

# Week 5: Introduction to Animal Movement Analyses in R

Lecture: Tuesday, 13 April 2021 and Thursday, 15 April 2021

8:30 am – 10:20 pm EDT



Instructor:  
Jared Stabach (StabachJ@si.edu)

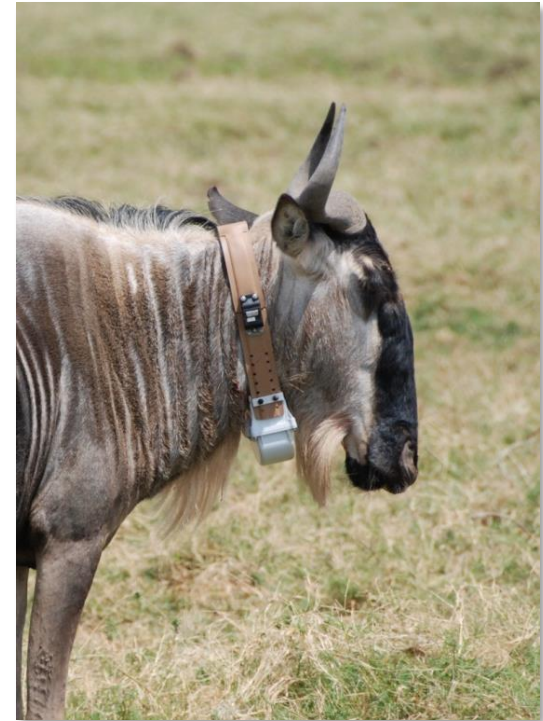
# Week 5: Introduction to Movement Analyses

Lecture: Tuesday, 06 April 2021 and Thursday, 08 April 2021  
8:30 am – 10:20 pm EDT

## Tuesday

- Data Import & Cleaning
  - Time Stamp Formatting
  - Duplicate Removal
  - Assessing Model Assumptions
  - Filtering Start Dates
  - Assessing Data Quality
  - Calculating Fix Success
- Data Visualization
  - Animating Animal Movements (Super Fun!)

We will use GPS collar data, collected on white-bearded wildebeest (*Connochaetes taurinus*) during my doctoral work from 2010-2013 for all exercises. Methods, however, should be broadly applicable to other taxa.



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Jared Stabach ([StabachJ@si.edu](mailto:StabachJ@si.edu))



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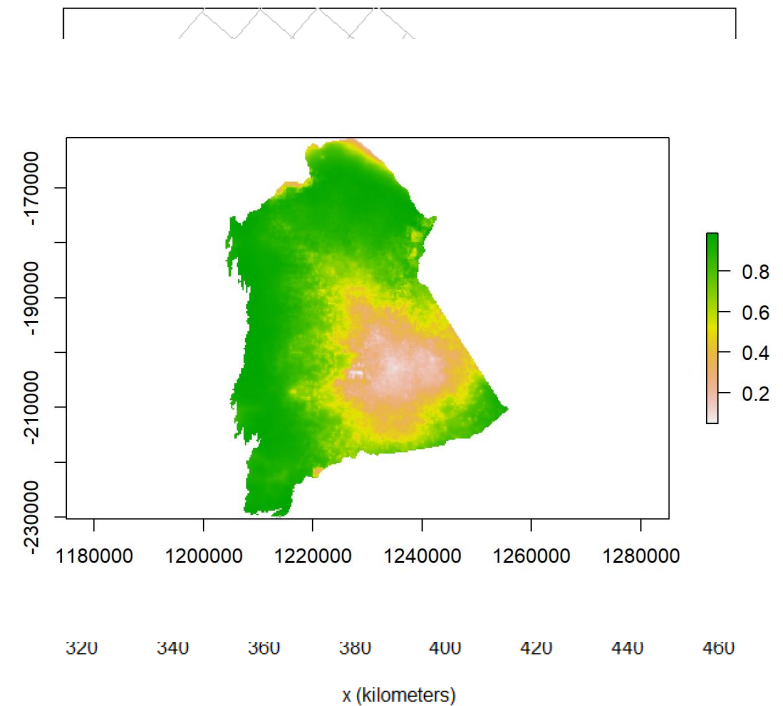
# Week 5: Introduction to Movement Analyses

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## Thursday

- Calculate Home Range
  - Minimum Convex Polygon (MCP)
  - Autocorrelated Kernel Density Estimation (AKDE)
- Resource Selection Function (RSF) Analysis
  - 'Use' vs 'Availability' design
  - Define 'Availability';
  - Create spatial database
  - Perform logistic regression
  - Assess response curves
  - Generate a prediction map



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At least 58 packages in R to analyze animal tracking data

How to know which to use?

