



MARE-Madeira 2025

Optimizing animal tracking projects

Using the 'movedesign' application



Inês Silva

✉ i.simoes-silva@hzdr.de



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.”



■ Fine-scale processes:

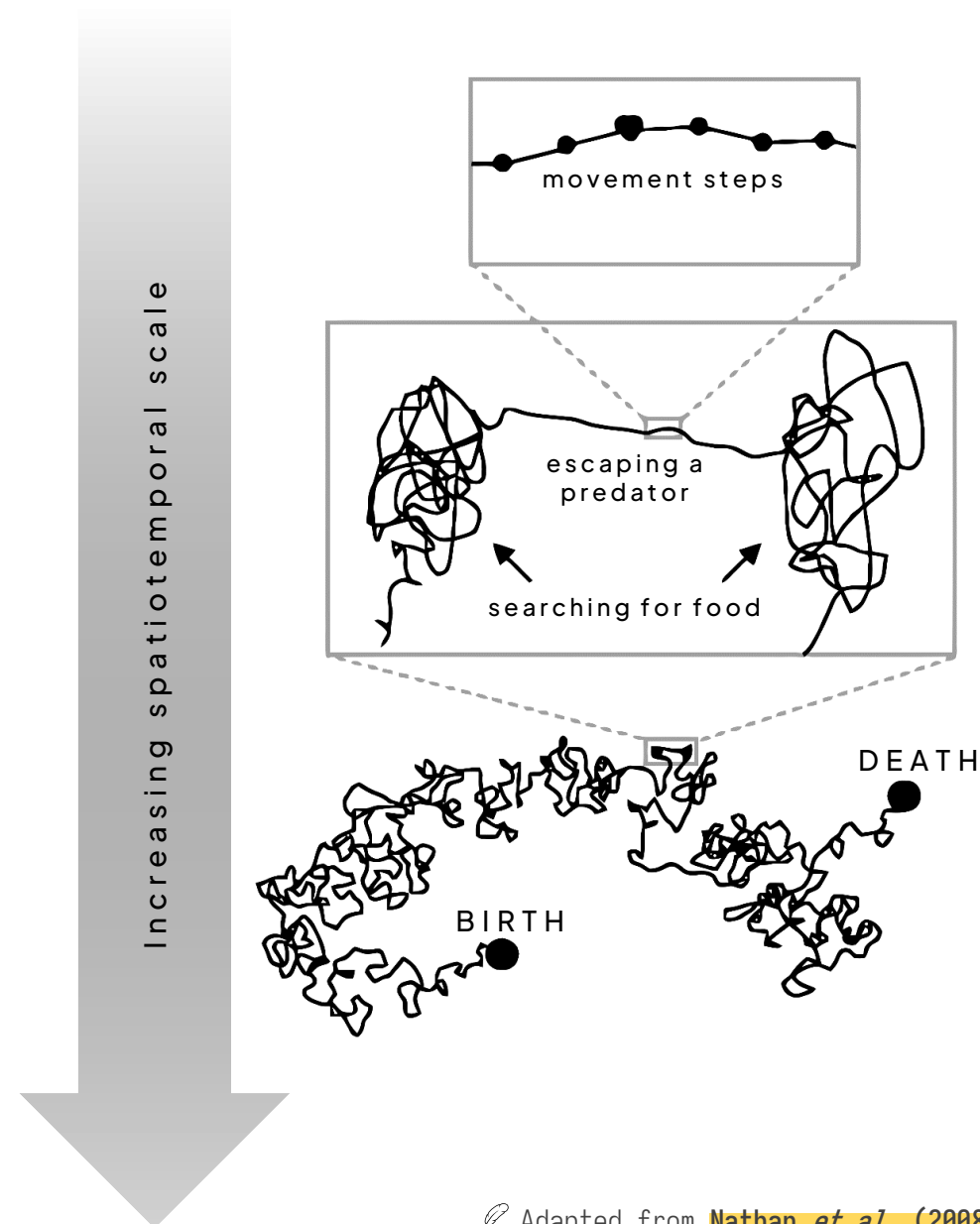
SPEED/DISTANCE — capture how far animals travel (and rate at which these distances are covered).

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

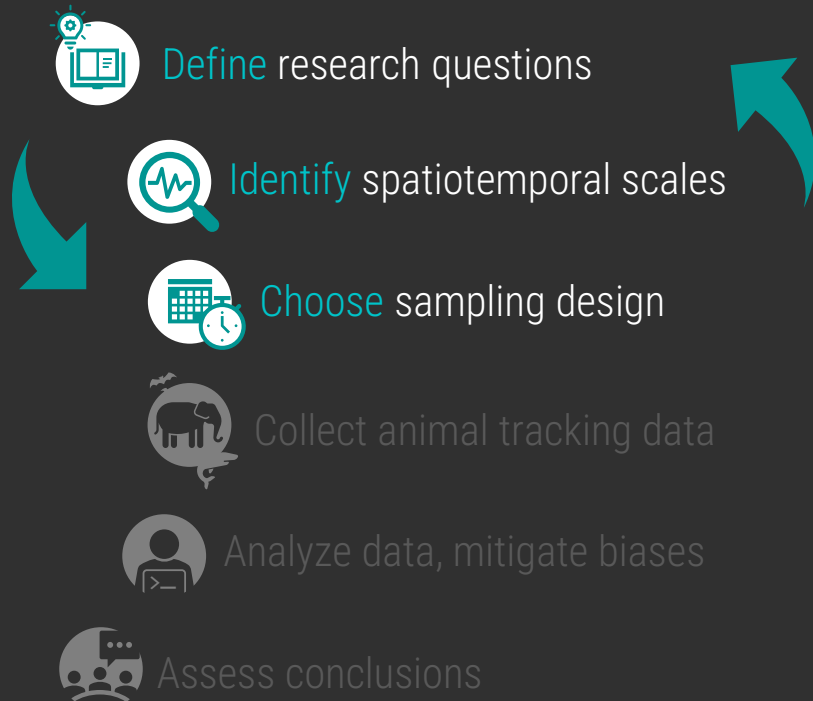
■ Large-scale processes:

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**.



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





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

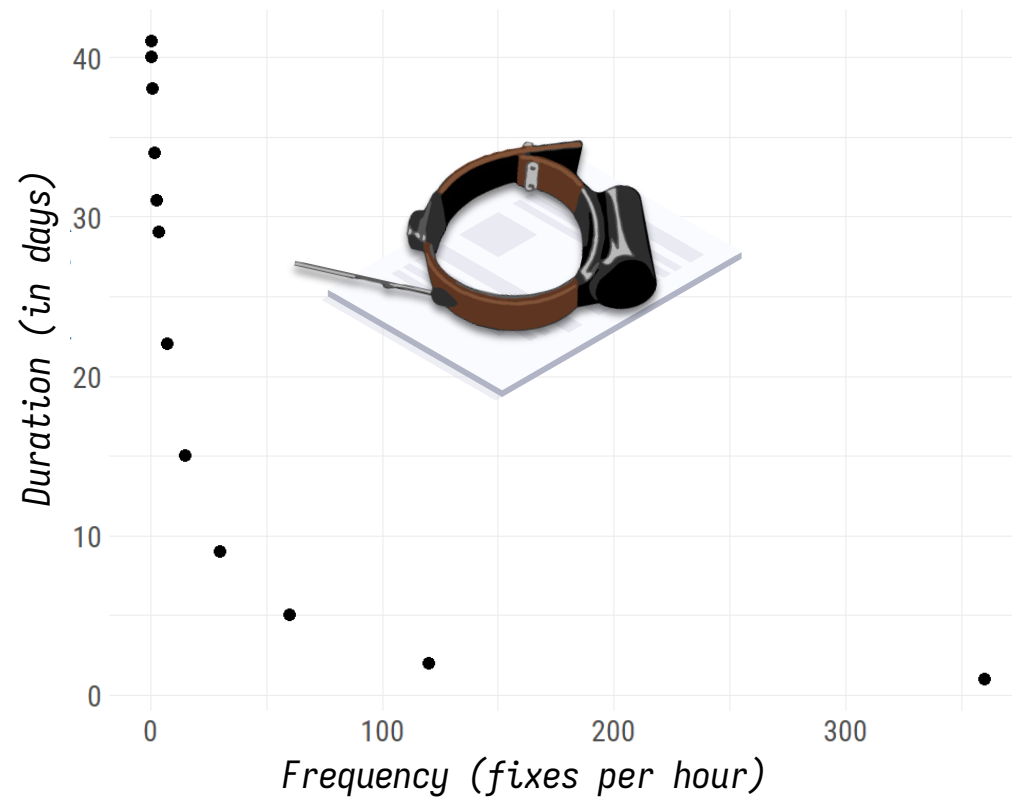


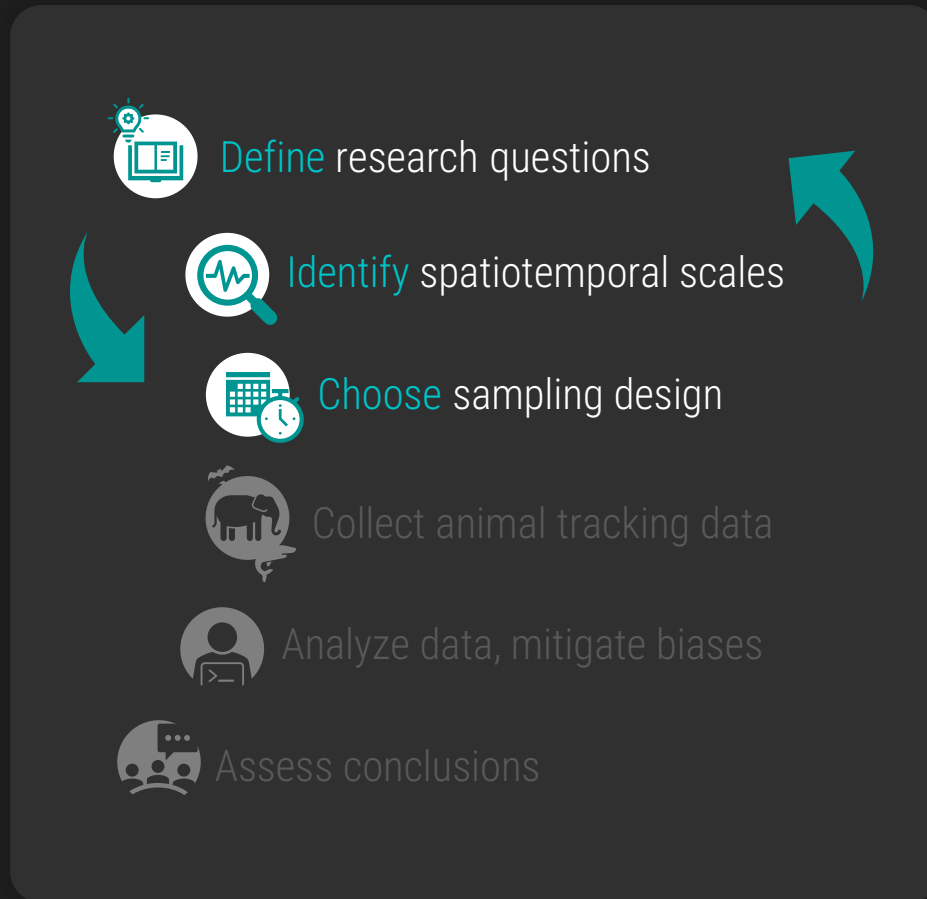
VHF
Radio-telemetry

GPS
technology



Choosing a higher fix rate leads to lower battery life.



