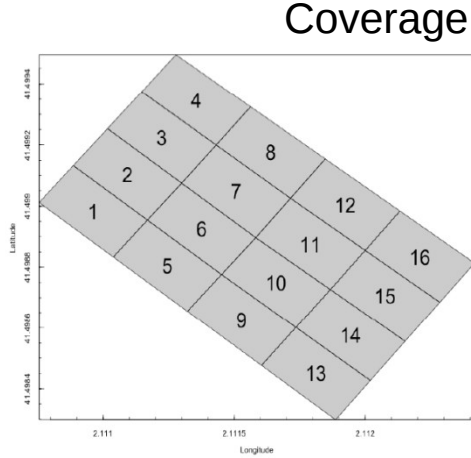
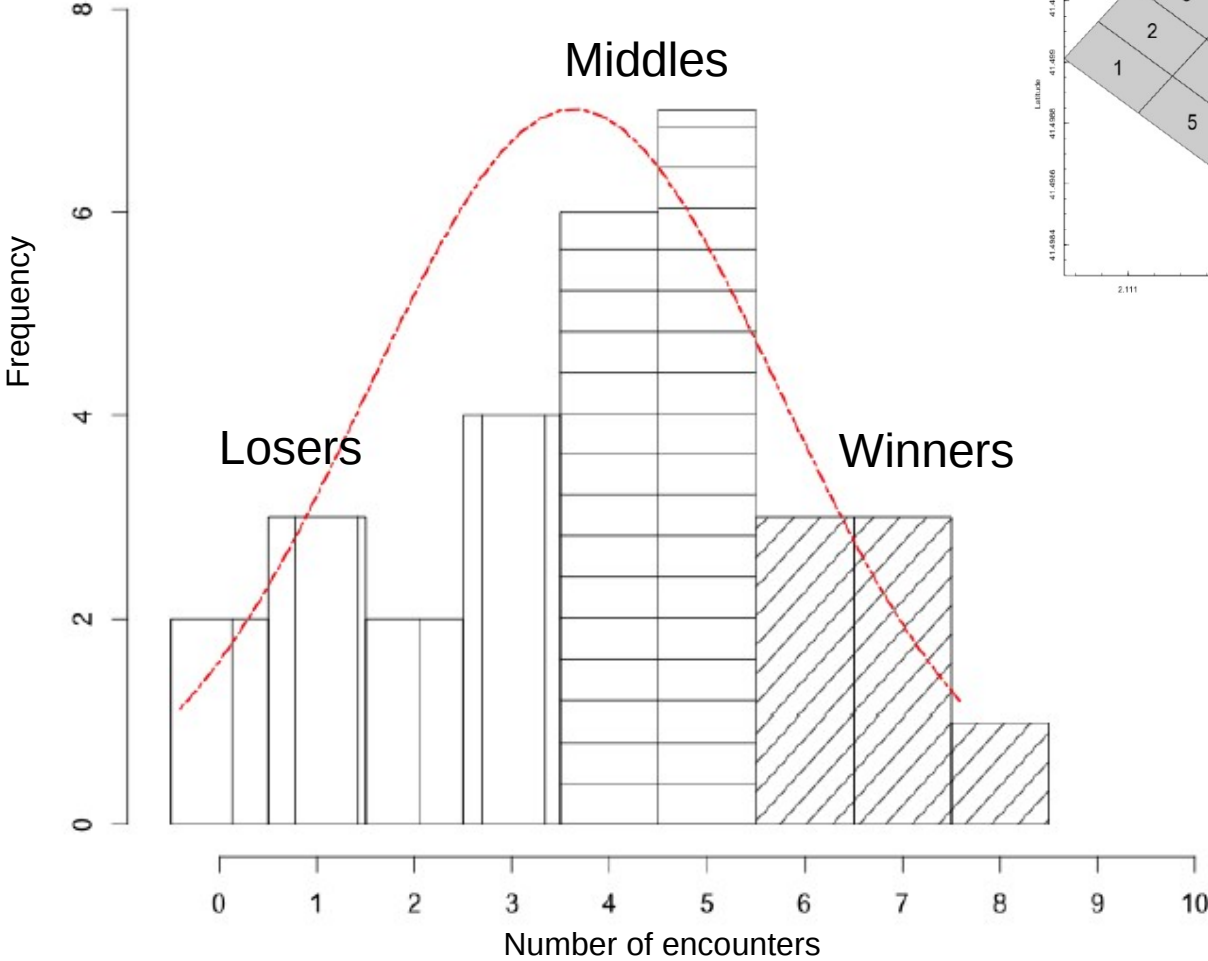
Human Search Strategies: systematic rules