Journal of Animal Ecology

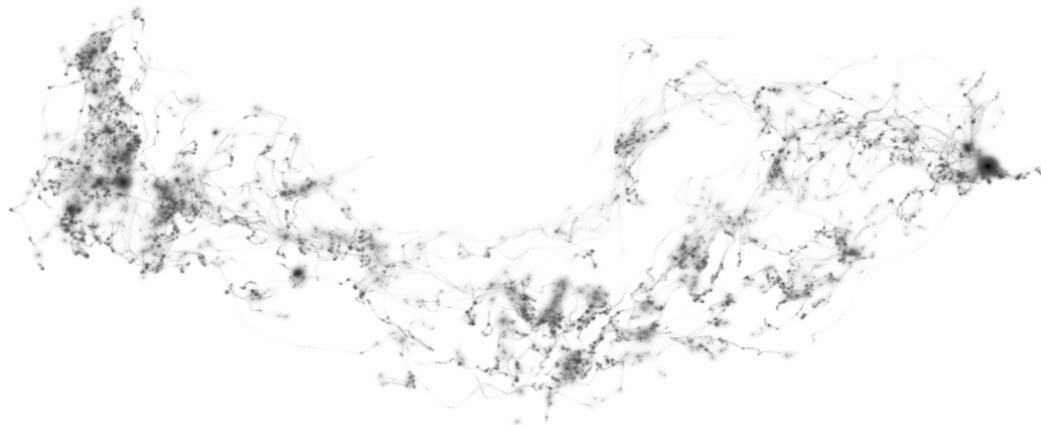


BIOLOGGING: REVIEW

## Navigating through the R packages for movement

Rocío Joo , Matthew E. Boone, Thomas A. Clay, Samantha C. Patrick, Susana Clusella-Trullas, Mathieu Basille

First published: 06 October 2019 | <https://doi.org/10.1111/1365-2656.13116> | Citations: 3



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## Where to get data.....



