

MARE-Madeira 2025

# *Optimizing animal tracking projects*



Using the '**move design**' application

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What can I do with the data  
I've already collected?

What are your priorities?

I want to know everything!



Define research questions



Identify spatiotemporal scales



Choose sampling design



Collect animal tracking data



Analyze data, mitigate biases



Assess conclusions



Define research questions



Identify spatiotemporal scales



Choose sampling design



Collect animal tracking data



Analyze data, mitigate biases



Assess conclusions



Cost of devices (& data transfer), challenges during deployment, and technological limitations,

can all constrain study design.

“

To consult the statistician after an experiment is finished is often merely to ask him to conduct a postmortem examination. He can perhaps say what the experiment died of.”



## Introduction

### ■ Fine-scale processes:

**SPEED/DISTANCE** — capture how far animals travel (and rate at which these distances are covered).

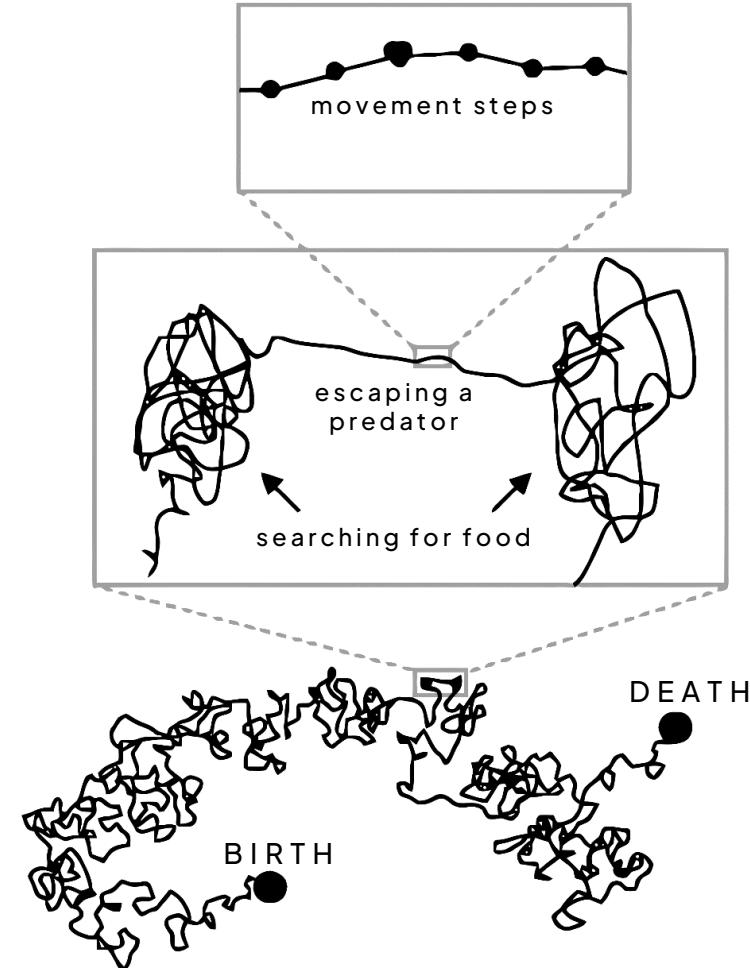
- To link **behavior and energetics**,
- As **indicators** of anthropogenic disturbance.

### ■ Large-scale processes:

**HOME RANGE** — capture the area repeatedly used throughout an animal's **lifetime**.

- For **protected area delineation**,
- To reduce **human-wildlife conflict**,
- To control the **spread of infectious diseases**.

Increasing spatiotemporal scale

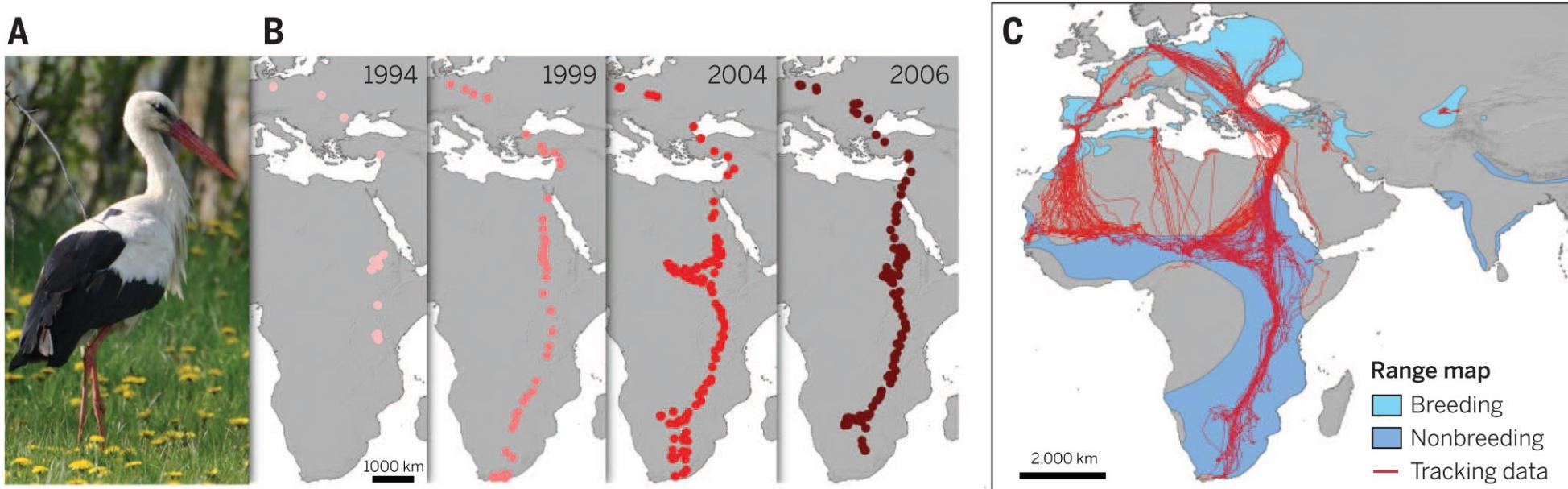


❧ Adapted from **Nathan et al. (2008)**



## Introduction

A white stork (*Ciconia ciconia*) (**A**) was tagged with a GPS tracking device as a 3-year-old nonreproductive juvenile in Germany in 1994, and was tracked until her death in 2006 (**B**).



🔗 Adapted from **Kays et al. (2015)**

DOI: [10.1126/science.aaa2478](https://doi.org/10.1126/science.aaa2478)

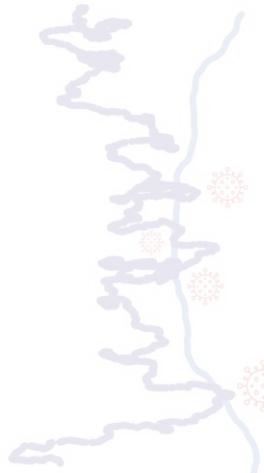


## Introduction

Nathan et al. (2022)

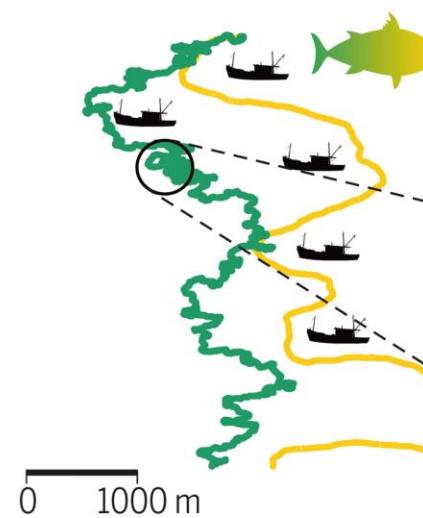
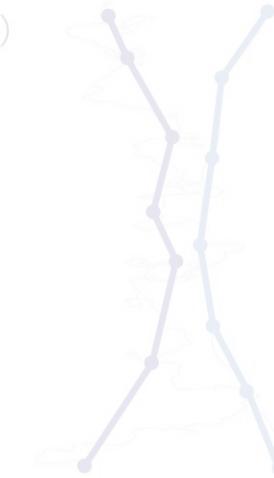
### Higher resolution

(5 s intervals)

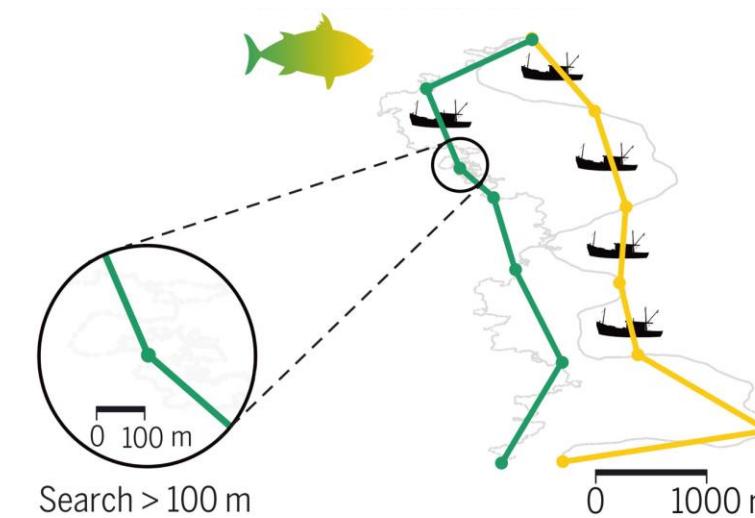


### Lower resolution

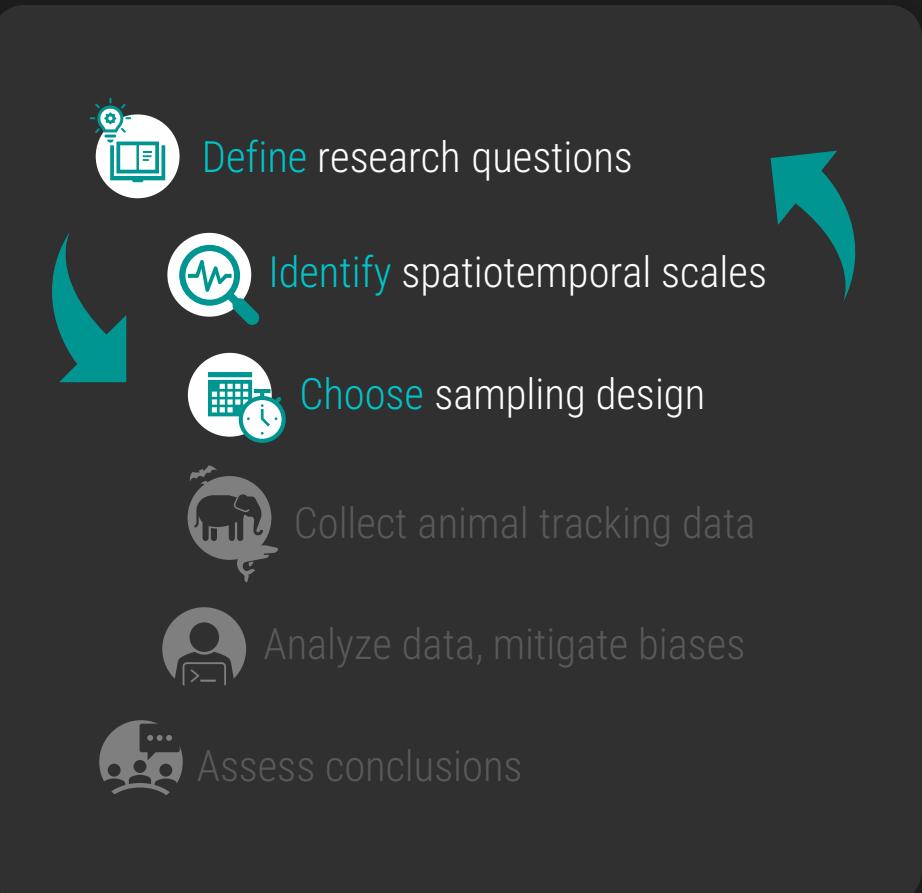
(30 min intervals)



Search < 100 m



Search > 100 m





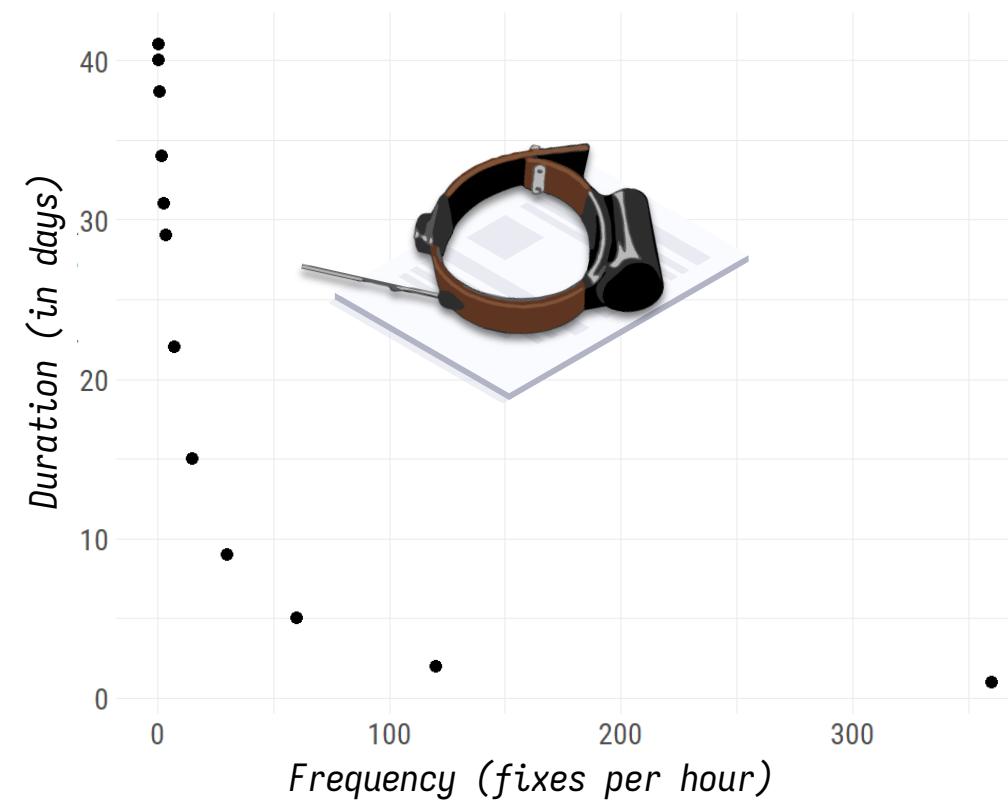
Trade-off between long battery life and high resolution of GPS devices.