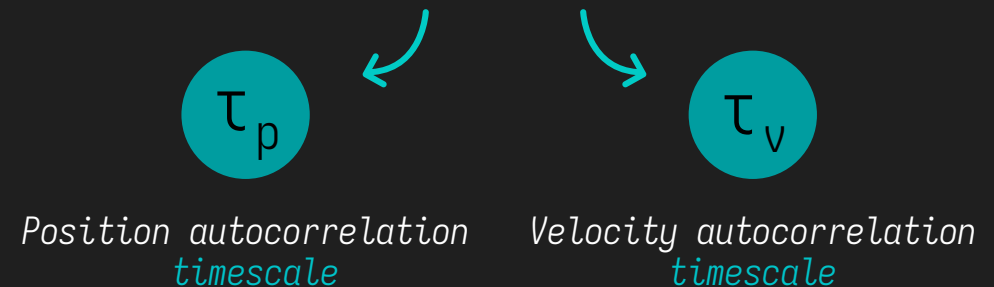


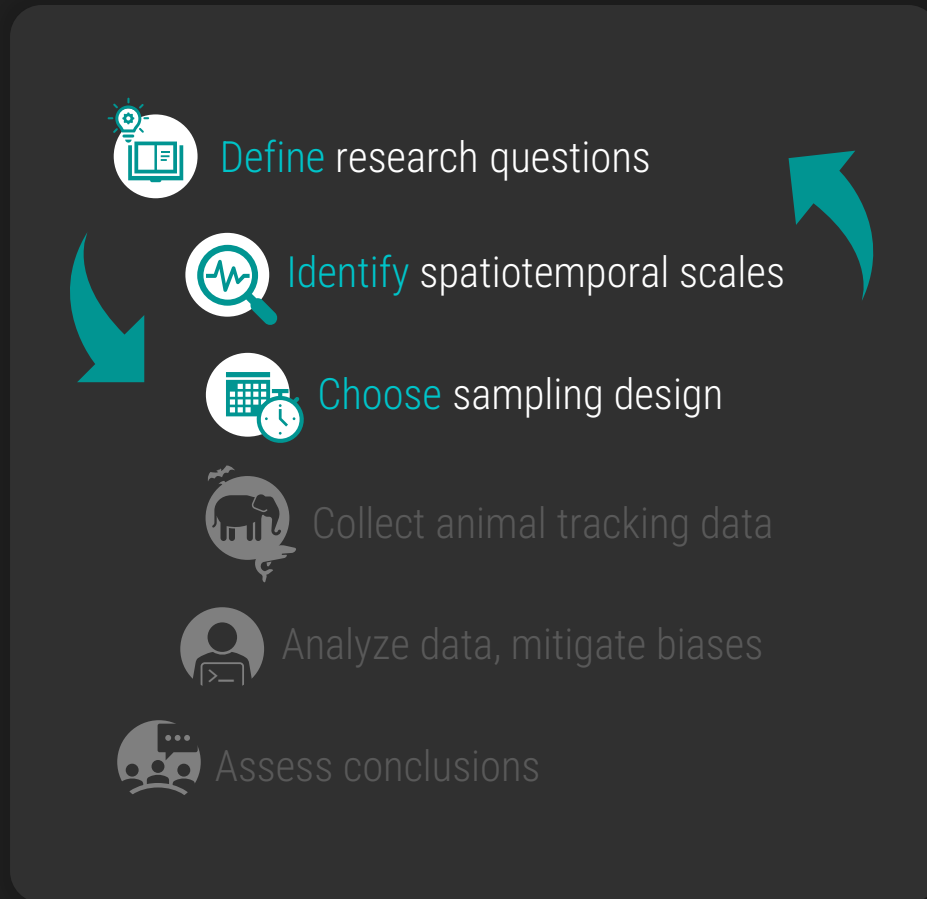
1.

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

2.

Summarize behavior using **characteristic timescales**,



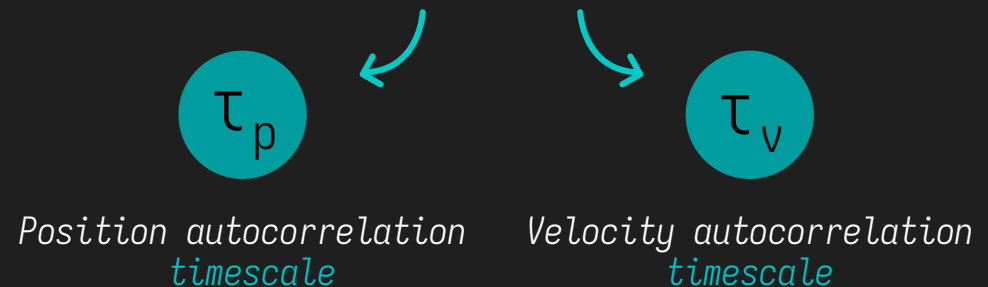


1.

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

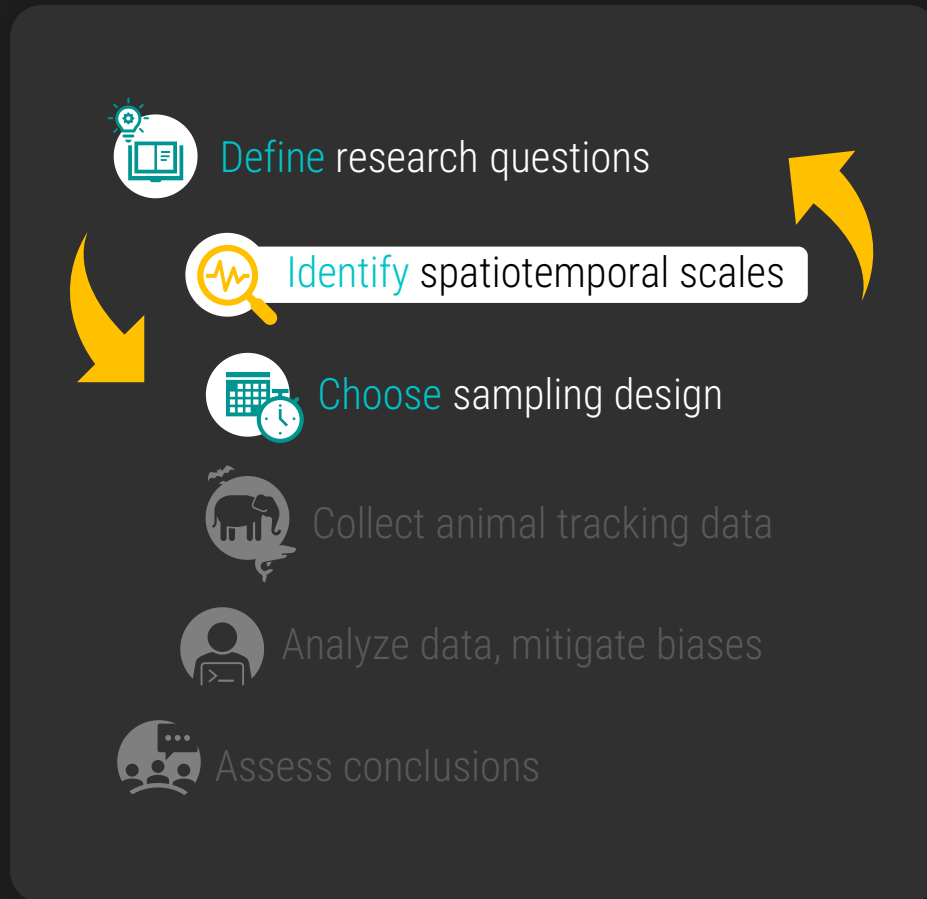
2.

Summarize behavior using **characteristic timescales**,



3.

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

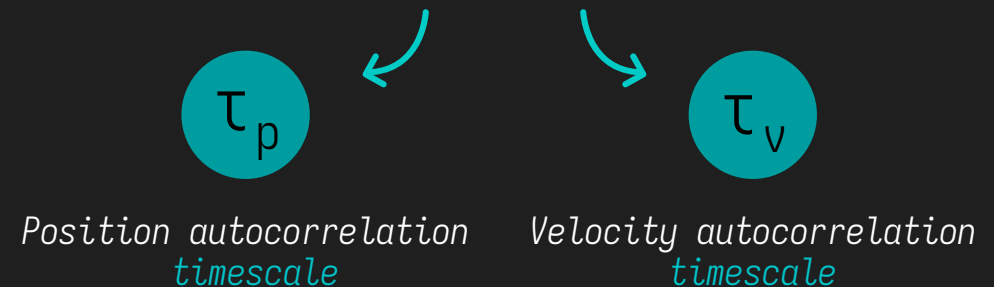


1.

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

2.

Summarize behavior using **characteristic timescales**,



3.