for animal tracking data

Data

Help

Tools

Archiving

News

Search documentation

About

Login

Search

Search Advanced Search

wildebeest

All Sensor Types

☐ Only studies where I can see data

Search

Search result

Sort by Animal Identifier

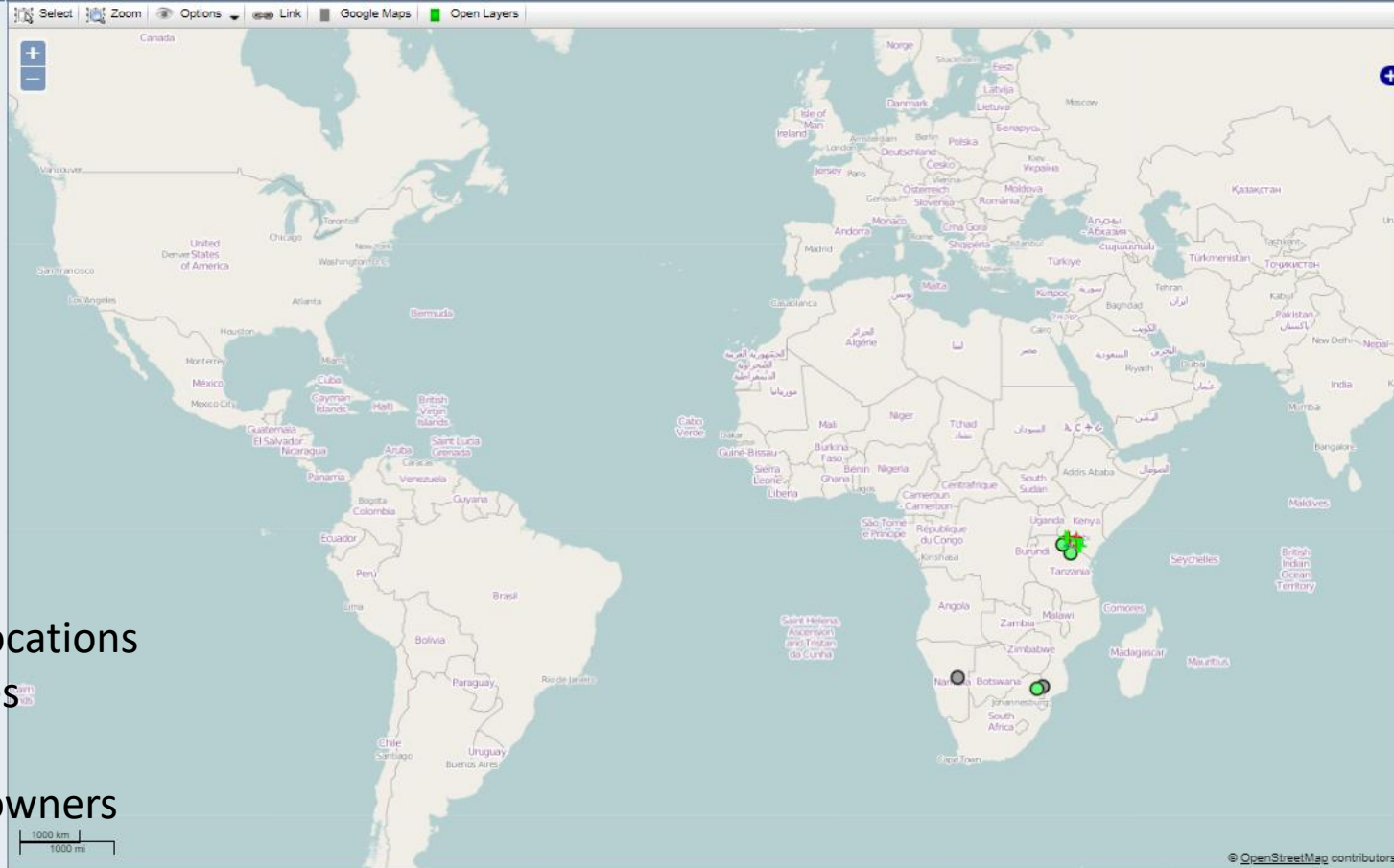
- ☐ African carnivores Vanak
- ☐ ICARUS in Namibia
- ☐ Kruger NP Plains game
- ☒ White-bearded wildebeest in Kenya (n=38)
- ☐ Wildebeest (Eastern white bearded) Morrison Tarangin
- ☐ Wildebeest (Western White-Bearded) Hopcraft Sereng