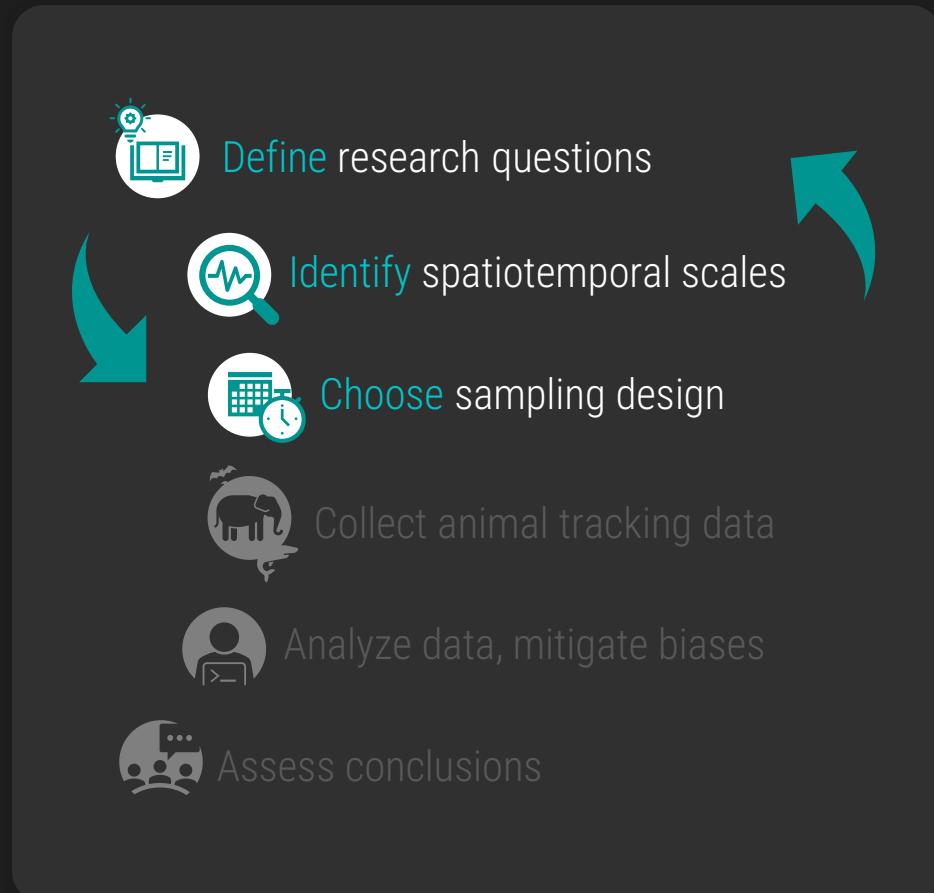


VHF  
Radio-telemetry



GPS  
technology

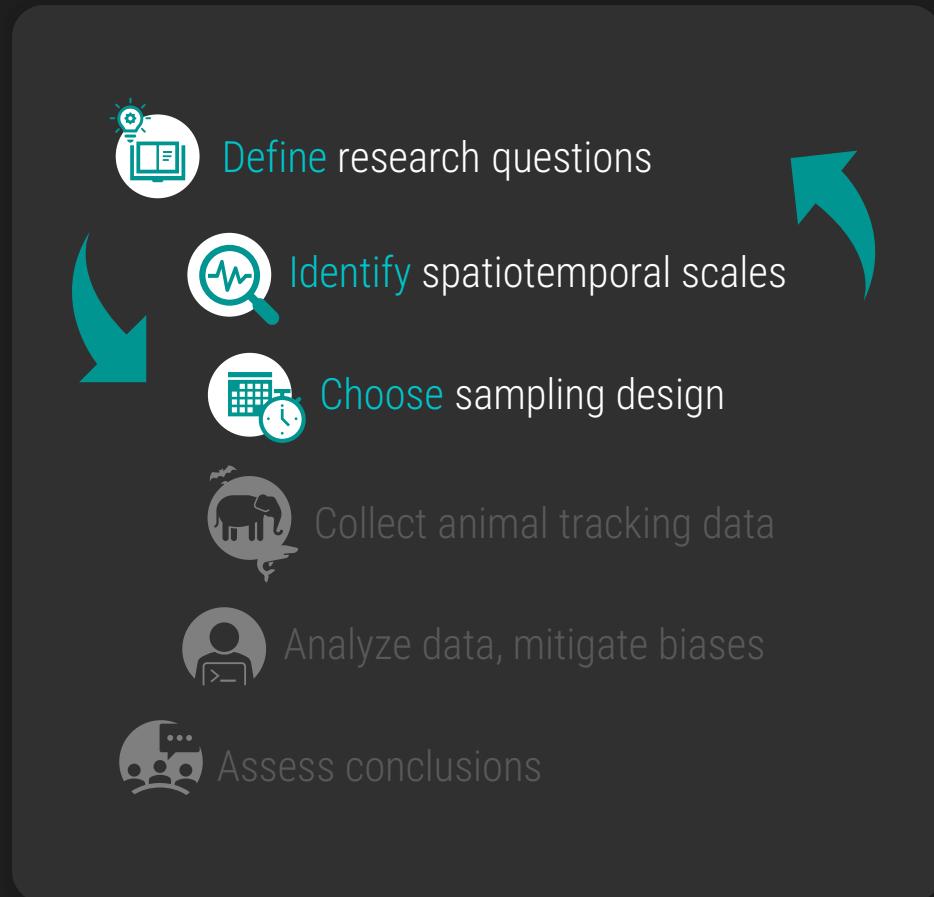




1.  
Animal movement paths are realizations of **continuous stochastic processes**,

2.  
Summarize behavior using **characteristic timescales**,

The diagram shows two teal circles representing timescales. The left circle contains the symbol  $\tau_p$  with the label 'Position autocorrelation timescale' below it. The right circle contains the symbol  $\tau_v$  with the label 'Velocity autocorrelation timescale' below it. Two teal arrows point from the text 'Summarize behavior using characteristic timescales,' towards the respective circles.



1. Animal movement paths are realizations of **continuous stochastic processes**,

2. Summarize behavior using **characteristic timescales**,

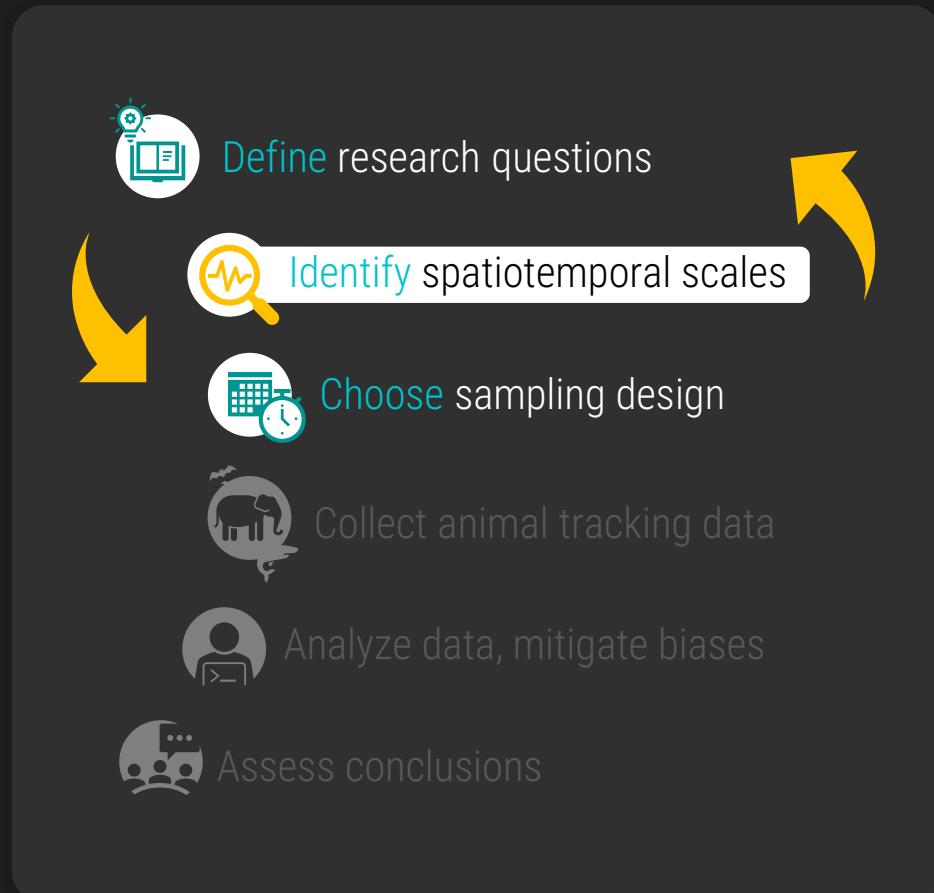
$\tau_p$

*Position autocorrelation timescale*

$\tau_v$

*Velocity autocorrelation timescale*

3. These **timescales** impose **constraints** on **sampling design** that *must* be met for sufficiently **large (effective) sample sizes**.



1. Animal movement paths are realizations of **continuous stochastic processes**,

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‘movedesign’  
Silva et al. (2023)

*Objectives:*

Develop a systematic approach, akin to statistical power analysis, to determine optimal sampling parameters in animal tracking projects.

*Analytical targets:*

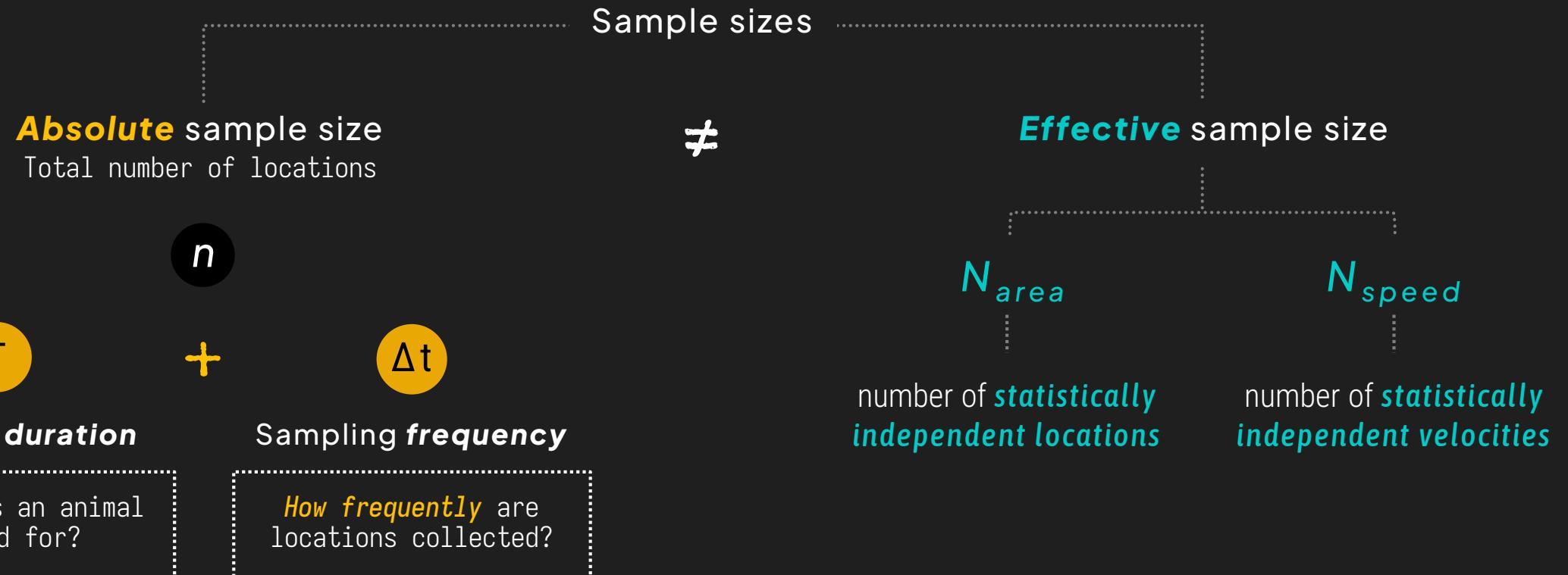
We considered three common estimates –home range area, speed and distance traveled.