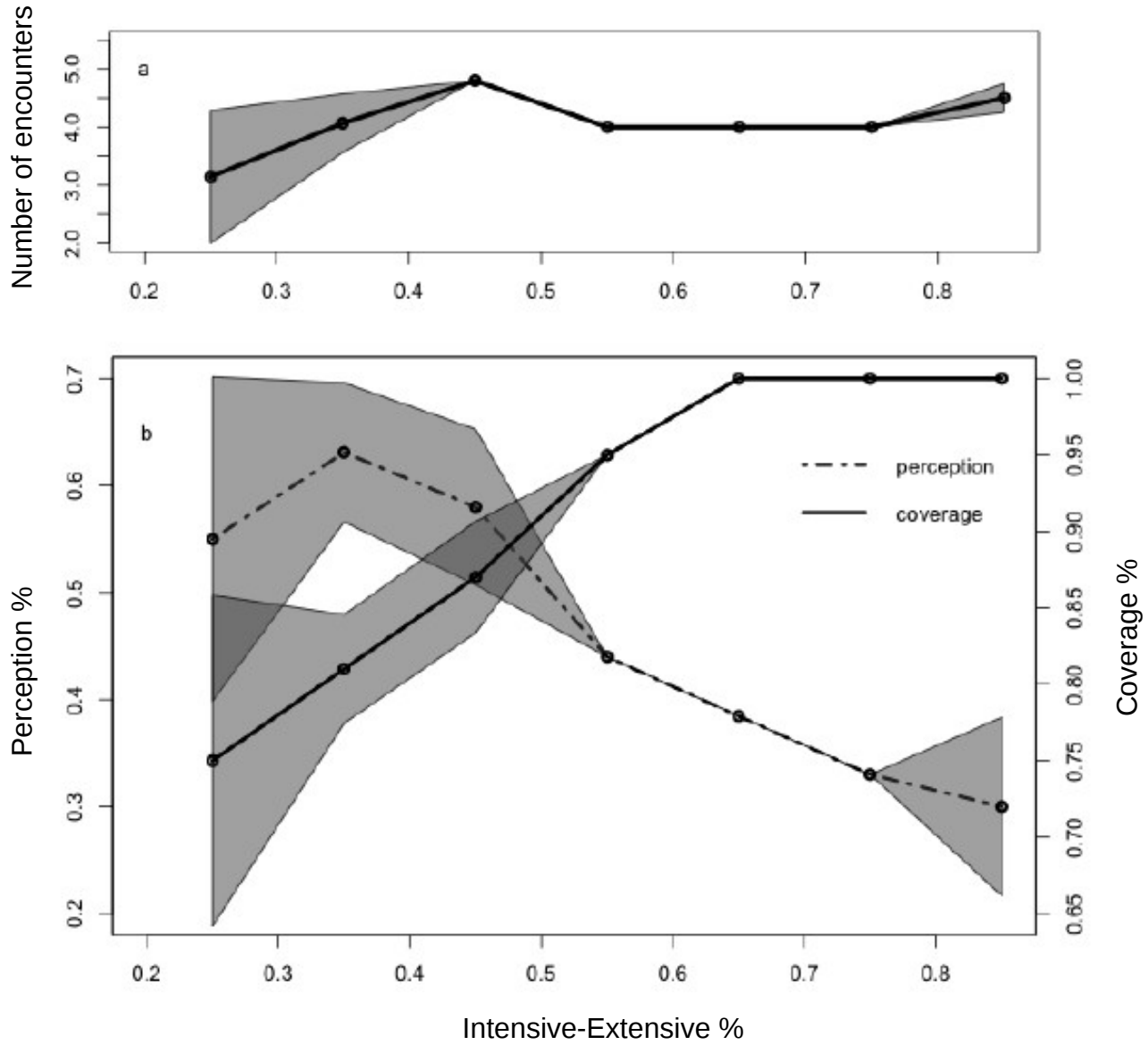
Human Search Strategies: systematic rules