**2.4 billion** locations

**5,915** studies

**1,025** taxa

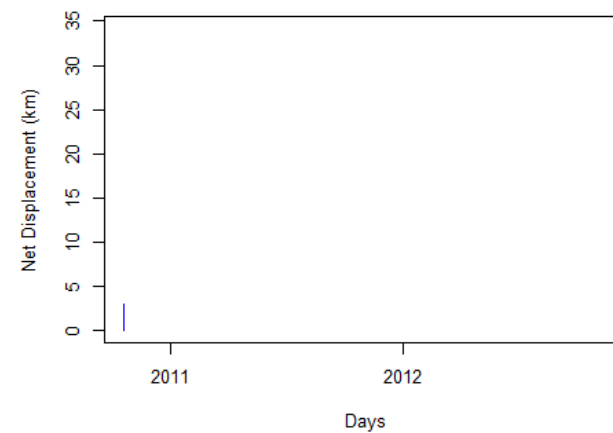
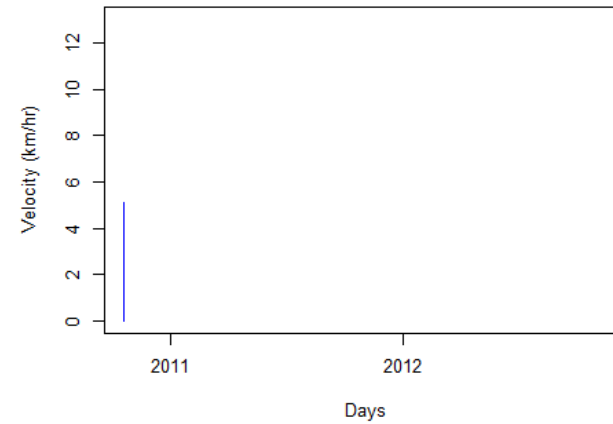
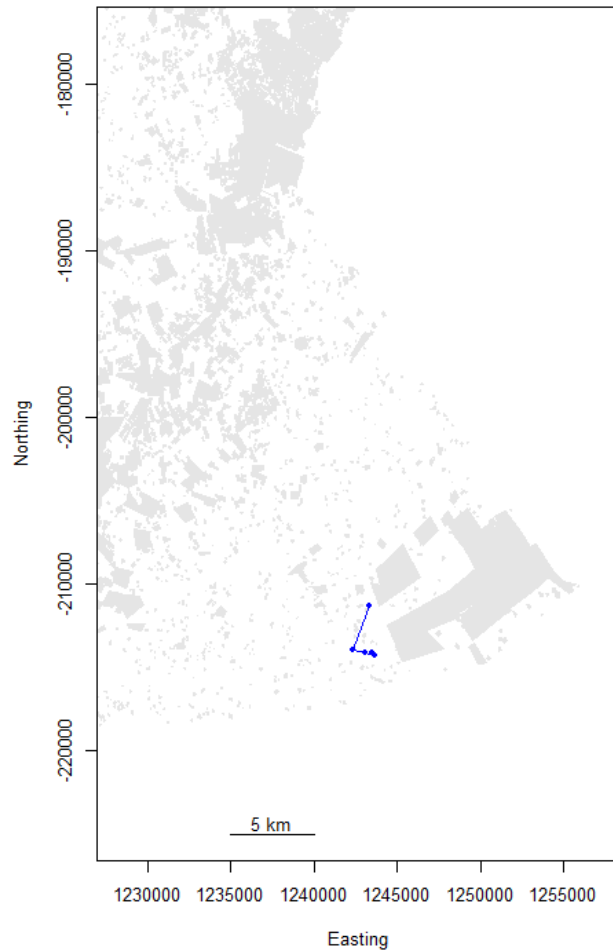
**3,000** data owners



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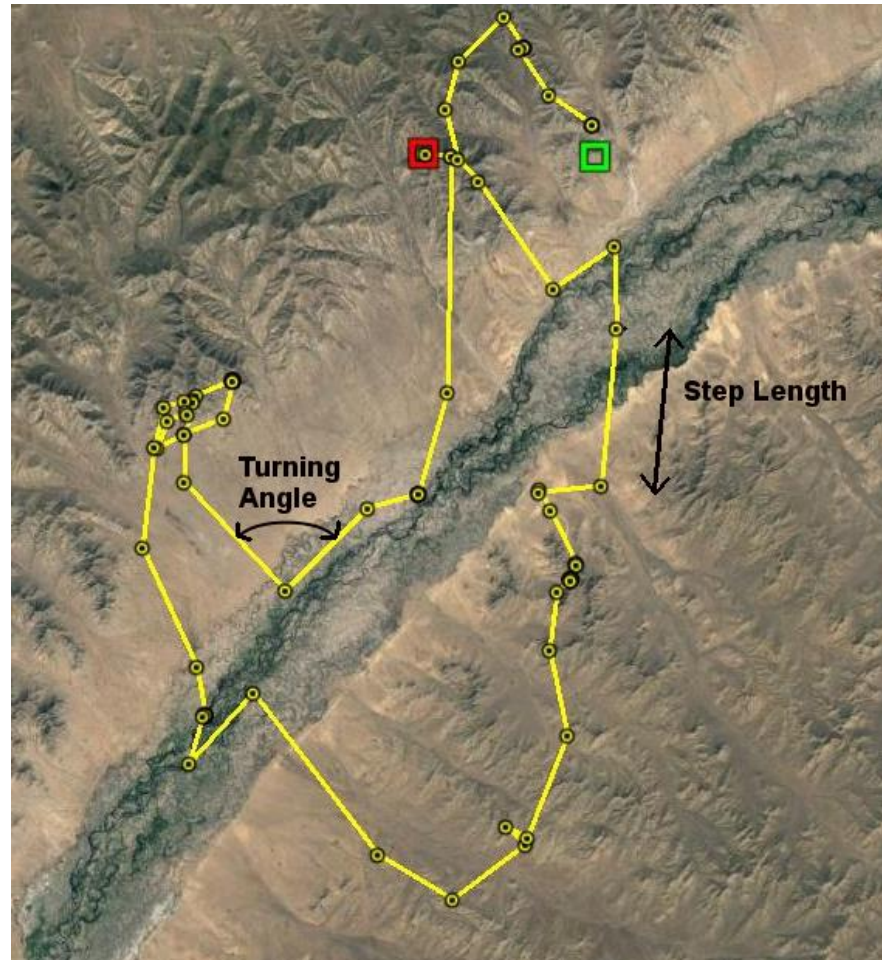
# Visualize Movement Data