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



'movedesign'
Silva et al. (2023)

Objectives:

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

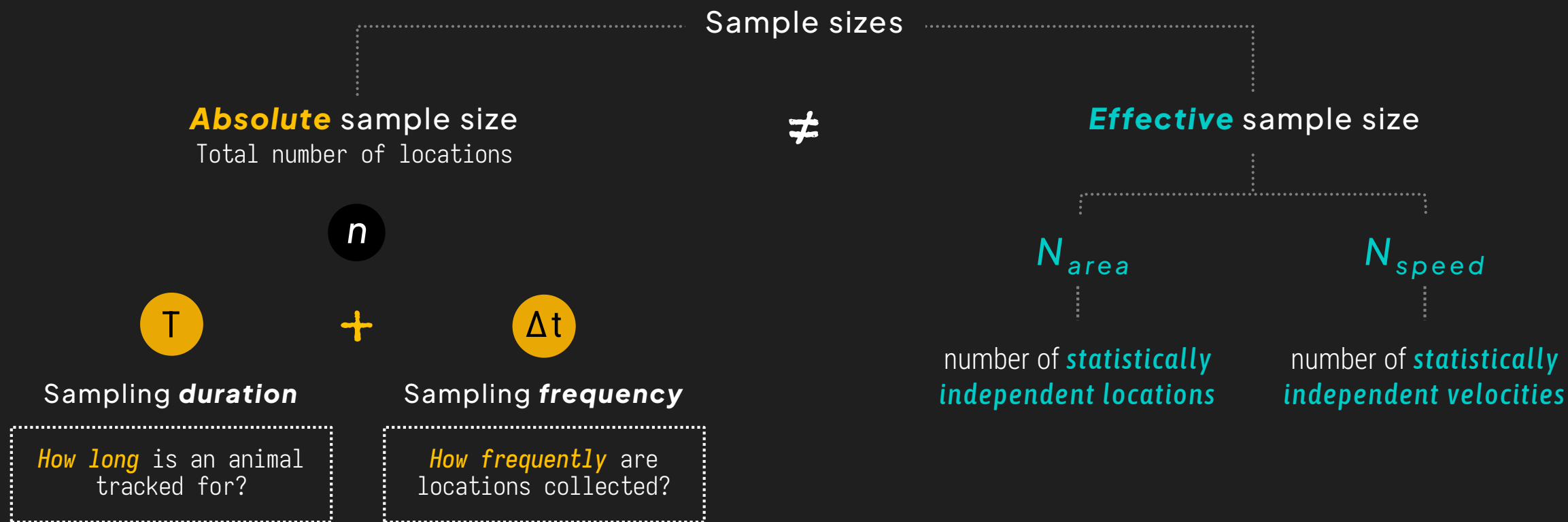
Analytical targets:

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 $N \ll n$

τ_p

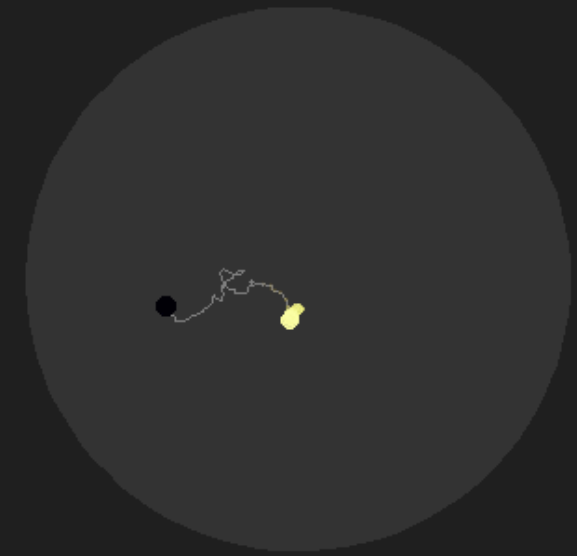
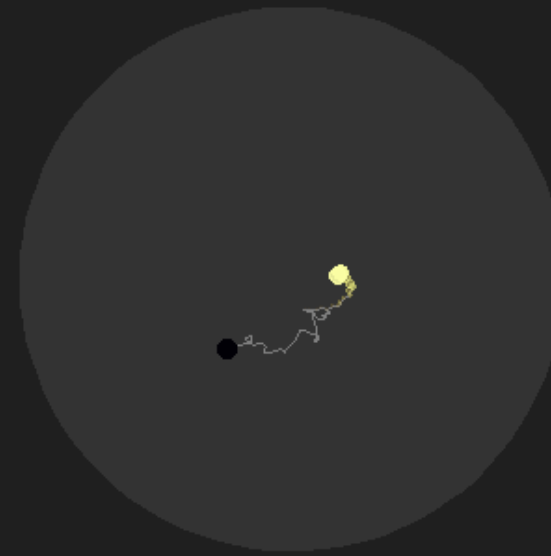
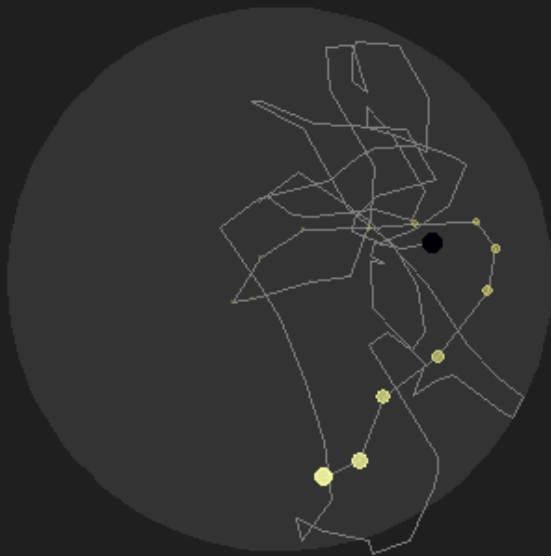
Position autocorrelation
timescale

$\tau_p = 1$ 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

τ_v

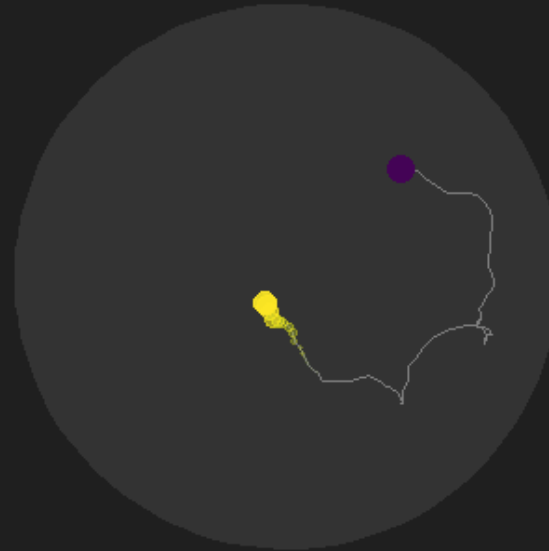
Velocity autocorrelation
timescale

$\tau_v = 1$ minute

$\tau_v = 1$ hour

$\tau_v = 12$ hours

$\tau_v = 1$ day



Δt

SPACE-USE

☐ HOME RANGE

SAMPLING FREQUENCY

How frequently are
locations collected?

.....▶

MOVEMENT BEHAVIOR

☒ SPEED & DISTANCE



Define research questions



Identify spatiotemporal scales



Choose sampling design



Collect animal tracking data



Analyze data, mitigate biases



Assess conclusions

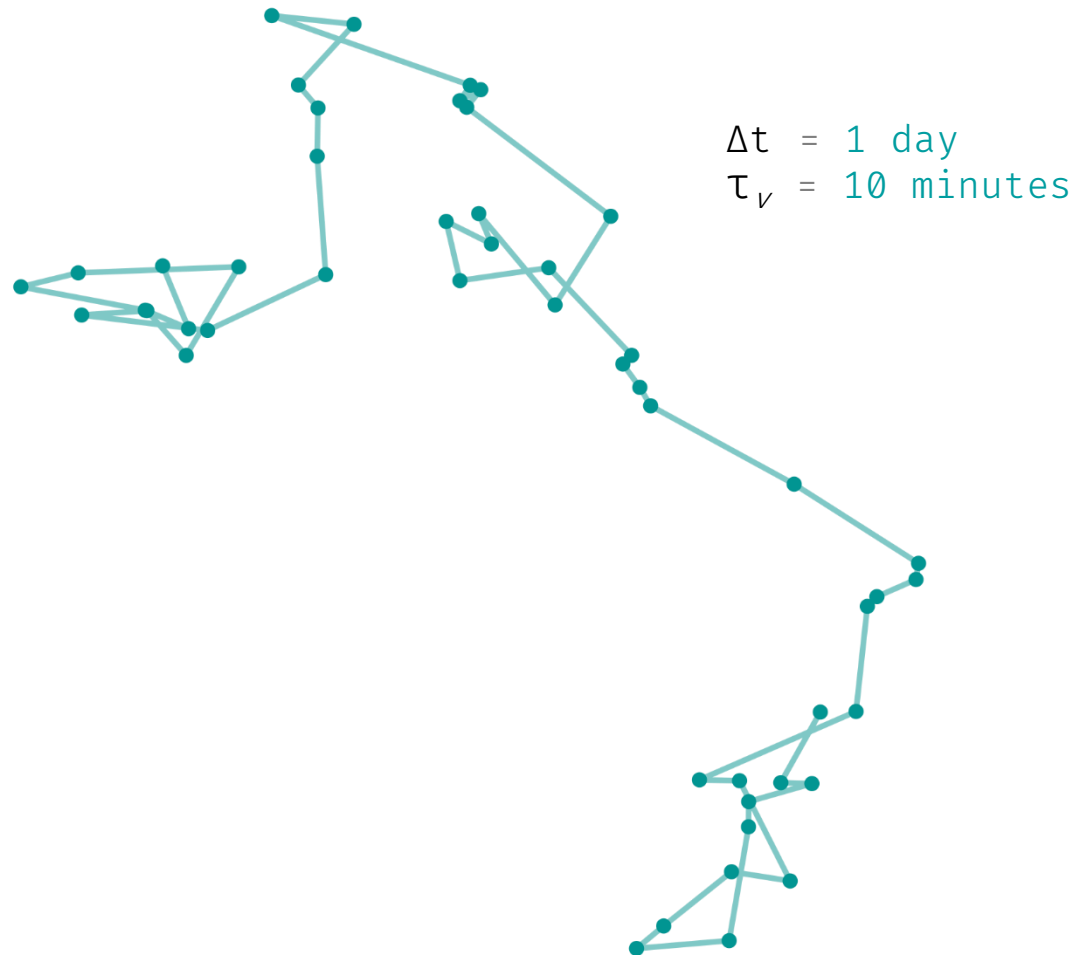
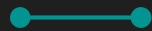


It is not physically possible for animal movement to be **uncorrelated**.
Now, the questions are:

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



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





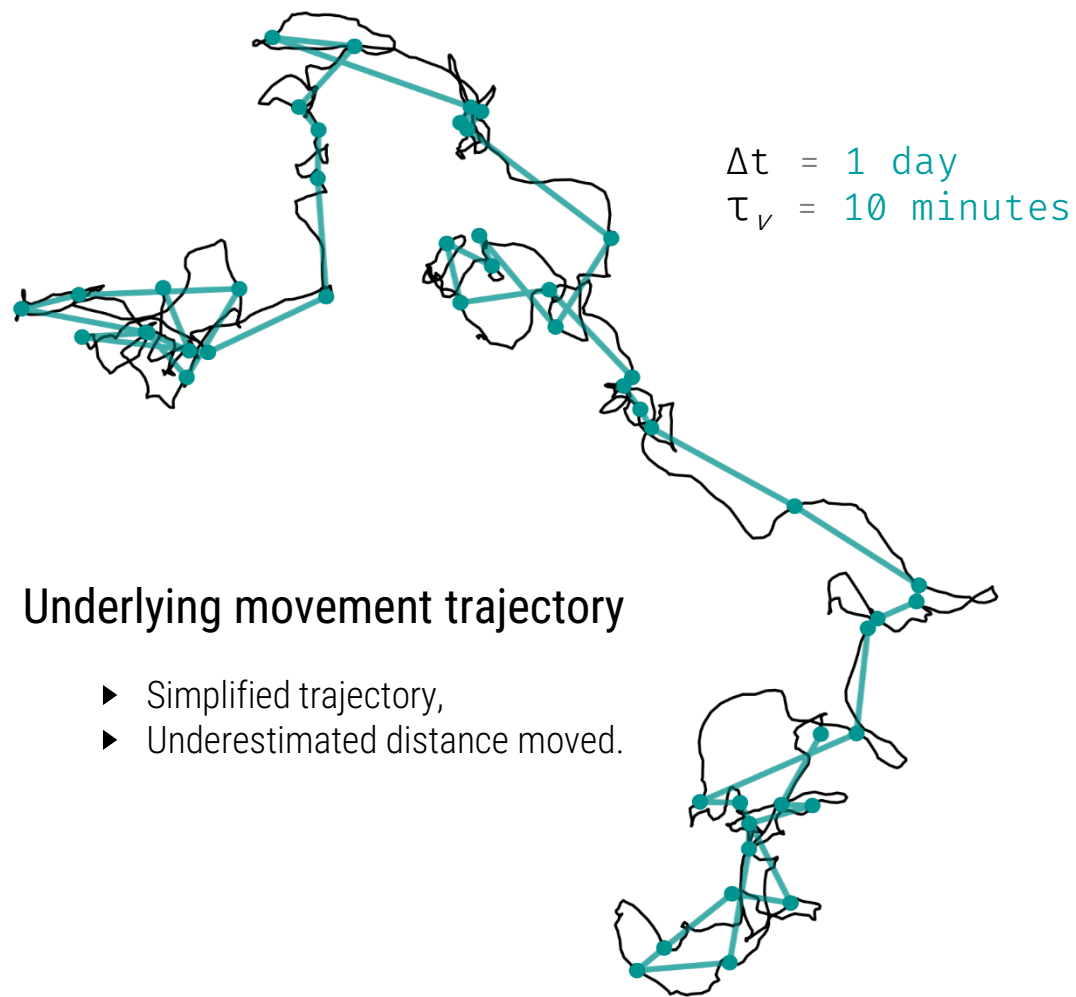
Simulated tracking dataset with a new location **once per day**, $\tau_v > \Delta t$.



We must carefully consider the frequency of data collection!

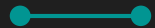


For the same Δt , this bias will be greater for individuals with more tortuous movement (shorter τ_v).





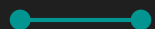
Simulated tracking dataset with
a duration of 6 months, $\tau_p > T$.



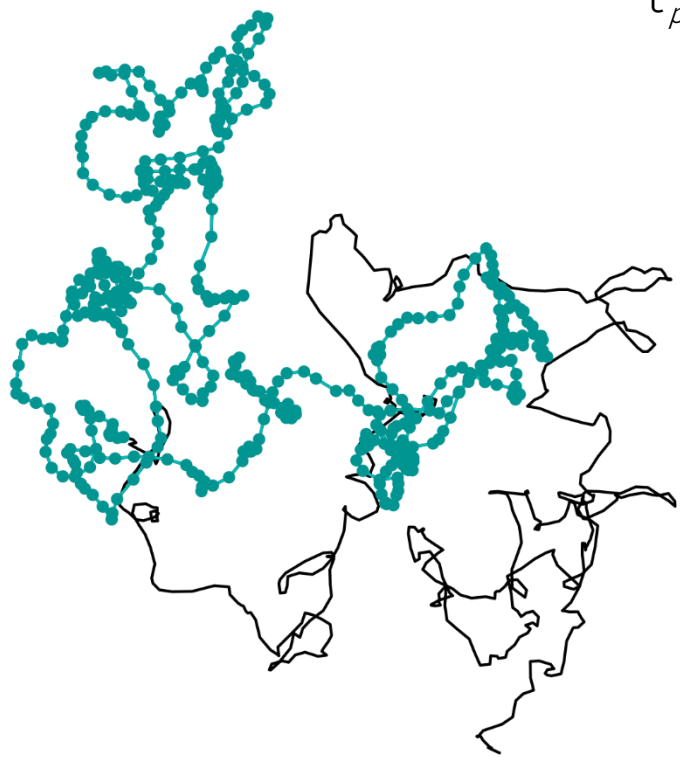
$$T = 6 \text{ months}$$
$$\tau_p = 8 \text{ months}$$



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



$T = 12$ months
 $\tau_p = 8$ months



—— Movement trajectory for the following 6 months.



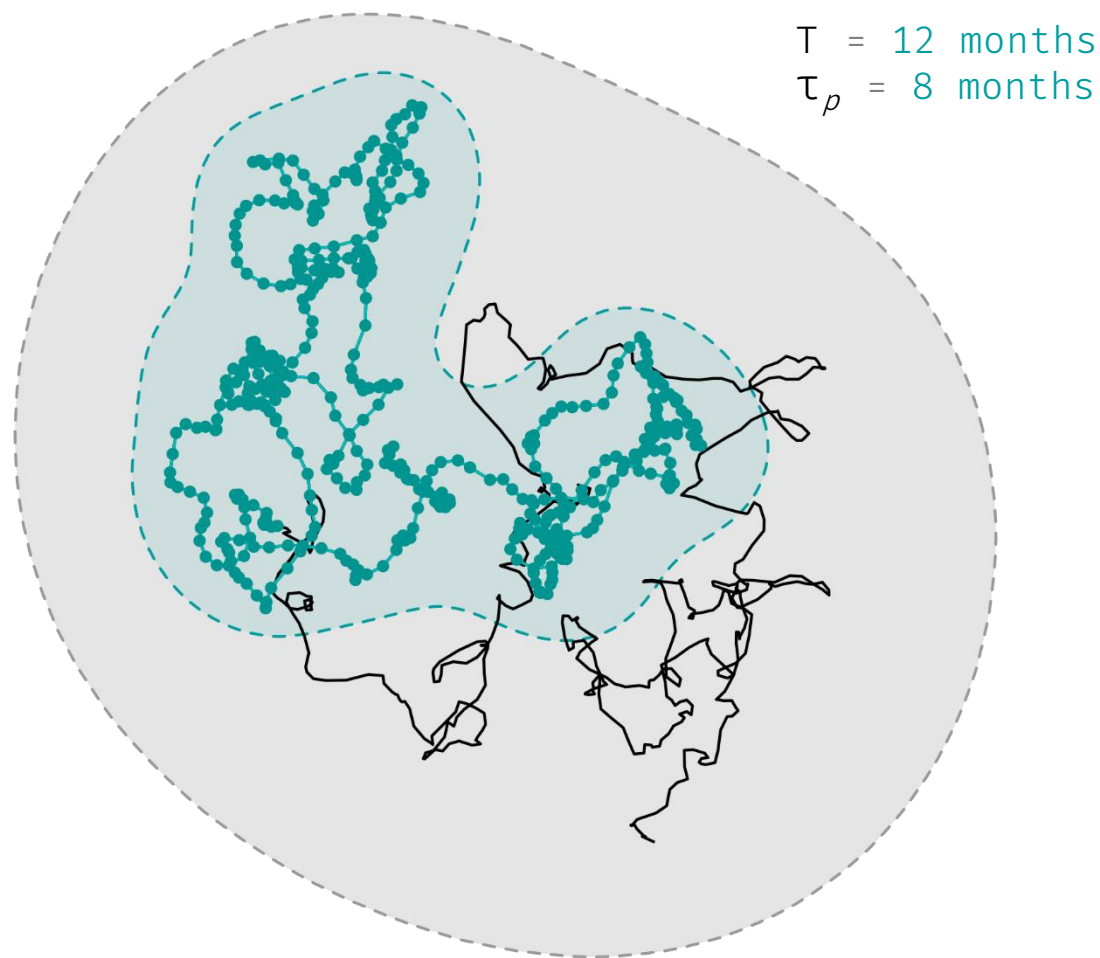
Simulated tracking dataset with a duration of **6 months**, $\tau_p > T$.



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 (τ_p).



$T = 12$ months
 $\tau_p = 8$ months

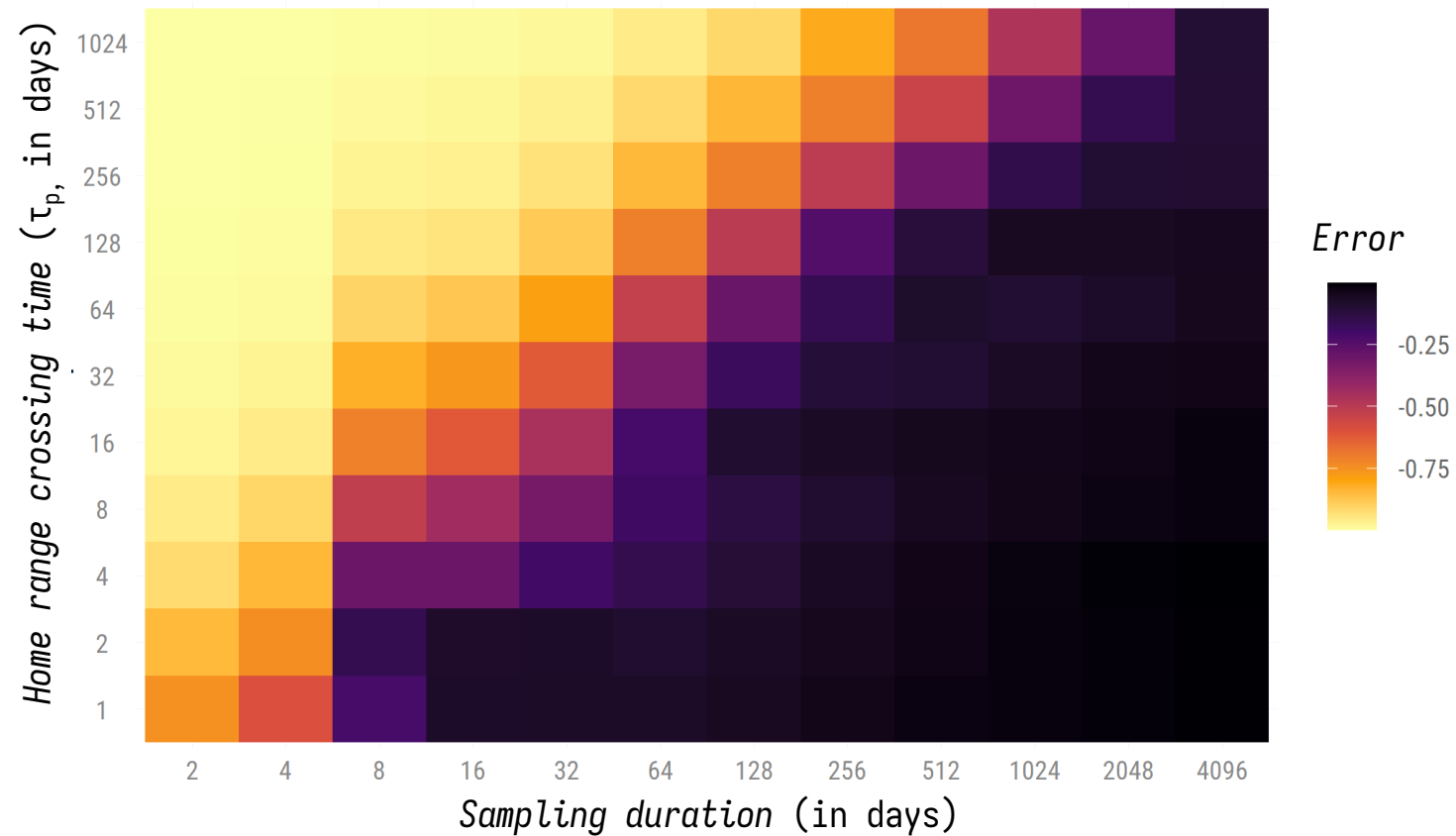
—— Movement trajectory for the following **6 months**.

