

ECOBHVR 2025

Too few, too many, or just right?

Optimizing sample sizes in animal tracking projects



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Introduction



Adapted from Purgar *et al.* (2022)



Introduction

The cost of a single **GPS unit** can reach several thousand dollars, and each additional tracked individual adds **logistical overheads** (e.g., field time, personnel, travel).

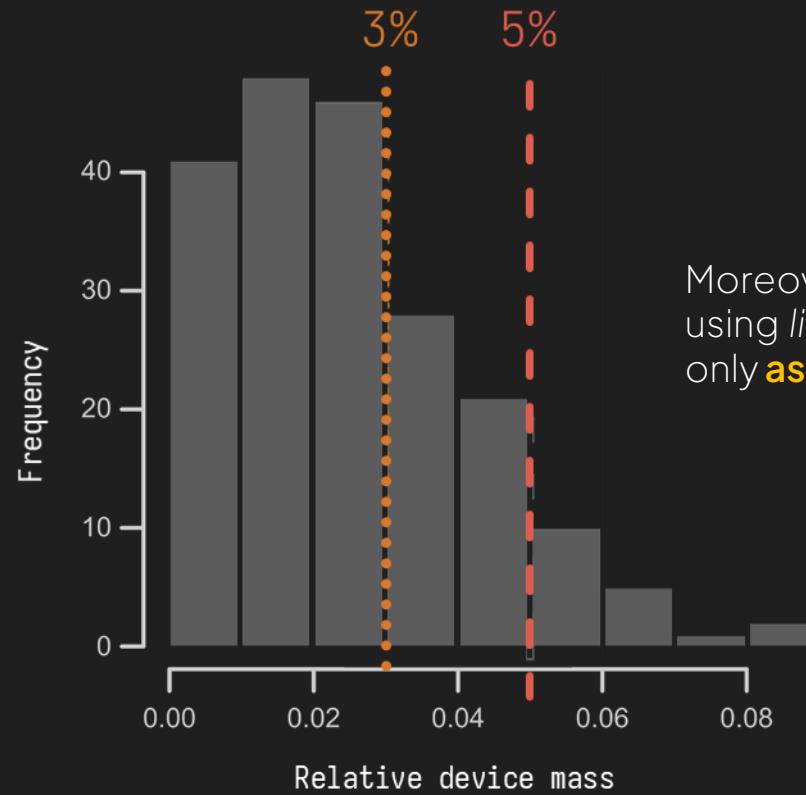
📷 PR Ganapathy



📷 Robert Ronconi



📷 Thanisha Kumar

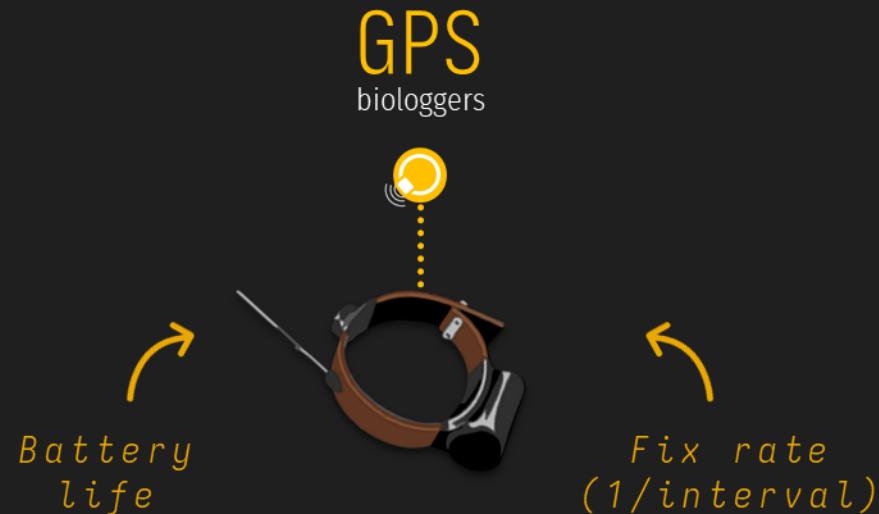


Moreover, **animal welfare guidelines** require using *lighter, less invasive tags*, and tracking only **as many individuals as necessary**.