Like any statistical tool, these methods still require sufficiently large **effective sample sizes** to achieve high accuracy.



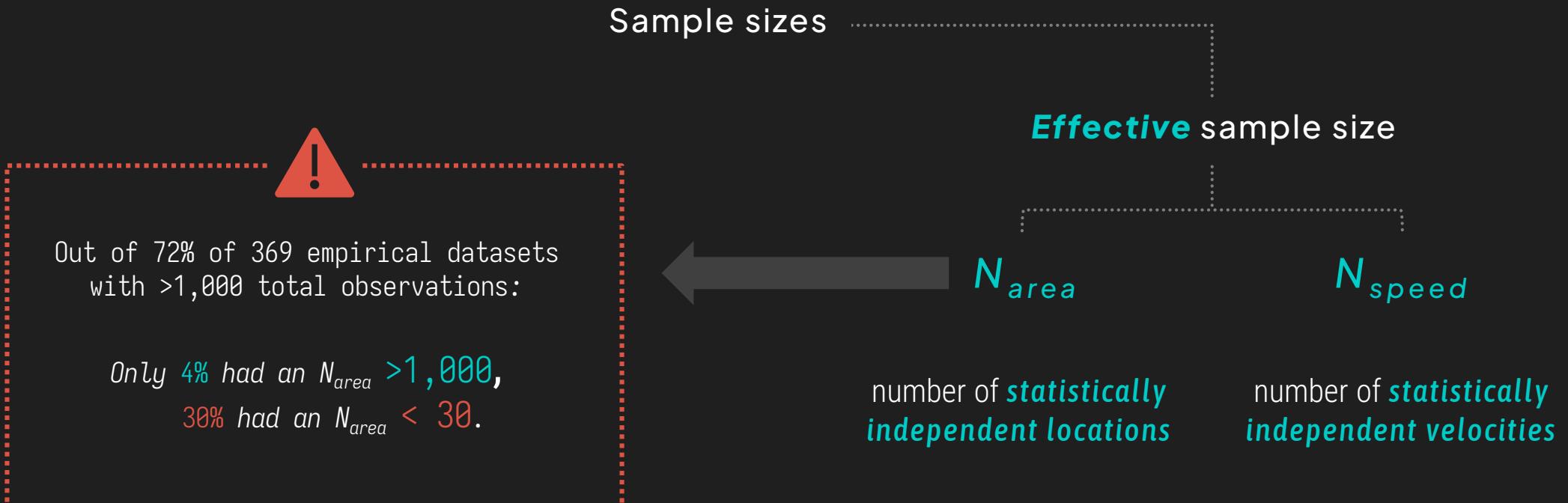
A successful animal tracking project requires a sampling schedule that leads to sufficiently **large (effective) sample sizes**.



For autocorrelated data,  $N < n$ , and often  $N \ll n$



A successful animal tracking project requires a sampling schedule that leads to sufficiently **large (effective) sample sizes**.



Noonan *et al.* (2019)

DOI: 10.1002/ecm.1344



## Understanding sample sizes

$\tau_p$

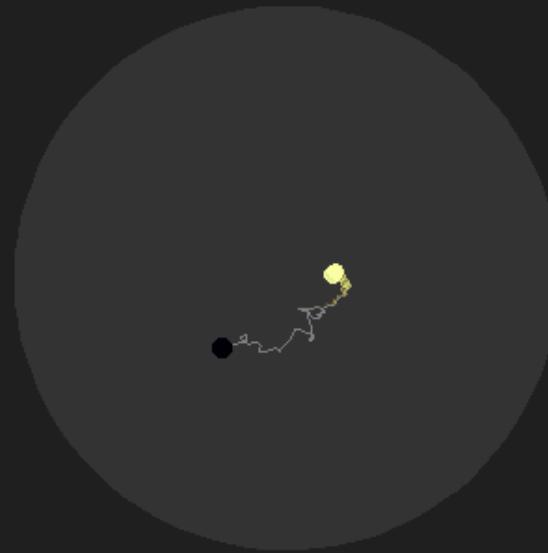
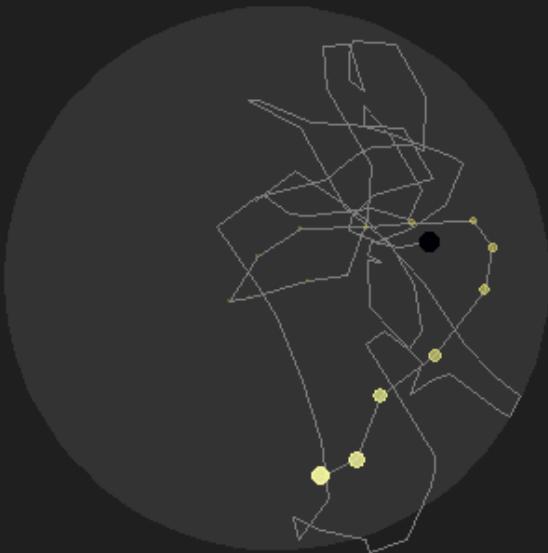
*Position autocorrelation  
timescale*

$\tau_p = 1$  hour

$\tau_p = 1$  day

$\tau_p = 5$  days

$\tau_p = 10$  days



T

SPACE-USE  
HOME RANGE

SAMPLING DURATION  
How long is an animal  
tracked for?

MOVEMENT BEHAVIOR  
SPEED & DISTANCE



## Understanding sample sizes

$\tau_v$

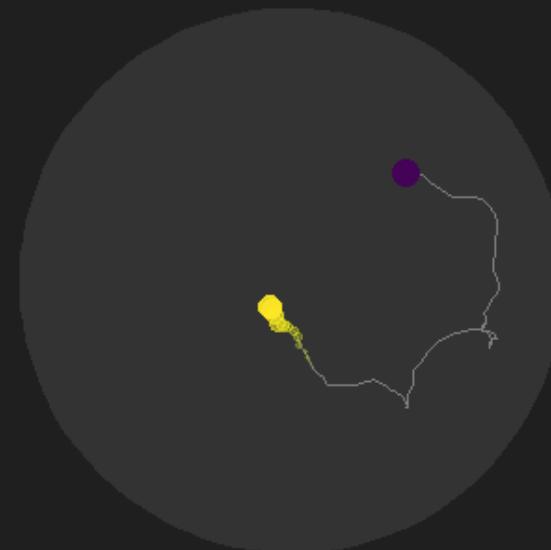
Velocity autocorrelation  
timescale

$\tau_v = 1$  minute

$\tau_v = 1$  hour

$\tau_v = 12$  hours

$\tau_v = 1$  day



$\Delta t$

SPACE-USE

HOME RANGE

SAMPLING FREQUENCY

How frequently are  
locations collected?

MOVEMENT BEHAVIOR

SPEED & DISTANCE



## Conceptualization



Define research questions



Identify spatiotemporal scales



Choose sampling design



Collect animal tracking data



Analyze data, mitigate biases



Assess conclusions



It is not physically possible for animal movement to be **uncorrelated**.

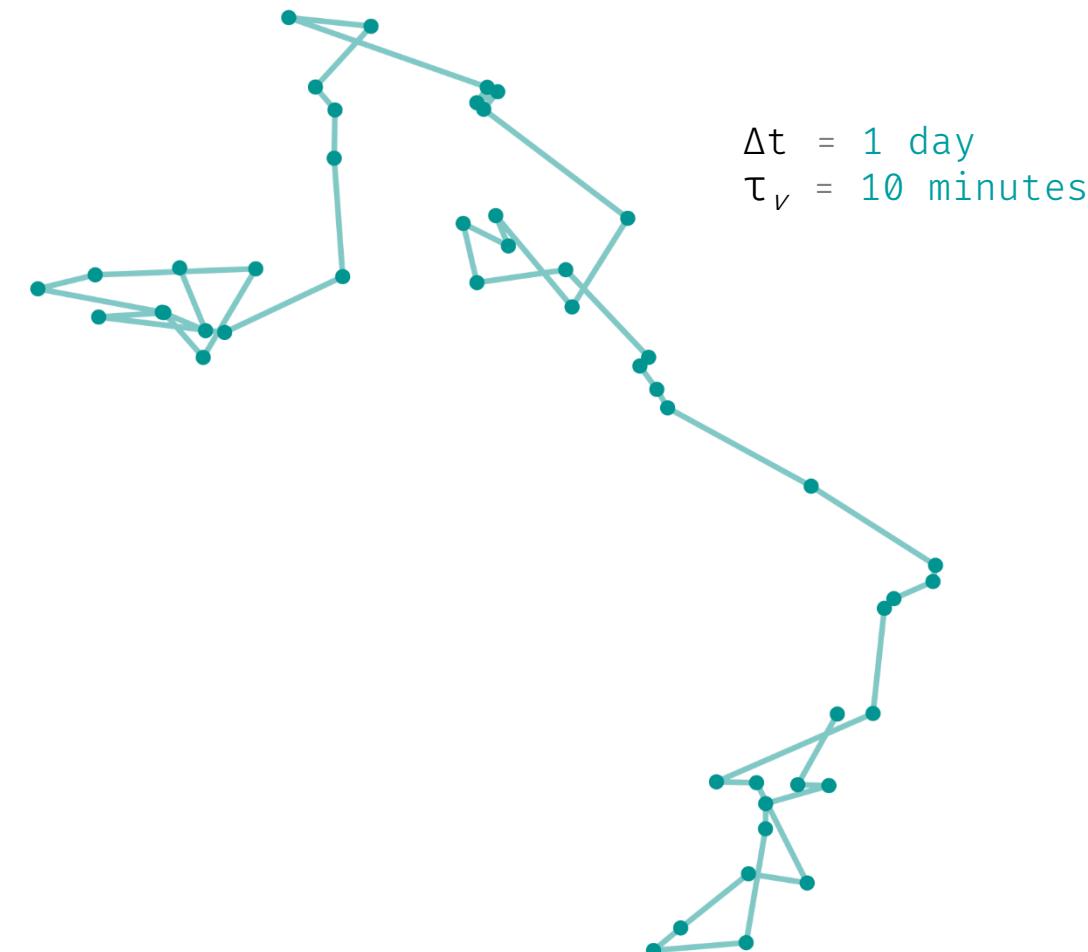
Now, the questions are:

- 1.** Can you detect a signature of these correlations in your data?
  
- 2.** And is this data sufficient to answer specific research questions?



## Conceptualization

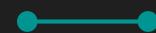
Simulated tracking dataset with a new location **once per day**,  $\tau_v > \Delta t$ .