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



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

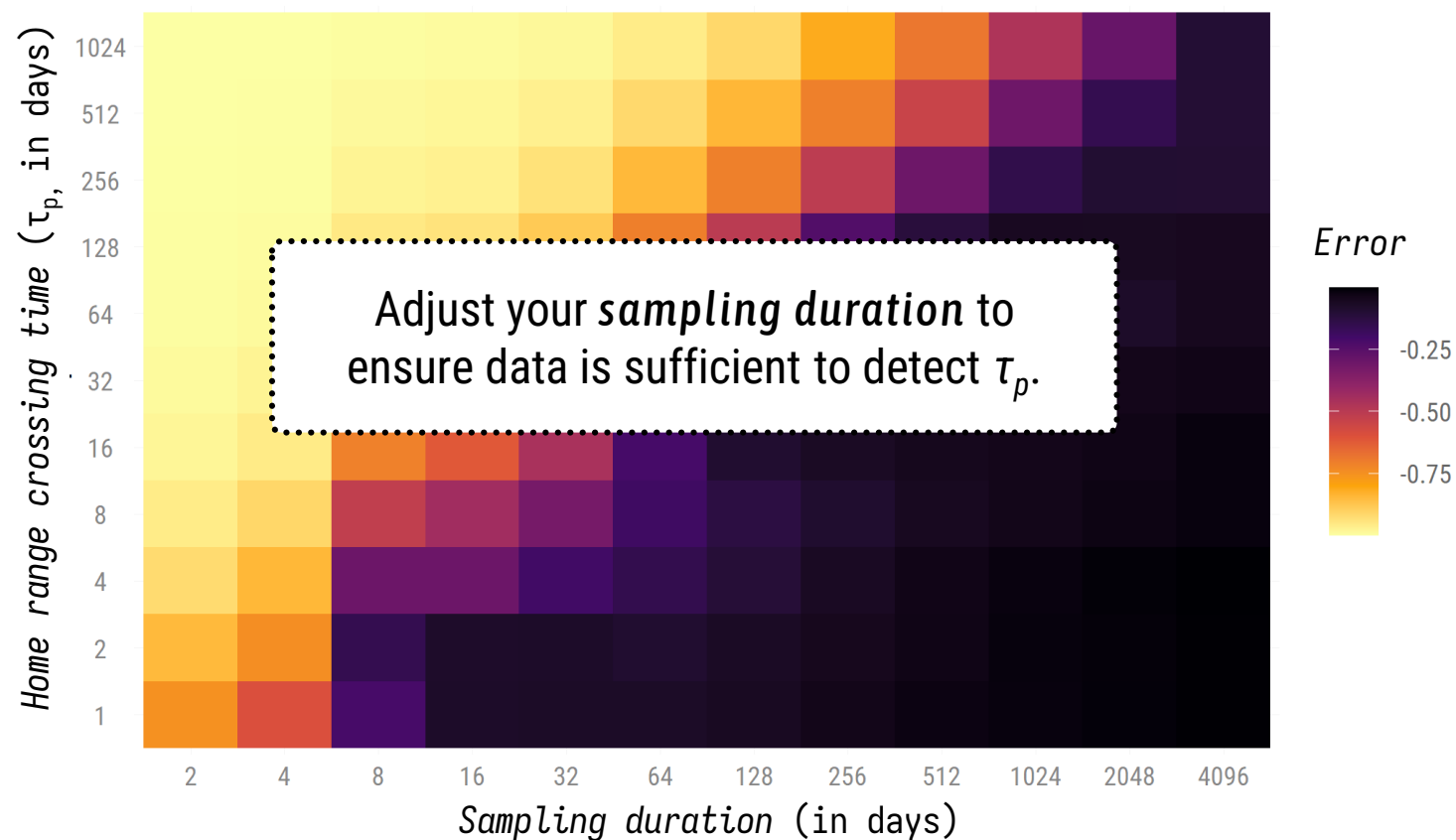
ℓ Fleming *et al.* (2015)





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

Fleming et al. (2015)





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

ℓ Fleming *et al.* (2015)

Given a relative target bias of $\approx 5\%$,

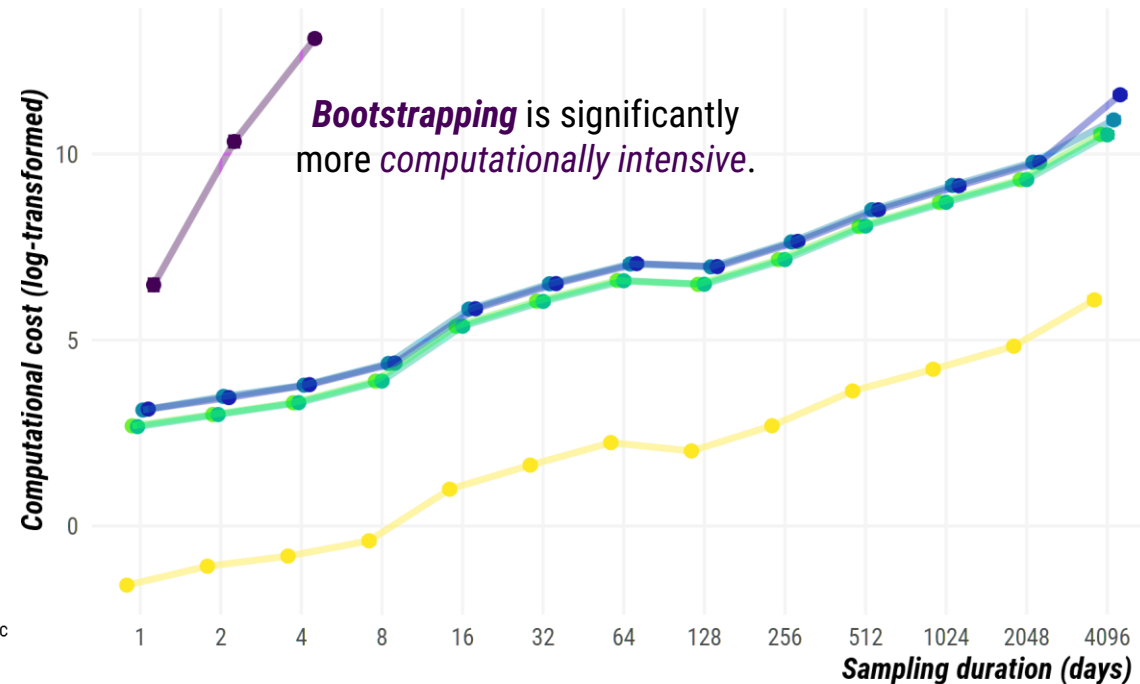
Minimum N_{area} for **ML** is ≈ 20 ;

Minimum N_{area} for **pHREML** is ≈ 4.5 ;

Minimum N_{area} for **bootstrapped pHREML** is ≈ 2.7 .

Method:

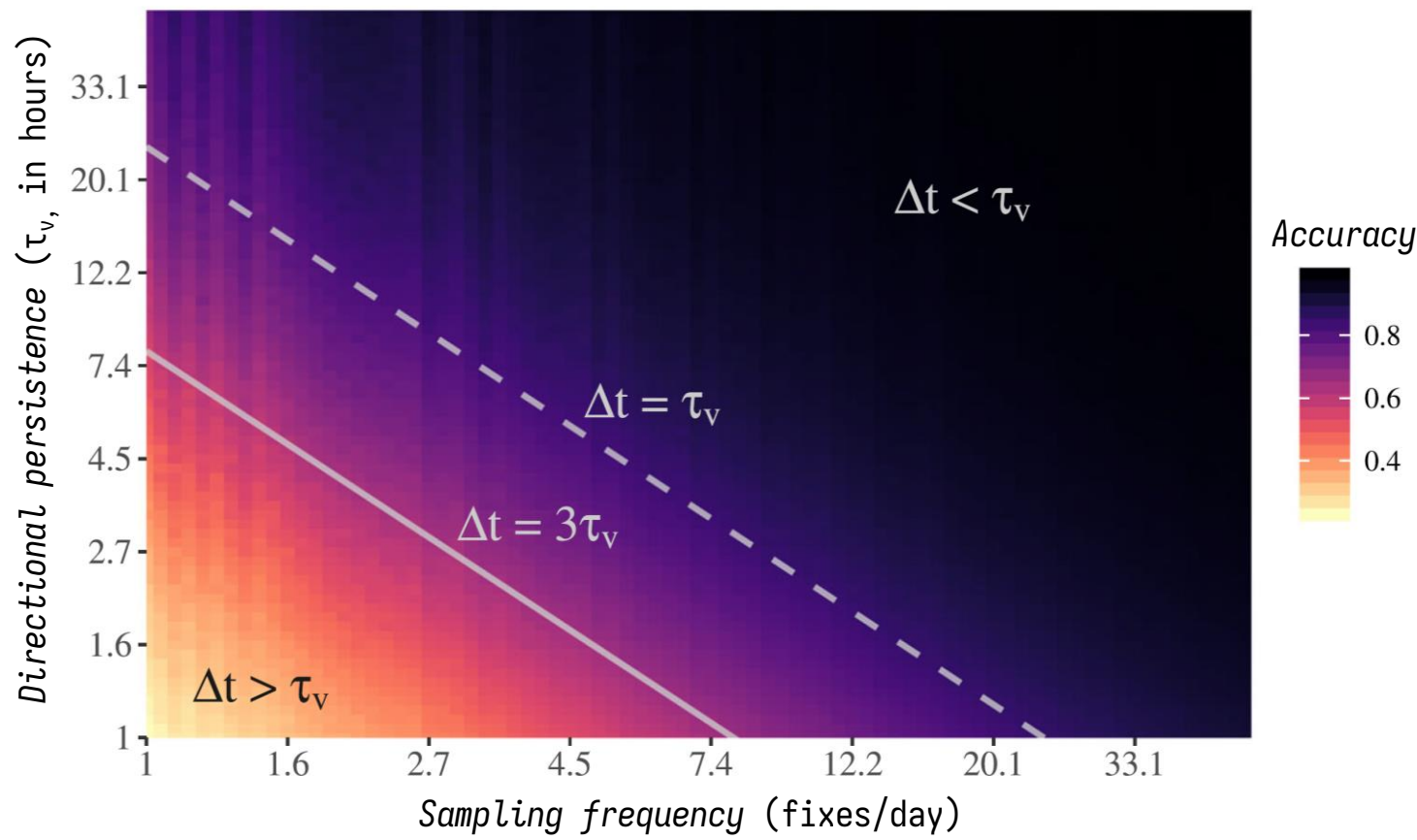
- KDE
- AKDE
- AKDE_c
- pHREML AKDE_c
- pHREML wAKDE_c
- Bootstrapped pHREML wAKDE_c