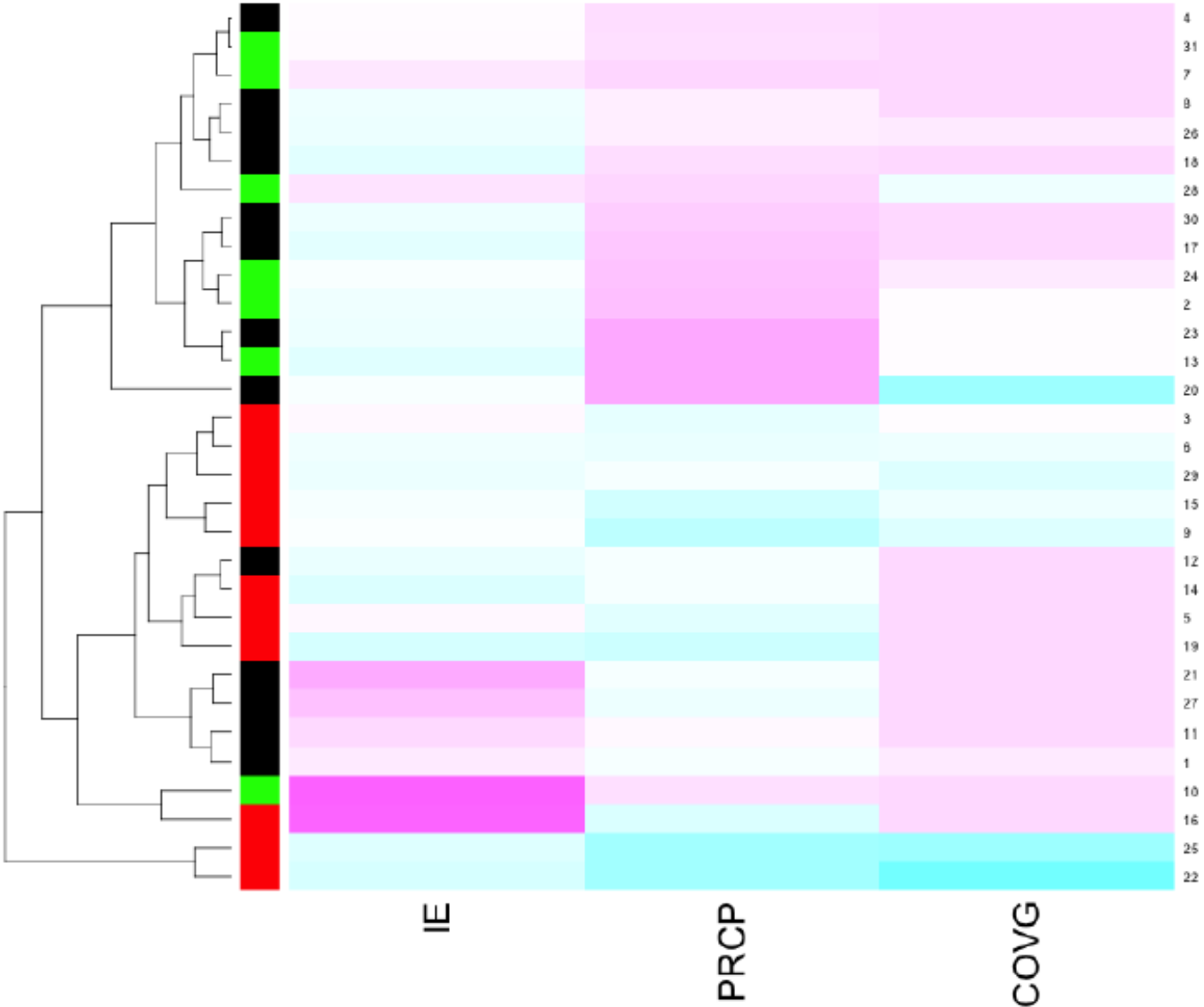
Human Search Strategies: systematic rules



