

MARE-Madeira 2025



Habitat-informed home ranges

Using the '`ctmm`' R package

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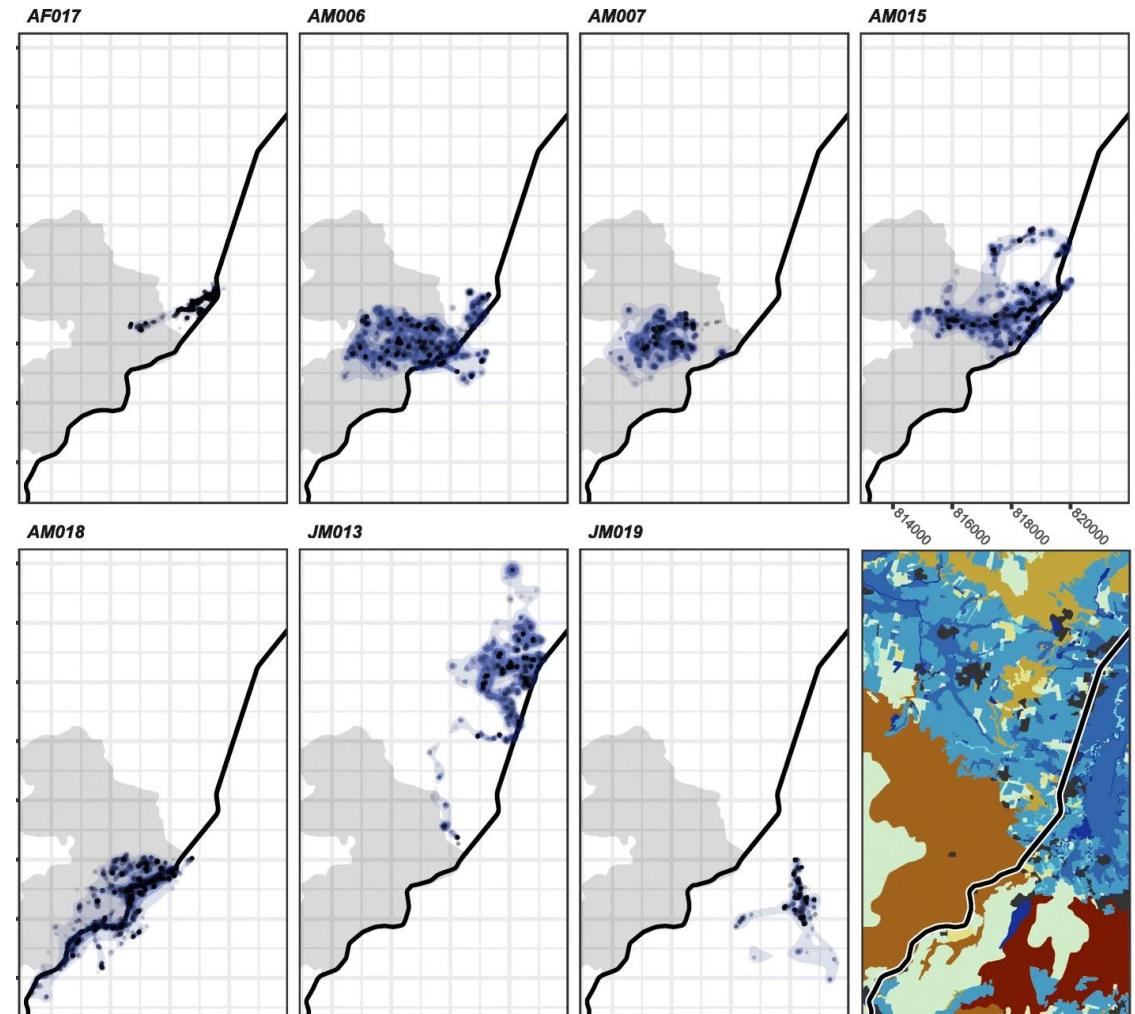


Introduction



“What **landscape features** do animals seek or avoid?”

“In which areas of the landscape are animals at **risk**? ”



Marshall et al. (2020)

DOI: 10.1186/s40462-020-00219-5



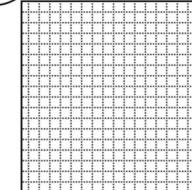
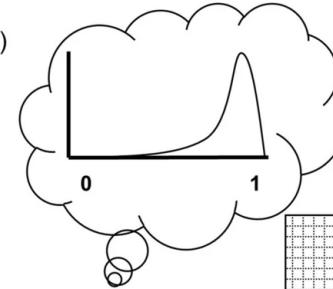
Introduction



I. preference (intrinsic)

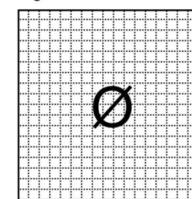
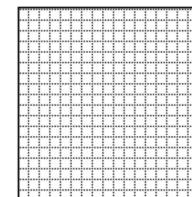


Preference (*intrinsic*)



Selection (extrinsic)
or avoidance

attempt to
occupy

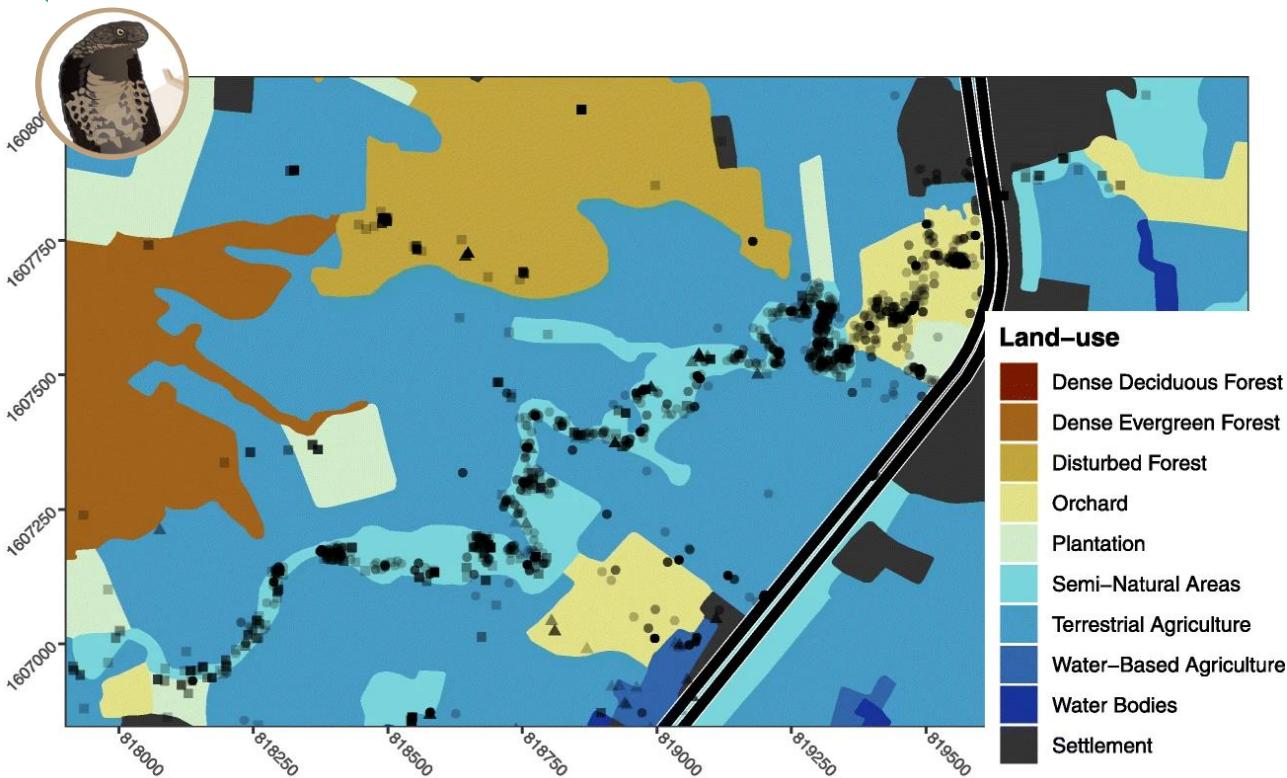


Use (*outcome*)





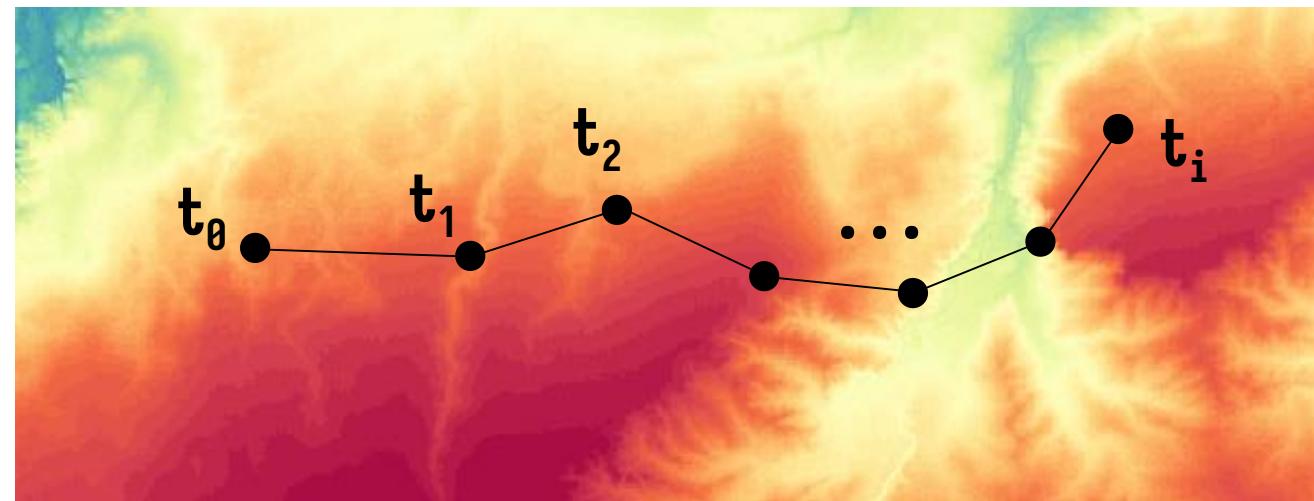
Reliance on **semi-natural areas**,
compared to the surrounding **matrix of agricultural fields**.



Marshall et al. (2020)
DOI: 10.1186/s40462-020-00219-5

≈ acting as movement corridors through the fragmented landscape?

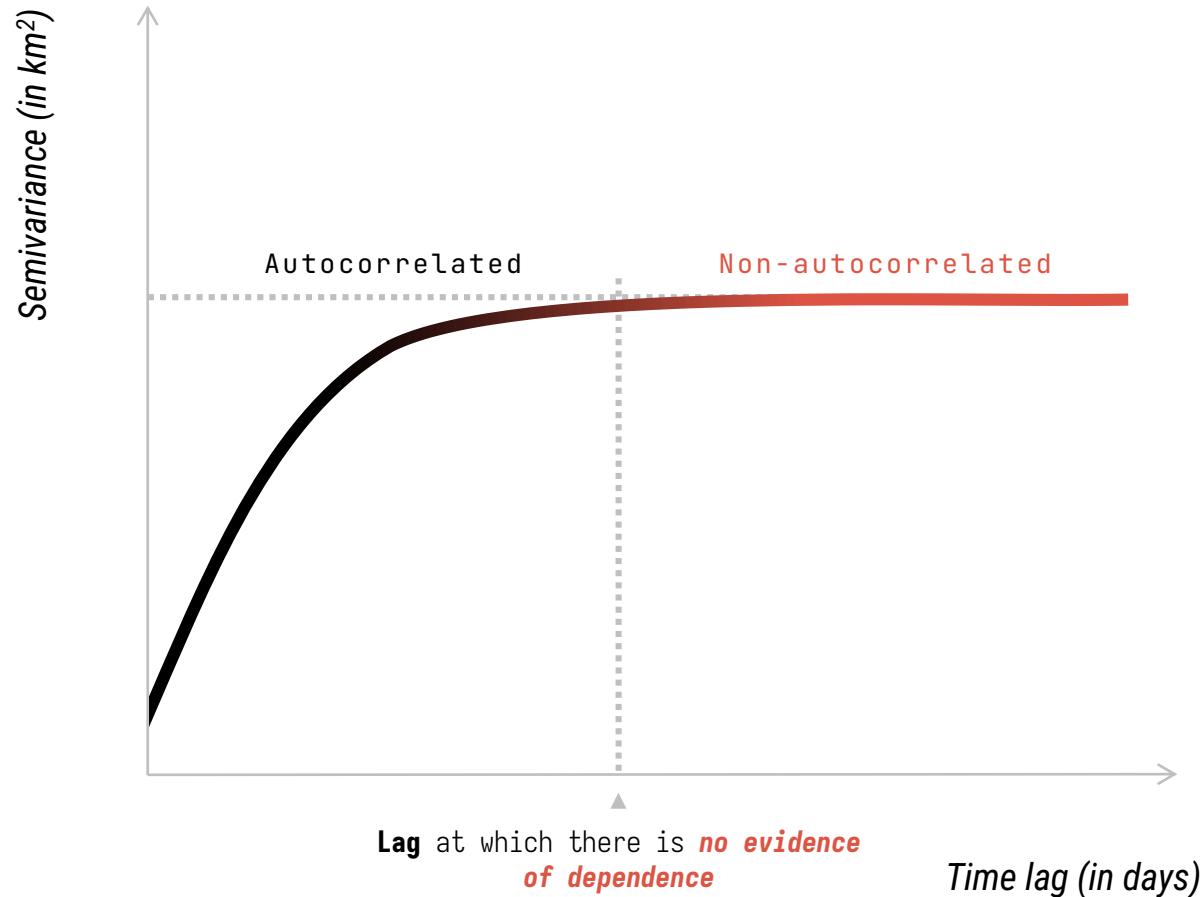
Assumes that successive positions are **independent**. However, ...



An animal's location at time t_i is a function of both resource selection and its location at time t_{i-1} .

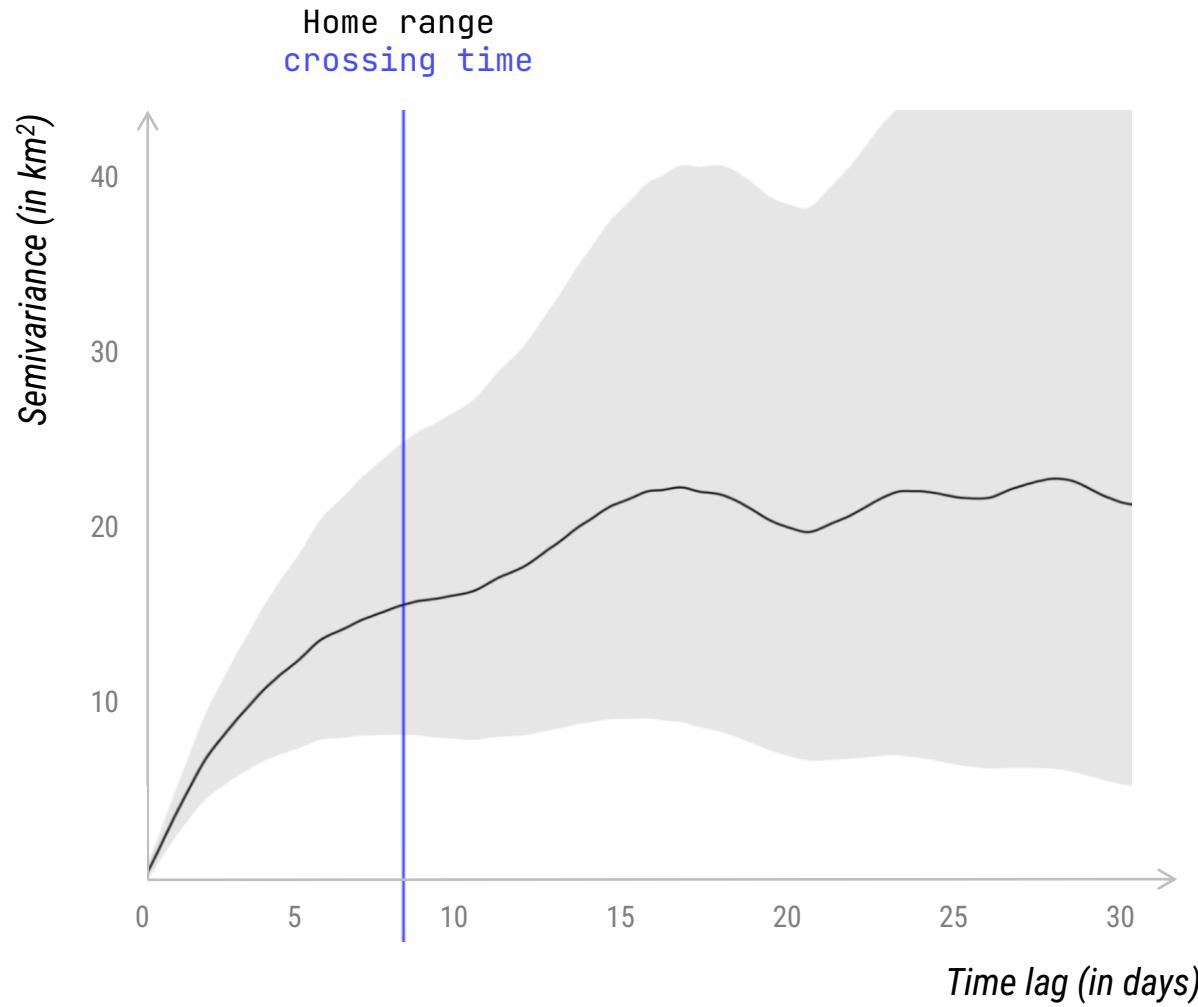


Resource selection functions





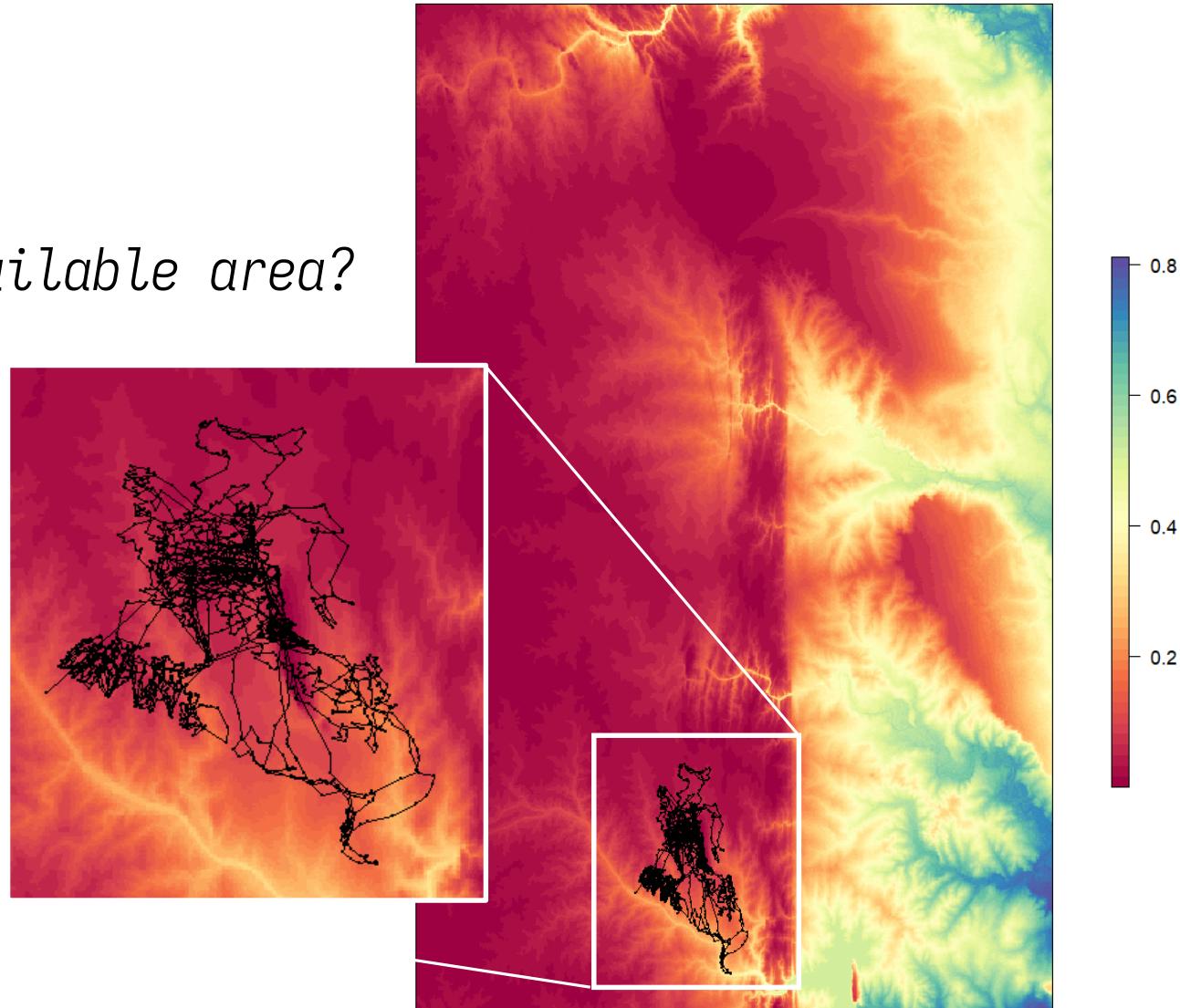
Resource selection functions



Lag > **blue**, we are fulfilling the assumptions of RSF.