## Conceptualization

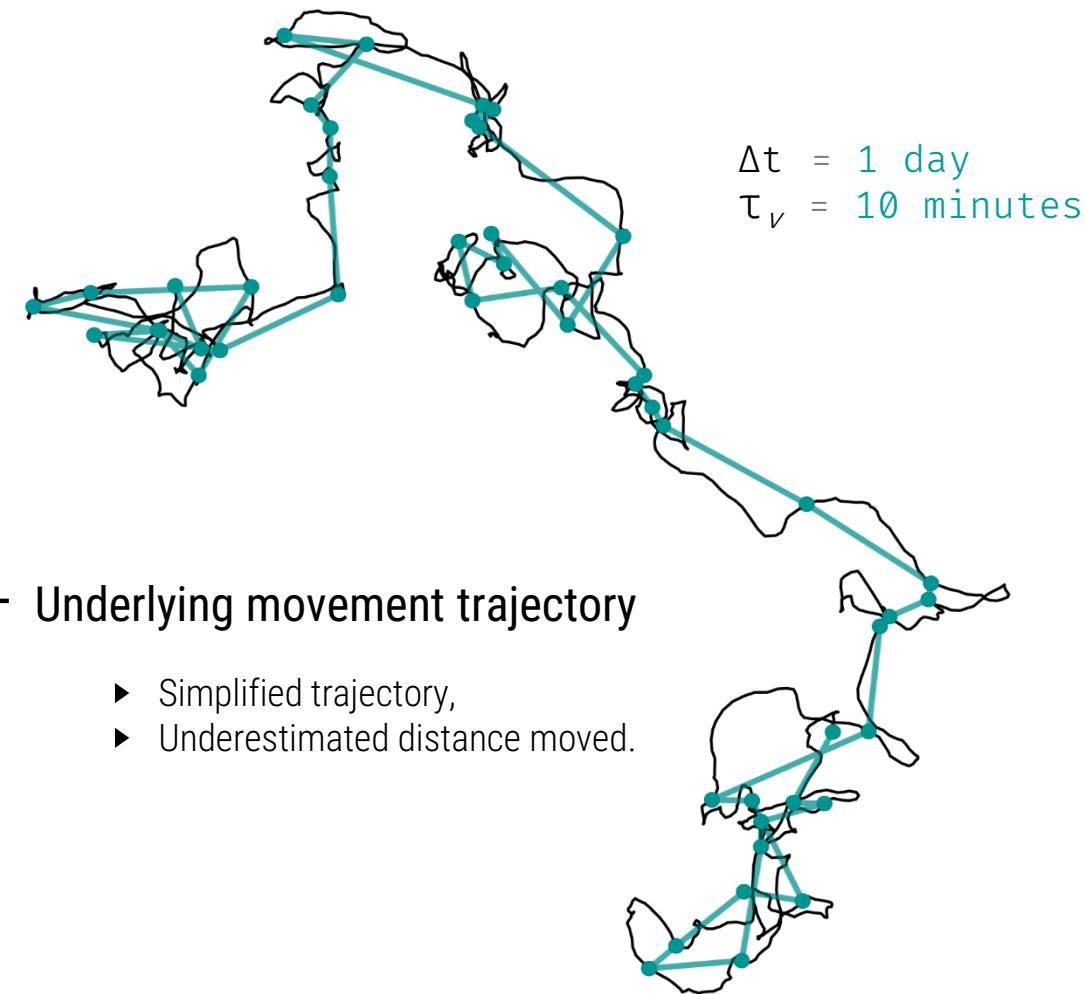
Simulated tracking dataset with a new location **once per day**,  $\tau_v > \Delta t$ .



We must carefully consider the frequency of data collection!



For the same  $\Delta t$ , this bias will be greater for individuals with more tortuous movement (shorter  $\tau_v$ ).





## Conceptualization

Simulated tracking dataset with  
a duration of **6 months**,  $\tau_p > T$ .



$T = 6 \text{ months}$   
 $\tau_p = 8 \text{ months}$

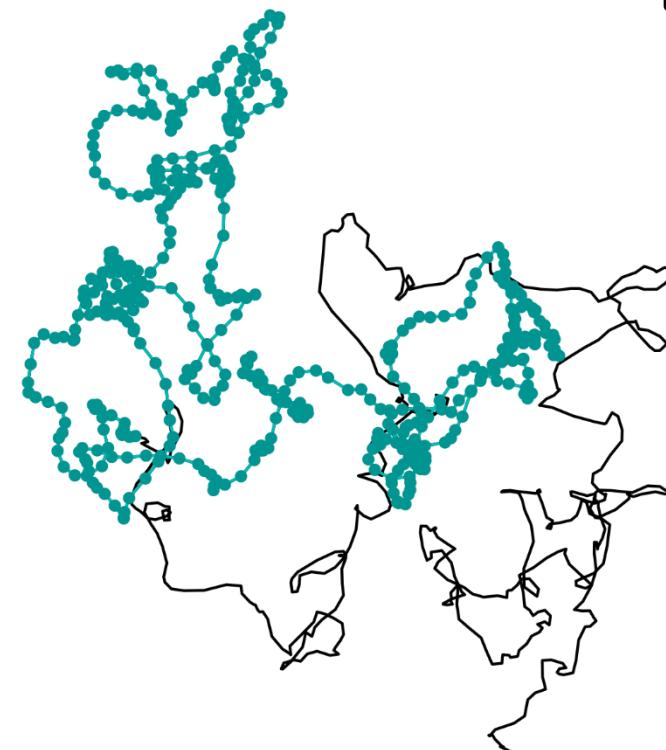


## Conceptualization

Simulated tracking dataset with a duration of **6 months**,  $\tau_p > T$ .



$T = 12 \text{ months}$   
 $\tau_p = 8 \text{ months}$



— Movement trajectory for the following **6 months**.



## Conceptualization

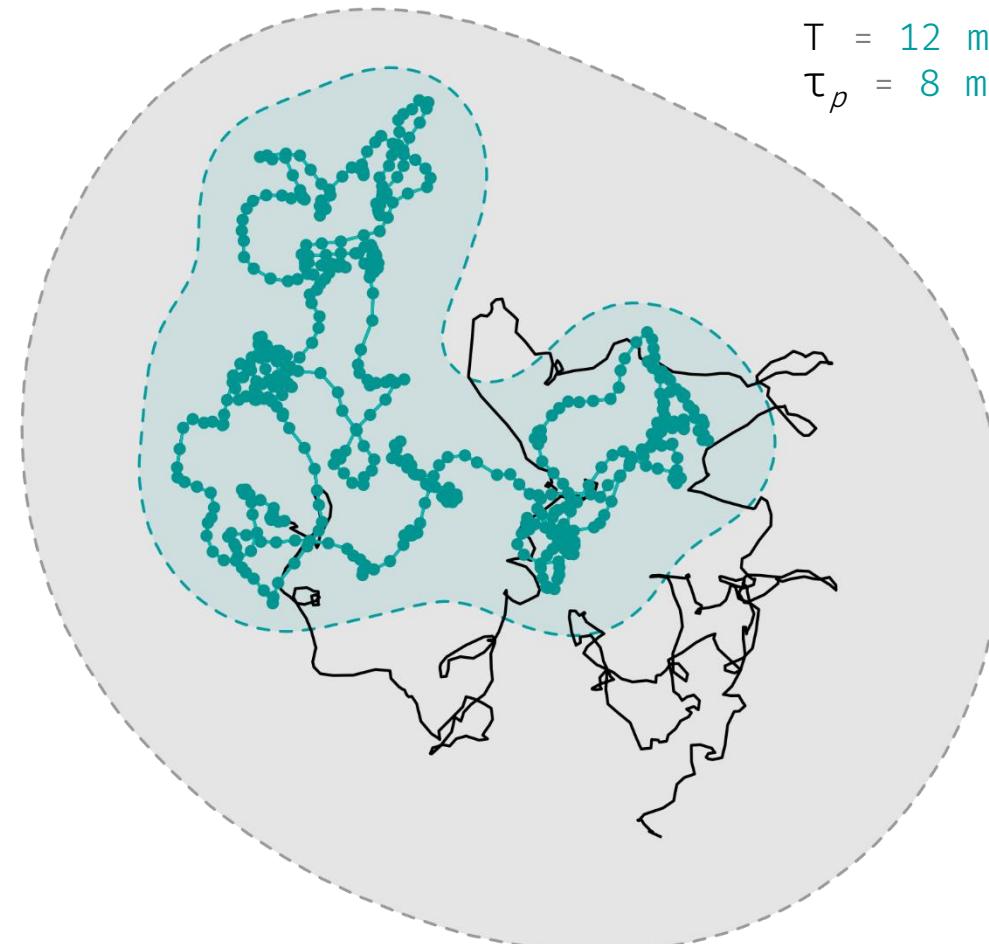
Simulated tracking dataset with a duration of **6 months**,  $\tau_p > T$ .



*We must carefully consider the duration of data collection!*



For the same  $T$ , the extent of this bias will be greater for individuals with longer crossing times ( $\tau_p$ ).



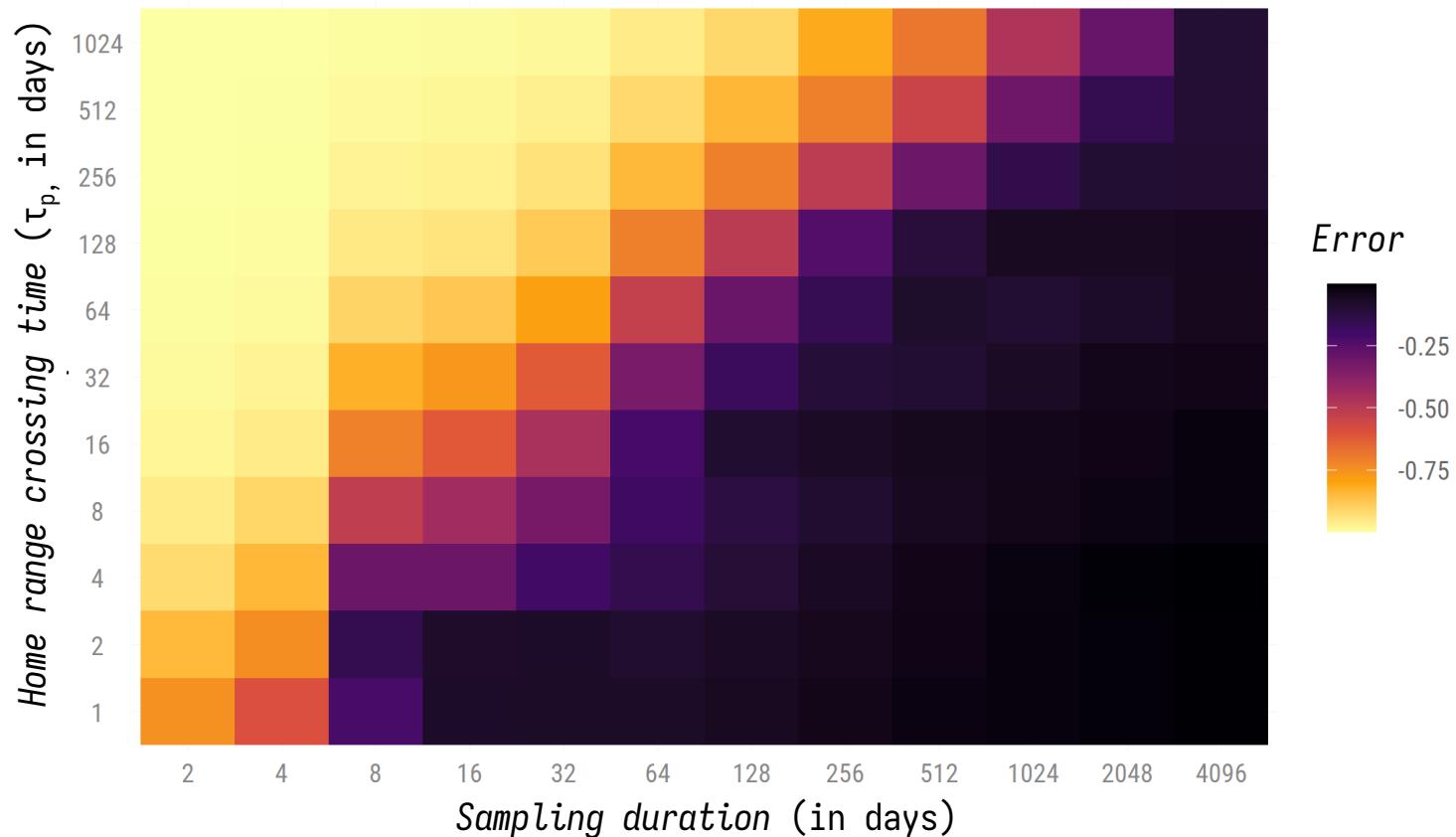
— Movement trajectory for the following **6 months**.

- ▶ Sampling missed used areas.
- ▶ Underestimated home range.