White-bearded wildebeest (*Connochaetes taurinus*)

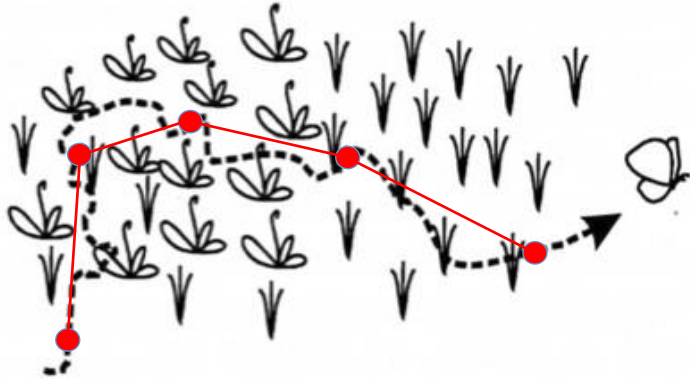


# Properties of a Movement Path

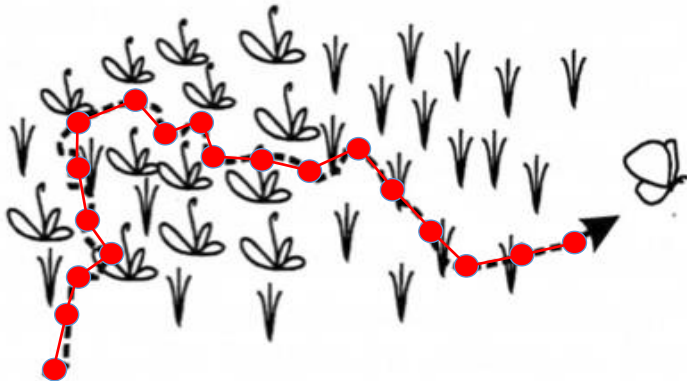
- Step length
- Turning angle
- Velocity
- Autocorrelation



# Things to be wary of



Different sampling rates give  
different answers





# Home Range

Burt (1943): the area occupied by an individual during normal everyday activities (i.e., mating, feeding, carrying for young)

A home range must contains all the necessary resources to insure survival and reproduction

Spatial expression of behaviors animals perform to survive and reproduce  
(Borger et al. 2008)