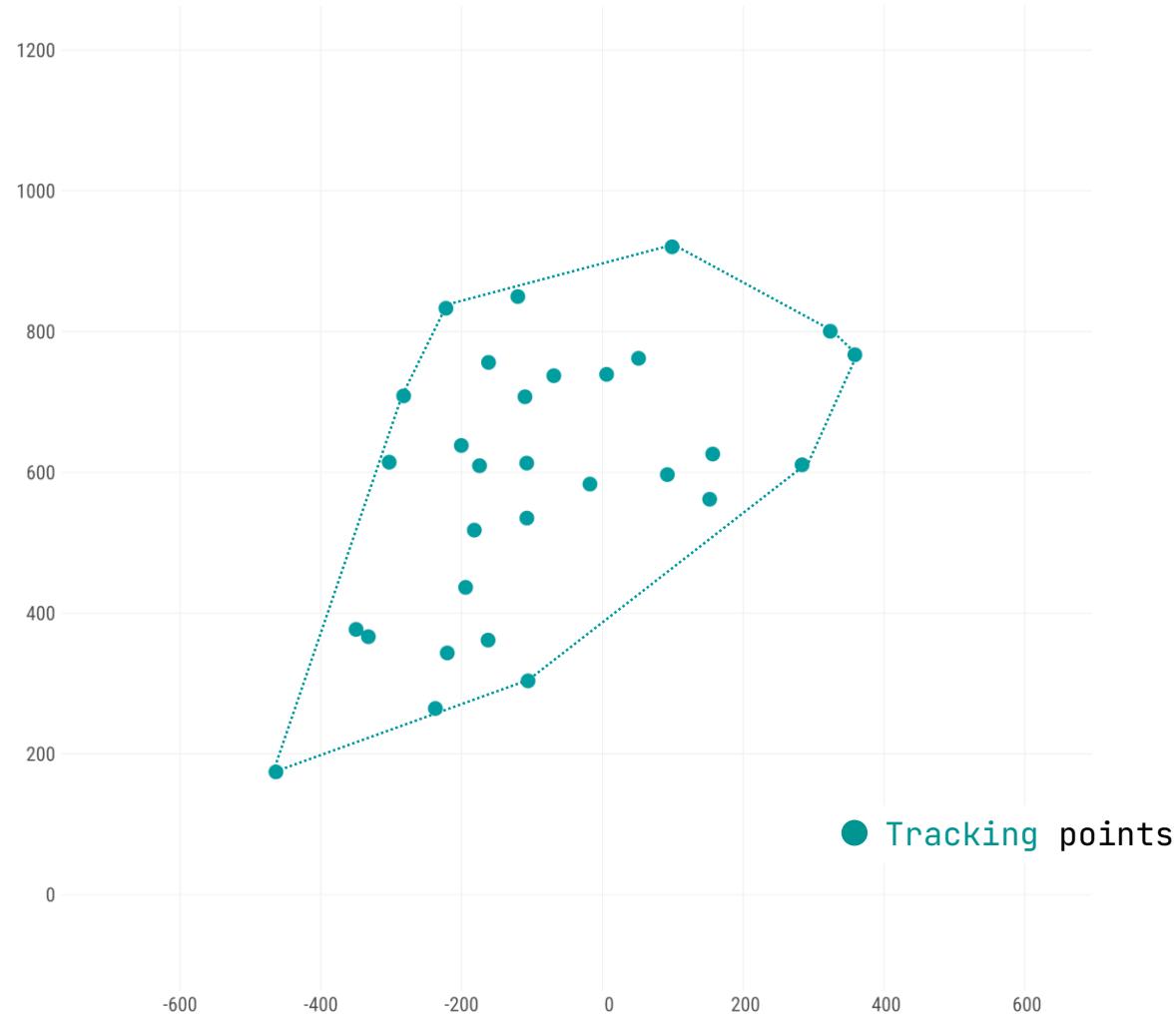


What is the available area?



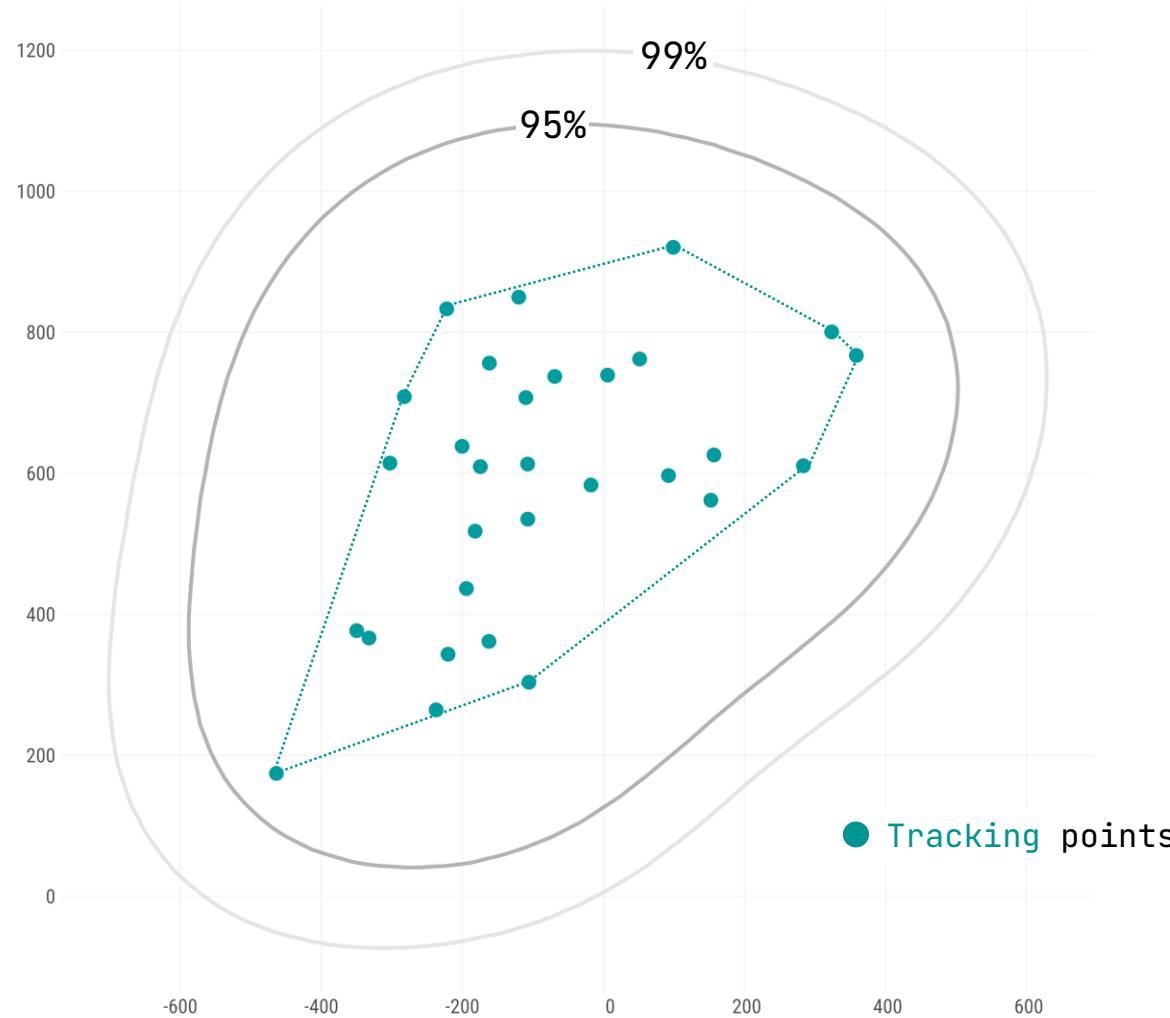


Resource selection functions



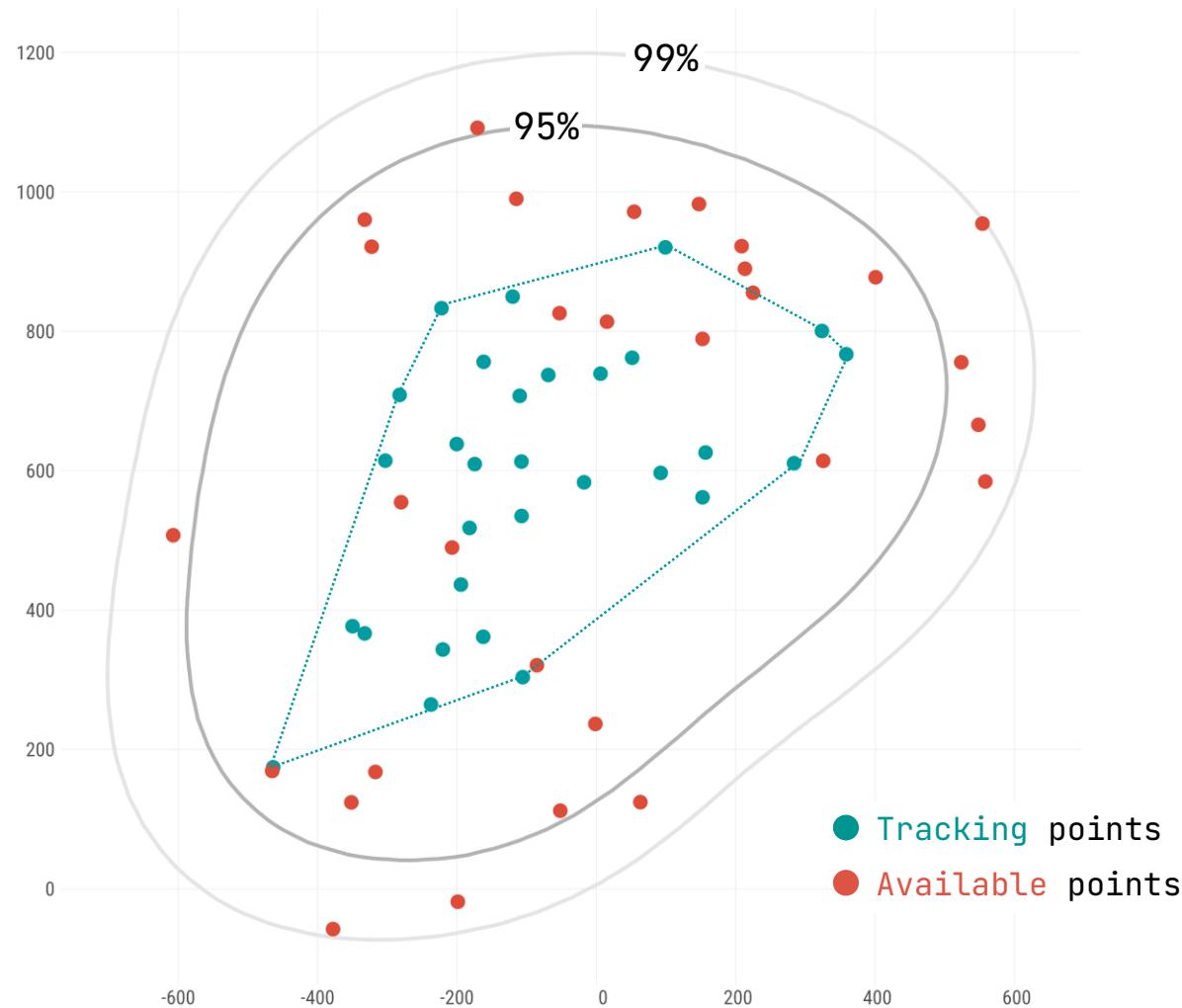


Resource selection functions





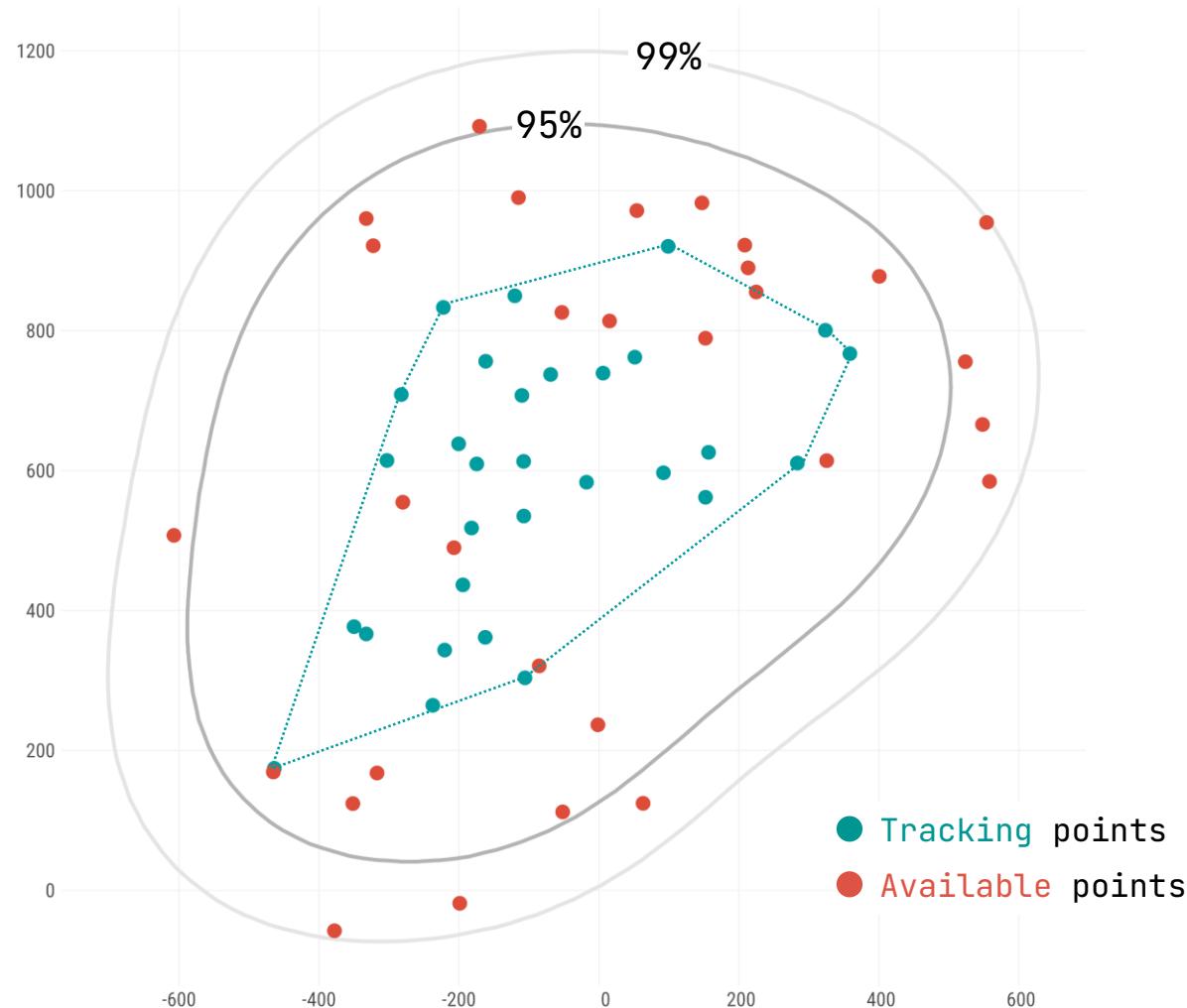
Resource selection functions



Framed as comparing data to
random available points.



Resource selection functions



Framed as comparing data to *random available* points.

However,

Available locations are *not* directly estimated, and depend on arbitrary choices (e.g., MCP, 99% home range area).

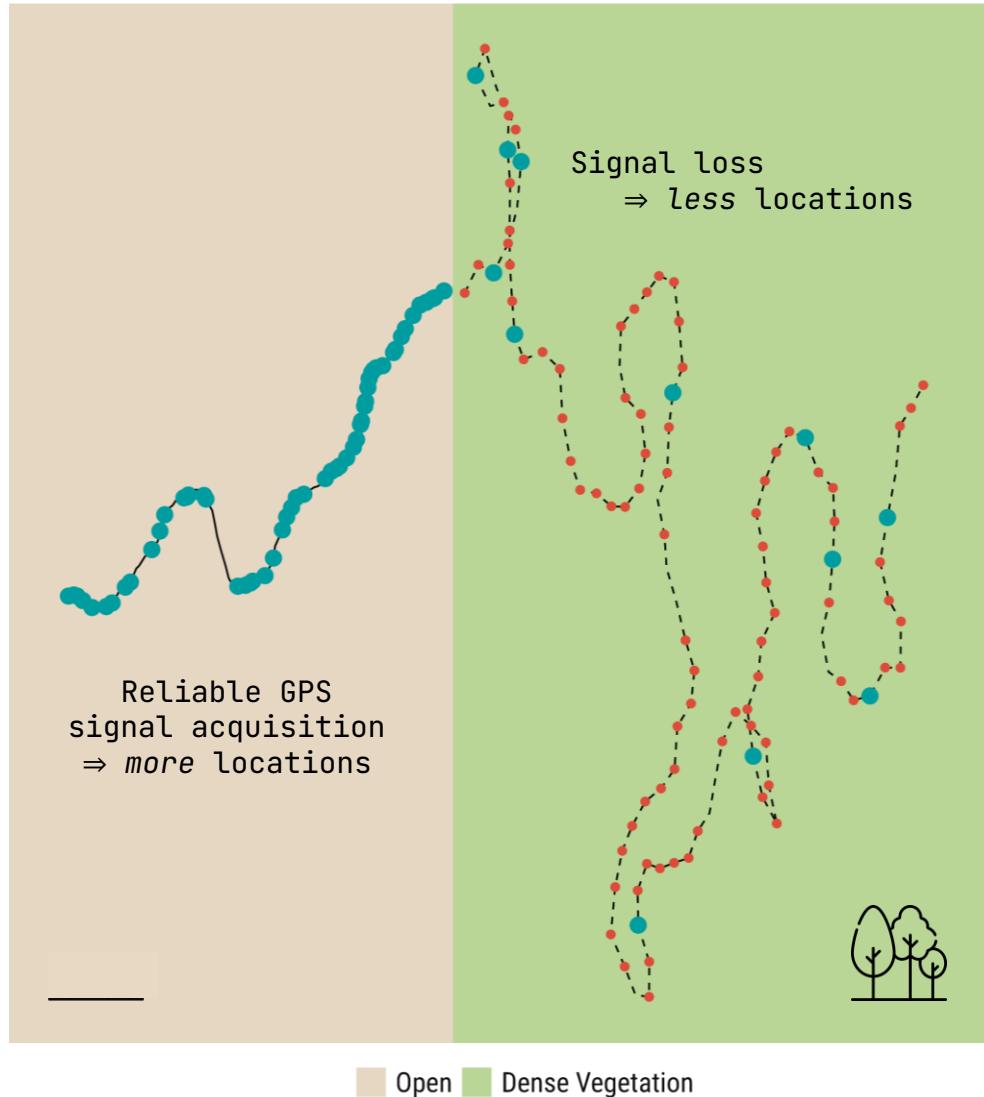
RSF results can be *sensitive* to how availability is defined.