2. Speed & distance — Continuous-time speed and distance (CTSD):

ℓ Noonan *et al.* (2019)



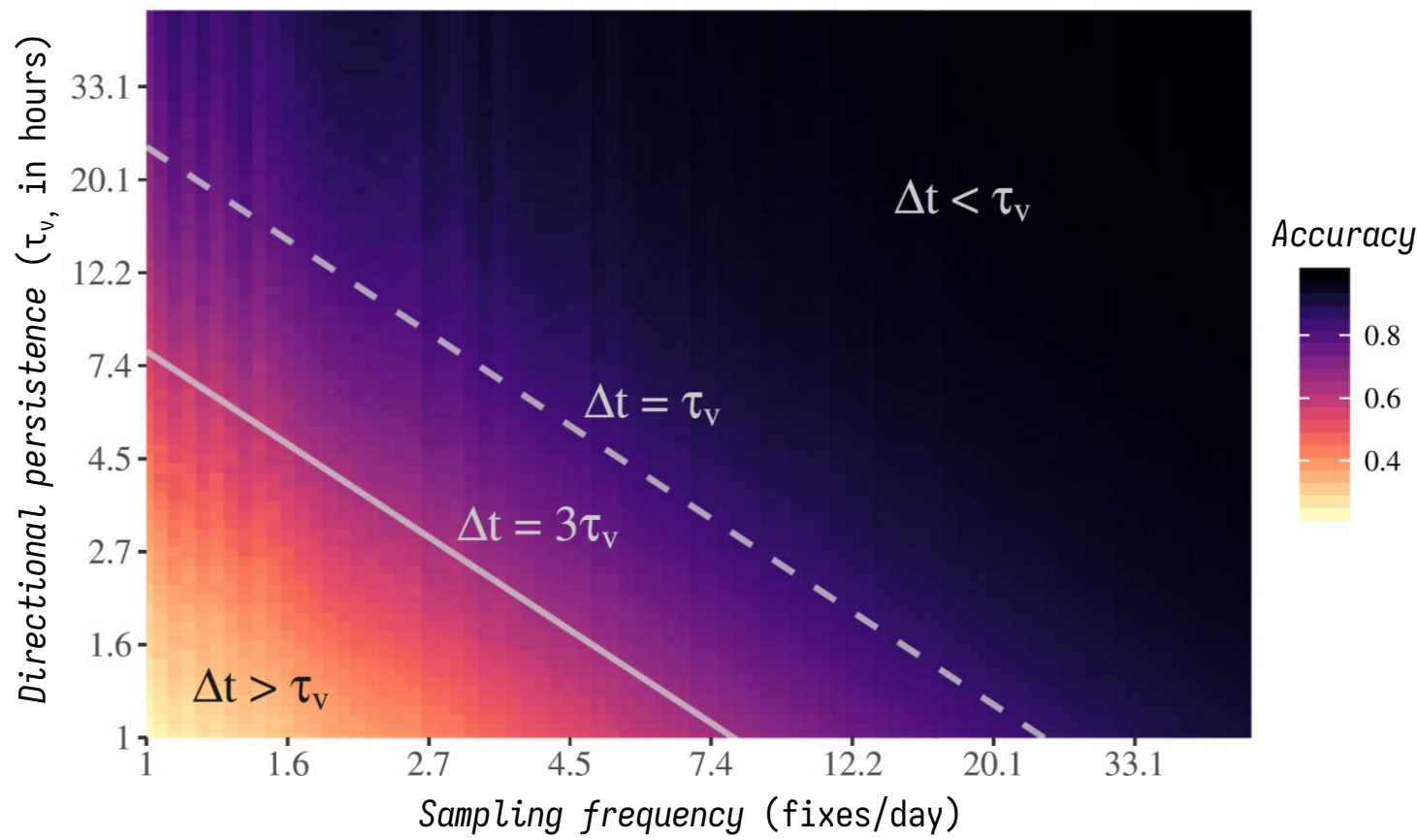
If $\Delta t > 3\tau_v$, no statistically significant signature of the animal's velocity will remain in the location data.

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



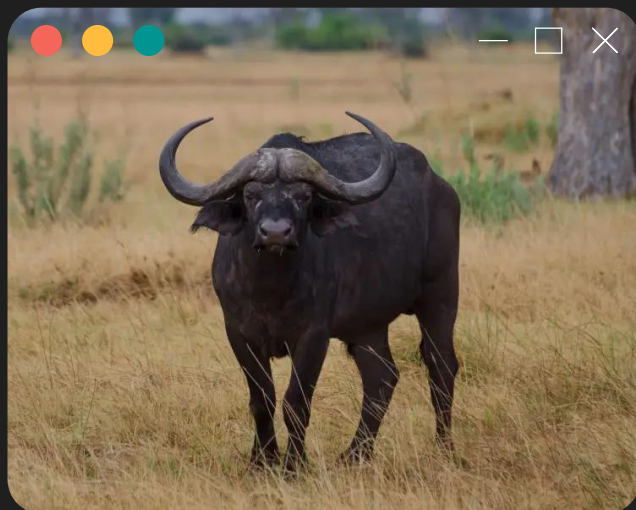
2. Speed & distance — Continuous-time speed and distance (CTSD):

ℓ Noonan *et al.* (2019)



Adjust your **sampling interval** to ensure data is of **sufficient resolution** to detect τ_v .

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



Position autocorrelation

10.3 days
8.2 – 12.8
(τ_p)

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

African buffalo

(*Syncerus caffer*)

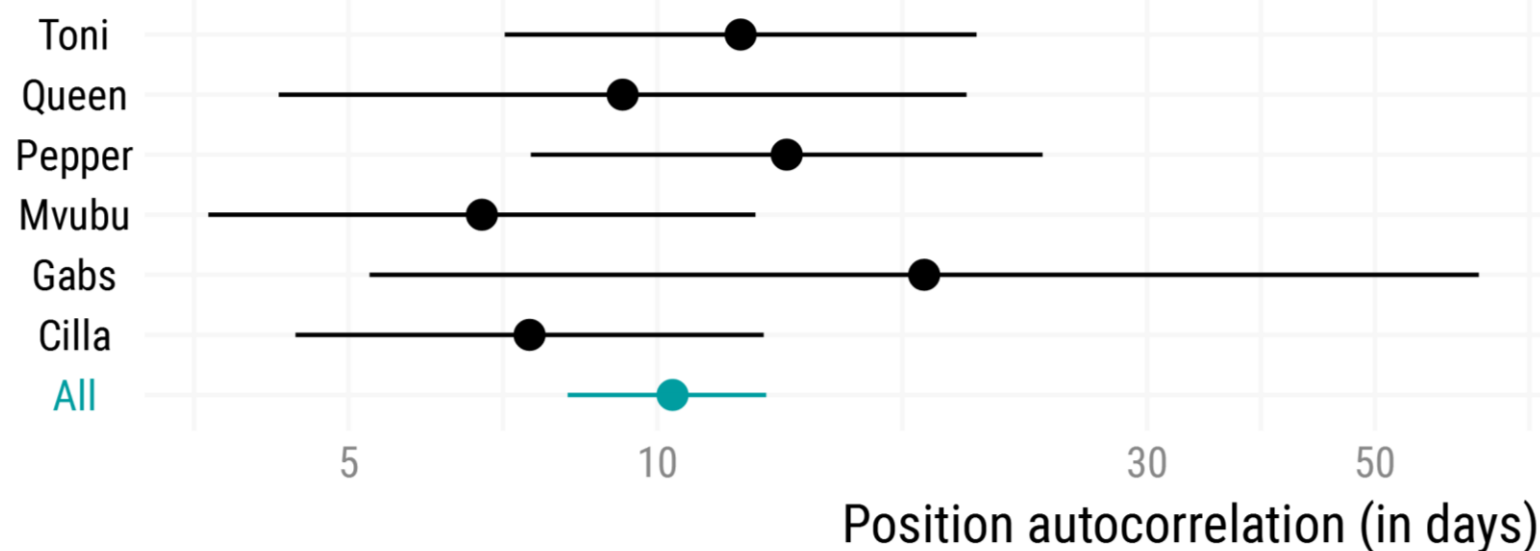
⌵ Timescales

🐾 Dataset

Show parameter:

🕒 Position autocorrelation (τ_p)

Velocity autocorrelation (τ_v)





Species



Upload, or
Select data



choose the same
(or a similarly
behaved) species

Model fitting & selection

extracting measures

- ▶ Spatial variance, σ_p
- ▶ Timescale parameters, τ :
 - Position autocorrelation, τ_p
 - Velocity autocorrelation, τ_v
- ▶ Velocity, σ_v