Intensive-Extensive Tradeoff



Search Analysis



WINNERS

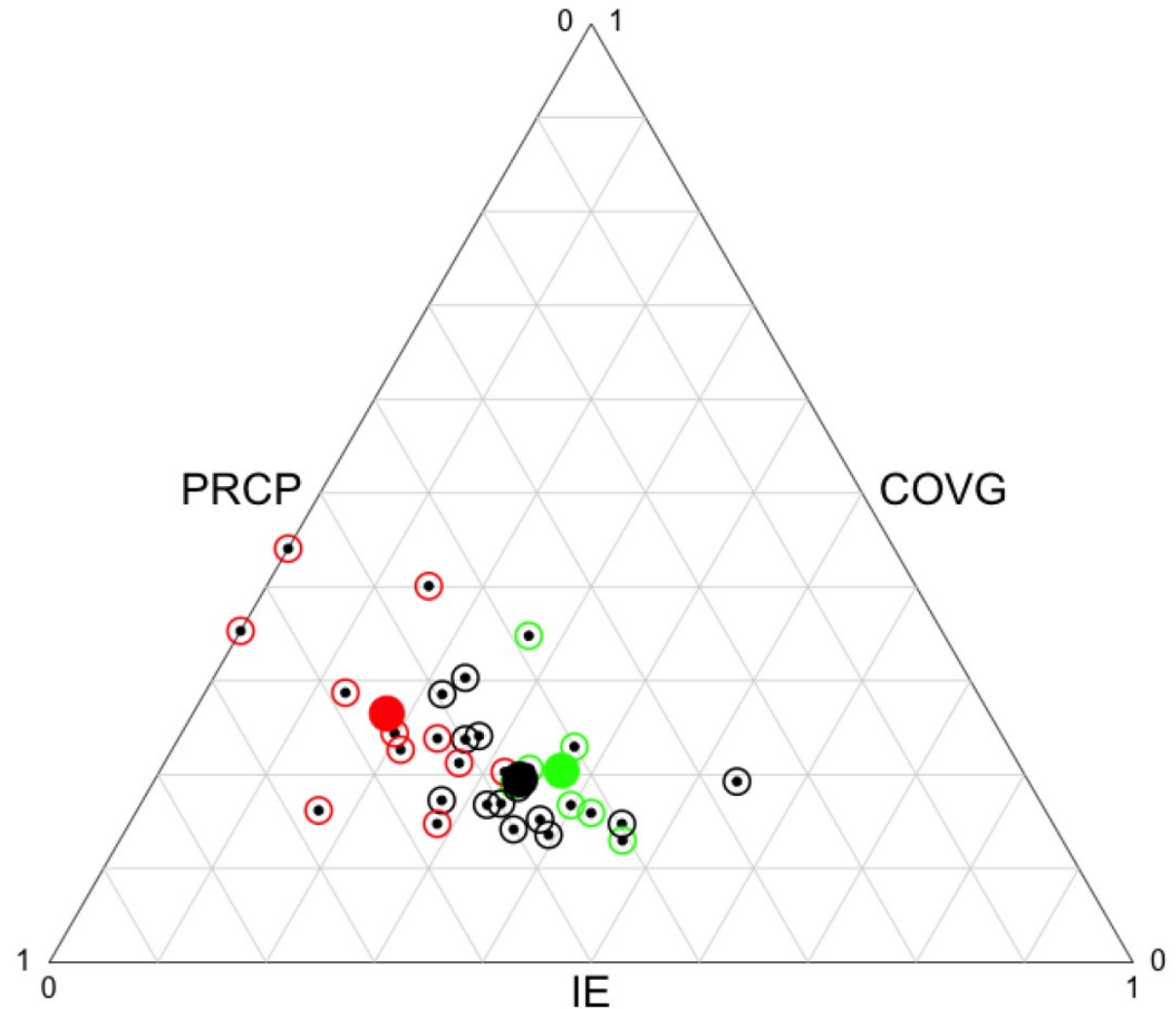
- High perception
- Medium coverage
- High Extensive

LOSERS (2 types)

- Low perception
- High coverage
- Too intensive

MIDDLES

- High perception
- Good coverage
- Extensive (ballistic)



Marco, J. Msc. (UAB, 2014)

Campos et al. 2015 (in preparation)



MOVOLAB

THANKS



PubliBlanes.net