Uncertainty in the available area is not propagated into the RSF models.





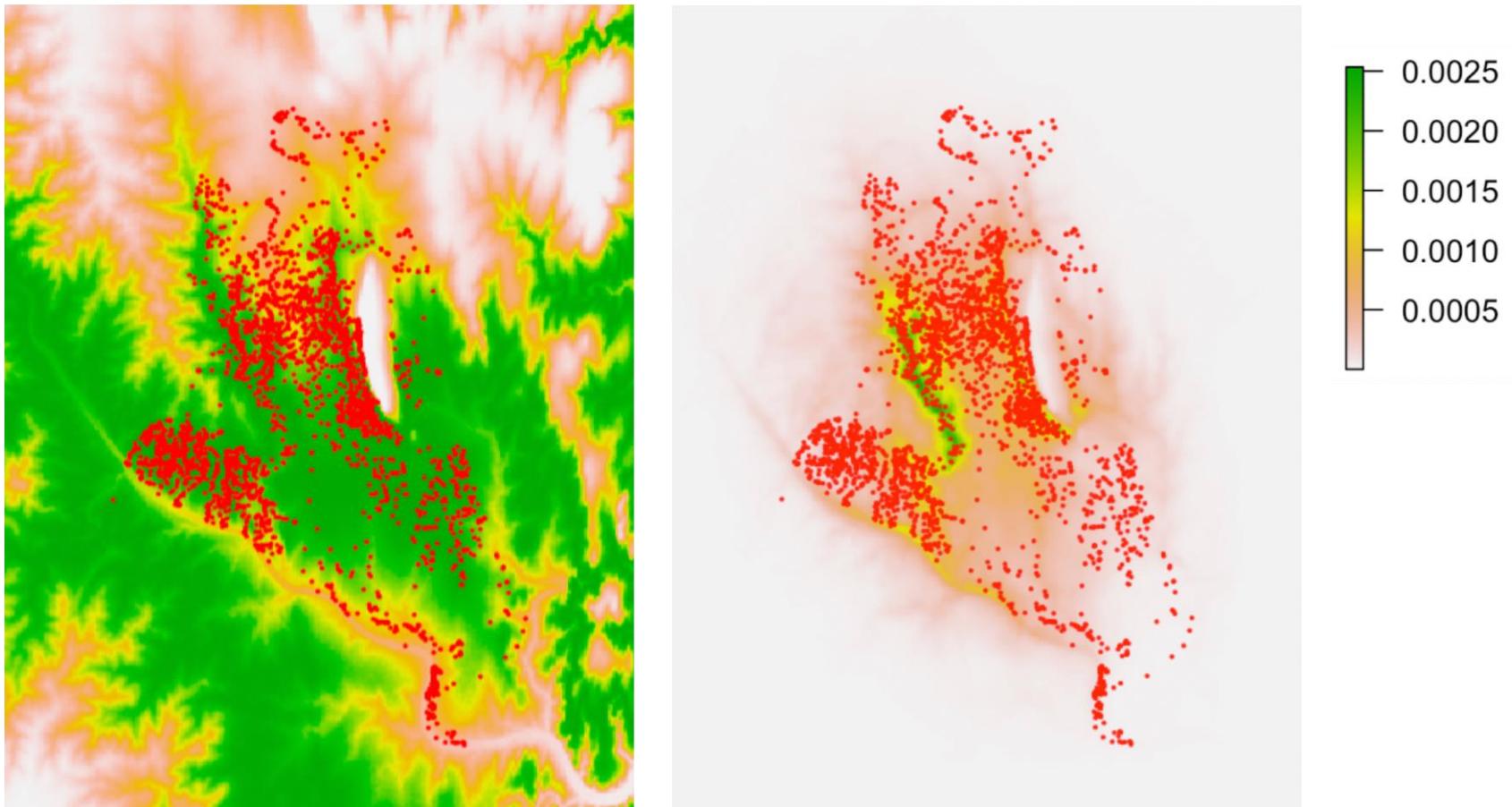
Resource selection functions



Failing to account for autocorrelation is more problematic if environmental covariates (e.g., **dense vegetation**) cause non-random data gaps.

What is the available area?

We estimate the available area in a **Poisson Point Process model**.



Conventional RSFs

Statistical issues:

1. No checks for numerical convergence
inconsistent parameter estimates
2. Assume IID data
spurious “significant” results
3. Assume uniformly weighted data
*can result in spurious results for,
e.g., habitat-dependent triangulation failure*
4. Assign a uniform distribution to available points
disregards movement process

Integrated RSFs

Solutions:

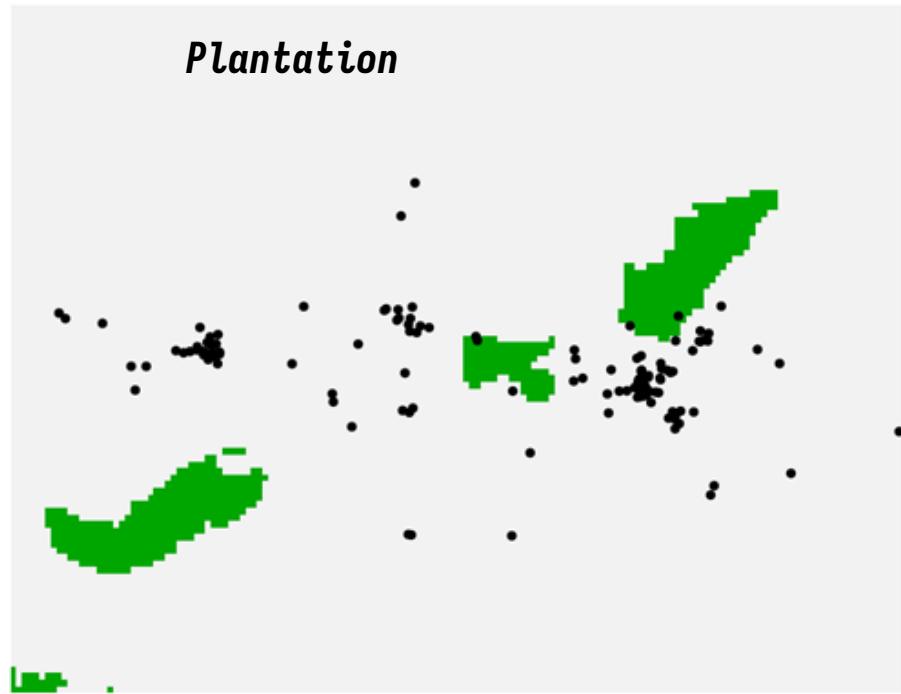
1. Automatically checks for numerical convergence
2. Down-weights data to reduce pseudoreplication
3. Re-weights data to mitigate differential autocorrelation
4. Samples availability according to null movement process



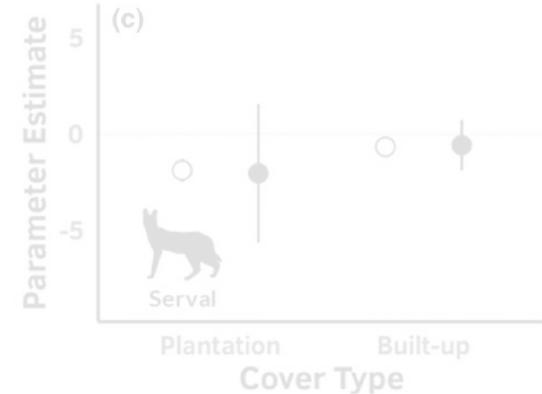
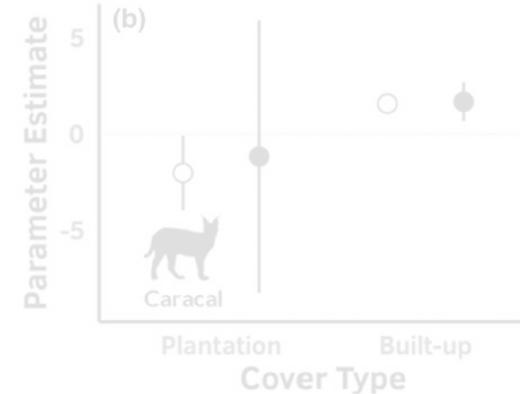
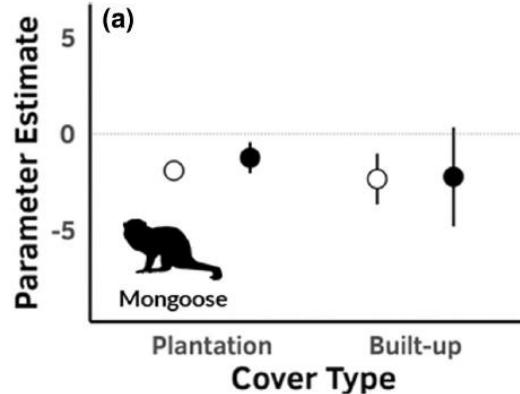
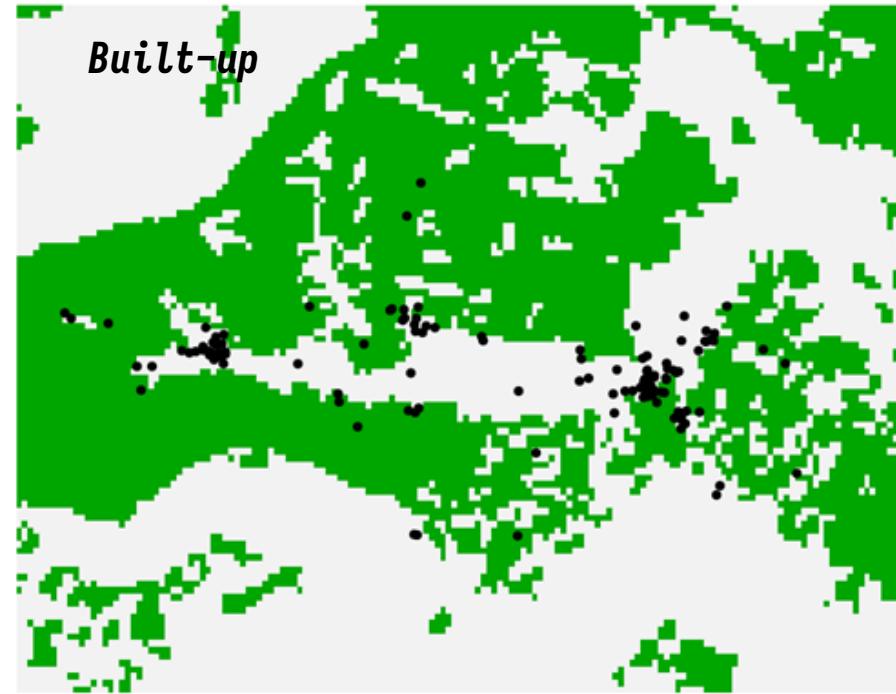
144 locations over 58 days



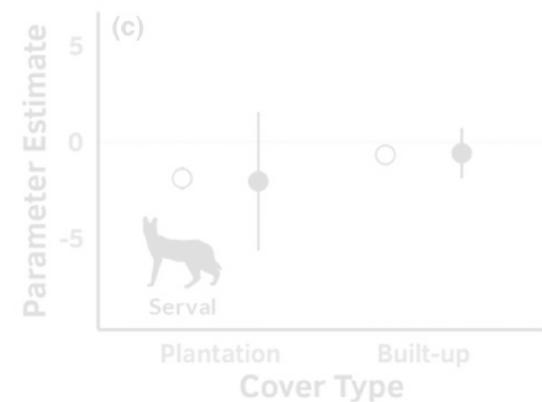
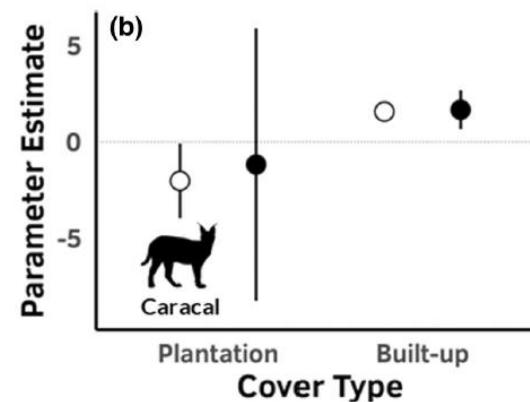
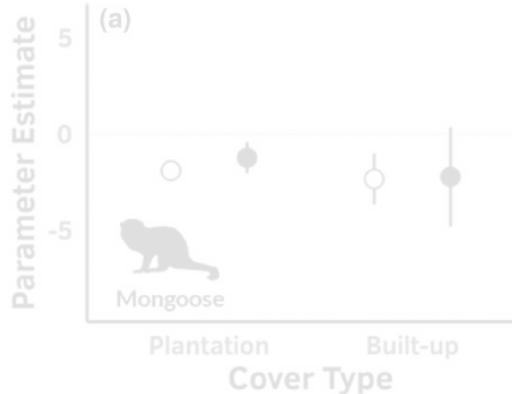
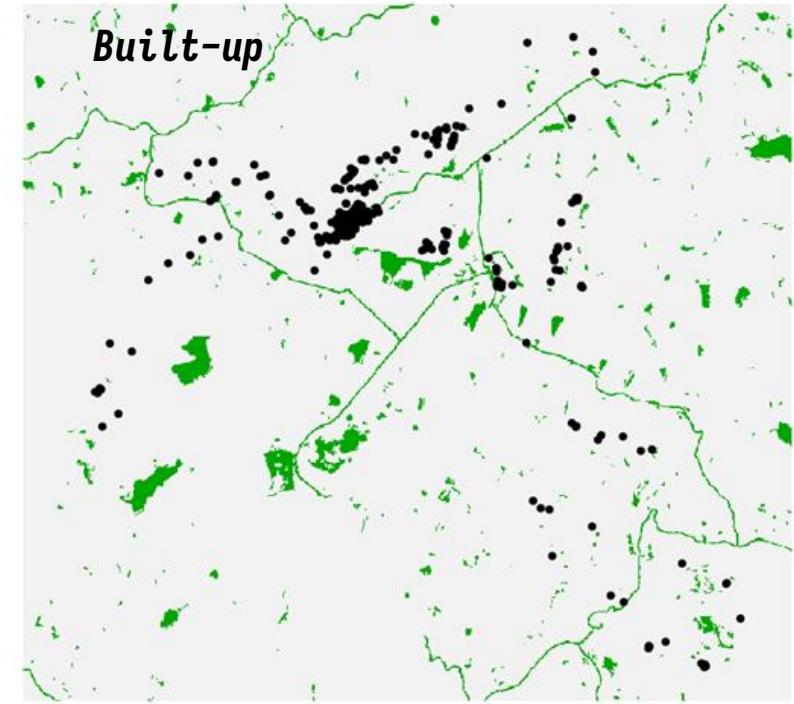
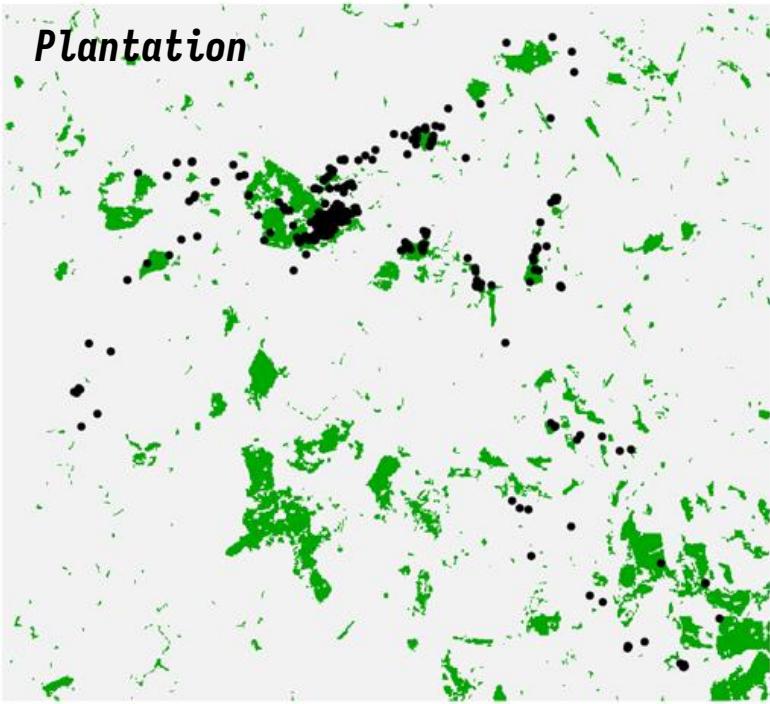
Plantation