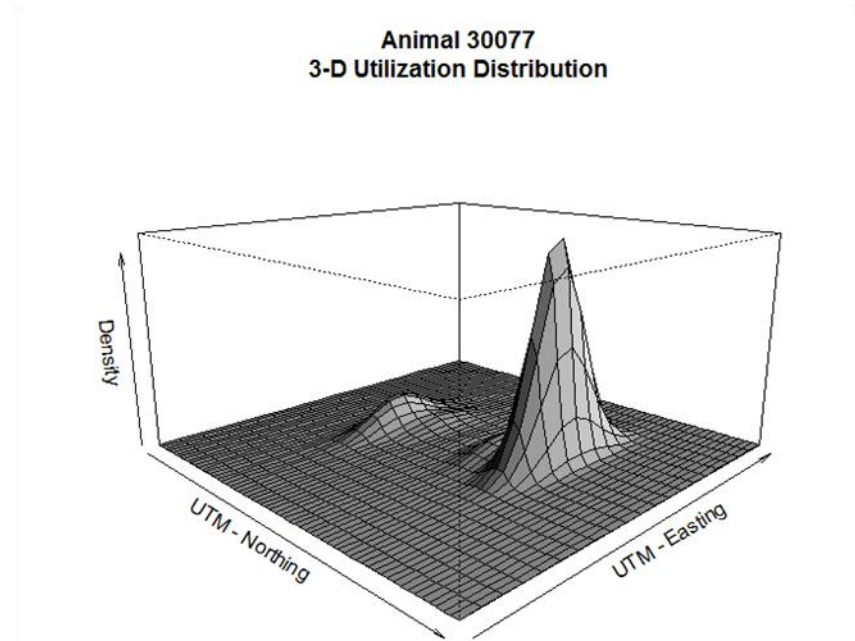
# Home Range Characteristics

- Individual spends majority of time here
- Vary in size
  - Gender
  - Seasonally
  - By region/habitat
- Irregular shape
- Increase in size over time, but should asymptote (resident assumption)
- Successful individual will have a high quality home range



## Multiple Home Range Estimators Exist

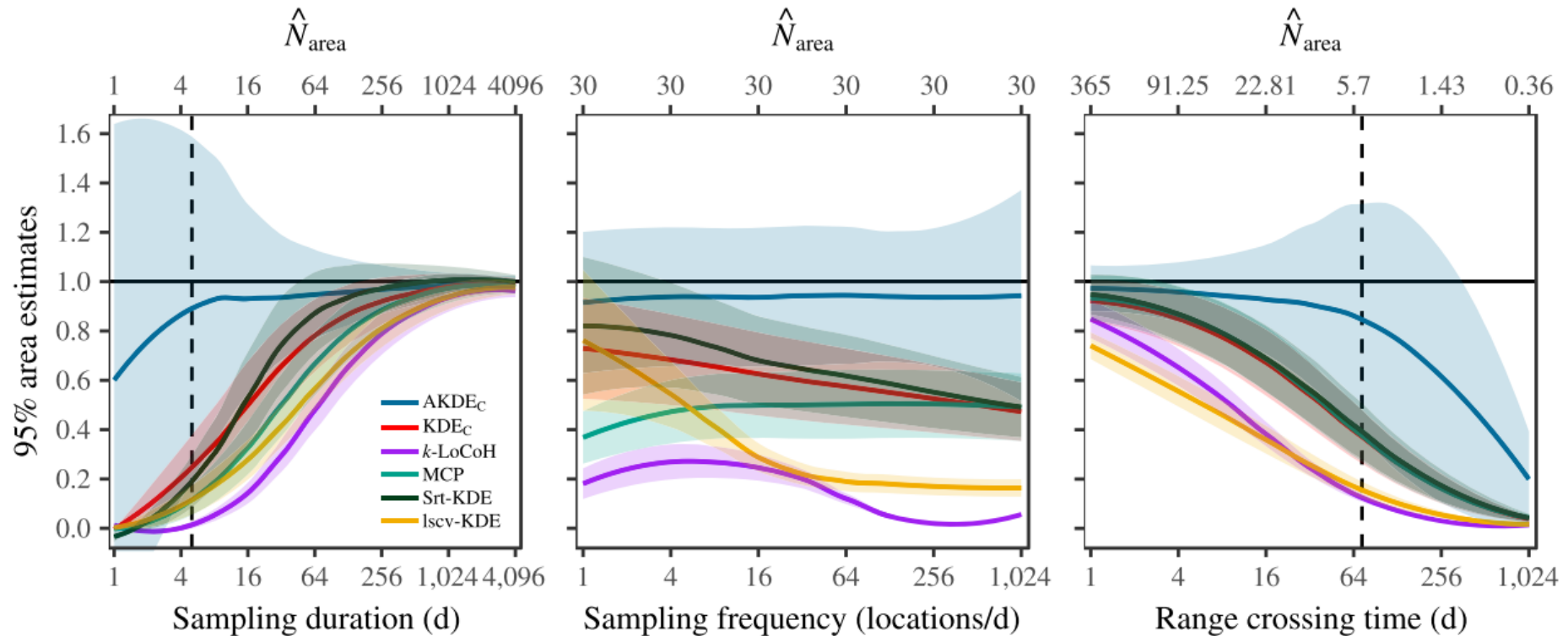
- Minimum Convex Polygons
- Harmonic Means
- Jennrich Turner
- Kernel Density
  - Brownian Bridge Kernel
  - Random Bridge Kernel
- Local Convex Hull (LoCoH)
- Characteristic Hull
- .....