📄 Adapted from Portugal & White (2018)

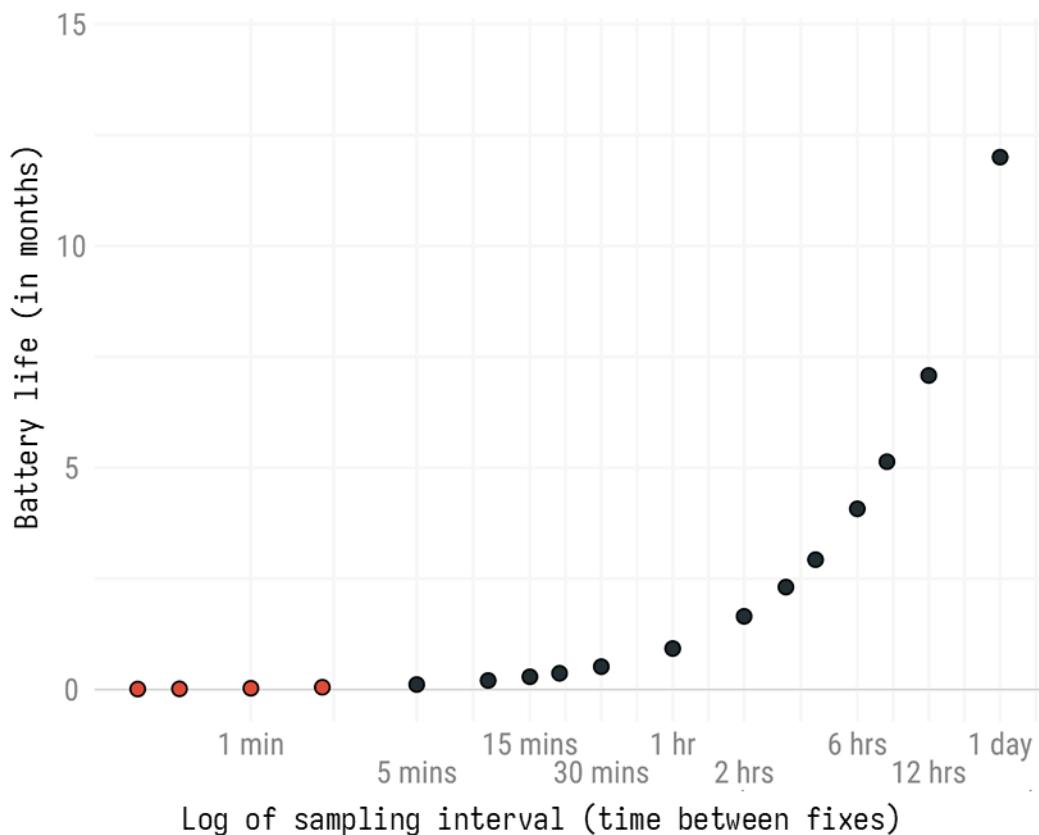


Introduction



Increasing resolution (e.g., sampling rate or fix frequency) consumes more power, **reducing battery life** and limiting tracking duration.

Trade-off between battery life and resolution:



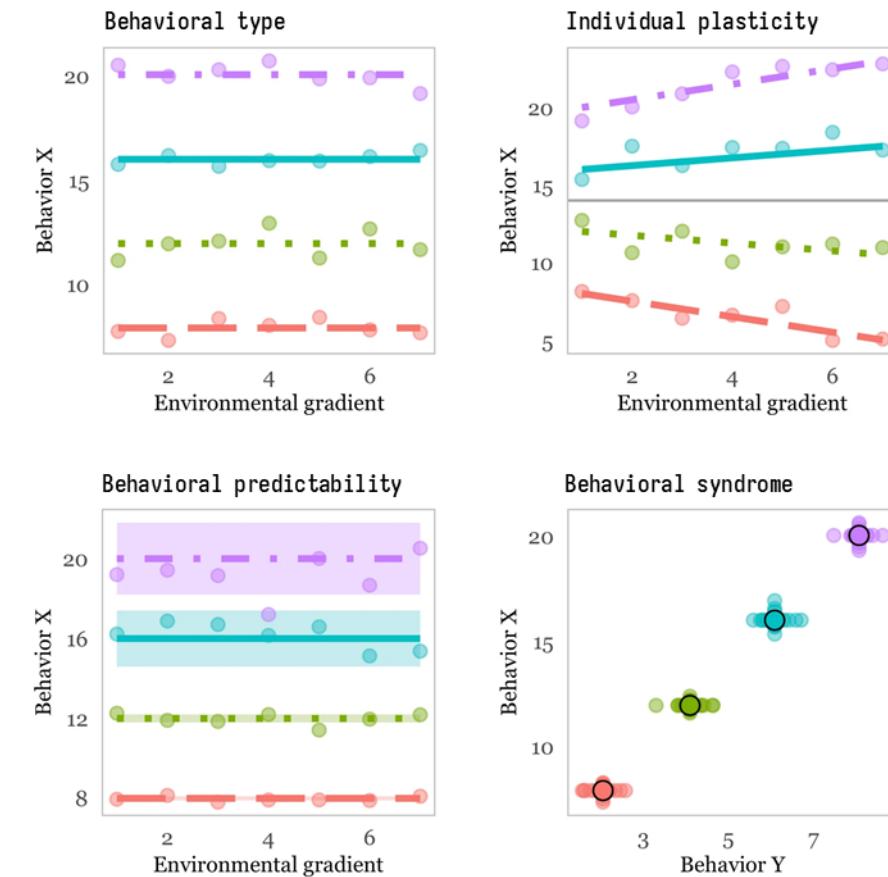
Silva et al. (2023)



Introduction



The intrinsic **behavioral variability** within populations increase the risk of producing biased or unreliable estimates.



¶ Hertel *et al.* (2020)