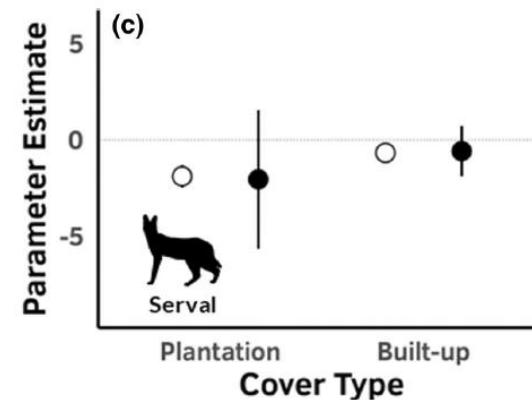
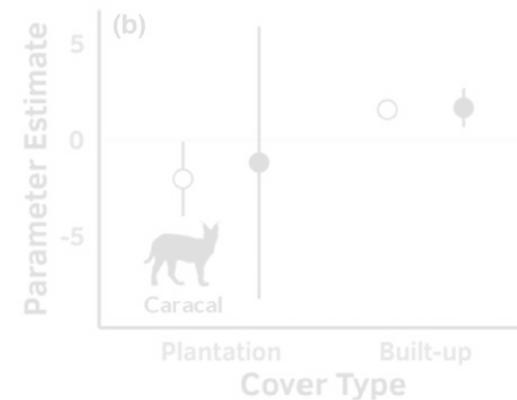
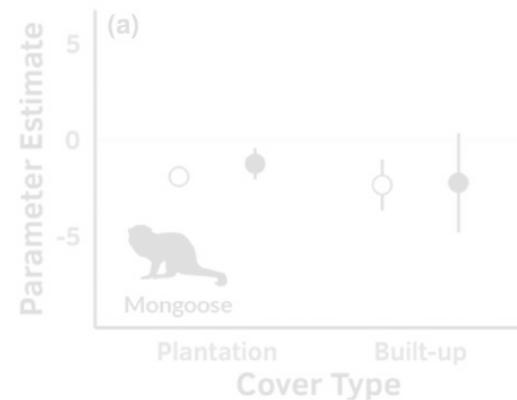
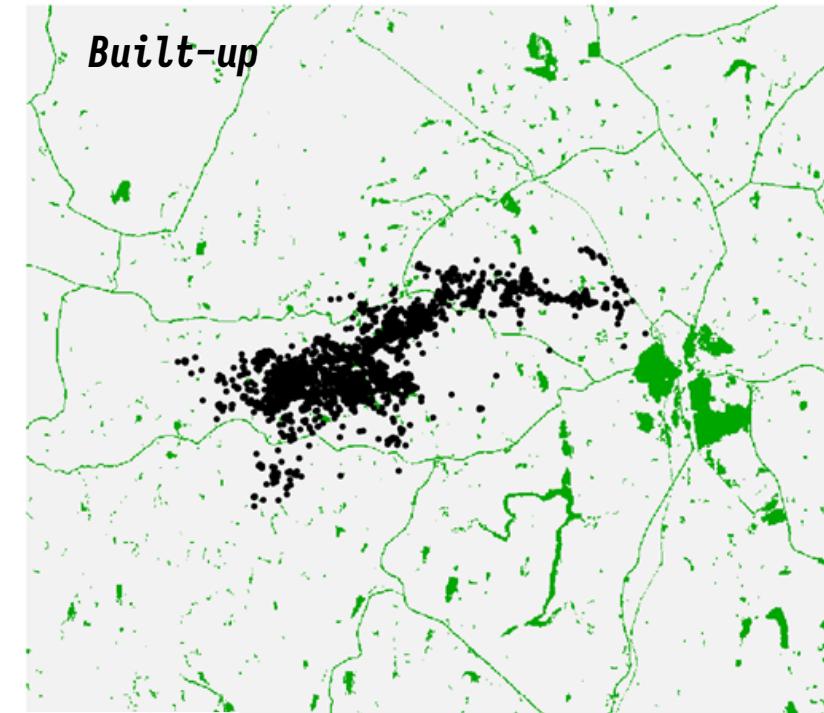
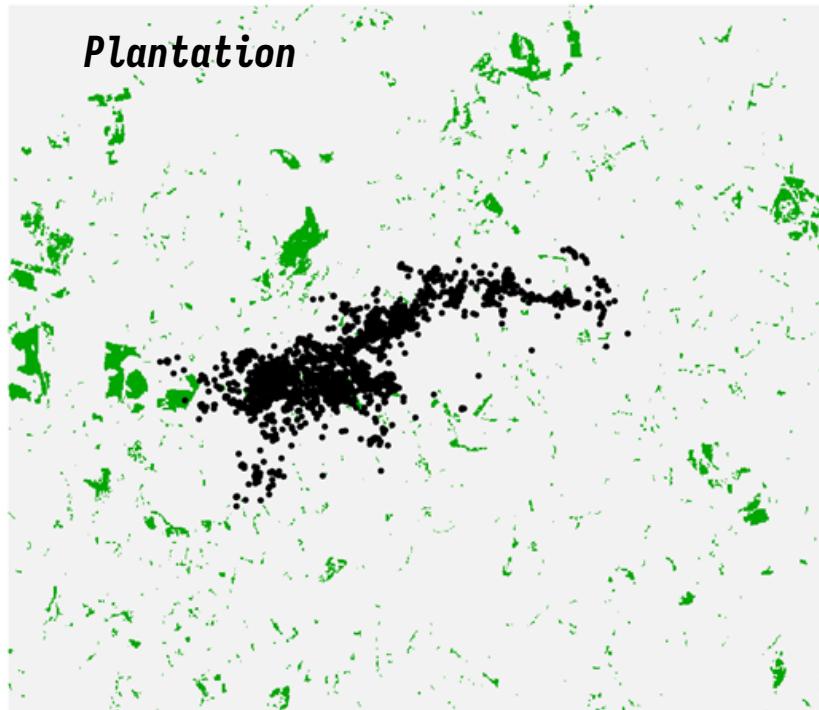
Built-up



504 locations over 85 days



3603 locations over 321 days





Integrated RSF

Implemented in ctmm via `rsf.fit()` and `rsf.select()`,

Advantages:

- Log-likelihood is downweighted to account for *autocorrelation* and *irregular sampling*,
- Available points are randomly sampled until numerical convergence,
- Available area is estimated (along with the parameters) so *uncertainty* is propagated.



Alston et al. (2023)

DOI: 10.1111/2041-210X.14025

Warnings:

- Will take **longer** to run than conventional RSF,
- Requires **range-resident animals** (no migration or dispersal).



KDE

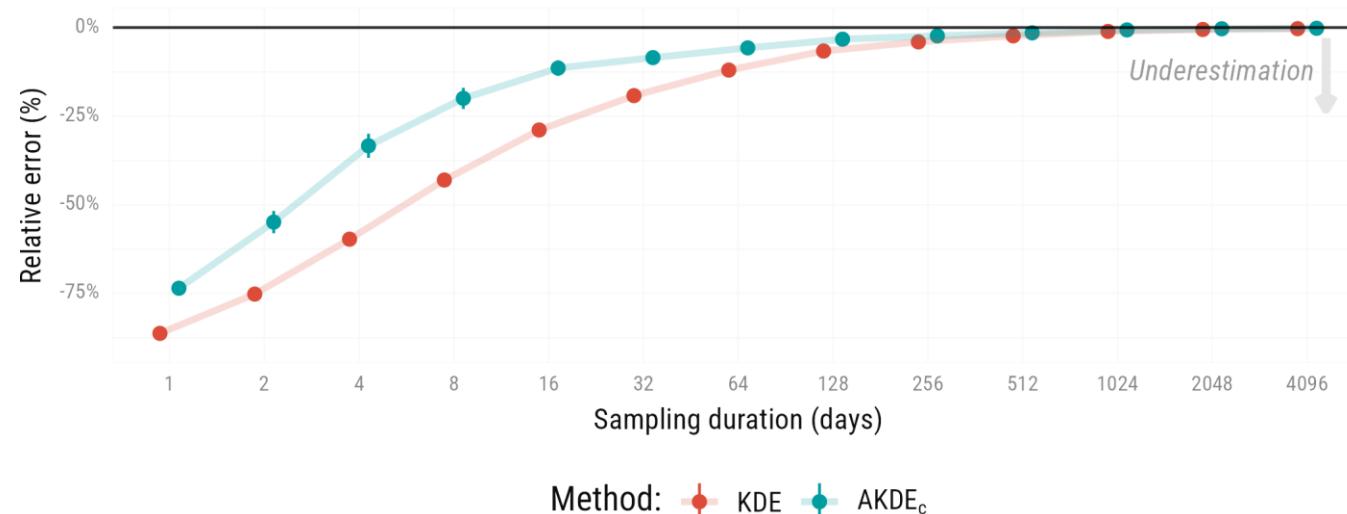
For IID data

- Non-parametric
- Optimal and asymptotically consistent
- Sensitive to autocorrelation (→ biased area)

AKDE

For autocorrelated data

- Non-parametric
- Optimal and asymptotically consistent
- Accounts for autocorrelation





KDE

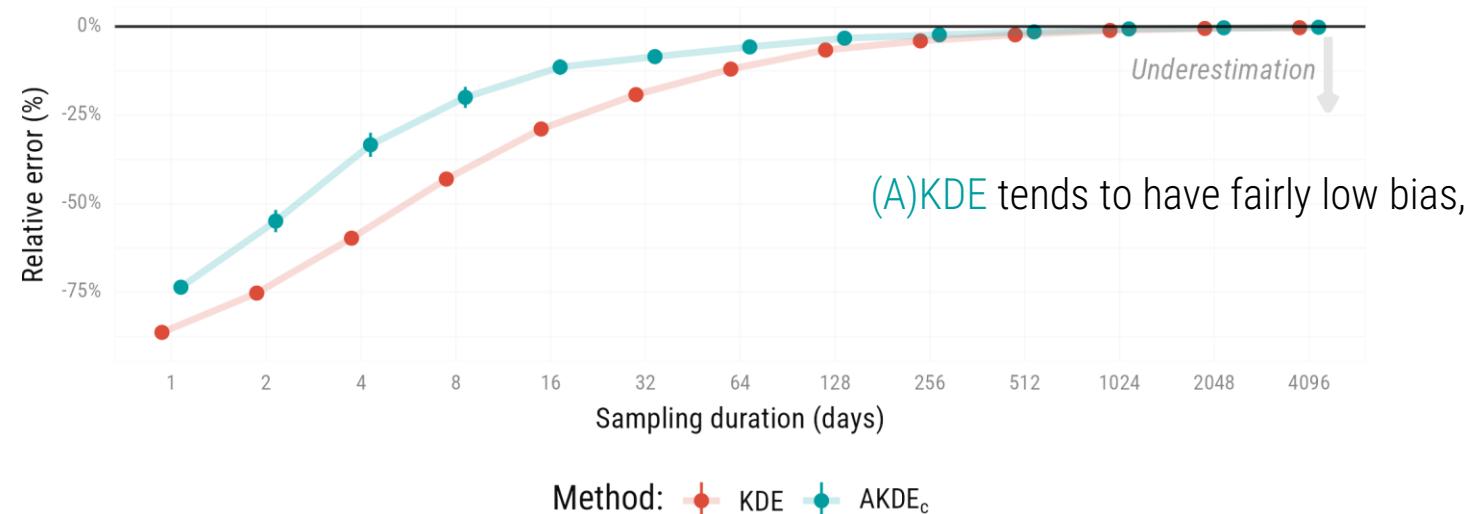
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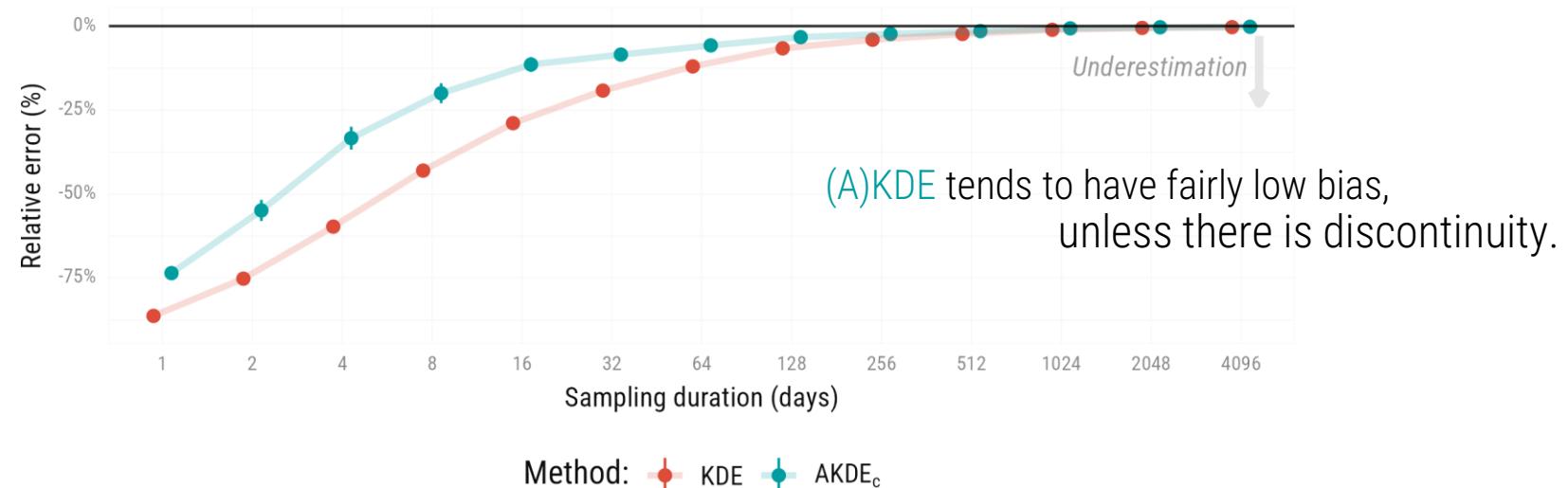
For IID data

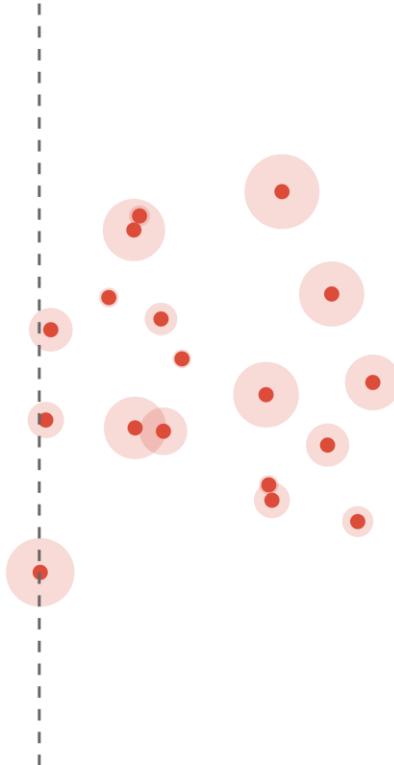
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AKDE

For autocorrelated data

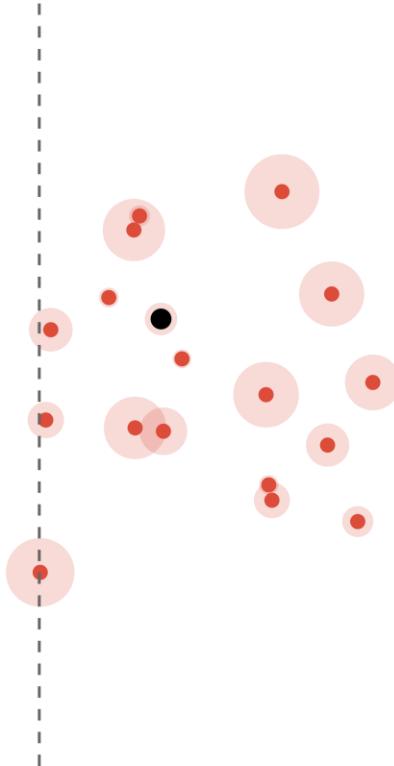
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- Optimal and asymptotically consistent
- Accounts for autocorrelation





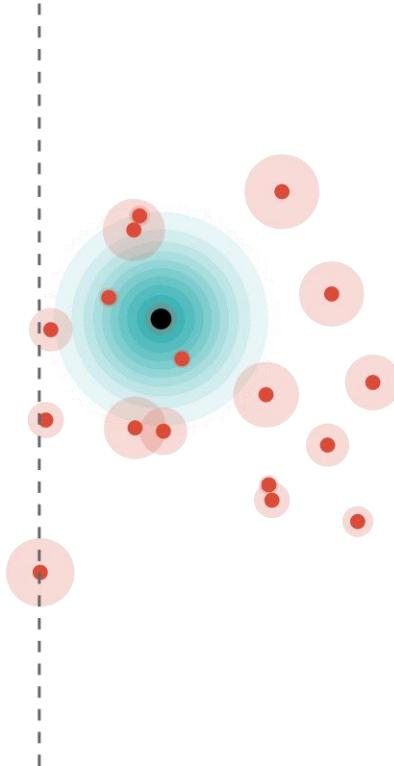
“Spillover bias”

in which home range estimates do not respect impassable movement boundaries (e.g., shorelines, fences), and occurs in all forms of kernel density estimation.



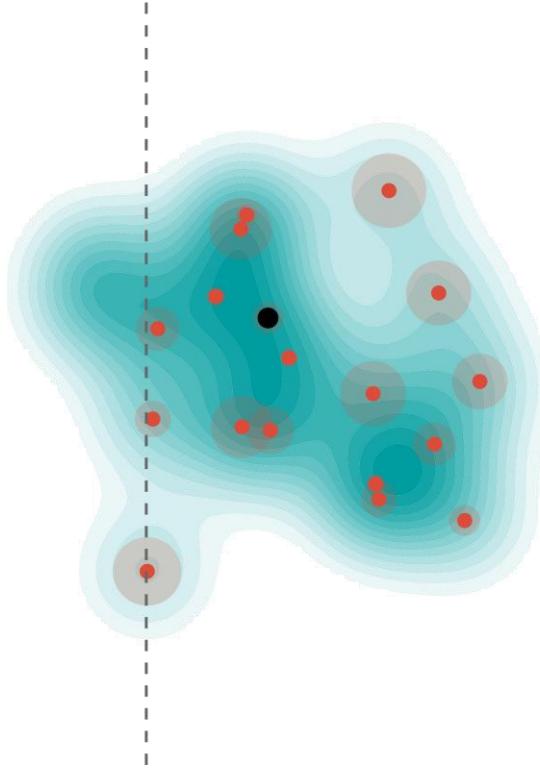
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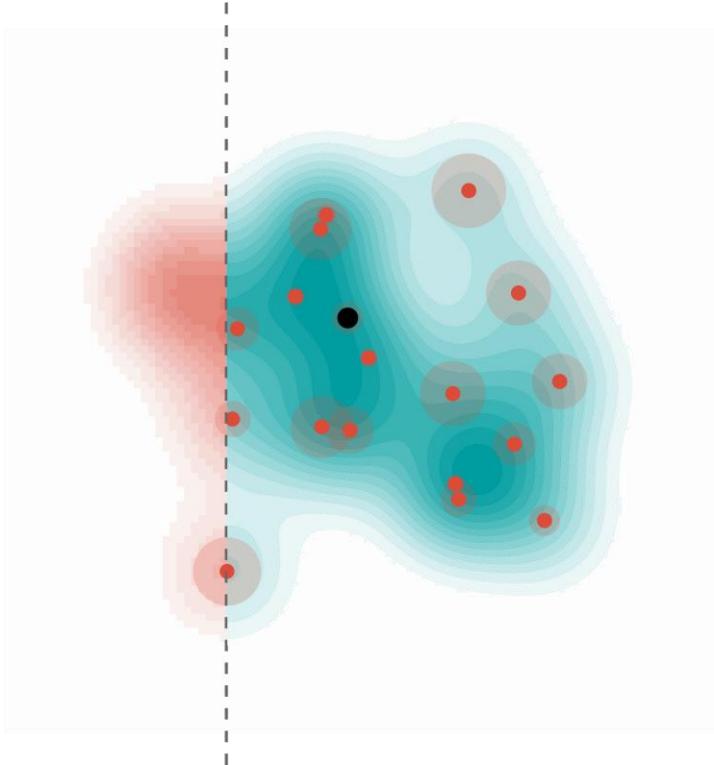
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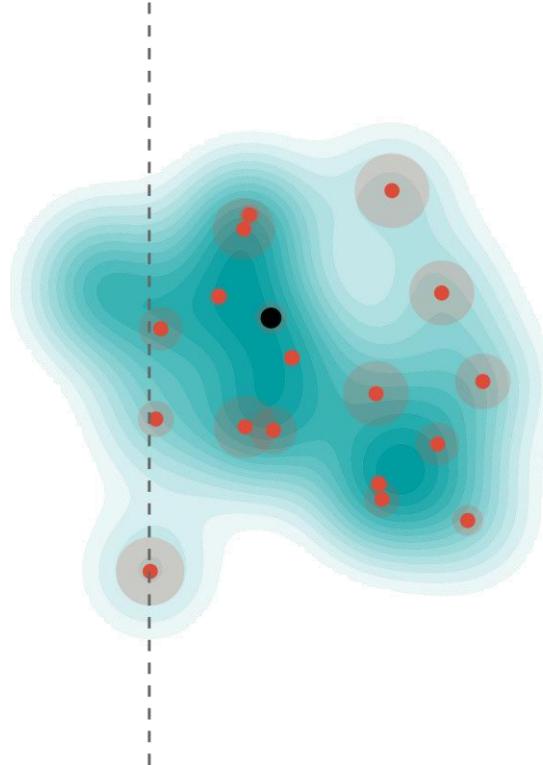
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“Spillover bias”

in which home range estimates do not respect impassable movement boundaries (e.g., shorelines, fences), and occurs in all forms of kernel density estimation.

i.e., assigning non-zero probability density to space the animal cannot reach.