All of these methods assume that data are IID  
(Independent and Identically Distributed)



# Quantitative Assessment of Home Range Estimators



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## A comprehensive analysis of autocorrelation and bias in home range estimation

MICHAEL J. NOONAN; MARLEE A. TUCKER; CHRISTEN H. FLEMING; THOMAS S. AKRE; SUSAN C. ALBERTS; ABDULLAH H. ALI; JEANNE ALTMANN; PAMELA CASTRO ANTUNES; JERROLD L. BELANT; DEAN BEYER; NIELS BLAUM KATRIN BÖHNING-GAESE; LAURYCULLEN JR.; ROGERIO CUNHA DE PAULA; JASJA DEKKER; JONATHAN DRESCHER-LEHMAN; NINA FARWIG; CLAUDIA FICHEL; CHRISTINA FISCHER; ADAM T. FORD; JACOB R. GOHEEN; RENE JANSSEN; FLORIAN JELTSCH; MATTHEW KAUFFMAN; PETER M. KAPPELER; FLAVIA KOCH; SCOTT LAPOINT; A. CATHERINE MARKHAM; EMILIA PATRICIA MEDICI; RONALDO G. MORATO; RAN NATHAN; LUIZ GUSTAVO R. OLIVEIRA-SANTOS; KIRK A. OLSON; BRUCE D. PATTERSON; AGUSTIN PAVIOLO; EMILIANO ESTERCI RAMALHO; SASCHA REOSNER; DANA G. SCHABO; NURIA SELVA; AGNIESZKA SERGIEL; MARINA XAVIER DA SILVA; ORR SPIEGEL; PETER THOMPSON; WIEBKE ULLMANN; FILIP ZIEZ BA; TOMASZ ZWIJACZ-KOZICA WILLIAM F. FAGAN; THOMAS MUELLER; AND JUSTIN M. CALABRESE



# Home Range Continuous-Time Movement Modeling (CTMM) Framework

## Methods in Ecology and Evolution



Methods in Ecology and Evolution 2016

doi: 10.1111/2041-210X.12559

### APPLICATION

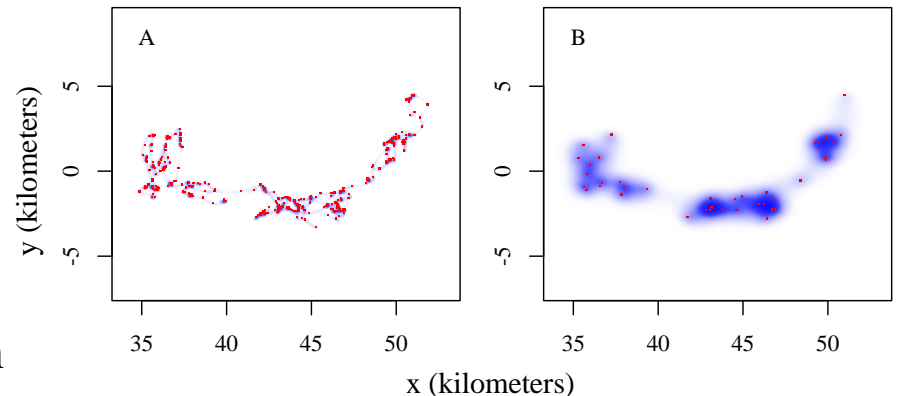
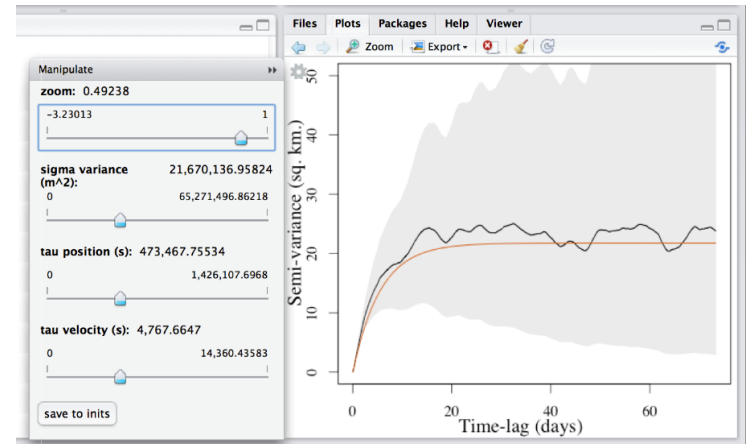
**ctmm: an R package for analyzing animal relocation data as a continuous-time stochastic process**