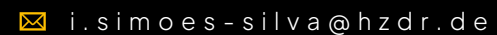


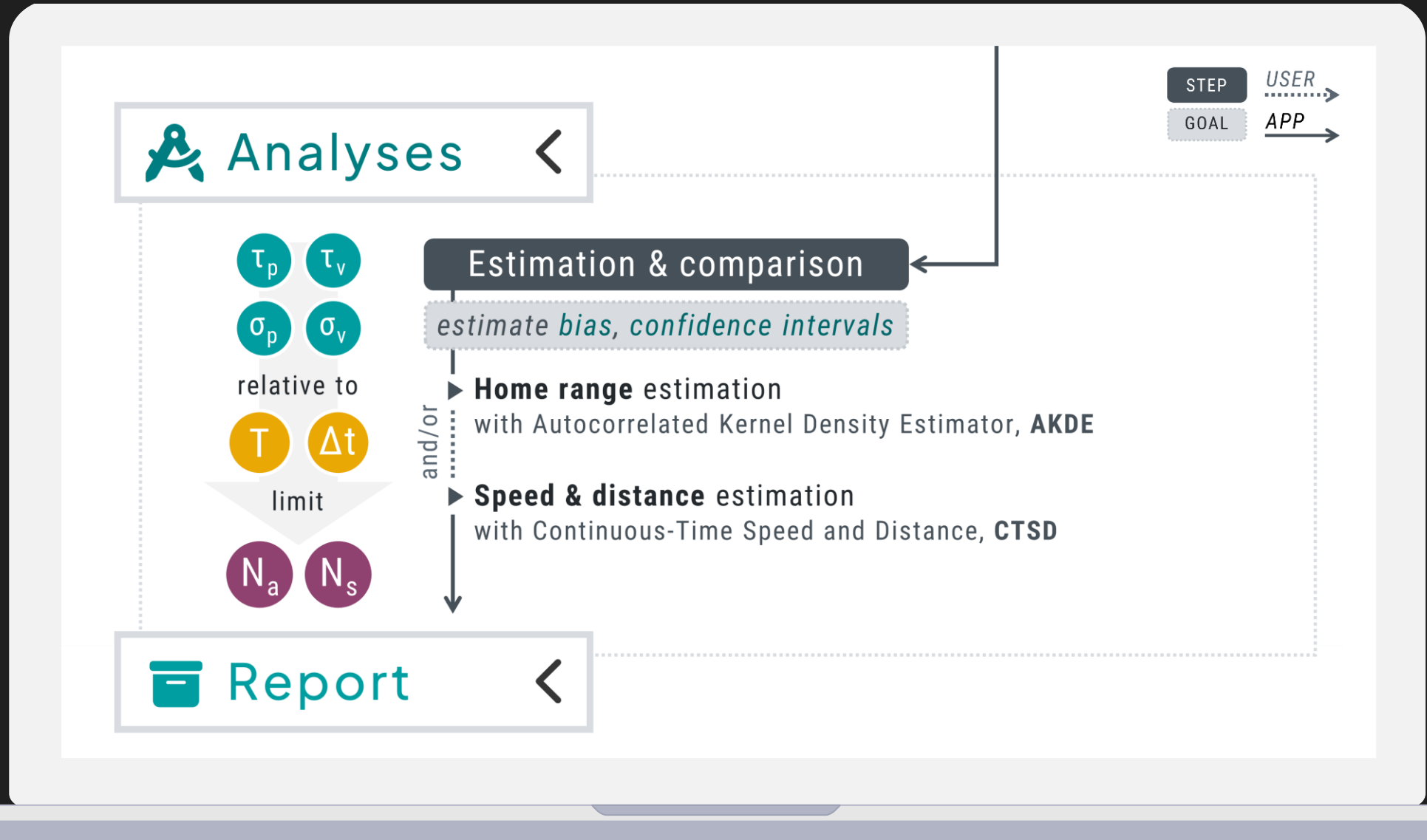
find or test
parameters
directly

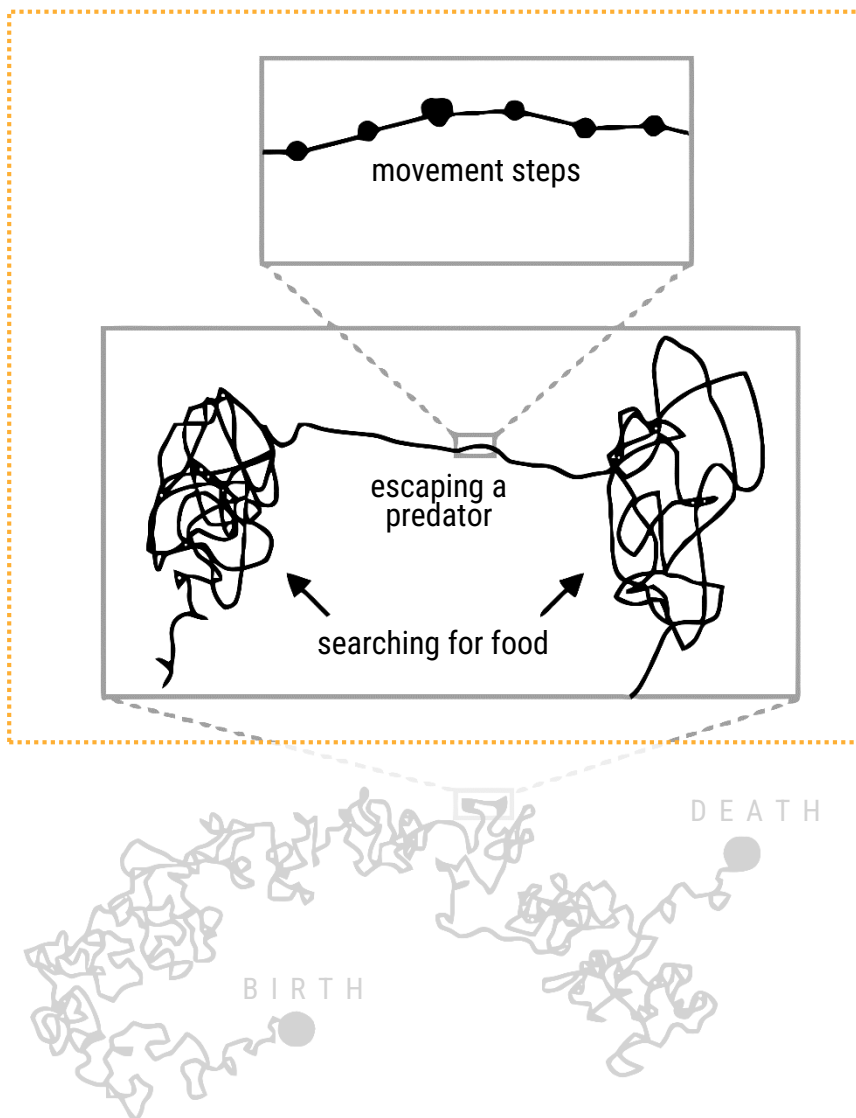


Simulate
data

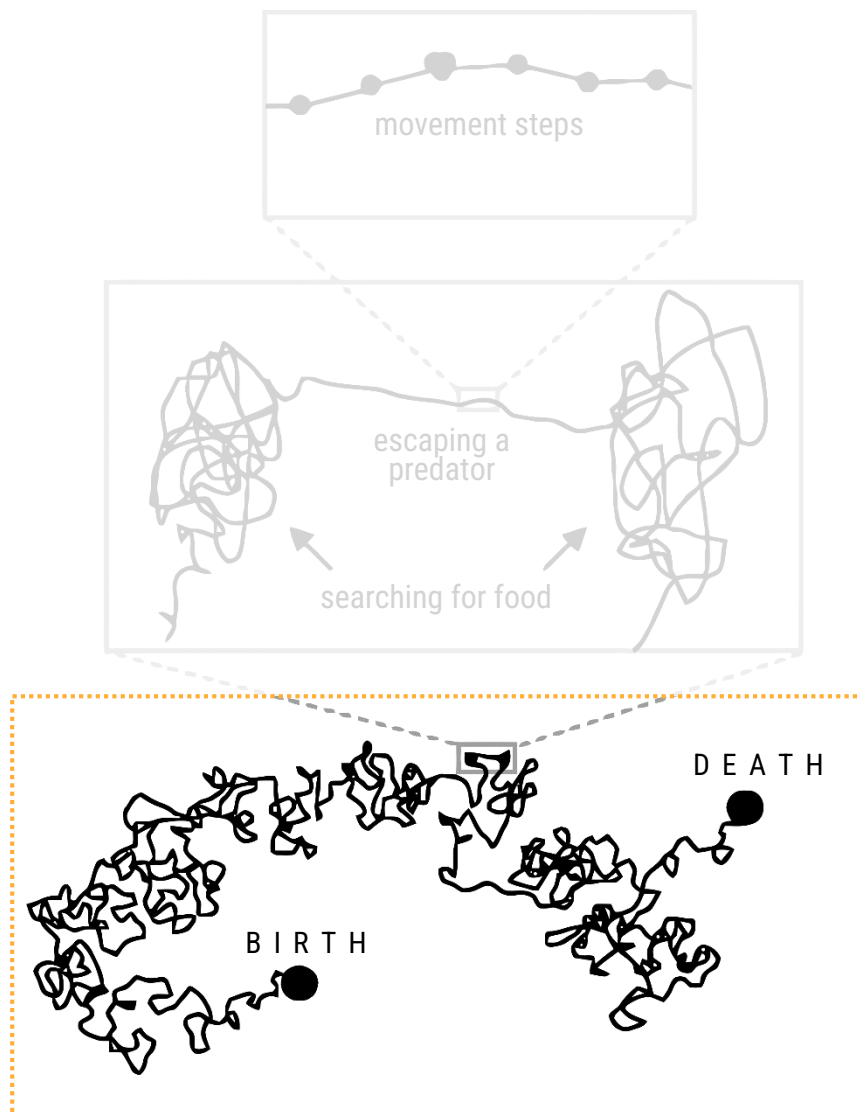






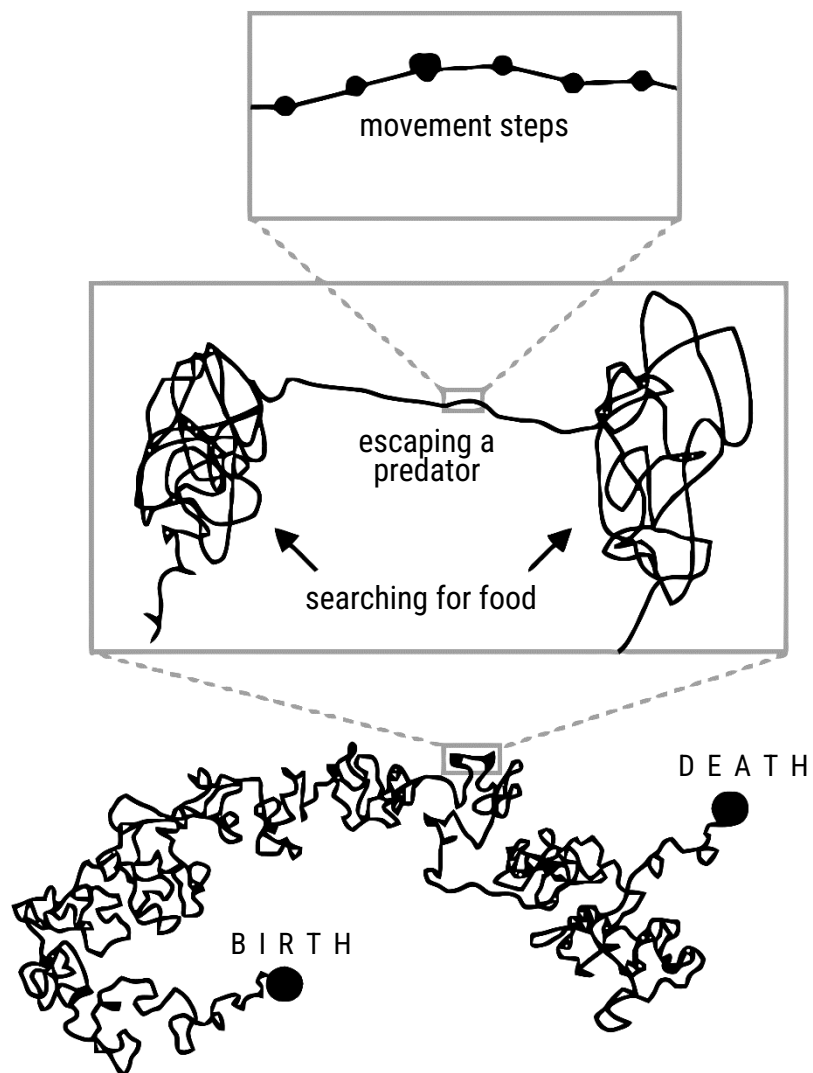


If the goal is **speed & distance estimation**,
adjust your **sampling interval (Δt)** to ensure data
is of sufficient resolution to detect τ_v .