“Discontinuity bias”

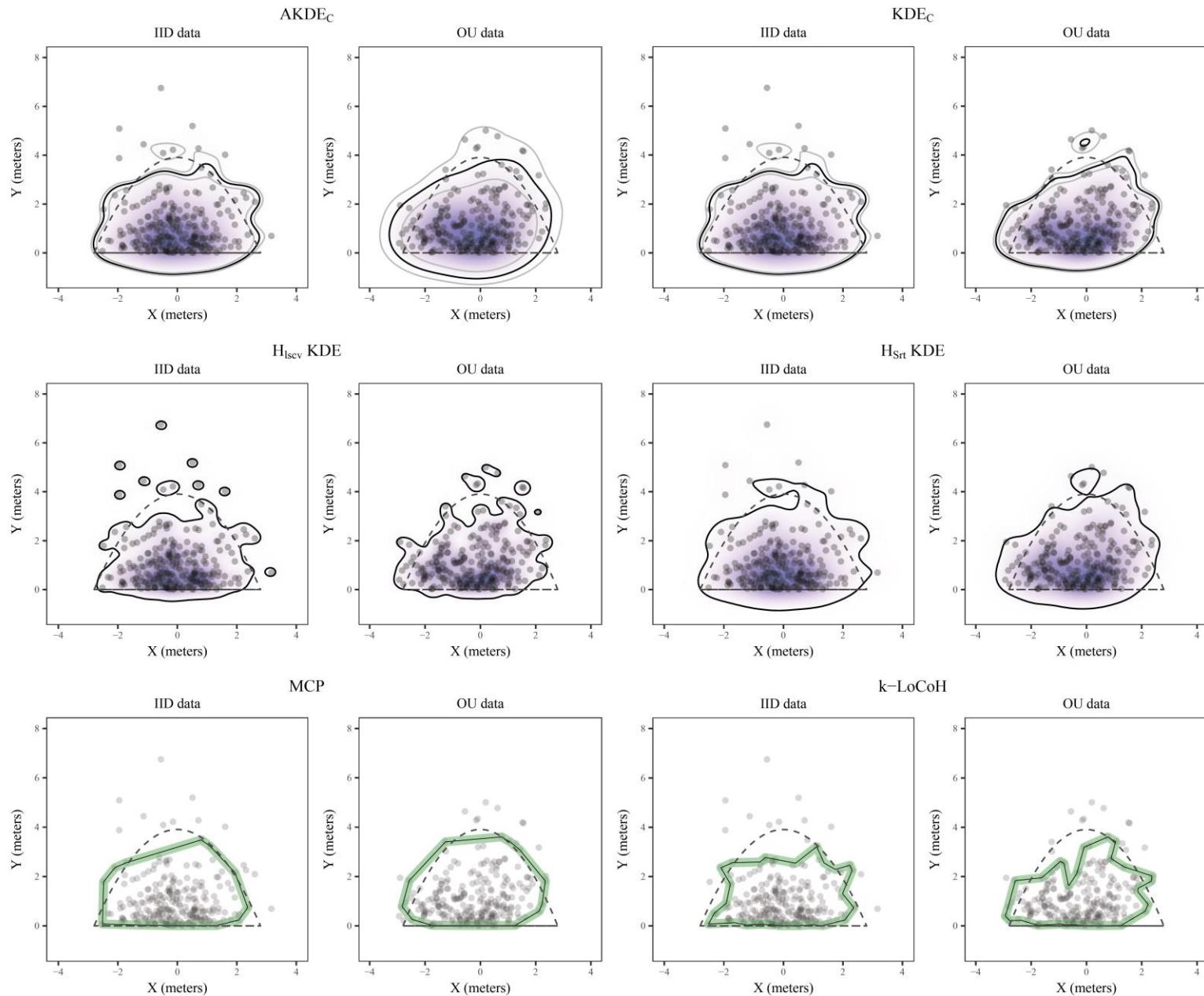
when animals show strong preferences for specific habitat types, constituent kernels may encompass areas which **are available, but unused**, by the tracked animal.



This bias will be exacerbated with shorter tracking durations, or when movement is *tightly constrained* by environmental characteristics.



Method comparison

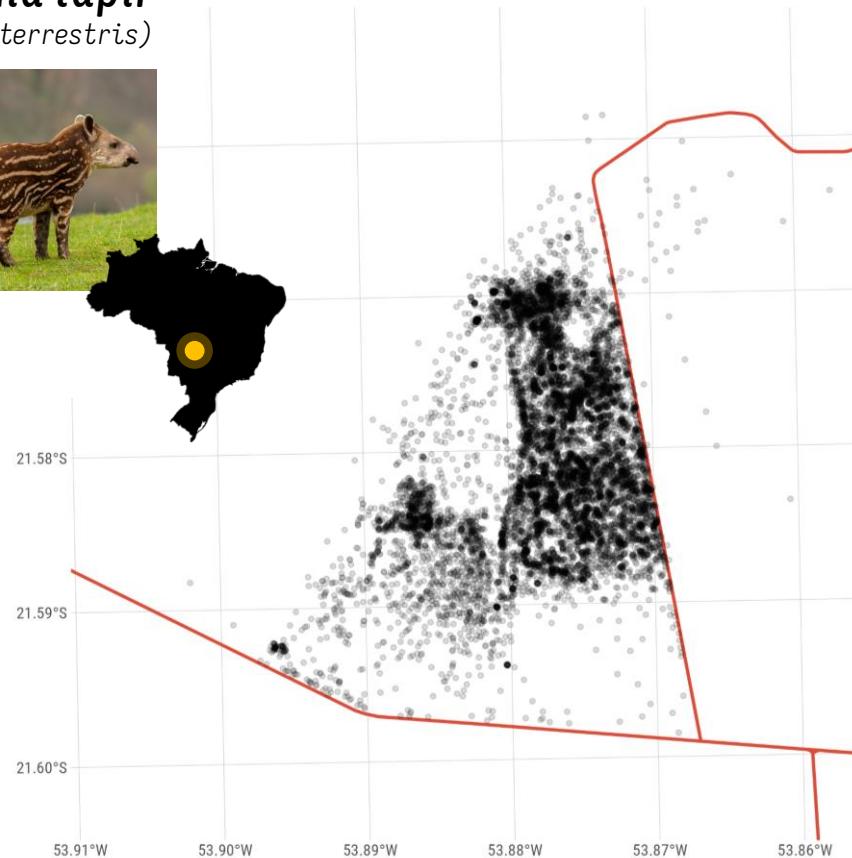


Lowland tapir

(*Tapirus terrestris*)



VU

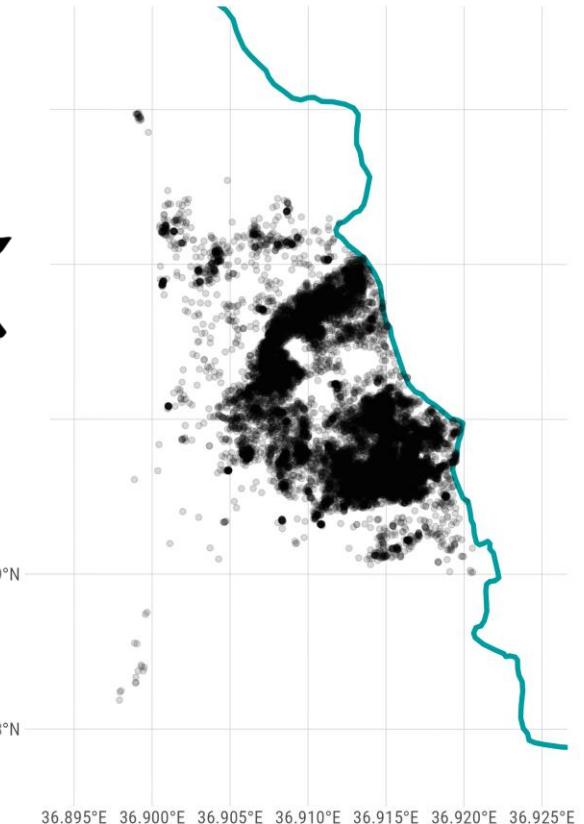
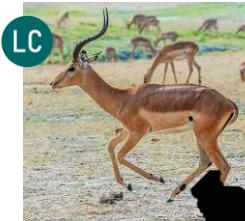


Impala

(*Aepyceros melampus*)



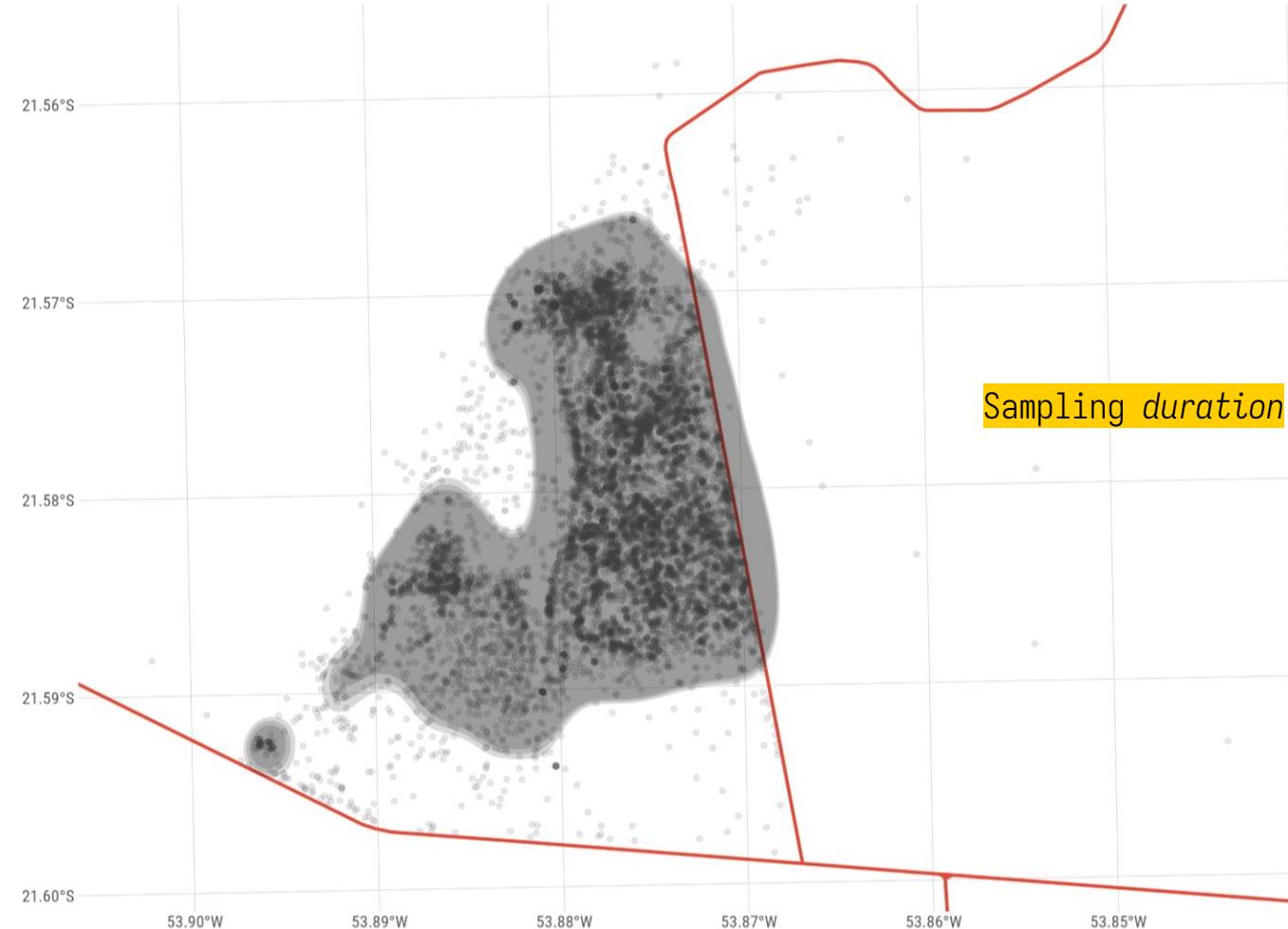
LC



Lowland tapir (*Tapirus terrestris*)



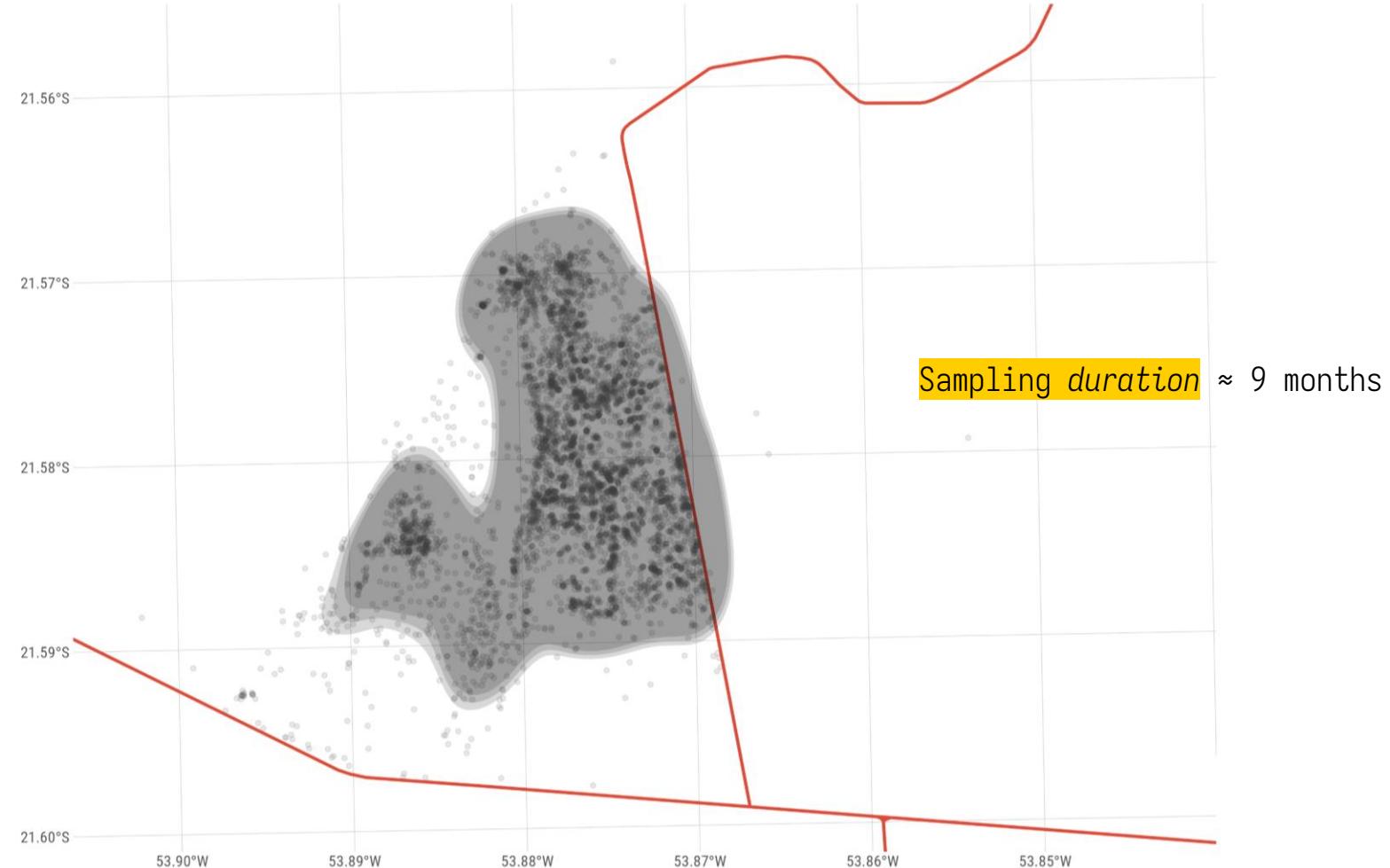
VU



Lowland tapir (*Tapirus terrestris*)



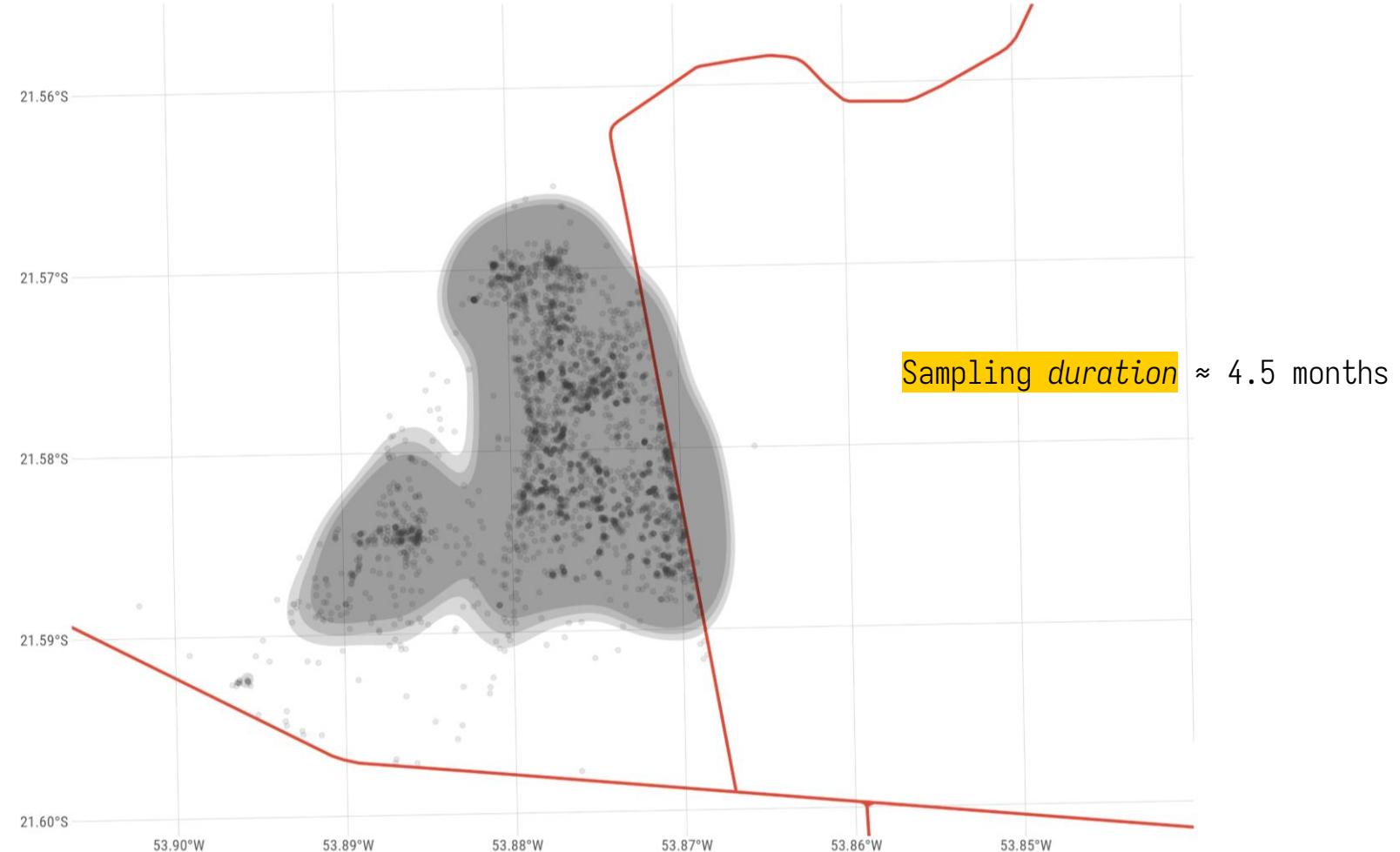
VU



Lowland tapir (*Tapirus terrestris*)