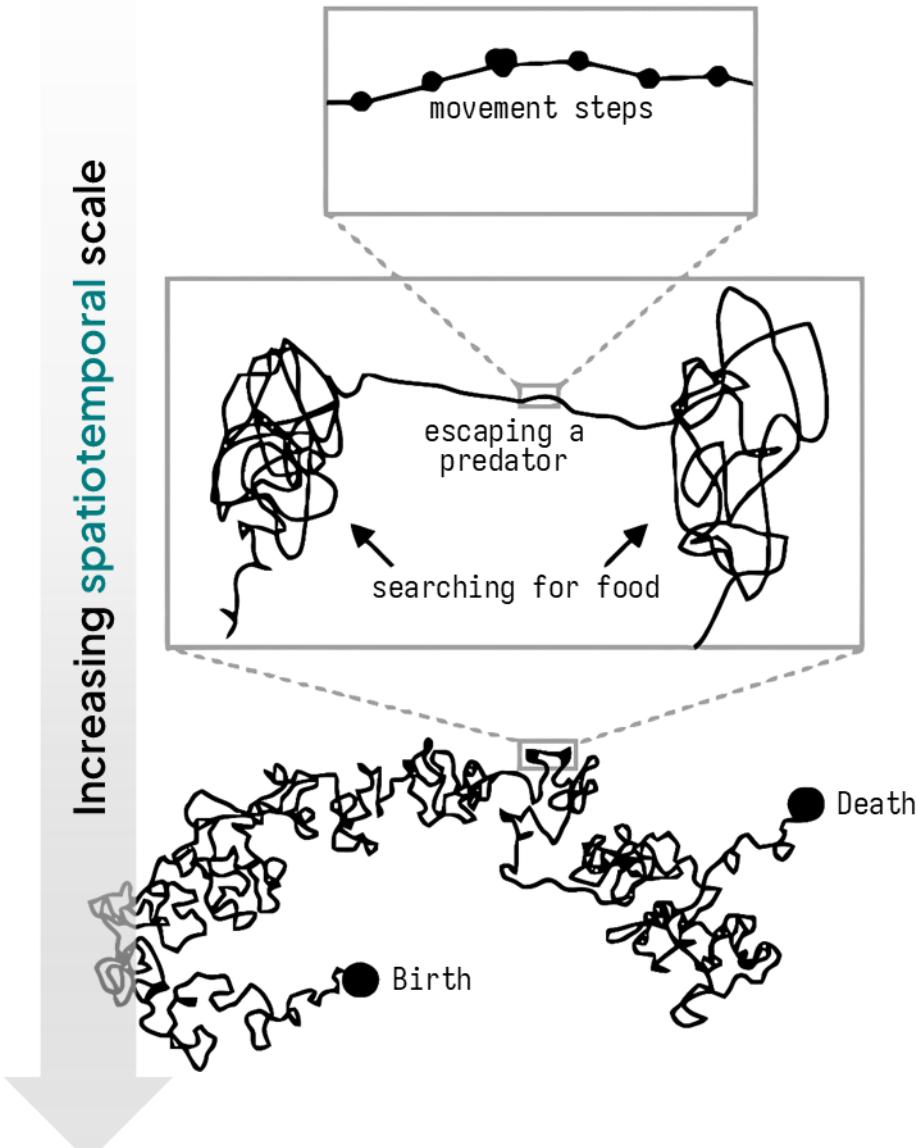


Introduction

Fine-scale processes

e.g., Movement speed
Distance traveled

Adapted from Nathan et al. (2008)



Large-scale processes

e.g., Home range area



Define research questions



Identify spatiotemporal scales



Determine sampling parameters



Collect animal tracking data



Analyze data, mitigate biases



Assess conclusions



1

Sampling duration:

How long should each individual be monitored?

2

Sampling frequency:

How often should locations be recorded?

3

Sample size:

How many individuals should be tracked?

Ideally, researchers would track *as many animals as possible, for as long as possible, and as frequently as possible.*



```
> install.packages(movedesign)
```

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

 Silva *et al.* (2023)
Silva *et al.* (2025)



Conceptualization



Define research questions



Identify spatiotemporal scales



Determine sampling parameters



Collect animal tracking data



Analyze data, mitigate biases



Assess conclusions

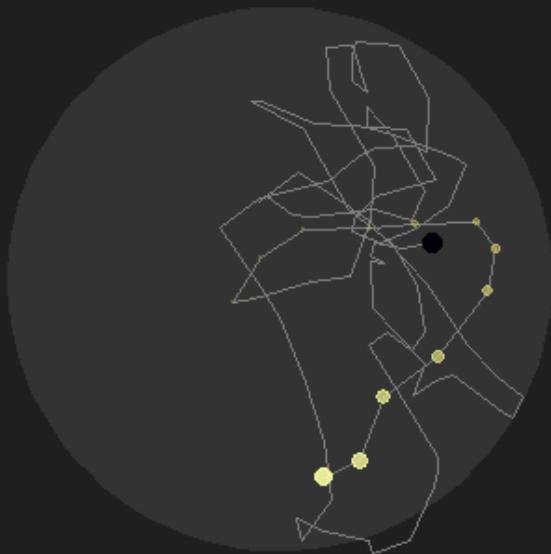
- 1** Animal movement paths are realizations of **continuous stochastic processes**,
- 2** Behavior can be summarized using **characteristic timescales**,


τ_p Position autocorrelation timescale τ_v Velocity autocorrelation timescale
- 3** These timescales impose **constraints on sampling design** that must be met for sufficiently large sample sizes.



τ_p

*Position autocorrelation
timescale*

 $\tau_p = 1 \text{ hour}$ $\tau_p = 1 \text{ day}$ $\tau_p = 5 \text{ days}$ $\tau_p = 10 \text{ days}$ 

T

SPACE - USE Home range
SAMPLING DURATION*How long* is an animal tracked for?**MOVEMENT BEHAVIOR** Speed & distance