## 1. Home ranges – **Autocorrelated Kernel Density Estimator (AKDE)**:

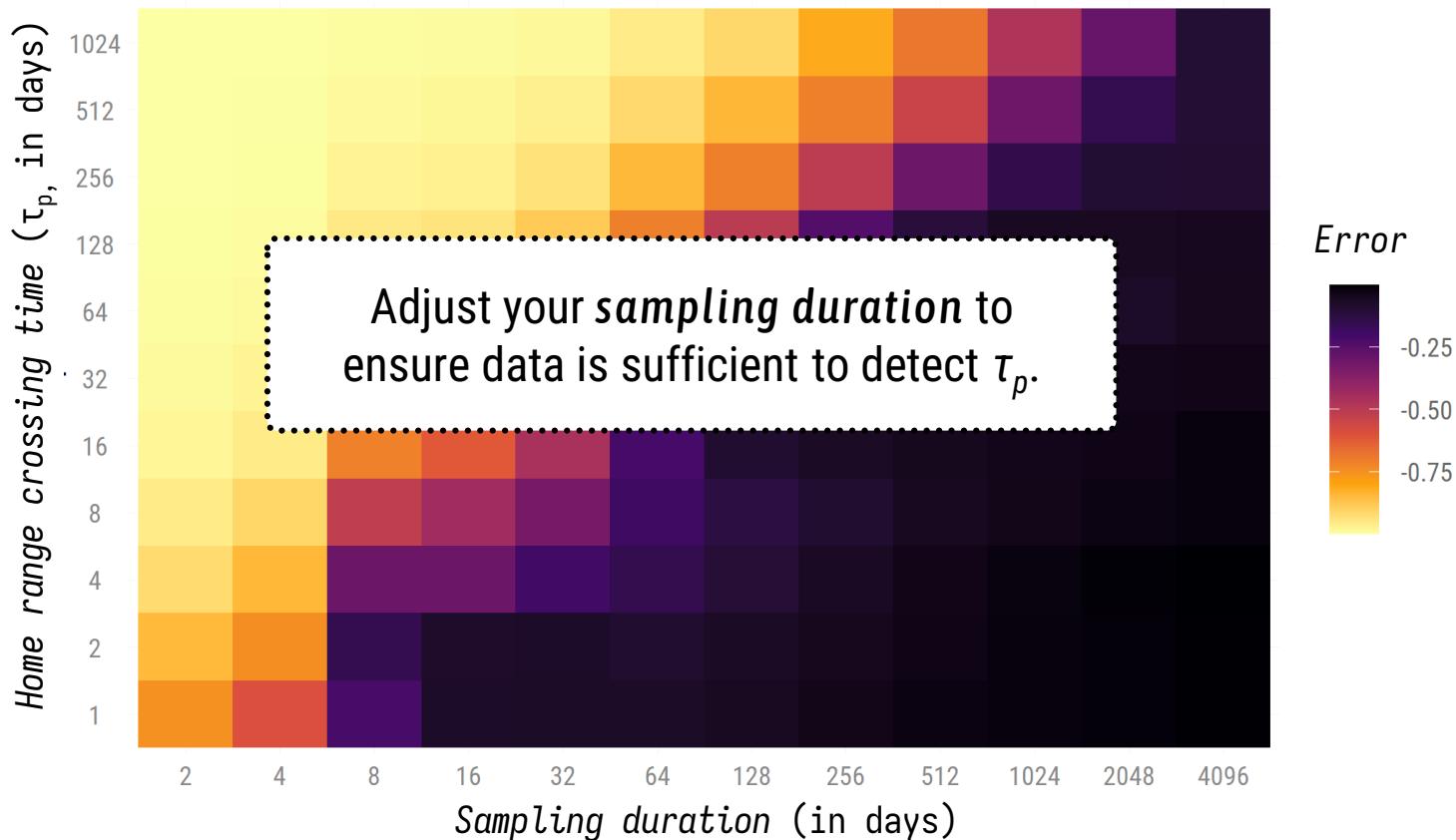
 Fleming et al. (2015)





## 1. Home ranges – **Autocorrelated Kernel Density Estimator (AKDE)**:

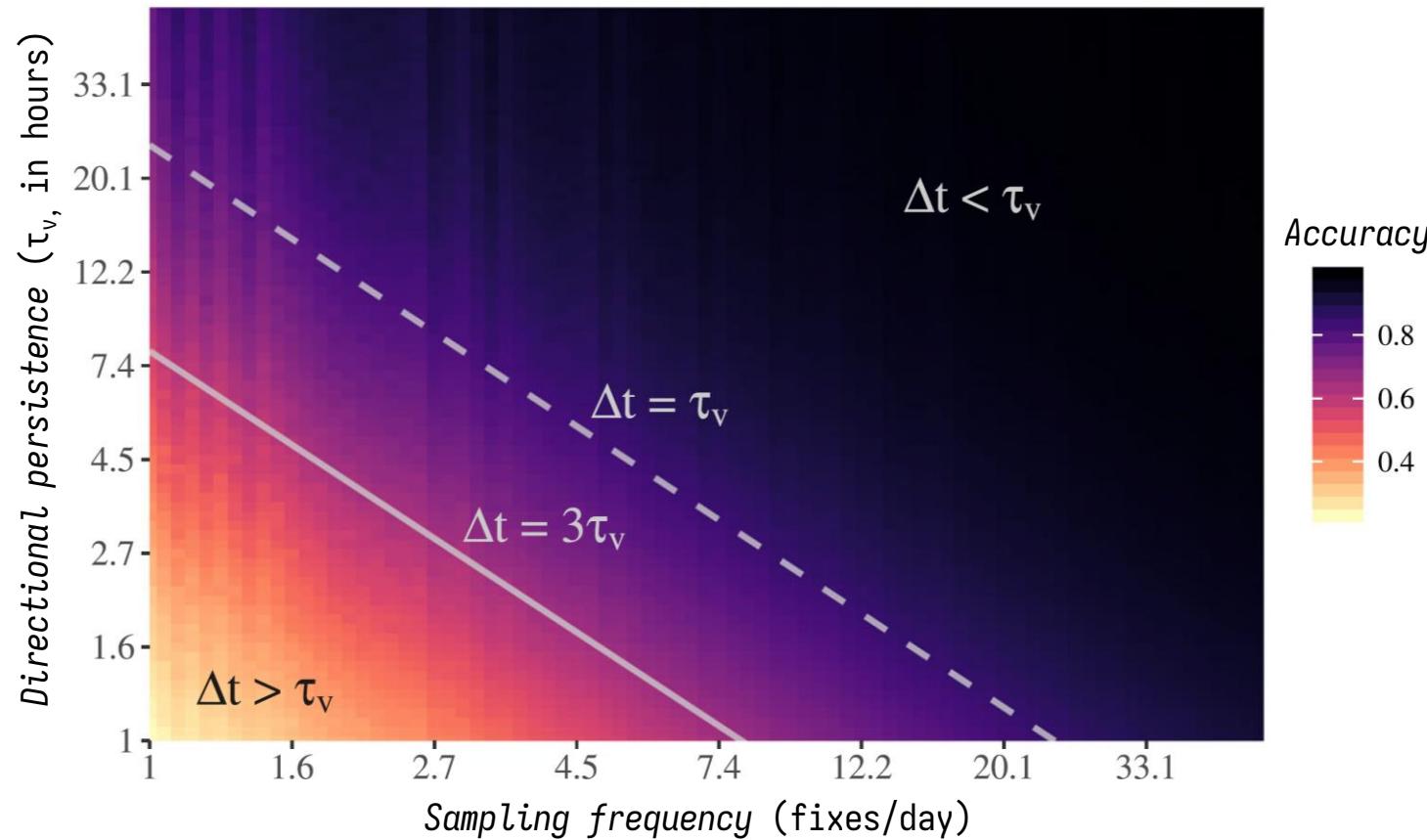
 Fleming et al. (2015)





## 2. Speed & distance – **Continuous-time speed and distance (CTSD)**:

ℓ Noonan et al. (2019)



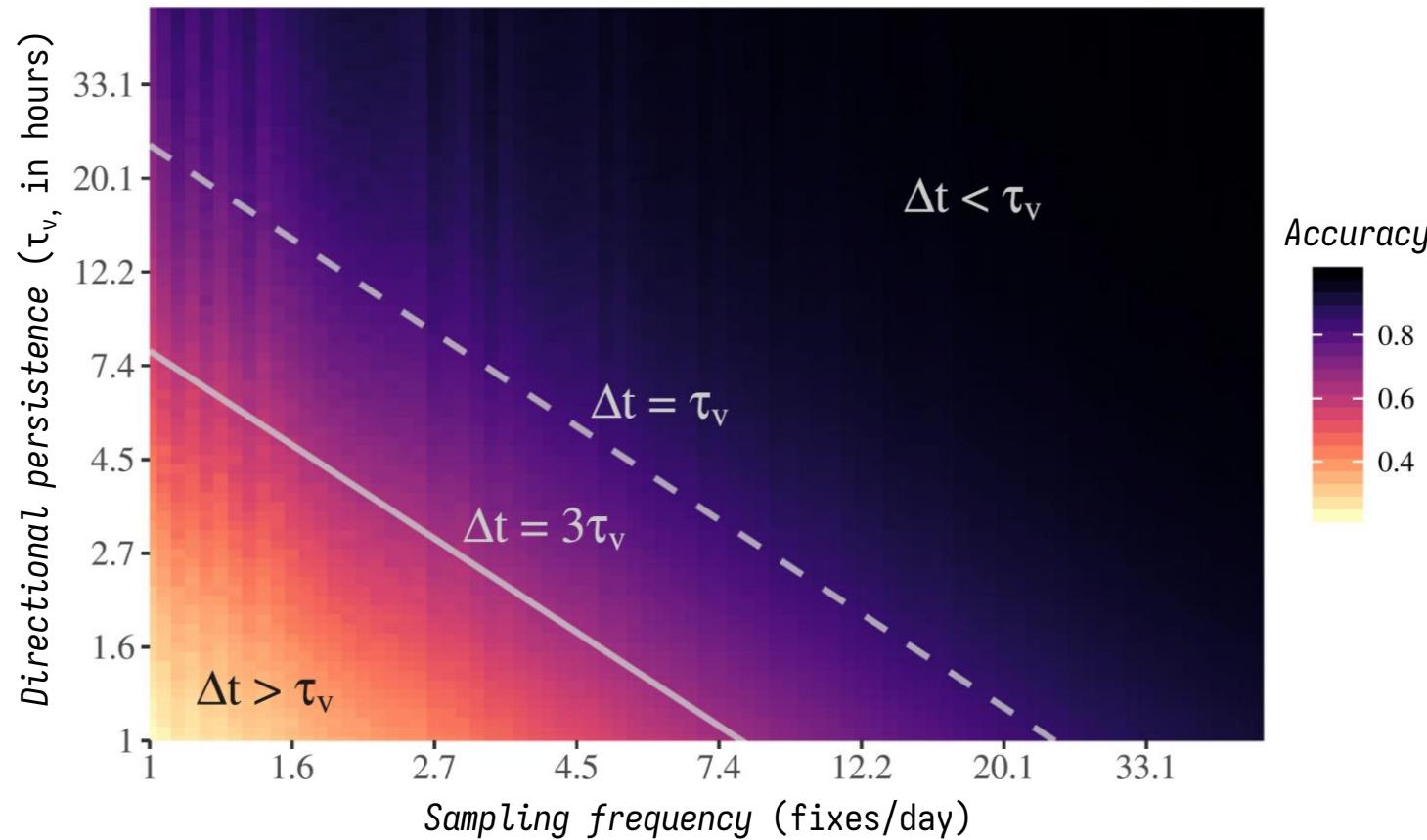
If  $\Delta t > 3\tau_v$ , no statistically significant signature of the animal's velocity will remain in the location data.

If  $3\tau_v > \Delta t > \tau_v$ , there will be some positive bias ( $\tau_v$  can not be accurately estimated).



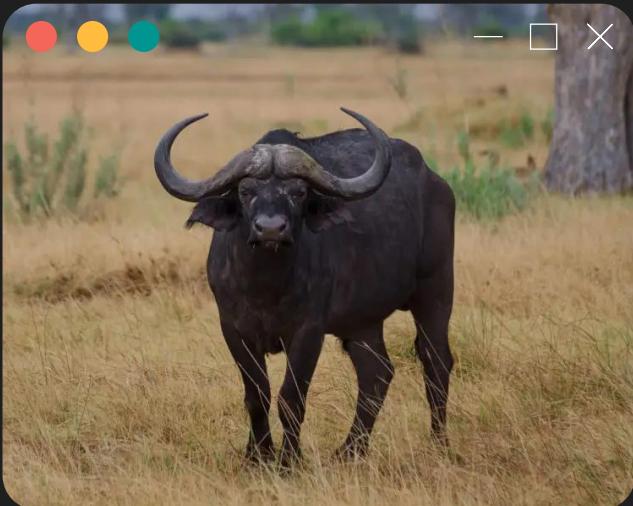
## 2. Speed & distance – **Continuous-time speed and distance (CTSD)**:

ℓ Noonan et al. (2019)



Adjust your **sampling interval** to ensure data is of **sufficient resolution** to detect  $\tau_v$ .

$$\Delta t \leq 3\tau_v$$



Position autocorrelation  
**10.3 days**  
8.2 – 12.8  
( $\tau_p$ )

These parameters are fairly  
**conservative** at the species-  
and population-level.