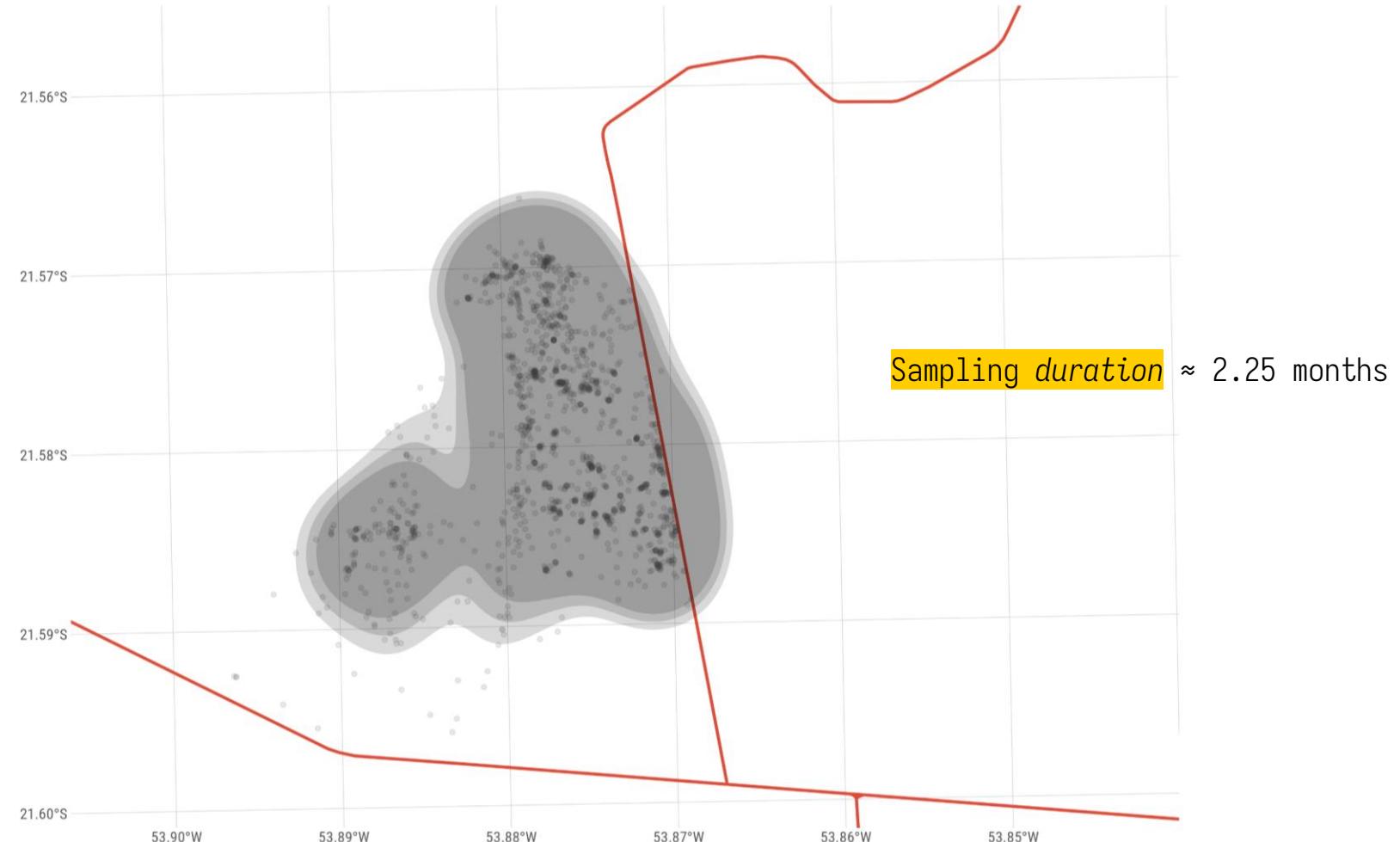
VU



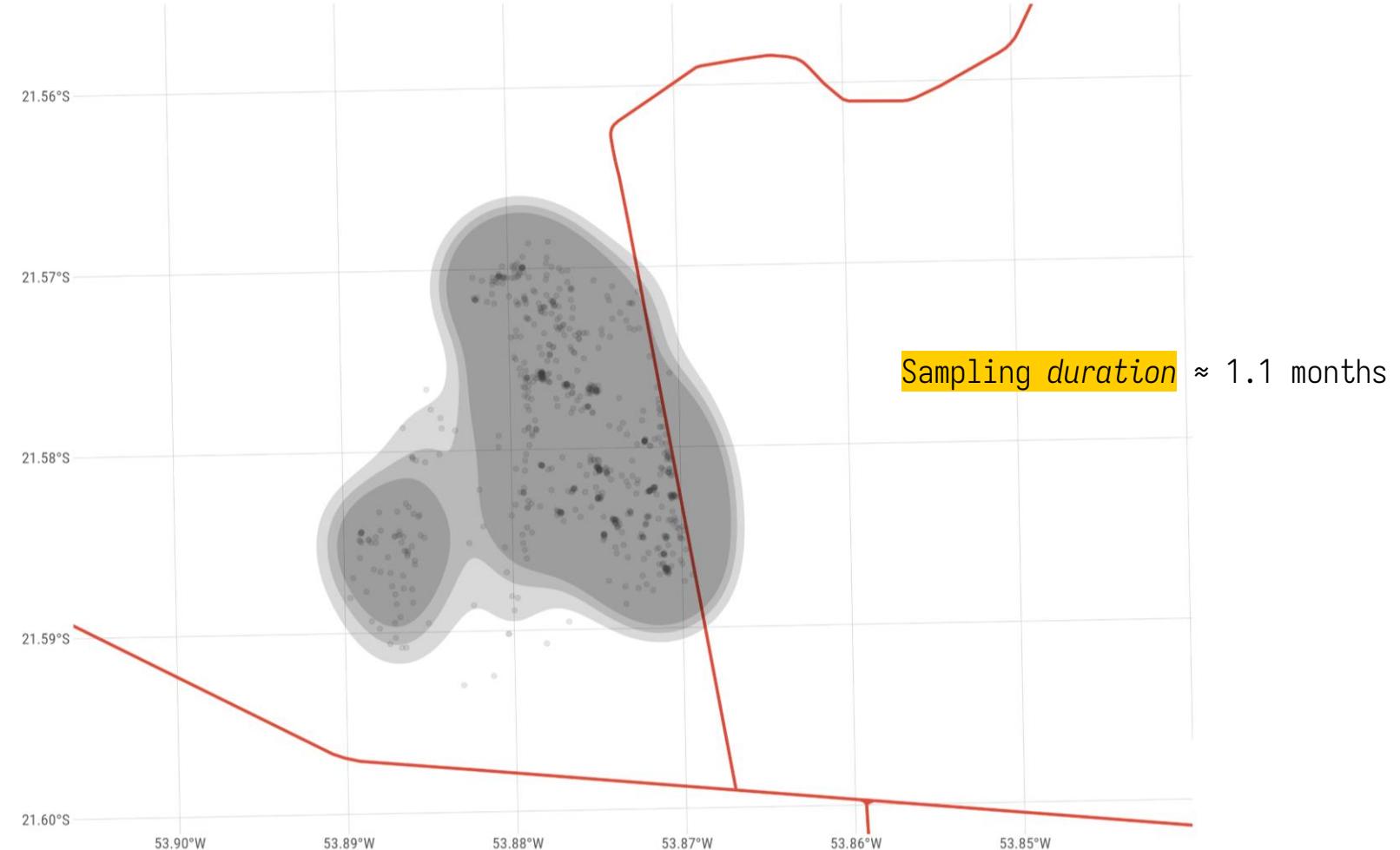
Lowland tapir (*Tapirus terrestris*)



VU



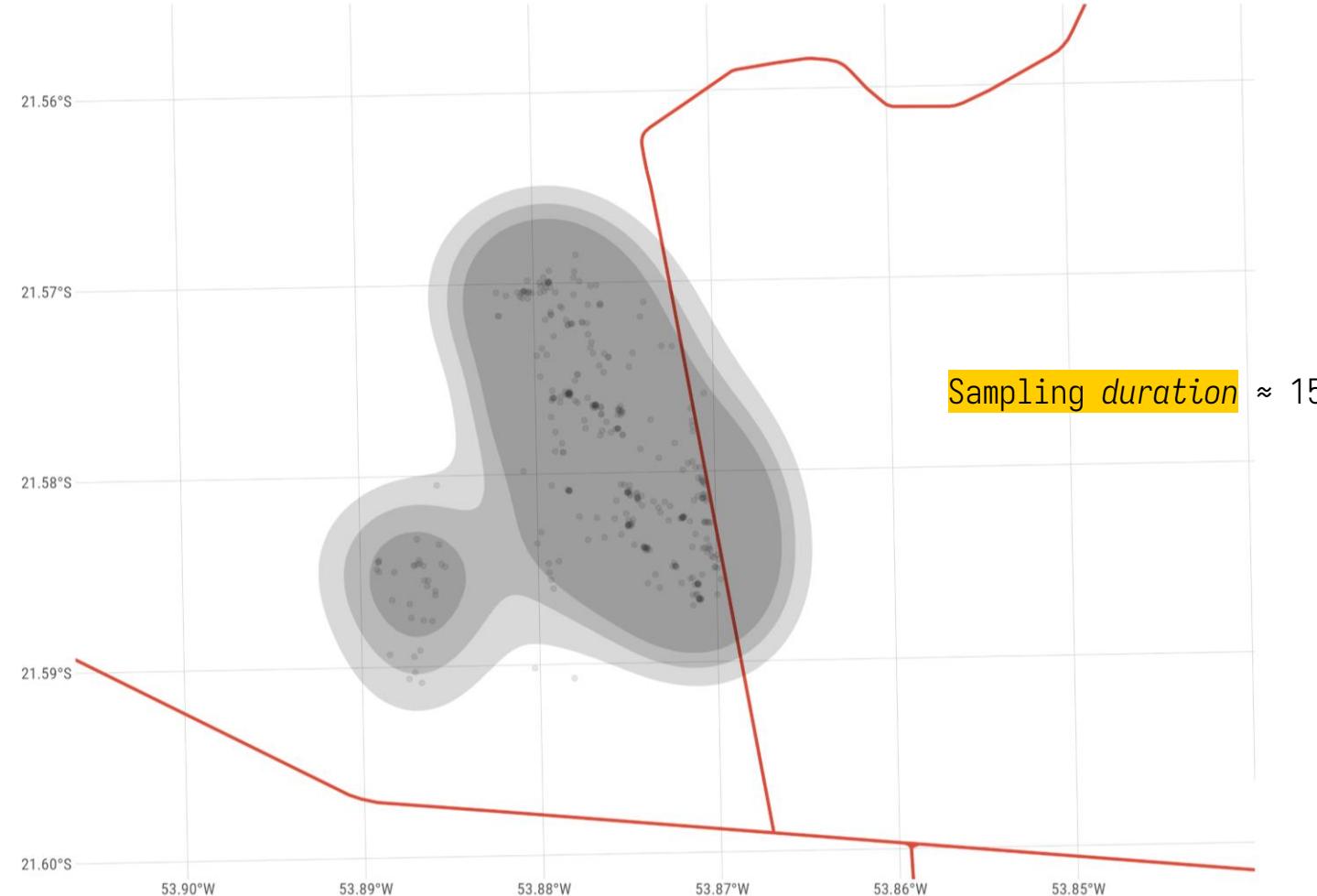
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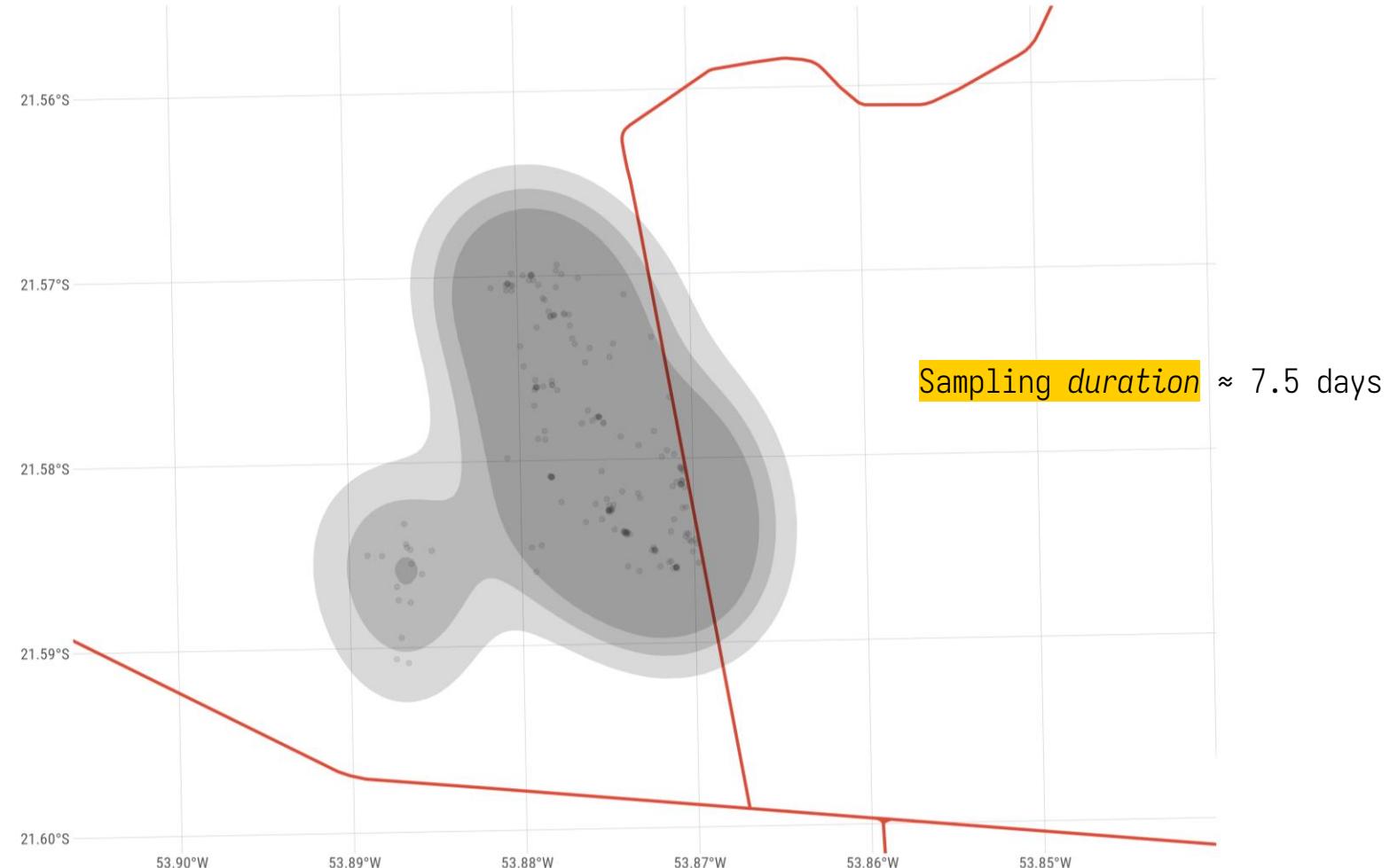
VU



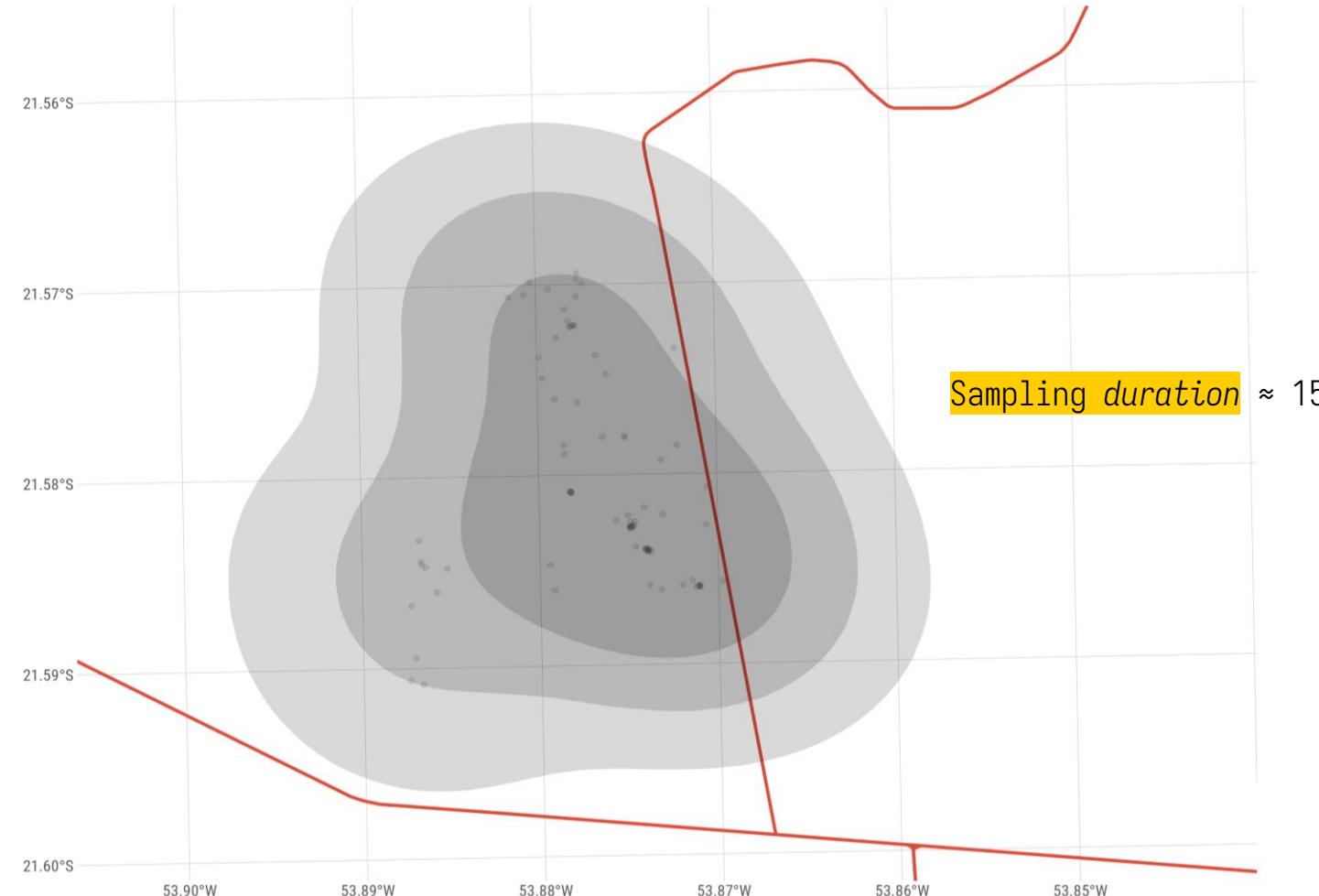
Lowland tapir (*Tapirus terrestris*)