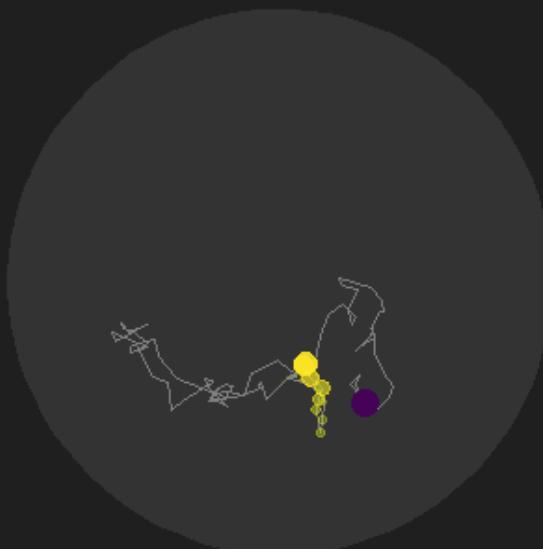
τ_v

*Velocity autocorrelation
timescale*

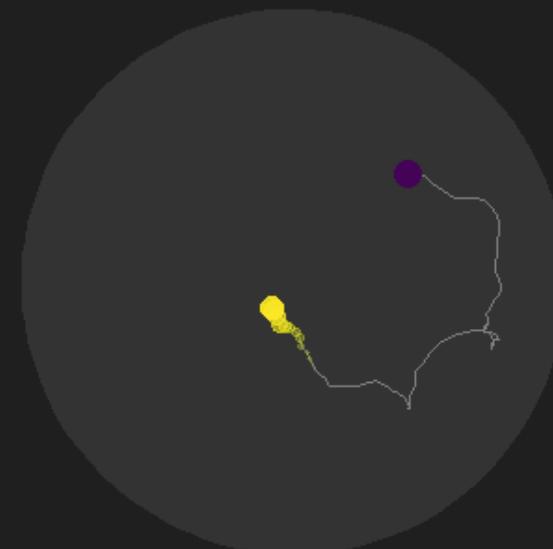
$\tau_v = 1$ minute



$\tau_v = 1$ hour



$\tau_v = 12$ hours



$\tau_v = 1$ day

 Δt

SPACE - USE

Home range

SAMPLING INTERVAL

How frequently are locations collected?