## African buffalo (*Syncerus caffer*)

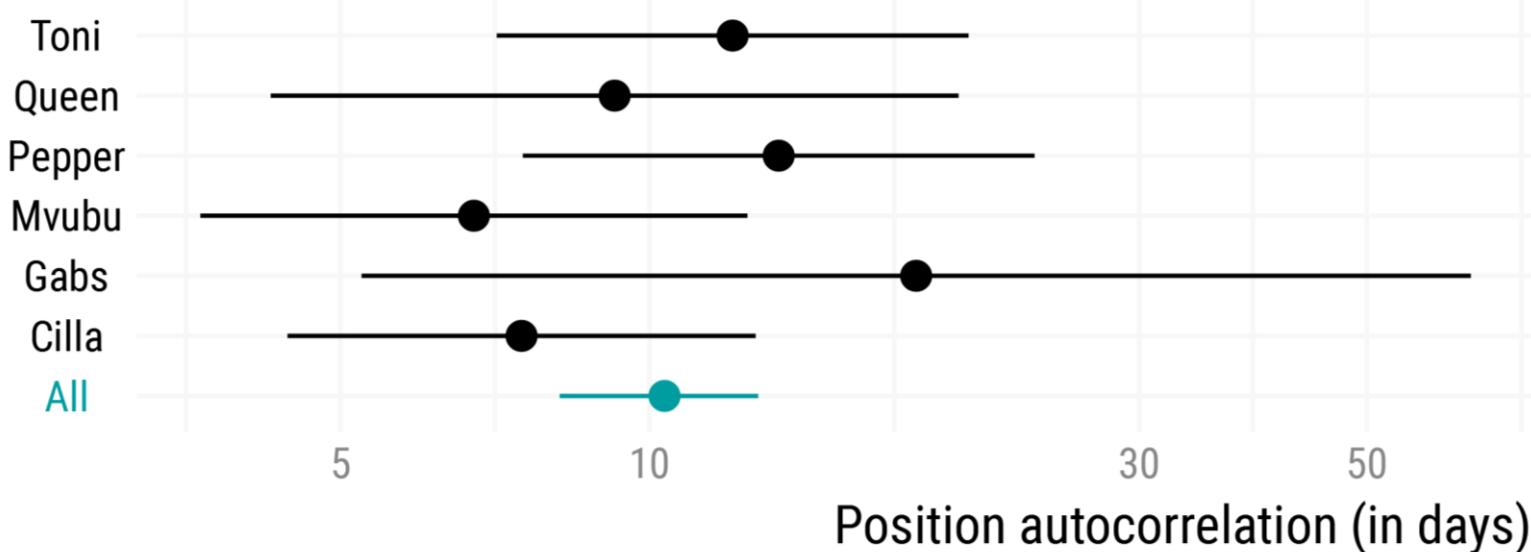
Timescales

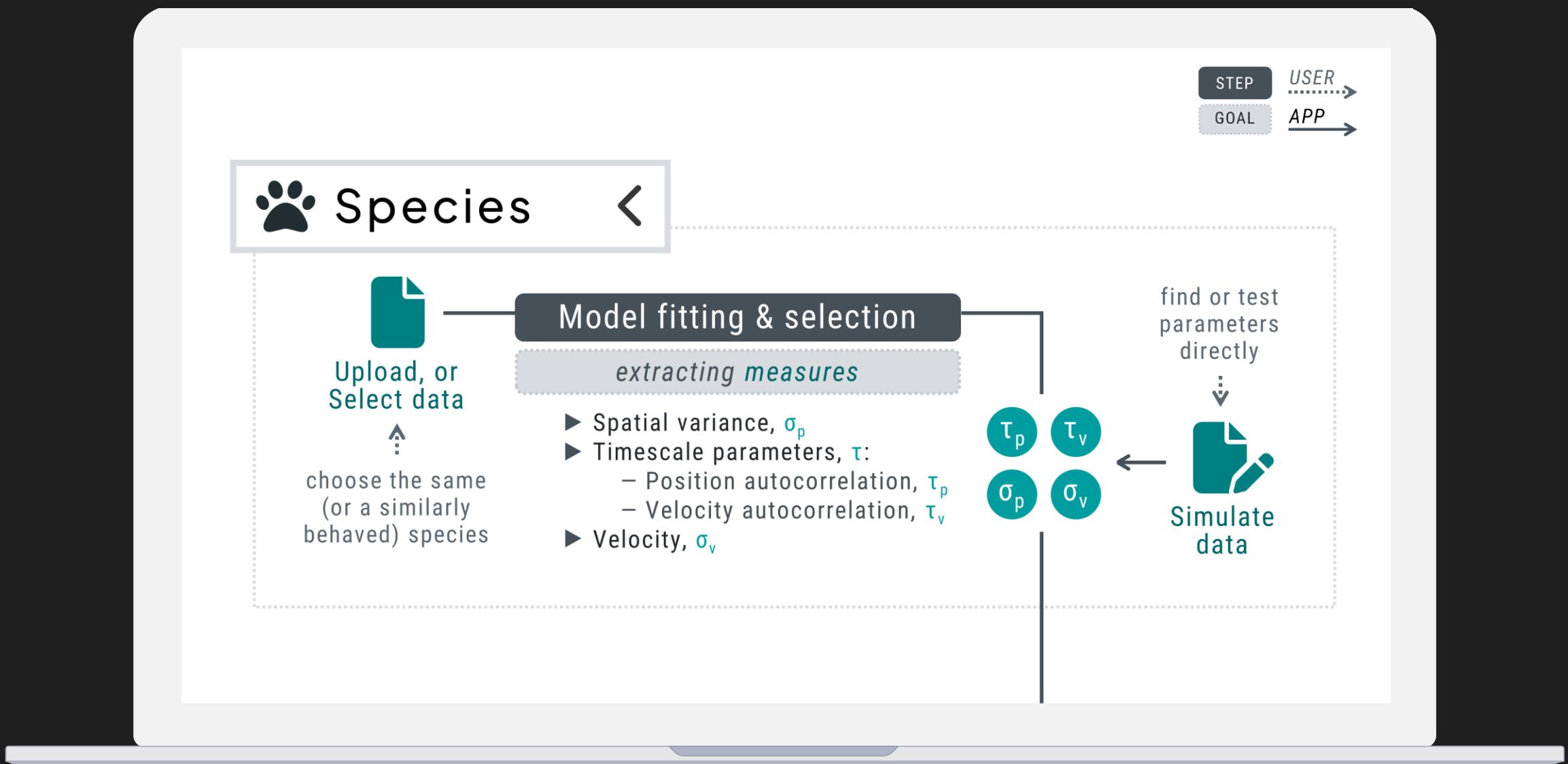
Dataset

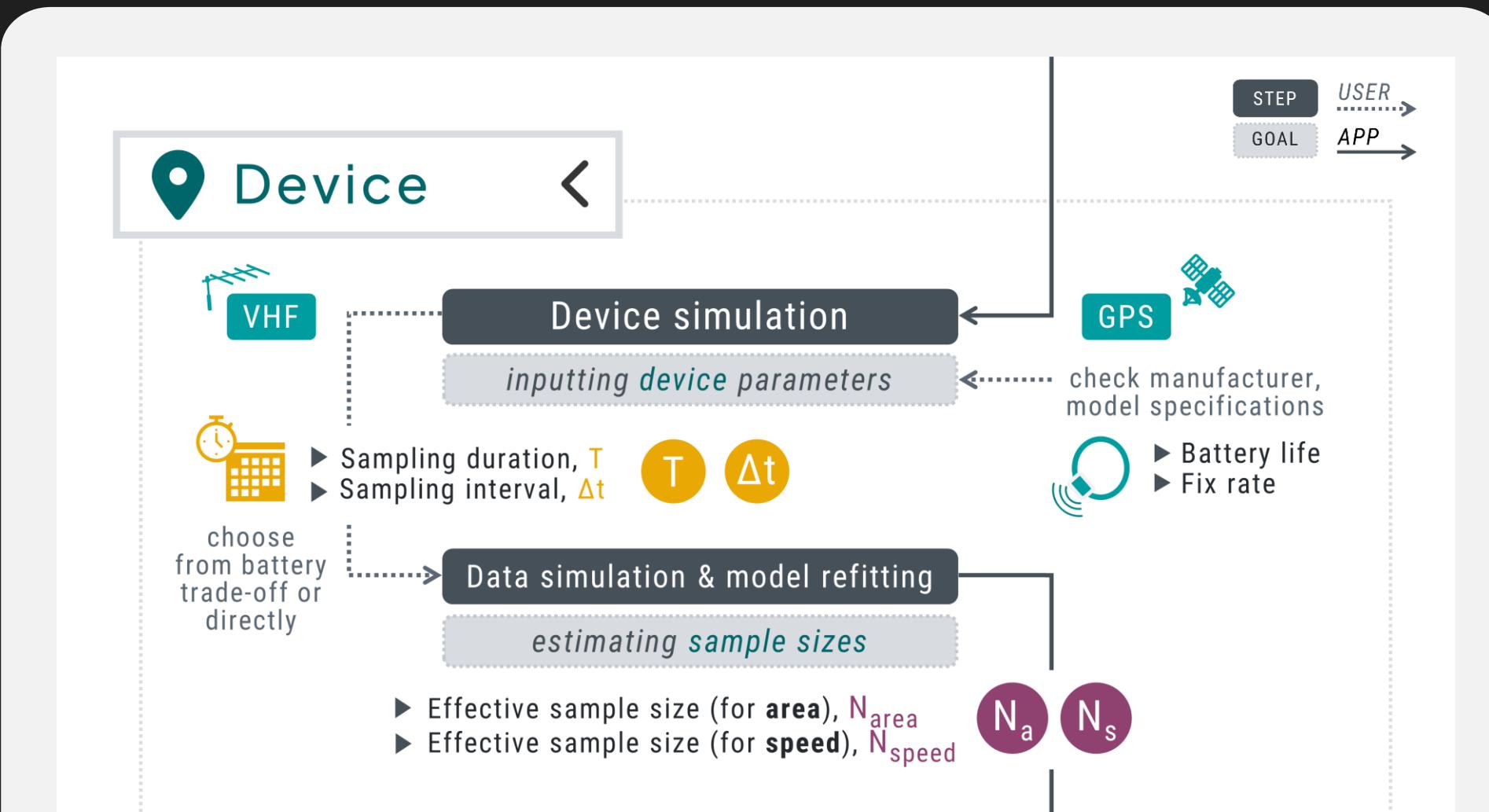
Show parameter:

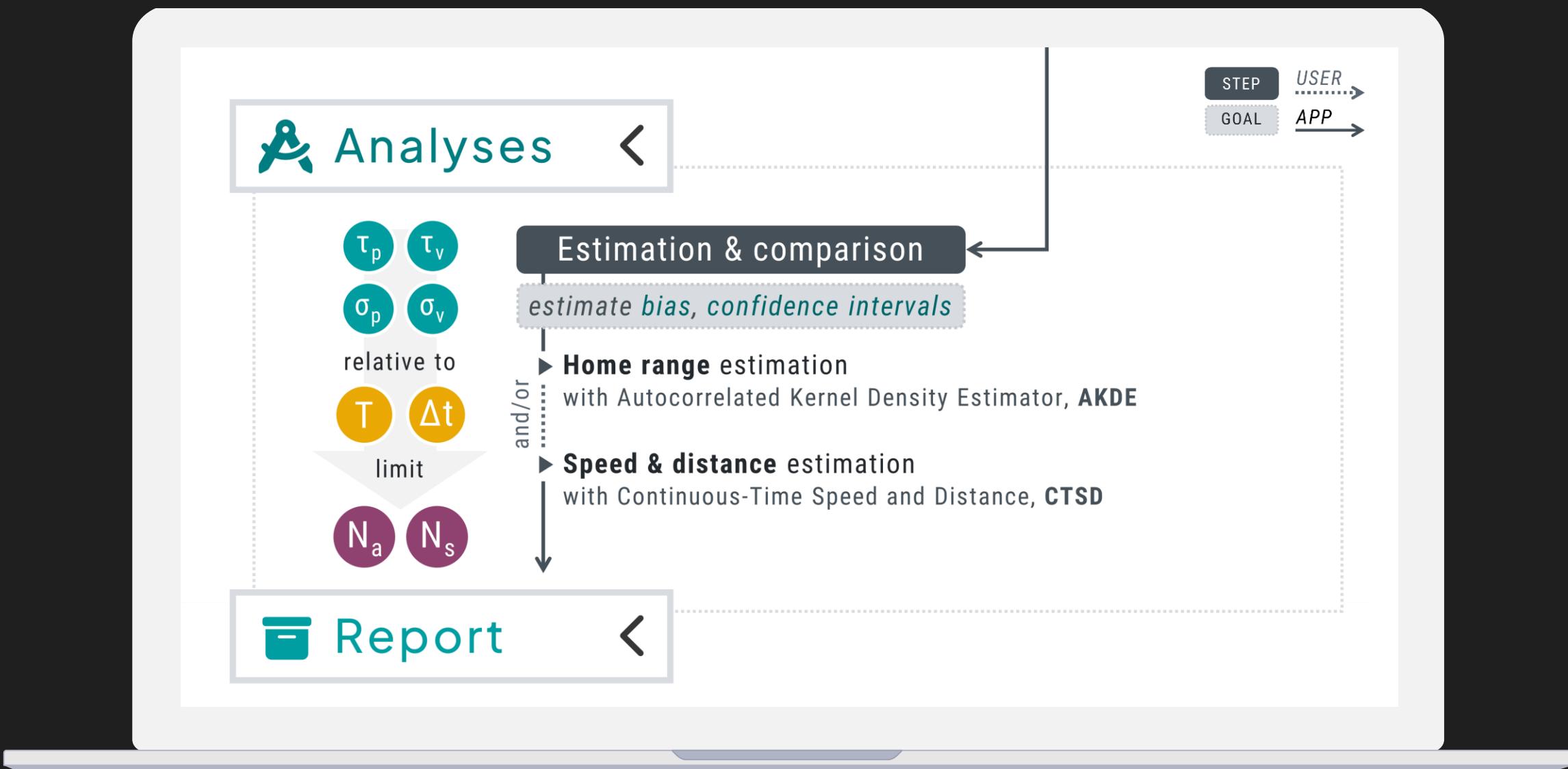
Position autocorrelation ( $\tau_p$ )