VU



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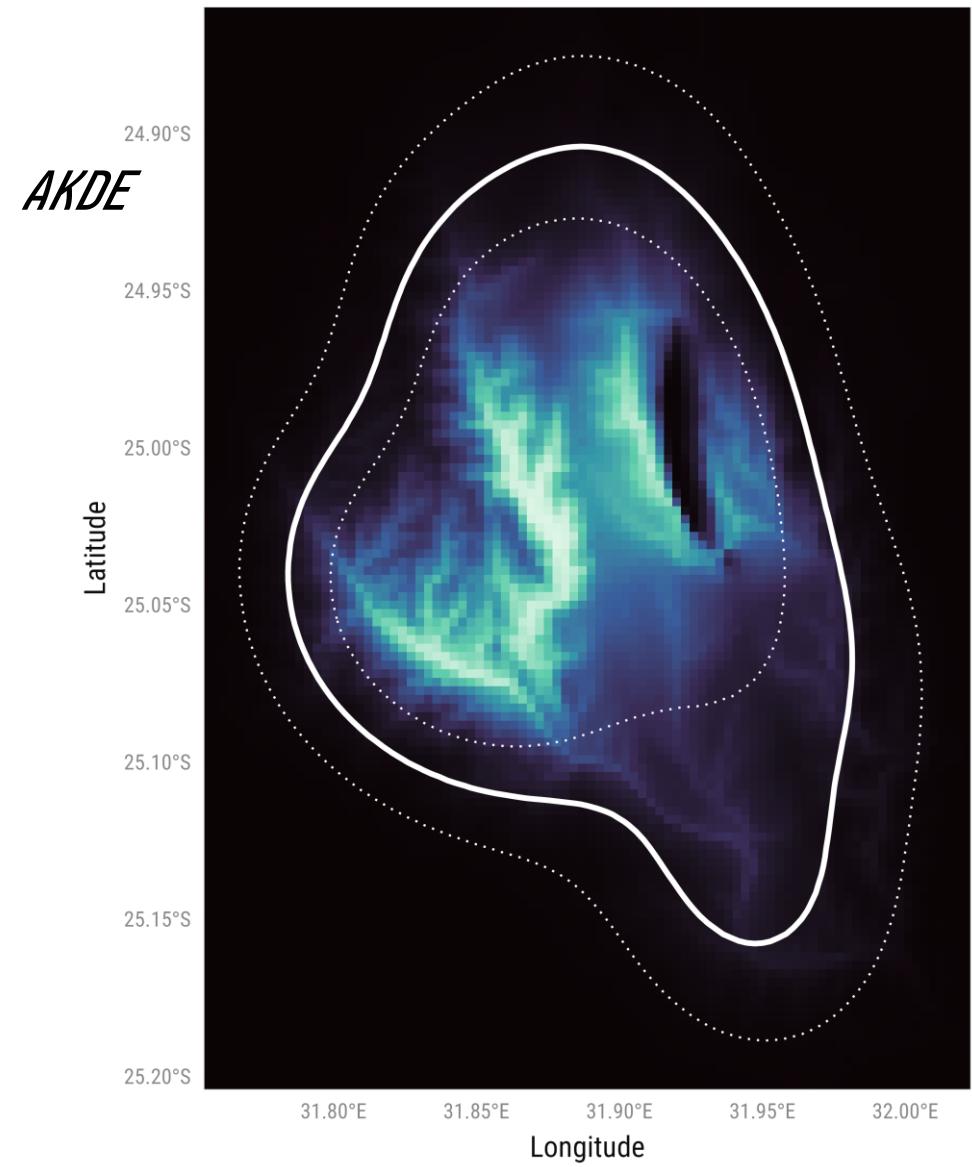
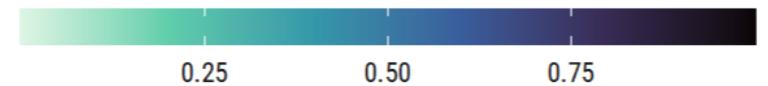
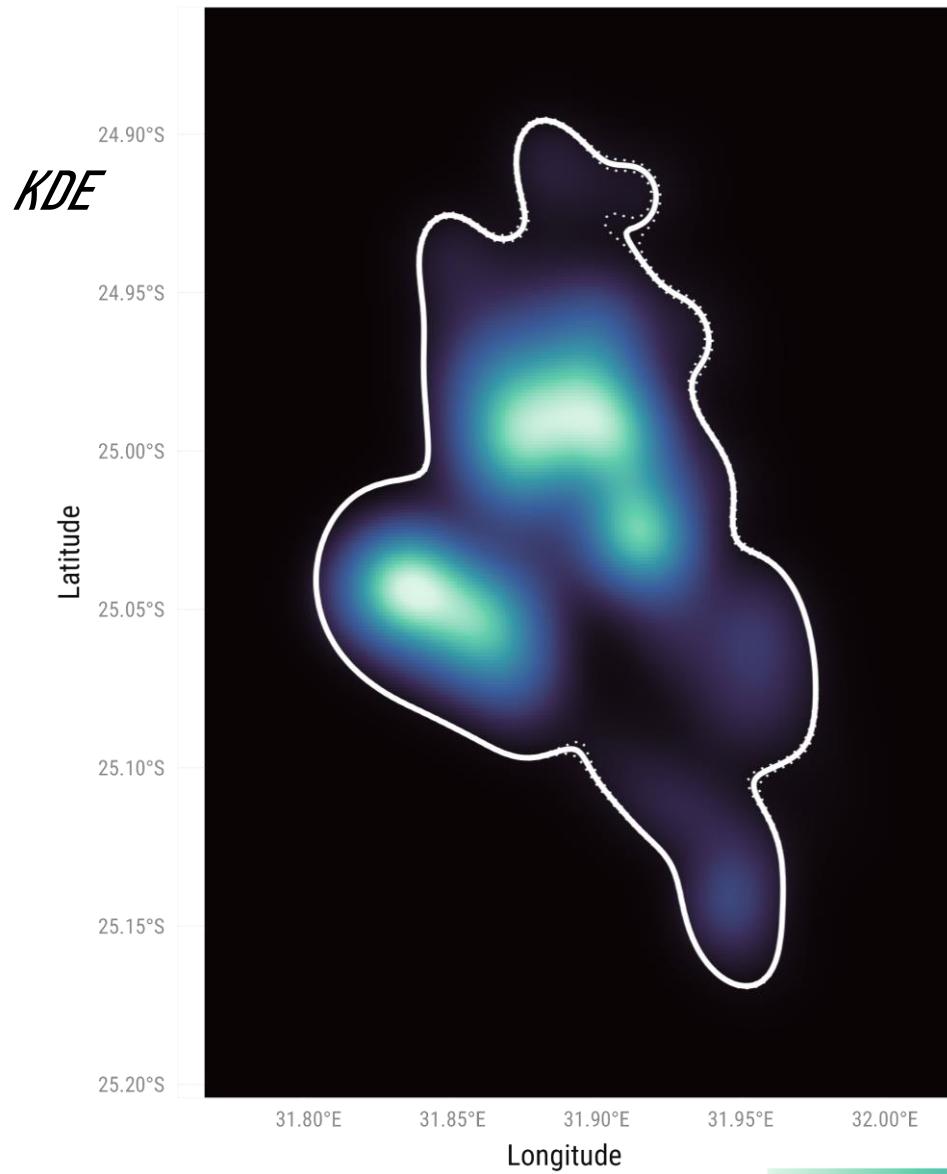
RSF models can provide **good habitat suitability estimates**, but usually provide poor home-range estimates. **(A)KDE** can provide **good home-range estimates**, when there is no discontinuity.

RSF + (A)KDE works better, respecting habitat preferences and barriers!

Incorporated into the `ctmm::akde()` function.



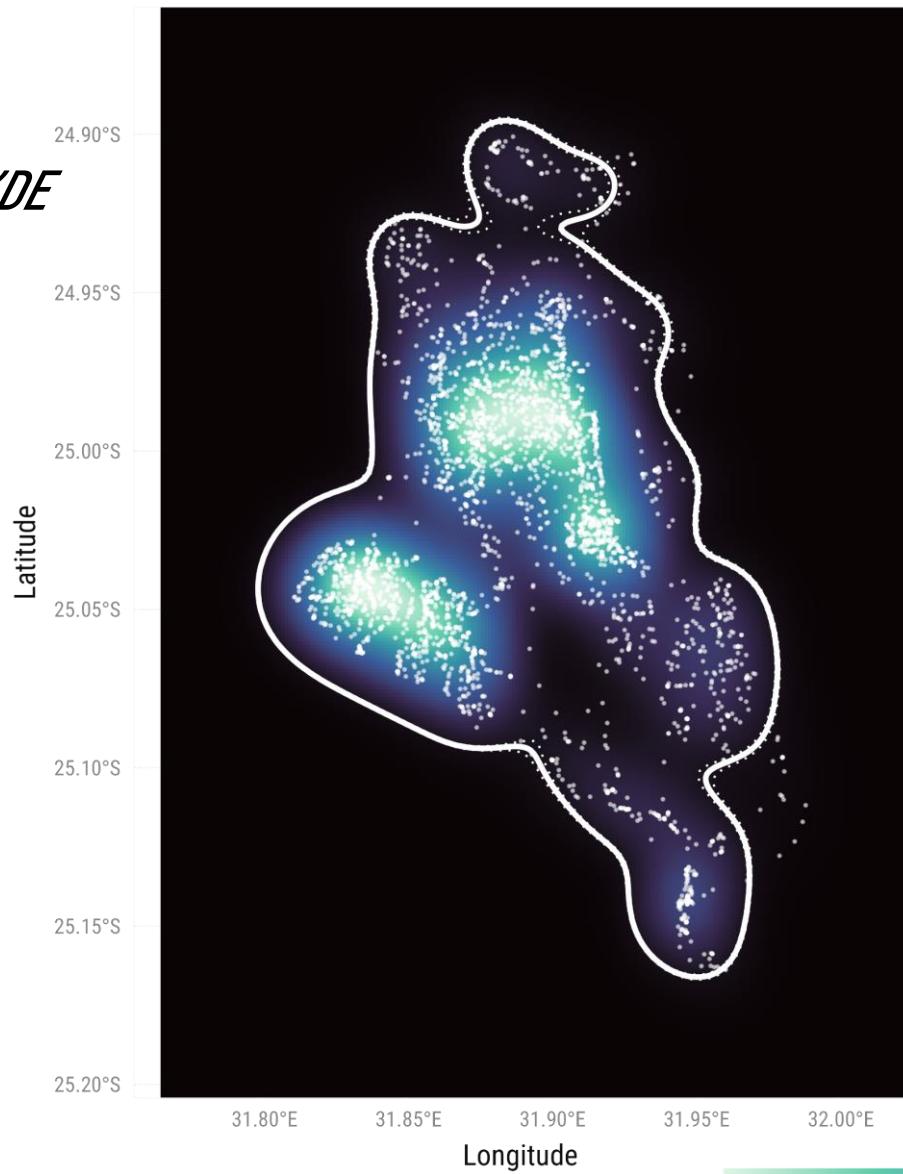
Example



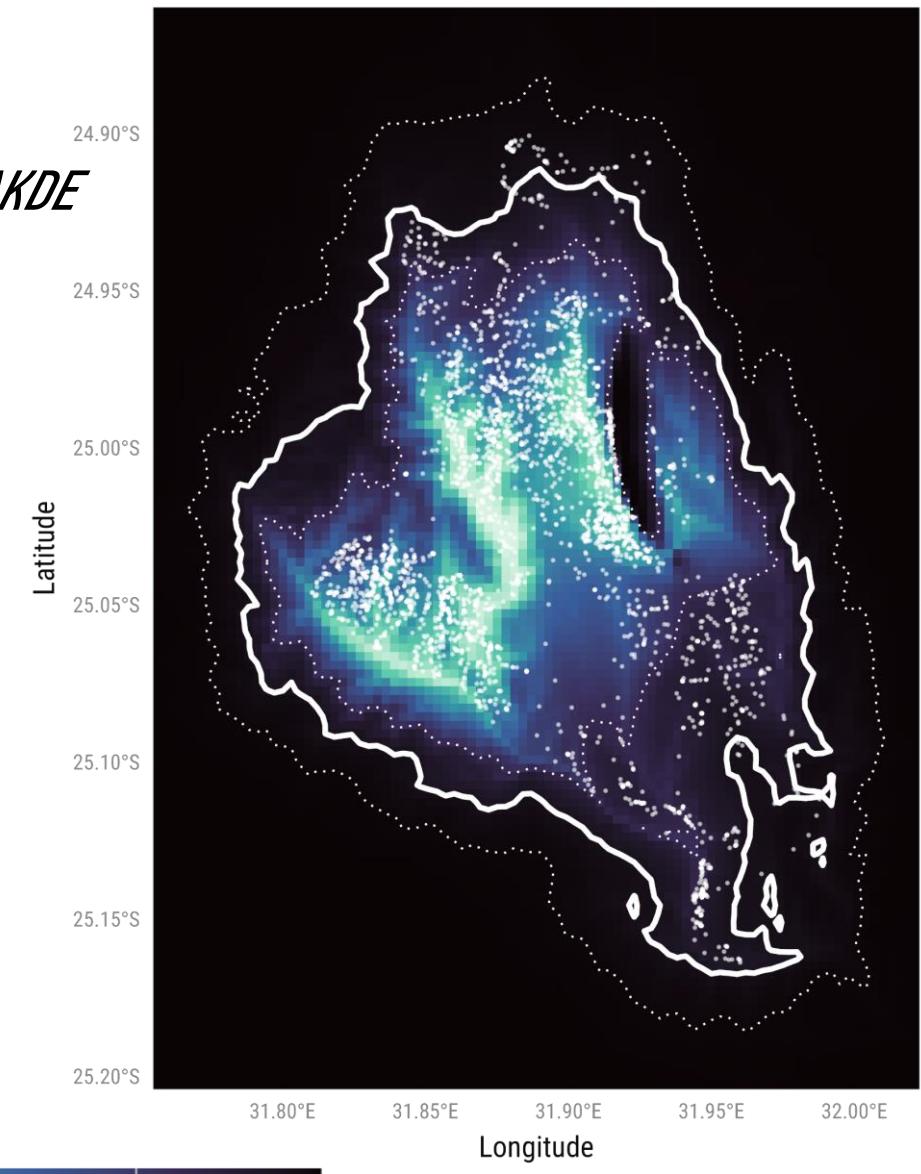


Example

RSF-KDE

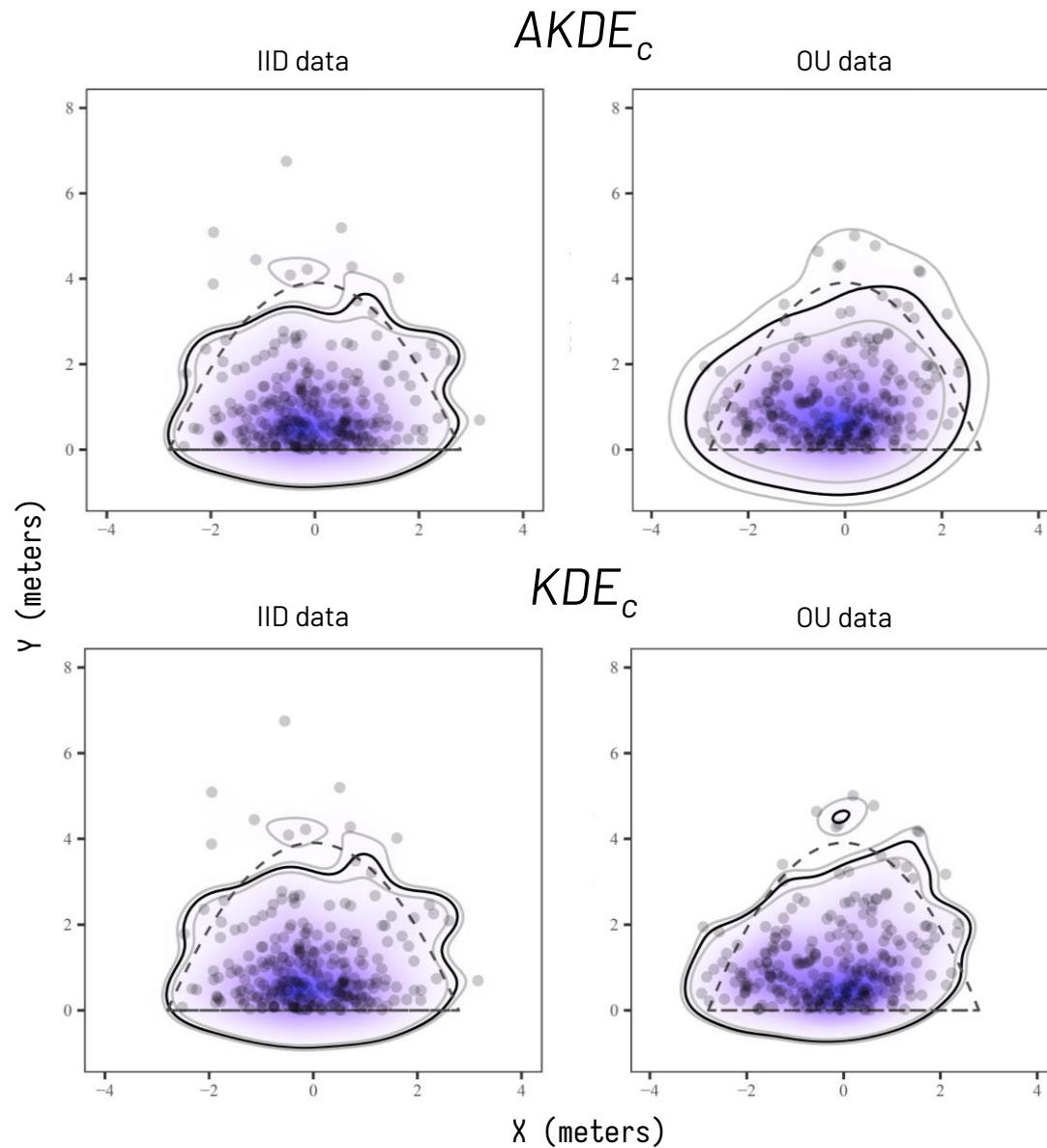


RSF-AKDE





Dealing with **hard boundaries**



“Spillover bias”

can lead to **overestimation** of home range areas, especially in (A)KDE.