Speed & distance



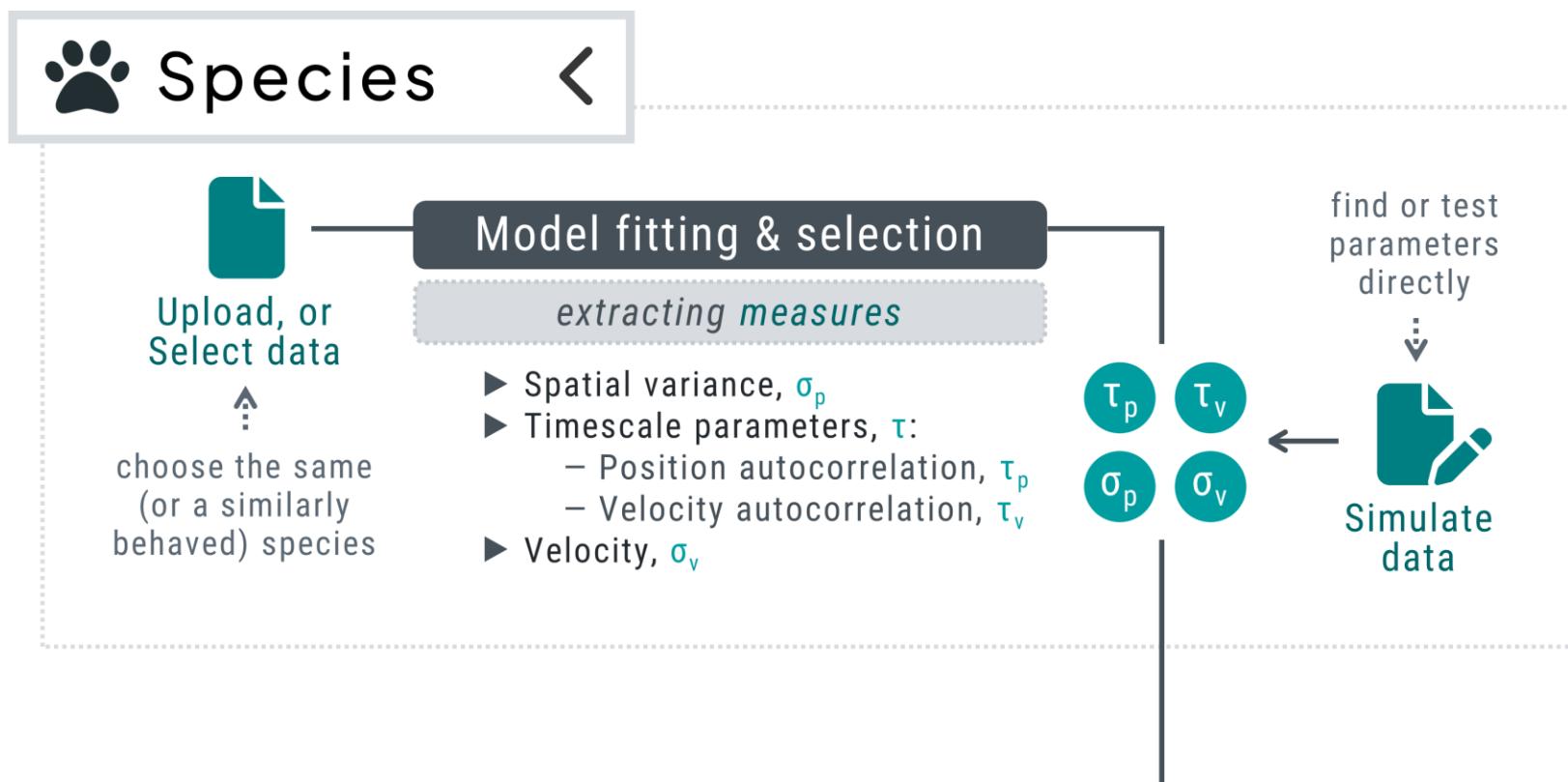
MOVEMENT BEHAVIOR

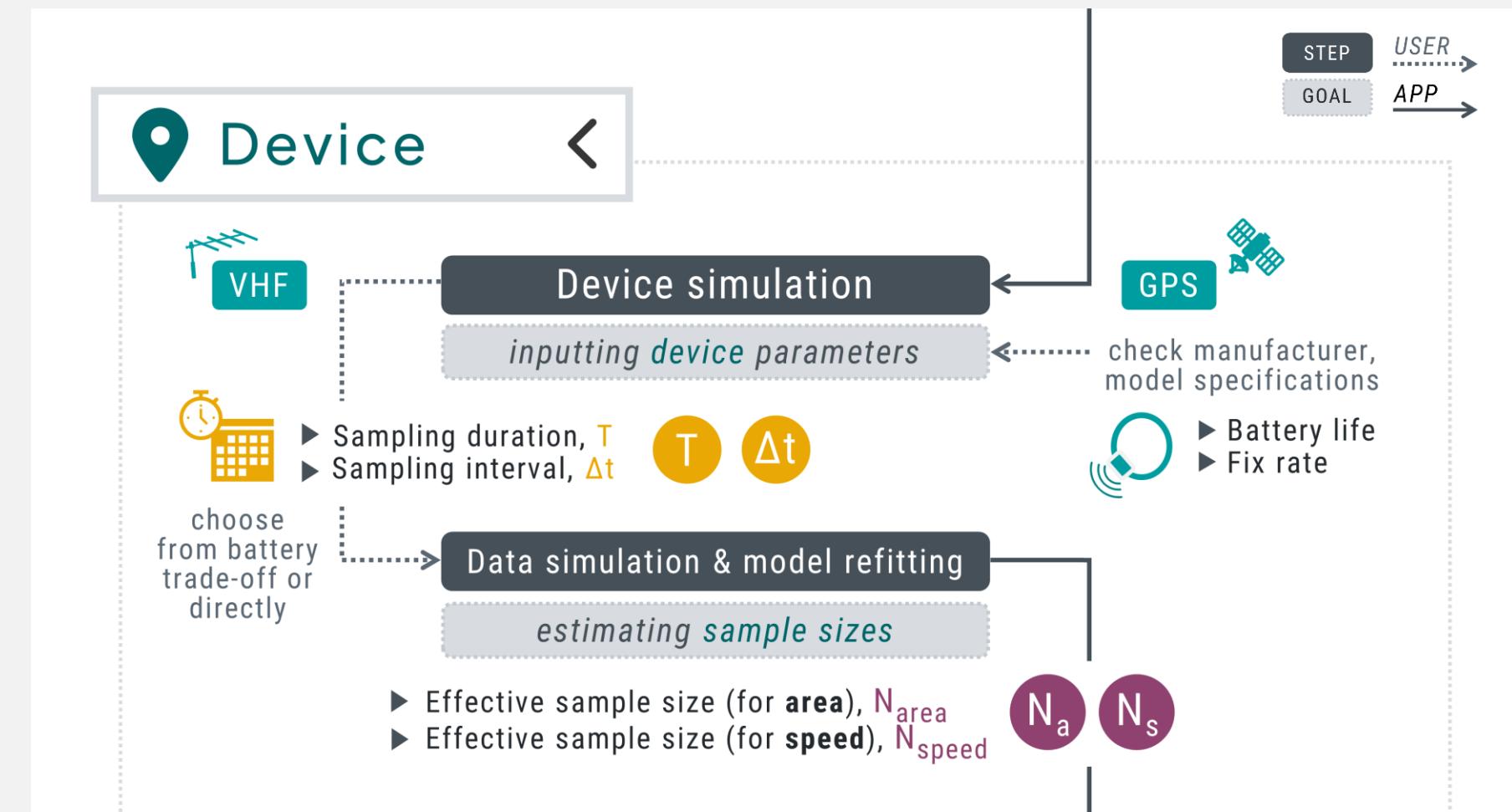


It is physically impossible for animal movement to be *uncorrelated*.

Now, this raises three key questions:

1. Can we *detect* a signature of these correlations in the data?
2. Is the sampling schedule sufficient to *produce reliable estimates* of key metrics?
3. Is the population sample size sufficient to *capture individual variation*?





Analyses <

 τ_p τ_v σ_p σ_v

relative to

 T Δt

limit

 N_a N_s

Estimation & comparison

estimate bias, confidence intervals

► **Home range estimation**

with Autocorrelated Kernel Density Estimator, **AKDE**

► **Speed & distance estimation**

with Continuous-Time Speed and Distance, **CTSD**



Report <

Extended workflow

Add *sources of bias* (if applicable):

- ▶ Location error
- ▶ Data loss (e.g., fix success rate)
- ▶ Device loss or mortality event
- ▶ Storage limits



Run simulations

*Test sampling designs,
Assess estimation accuracy and precision.*

YES



Simulate a *set number of tags* from the model fit

NO



Iteratively simulate more tags until threshold

Planning to track a specific number of animals?



Cross et al. (2013)

NT

Species
African buffalo
(*Syncerus caffer*)



Position autocorrelation (τ_p)

10.3 days
(8.2–12.8)

Velocity autocorrelation (τ_v)

32.2 minutes
(22.6–45.7)



Population-level inferences – mean home range areas

Species
African buffalo
(*Syncerus caffer*)