If the goal is **speed & distance estimation**,
adjust your **sampling interval (Δt)** to ensure data
is of sufficient resolution to detect τ_v .

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



If the goal is **speed & distance estimation**,
adjust your **sampling interval (Δt)** to ensure data
is of sufficient resolution to detect τ_v .

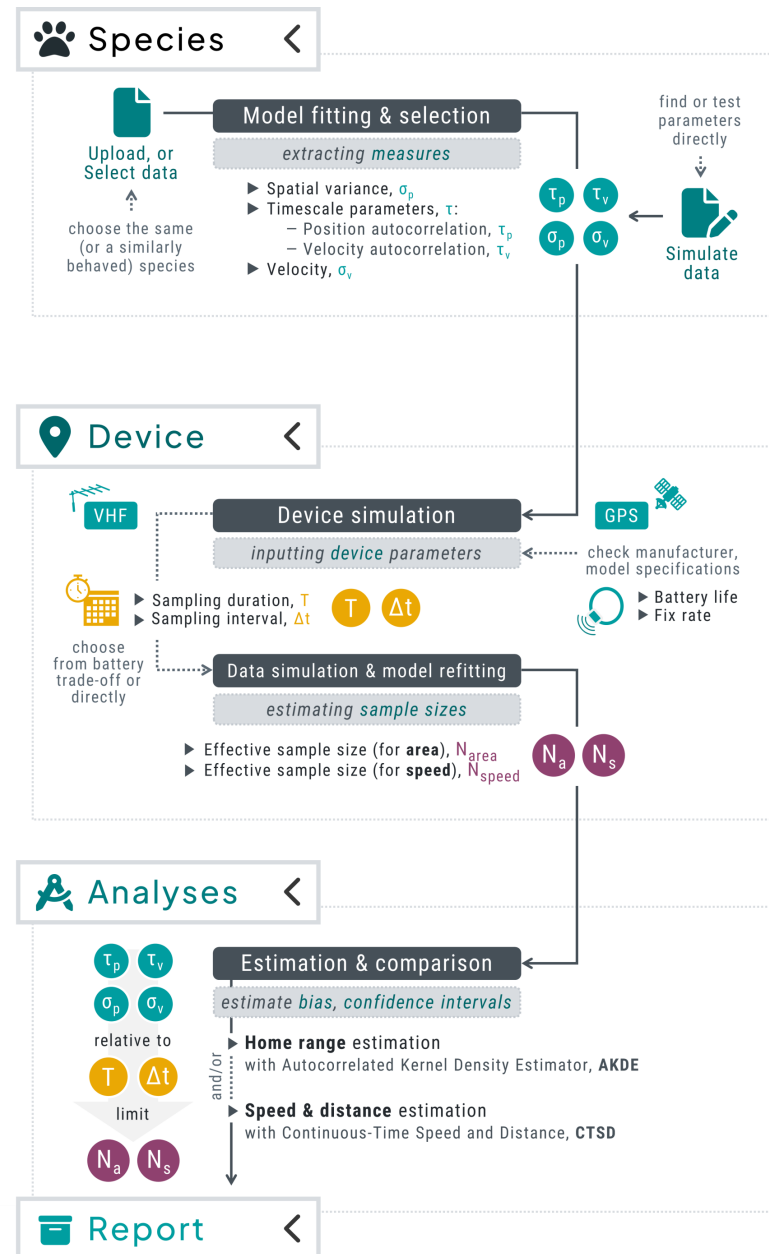
If the goal is **home range area estimation**,
adjust your **sampling duration (T)** to ensure data
is sufficient to detect τ_p .

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



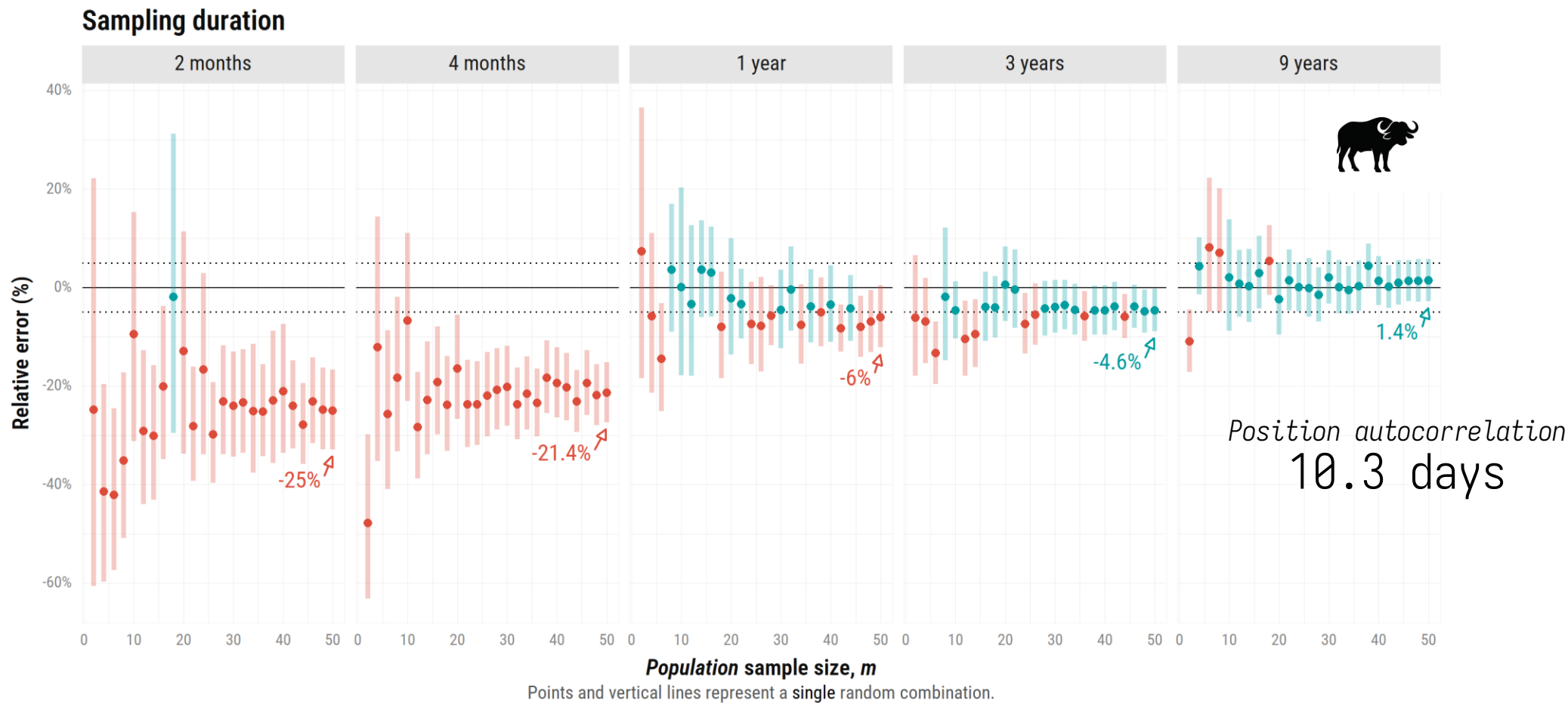
New version (v0.3.0):

- 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)



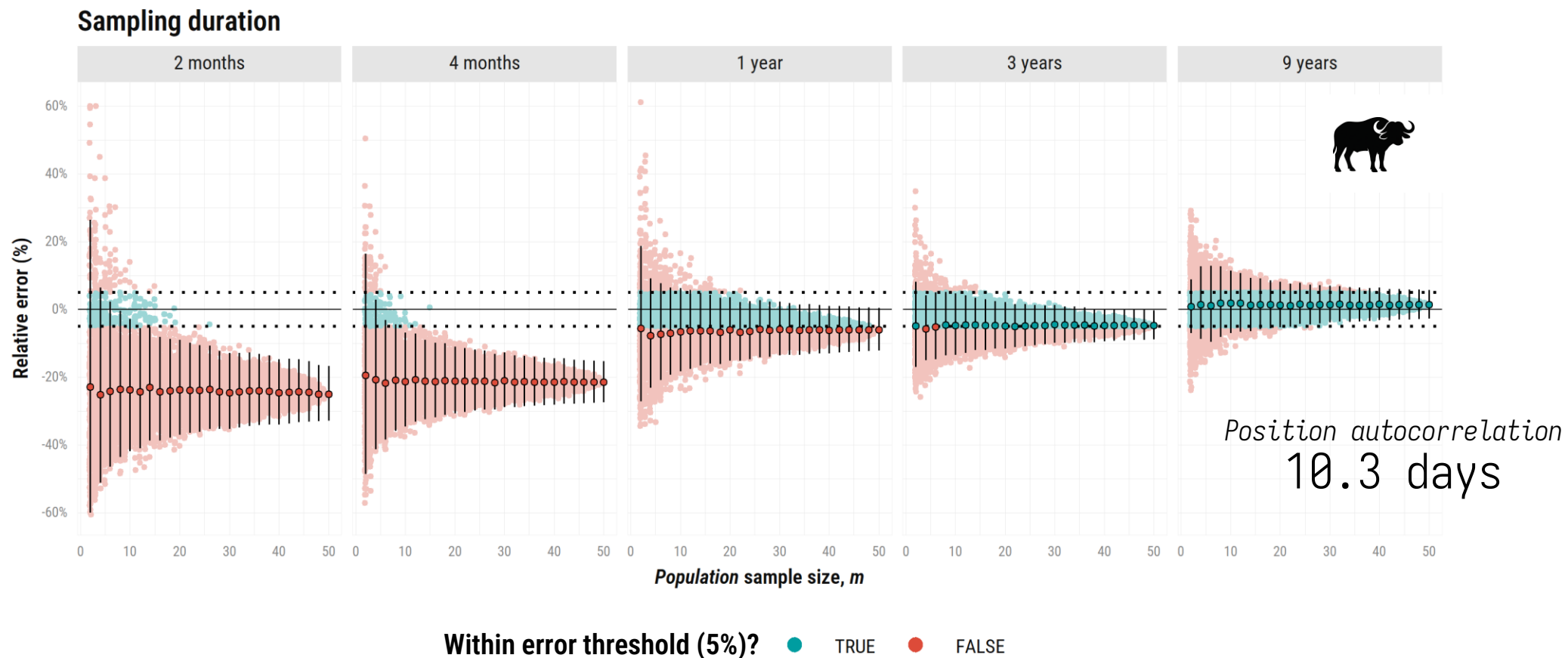


3.1. Population-level inferences — mean home range areas





3.1. Population-level inferences — mean home range areas





3.2. Population-level inferences — mean movement speed