This bias arises when animal movement is constrained by **impassable boundaries**, and is **proportional** to:

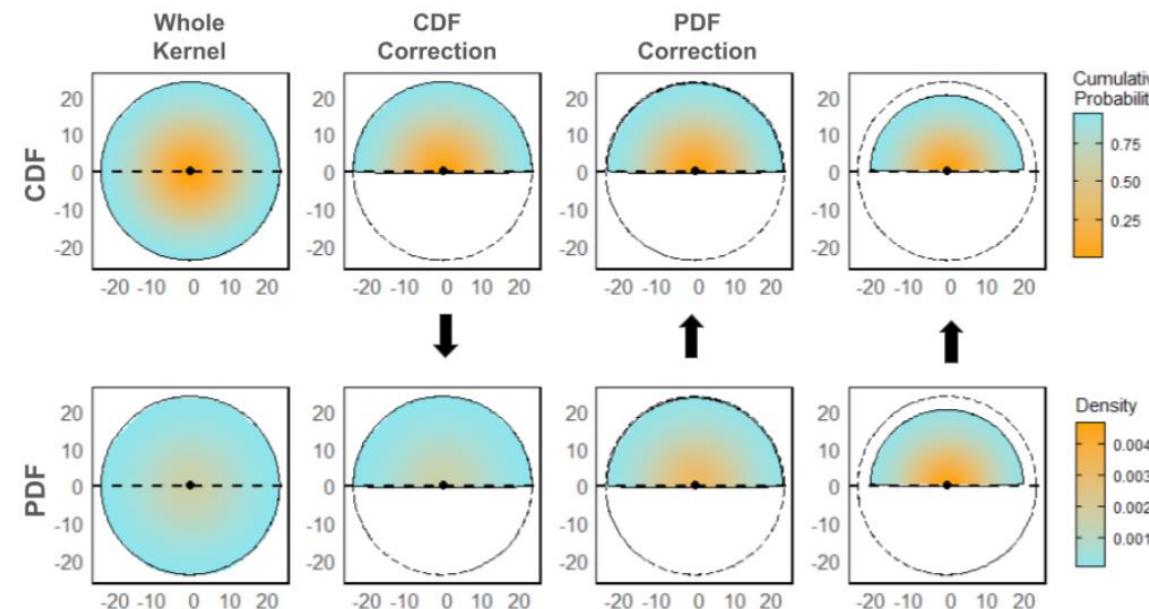
- **Size of the bandwidth.**
- **Amount of time** the animal spends near the boundary.

Adapted from **Noonan et al. (2019)**



Corrections to remove this “spillover bias”:

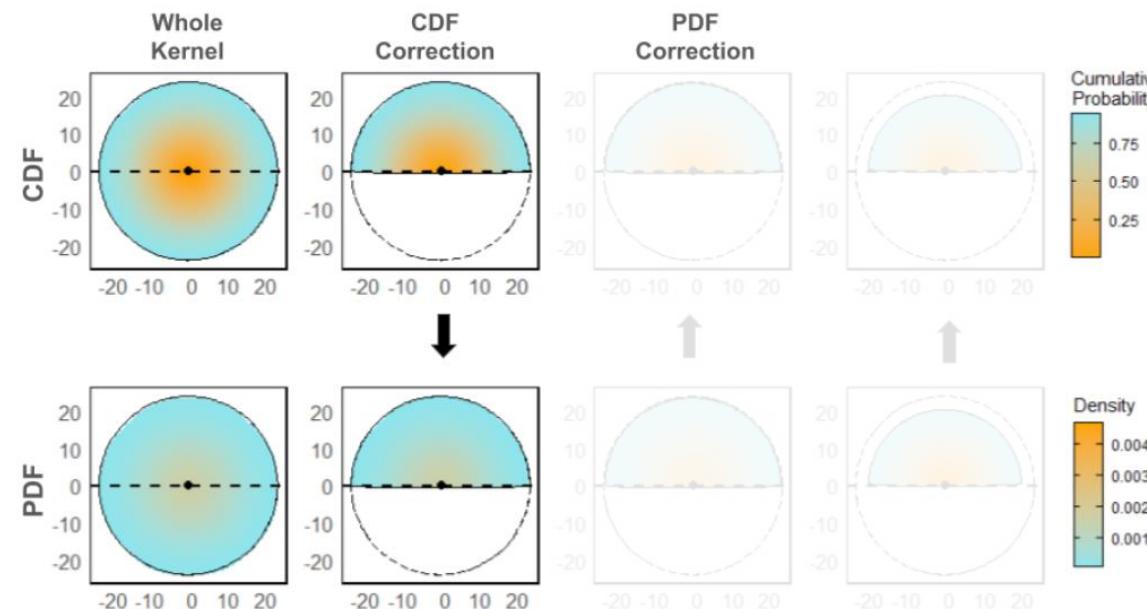
- **Local corrections** (applied to the constituent kernels)
- **Post-hoc corrections** (applied to the PDF or CDF)





Corrections to remove this “spillover bias”:

- **Local corrections** (applied to the constituent kernels)
- **Post-hoc corrections** (applied to the PDF or CDF)
often referred to as “masking” or “clipping”

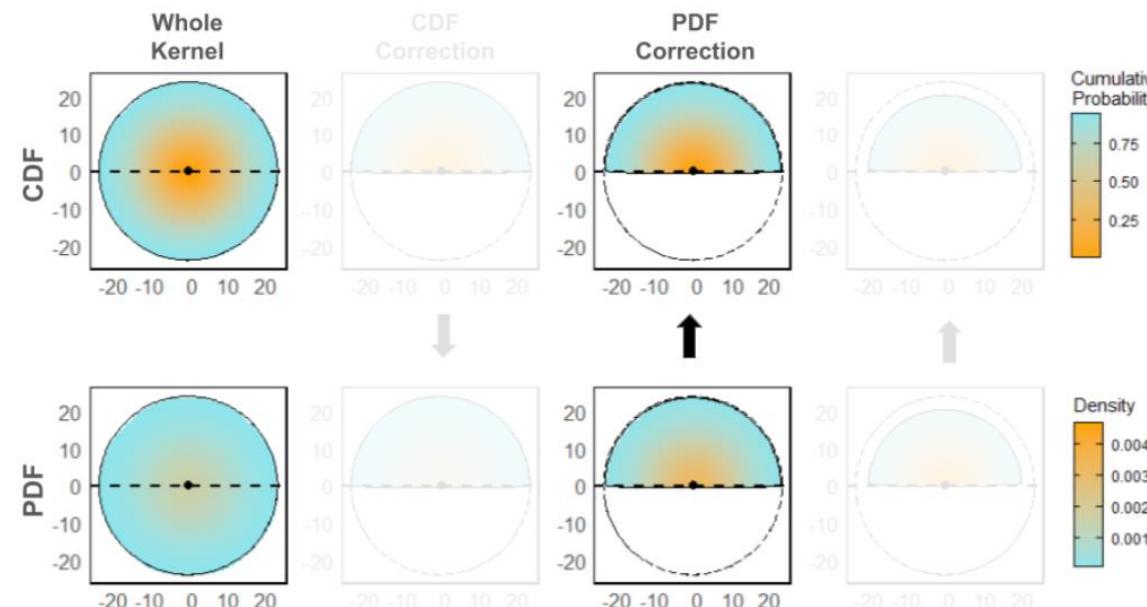


The removed probability mass of the spillover is
redistributed to the isopleth from which it was
removed on the available boundary side.



Corrections to remove this “spillover bias”:

- **Local corrections** (applied to the constituent kernels)
- **Post-hoc corrections** (applied to the PDF or CDF)
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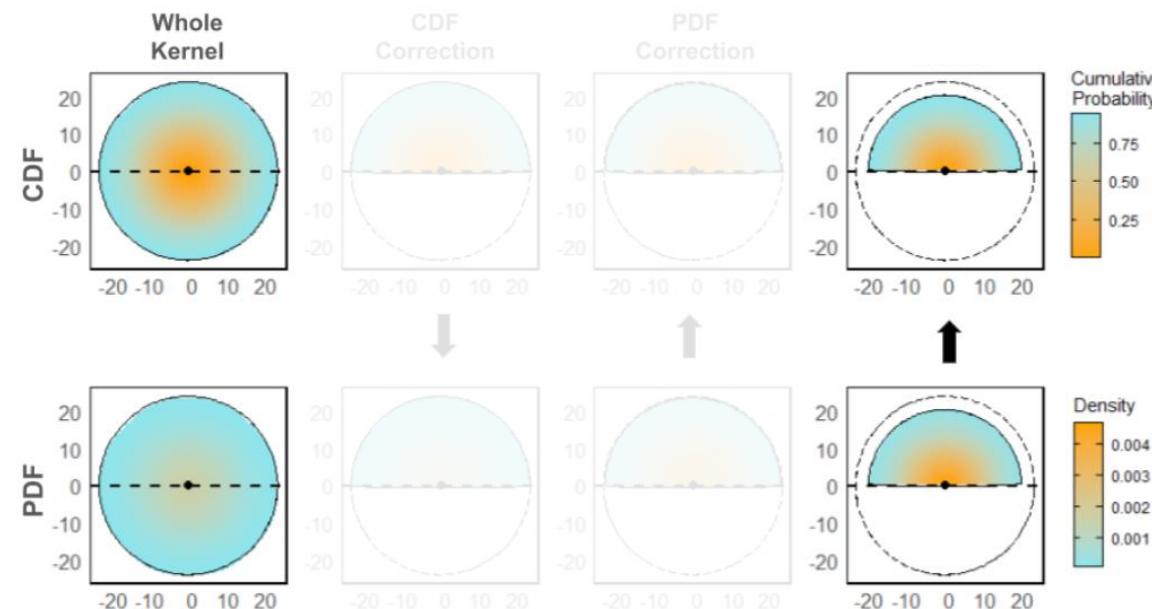


The removed mass of the range estimate is **redistributed throughout the entire remaining home range extent**, proportionally increasing the probability mass contained within each remaining cell.



Corrections to remove this “spillover bias”:

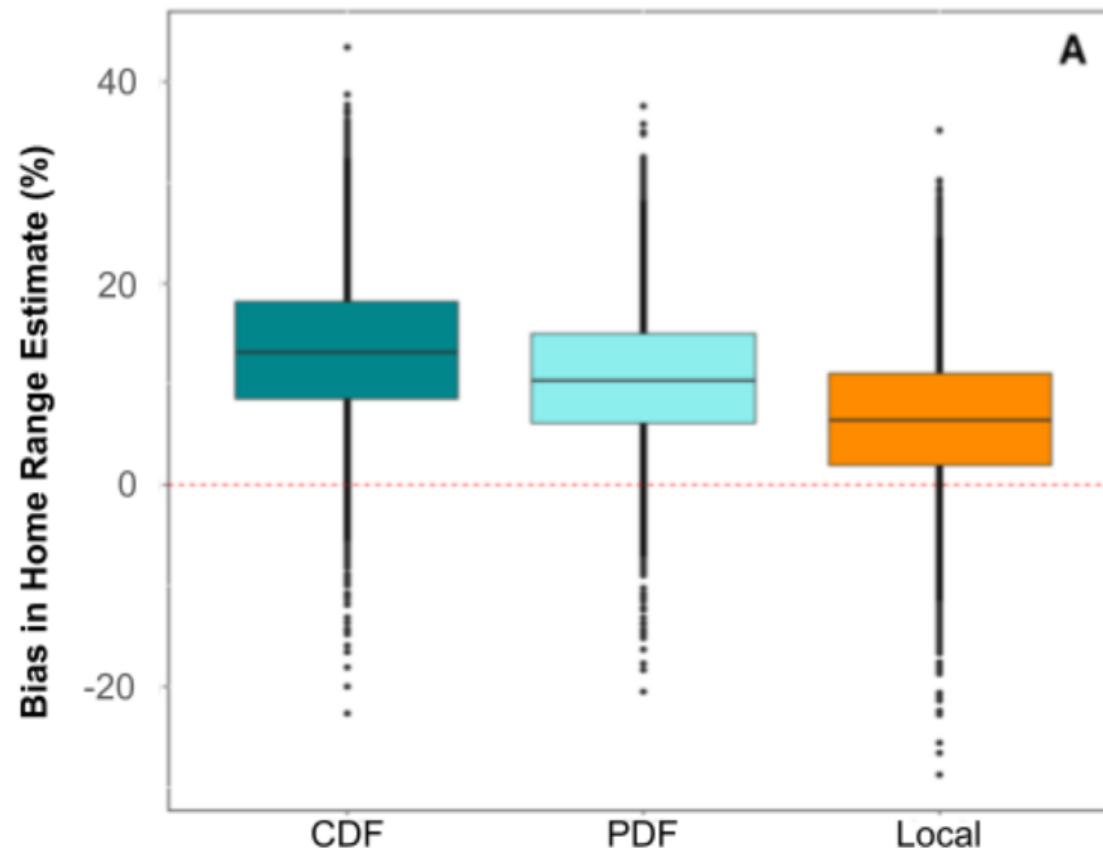
- **Local corrections** (applied to the constituent kernels)
- **Post-hoc corrections** (applied to the PDF or CDF)



Removed probability mass is then
redistributed within the constituent kernel
it was removed from before PDF integration.

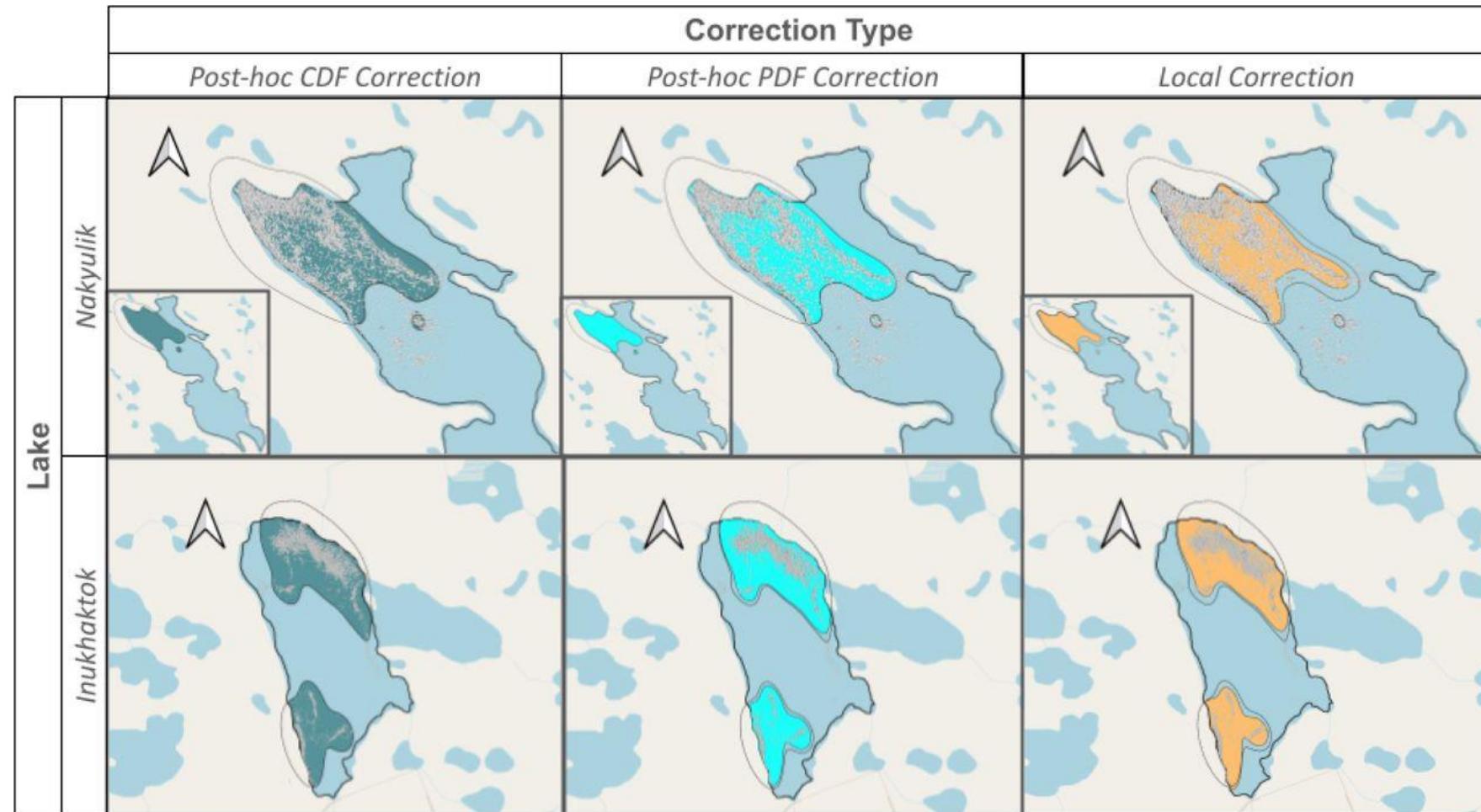


Dealing with **hard boundaries**





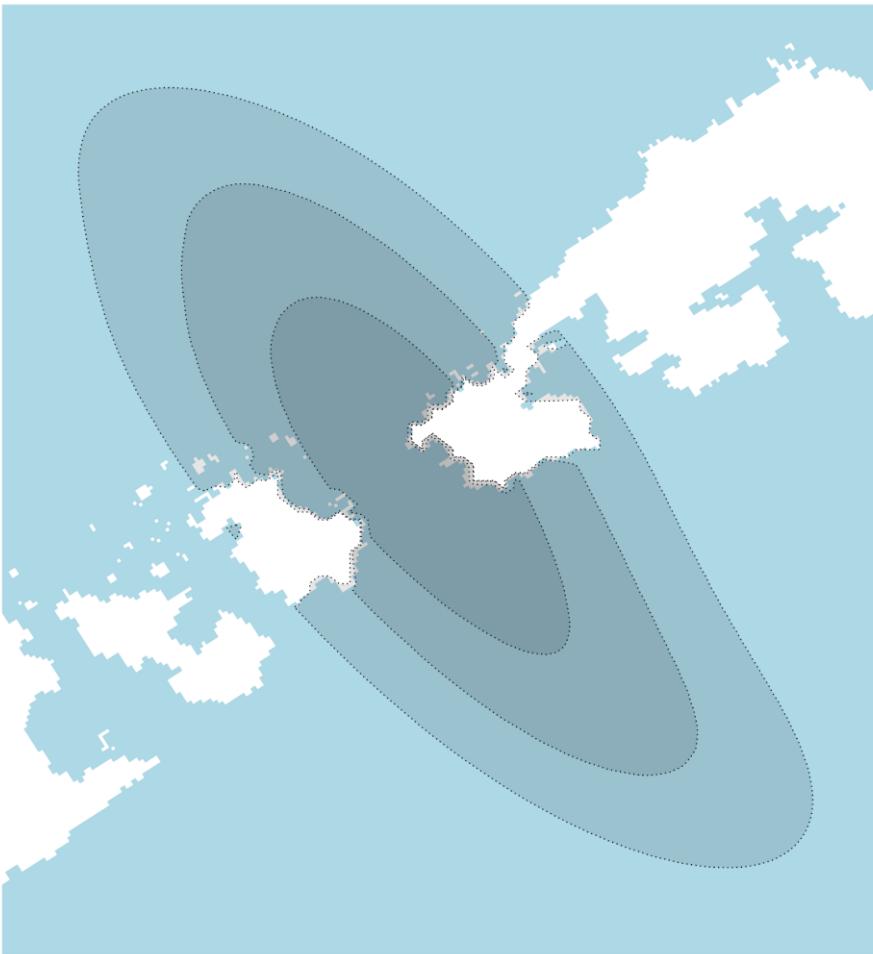
Dealing with hard boundaries





Locally-corrected AKDE

Each kernel's PDF is **truncated at movement boundaries** to avoid spillover. Removed probability mass is redistributed within the same kernel, ensuring the **total probability remains valid**.



Locally-corrected AKDE

Each kernel's PDF is **truncated at movement boundaries** to avoid spillover. Removed probability mass is redistributed within the same kernel, ensuring the **total probability remains valid**.