Sampling duration

2 months

4 months

1 year

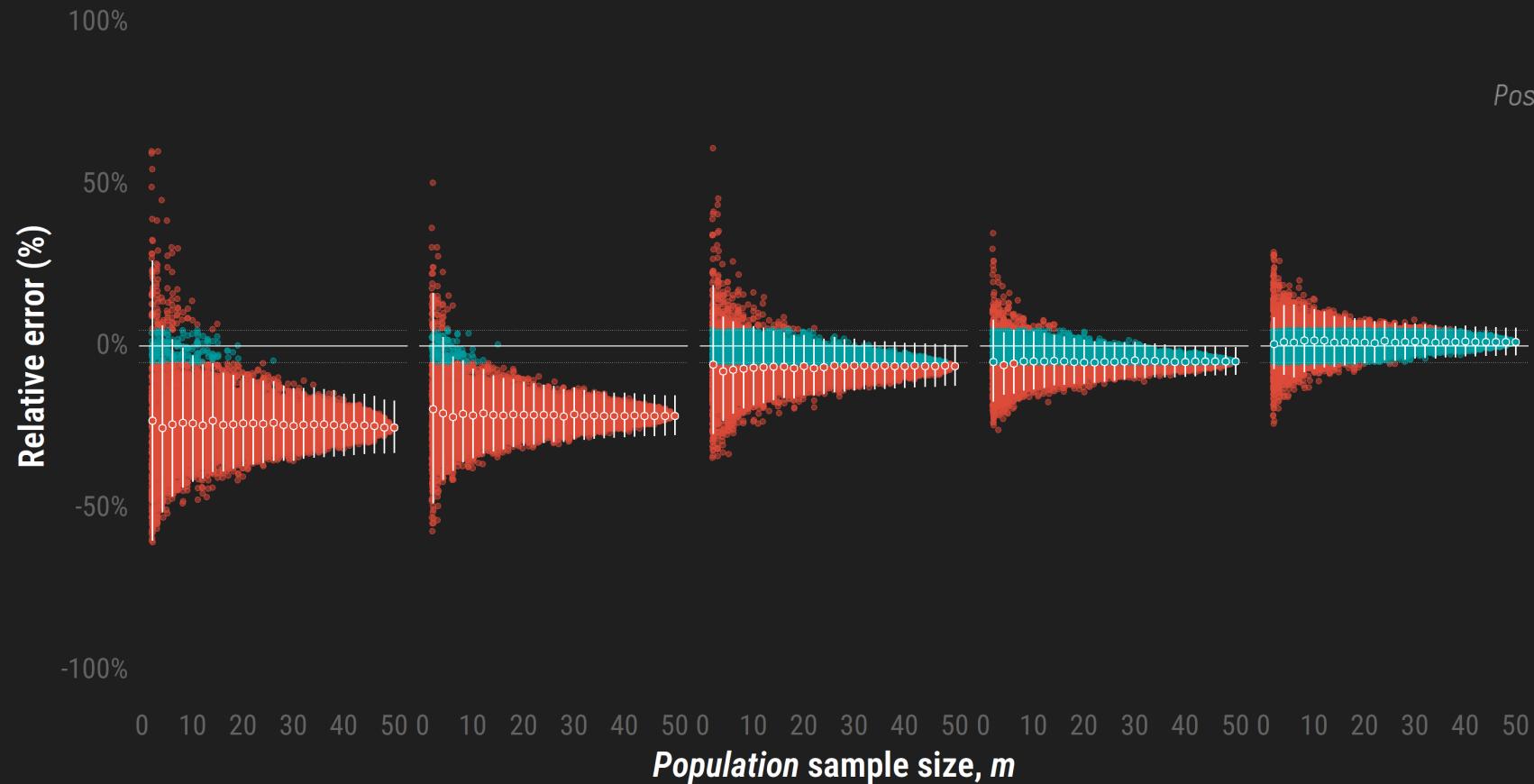
3 years

9 years



Error threshold
5%

Position autocorrelation
10.3 days
(8.2–12.8)





What if we want to obtain a reliable estimate (within a 5% error threshold) of both **mean home range area** and **mean movement speed**?



Within error threshold ($\pm 5\%$)? ● TRUE ● FALSE



Practical uses



1. *Justify resource allocation*

Support funding proposals with quantitative rationale,



2. *Refine study design*

Use pilot data to optimize methods before full data collection,



3. *Monitor reliability during data collection*

Iteratively assess if results are robust as more data becomes available,



4. *Evaluate research outcomes*

Assess any findings a posteriori and inform future research.

The latest release ([v0.3.2](#)) of the `movedesign` R Shiny application
is available on [CRAN](#) or [GitHub](#).

Contact

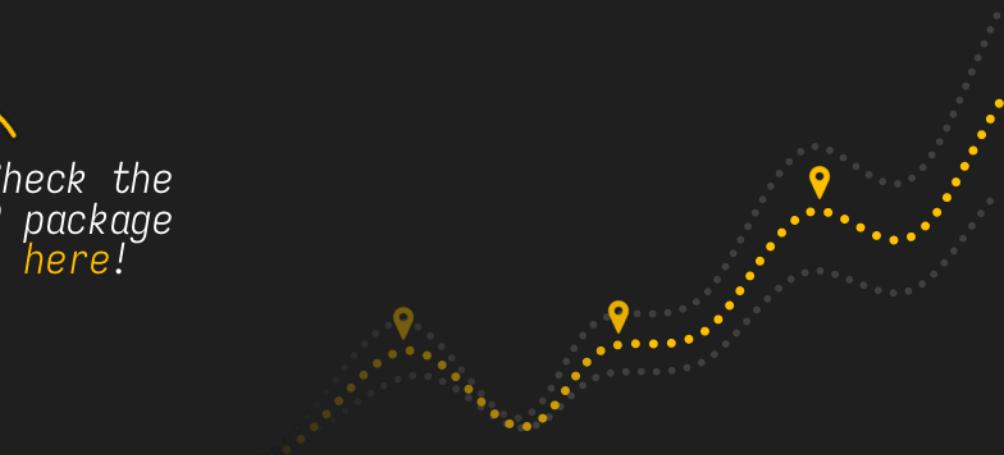
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Thank you! Any questions?



Check the
R package
[here!](#)



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