Velocity autocorrelation ( $\tau_v$ )





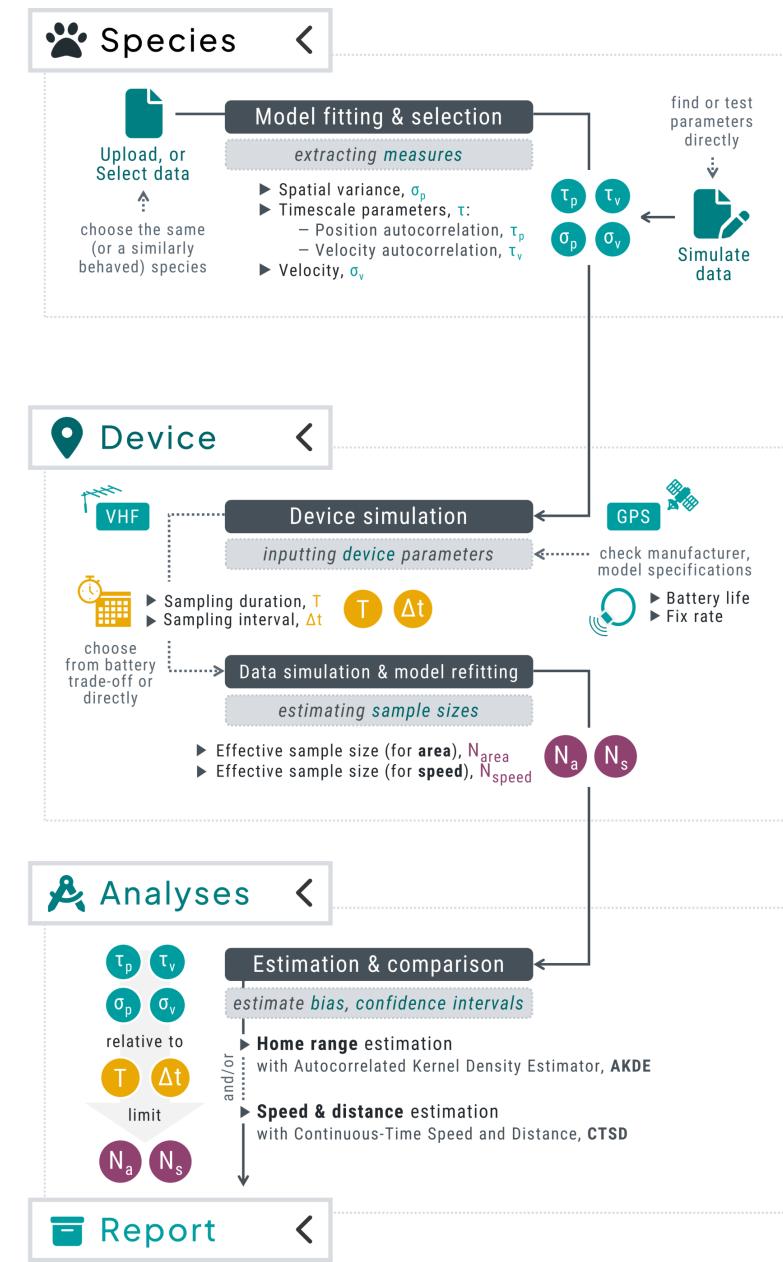






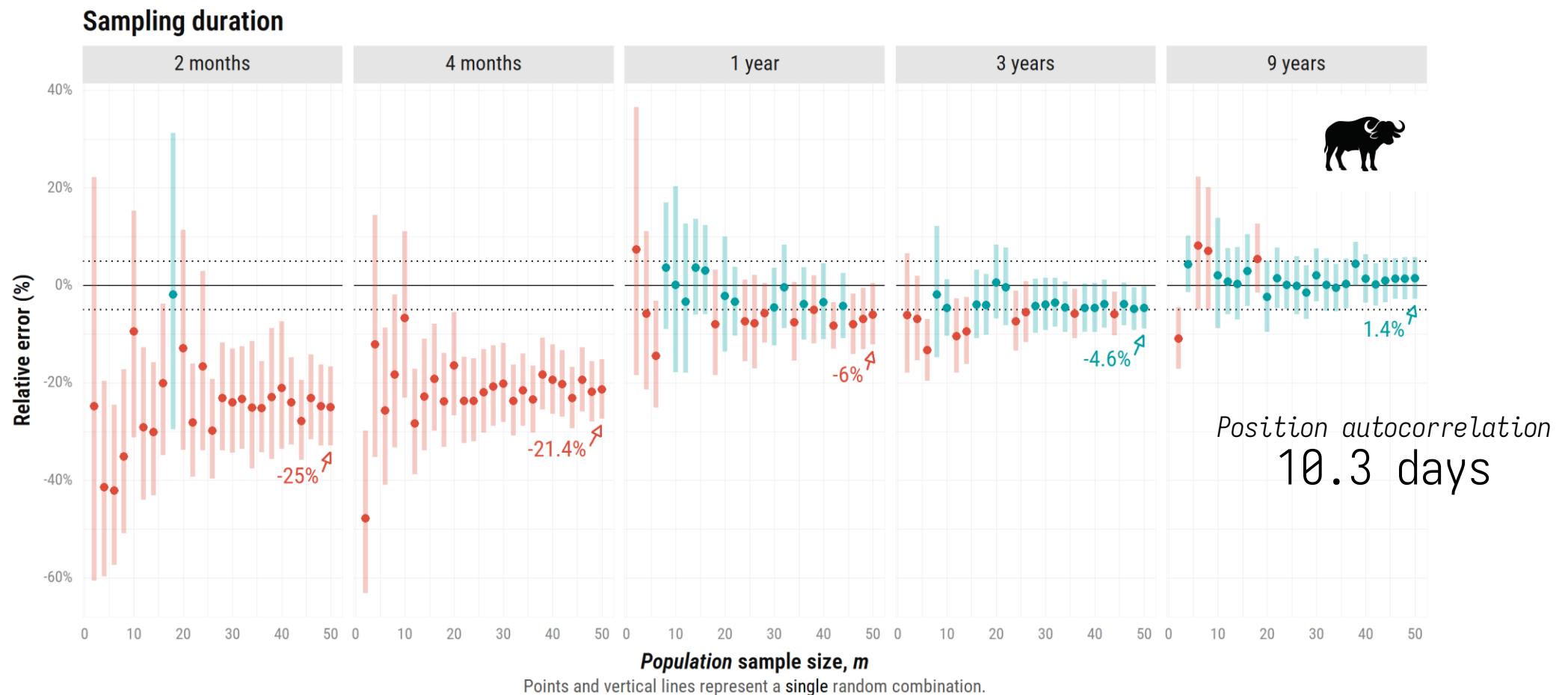
## New version (v0.3.0):

- Run multiple simulations for:
  - a predefined (*population*) sample size,
  - an iteratively higher sample size,until error is below a specified threshold.
- Get estimates for:
  - mean of sampled population,
  - compare means of two sampled populations.  
(e.g., males/females)



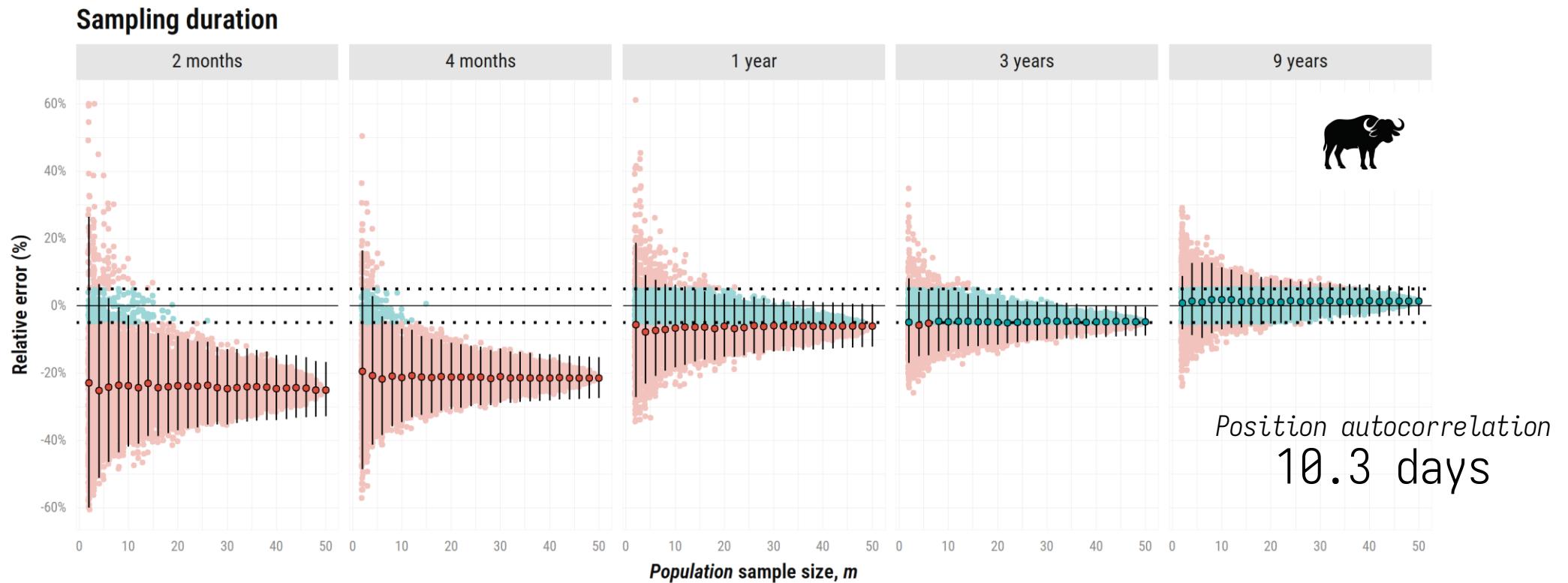


### 3.1. Population-level inferences – mean home range areas



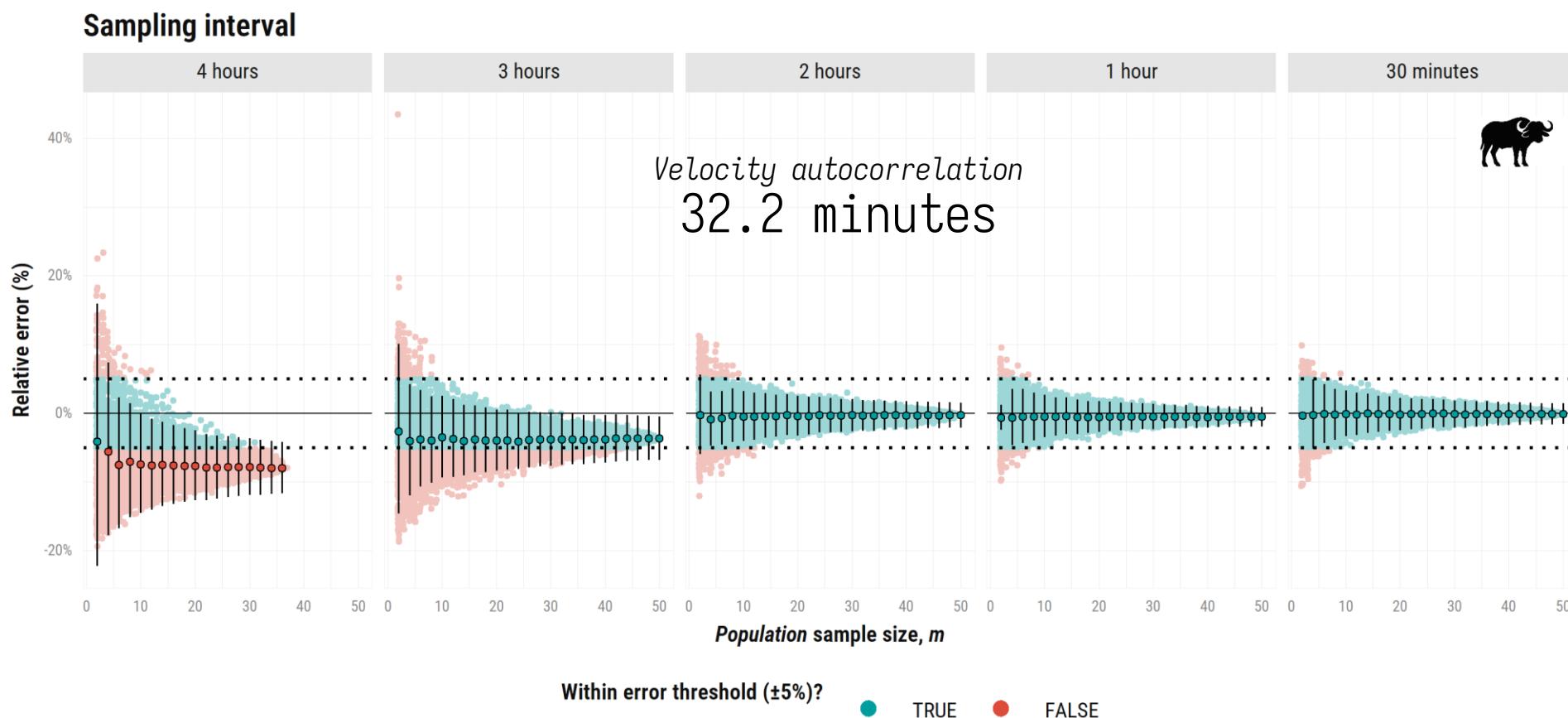


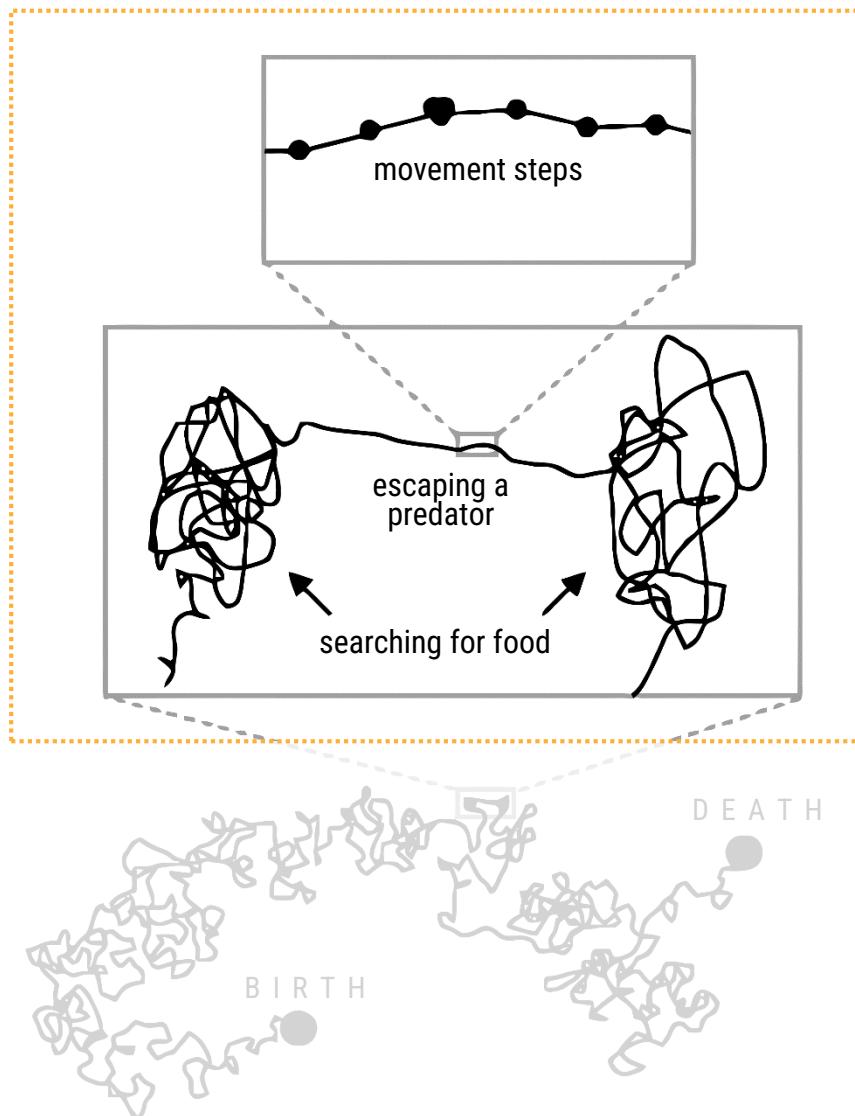
### 3.1. Population-level inferences – mean home range areas



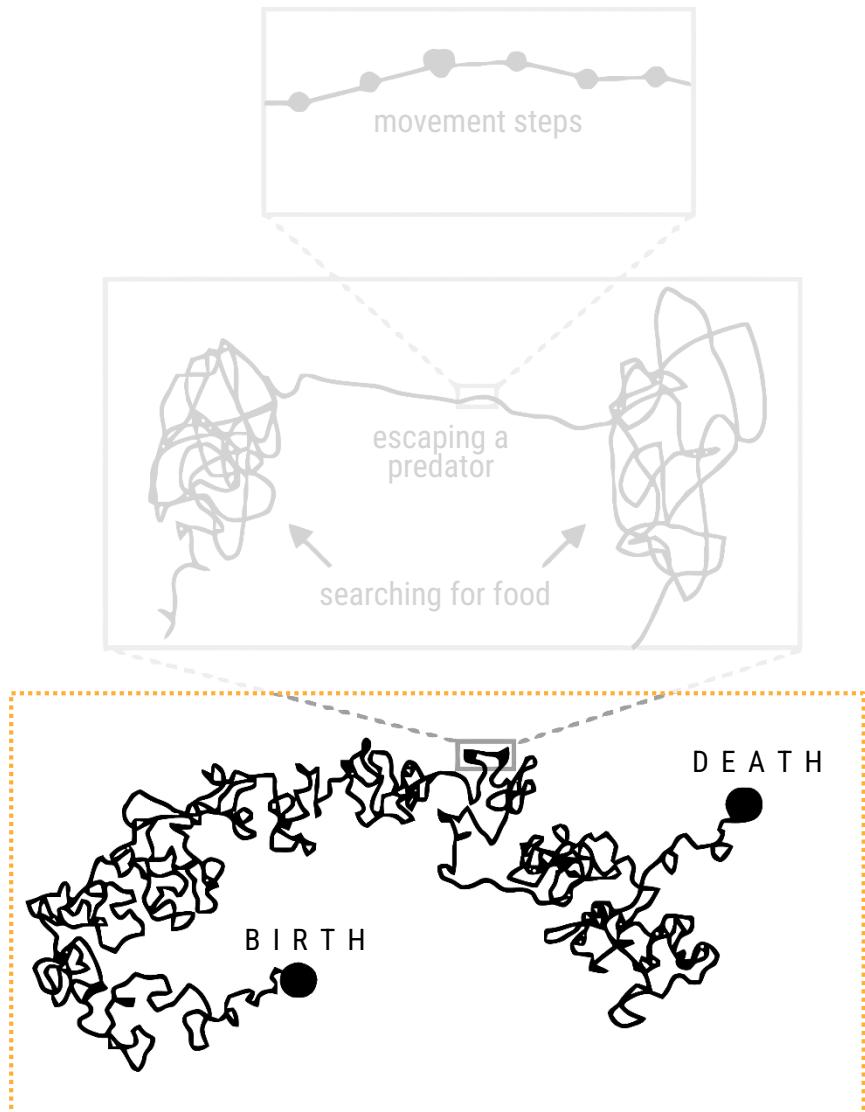


## 3.2. Population-level inferences – mean movement speed



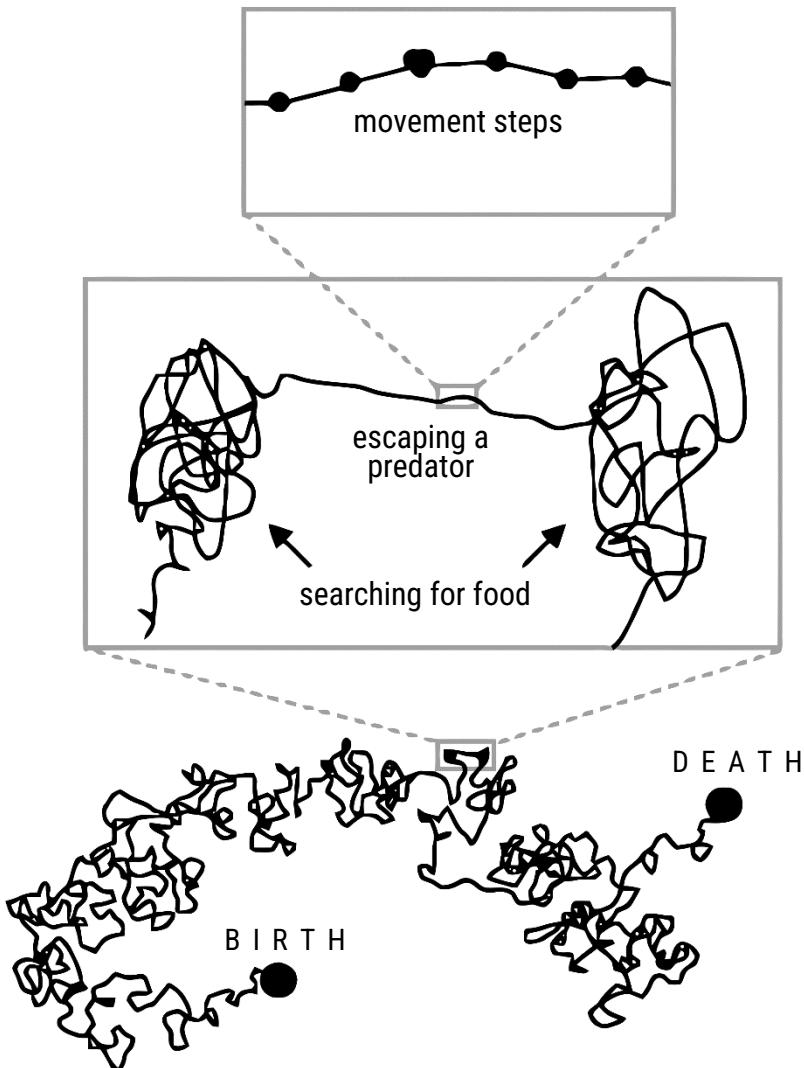


If the goal is **speed & distance estimation**,  
adjust your **sampling interval** ( $\Delta t$ ) to ensure data  
is of sufficient resolution to detect  $\tau_v$ .



If the goal is **speed & distance estimation**,  
adjust your **sampling interval** ( $\Delta t$ ) to ensure data  
is of sufficient resolution to detect  $\tau_v$ .

If the goal is **home range area estimation**,  
adjust your **sampling duration** ( $T$ ) to ensure data  
is sufficient to detect  $\tau_p$ .



If the goal is speed & distance estimation,  
adjust your **sampling interval** ( $\Delta t$ ) to ensure data  
is of sufficient resolution to detect  $\tau_v$ .

If the goal is home range area estimation,  
adjust your **sampling duration** ( $T$ ) to ensure data  
is sufficient to detect  $\tau_p$ .

If both,  
You may be able to address large-scale and fine-  
scale questions, but not always **concurrently**.



*Get informed:*

Collect **pilot data**, and/or look at published studies & datasets.

*Make smart decisions:*

Consider **spatiotemporal scales** relative to **questions**.

*Keep it simple:*

Use **even sampling rates** if possible. Don't get too clever!

*Try before you buy:*

Check it first with **simulated data**.



*Divide and conquer:*

Use different individuals to answer different questions.



*If you must sample unevenly...*

*Keep it simple:*

Use no more than **two** different sampling rates.

*Try before you buy:*

**Simulate** with uneven sampling first. Still works?

*Mind your math:*

Use sampling rates that are **integer multiples**.

*Be careful:*

Check for **artifacts** introduced by uneven sampling.

*Be realistic:*

No, you can't have it all!