Justin M. Calabrese<sup>1,2\*</sup>, Chris H. Fleming<sup>1,2</sup> and Eliezer Gurarie<sup>2</sup>

Visual diagnostics  
Model selection (AIC)

Home range  
estimation

Path reconstruction  
Occurrence distribution



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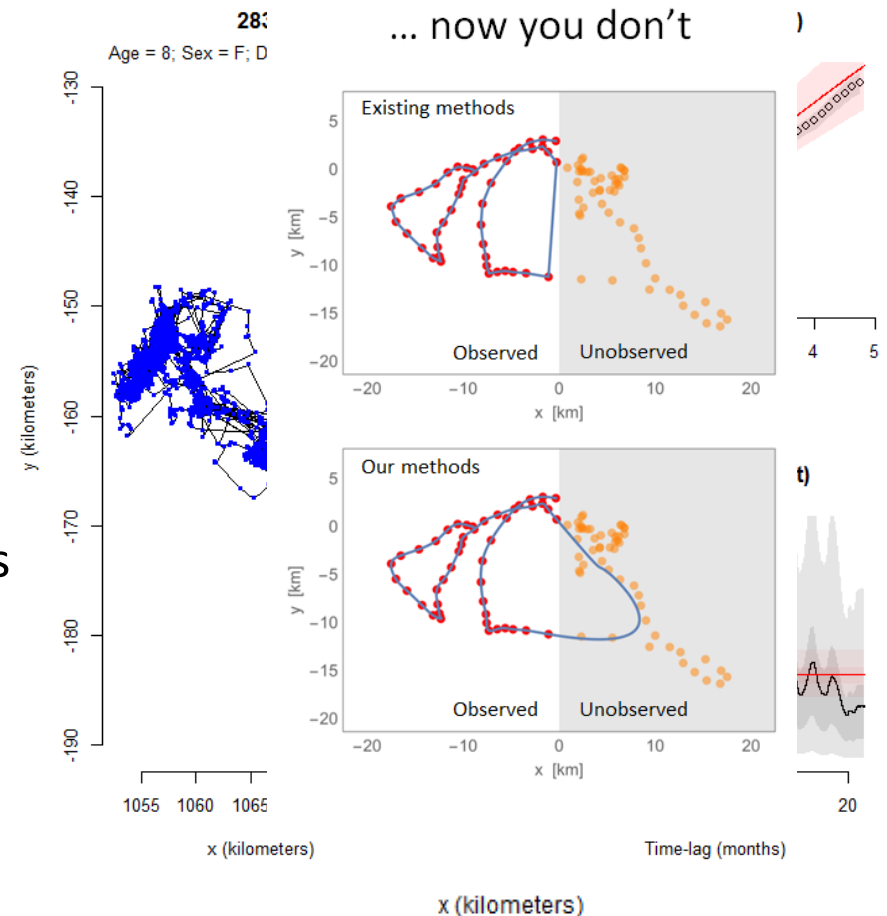
**ctmm: an R package for analyzing animal relocation data as a continuous-time stochastic process**

Justin M. Calabrese<sup>1,2\*</sup>, Chris H. Fleming<sup>1,2</sup> and Eliezer Gurarie<sup>2</sup>

Incorporate autocorrelation structure  
Insensitive to irregular sampling/data gaps  
Candidate model evaluation (AIC)  
Quantitatively evaluate range residency

Observed to outperform previous methods  
(MCP, KDE, LoCoH)  
Confidence interval estimation  
Conservation implications

Now you see me...  
... now you don't



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# Resource Selection Function Analysis

Describe the relationship between an animal and its environment

Point Process Model: Points Instead of Paths – A first step

Does the animal use a habitat type more than expected by random

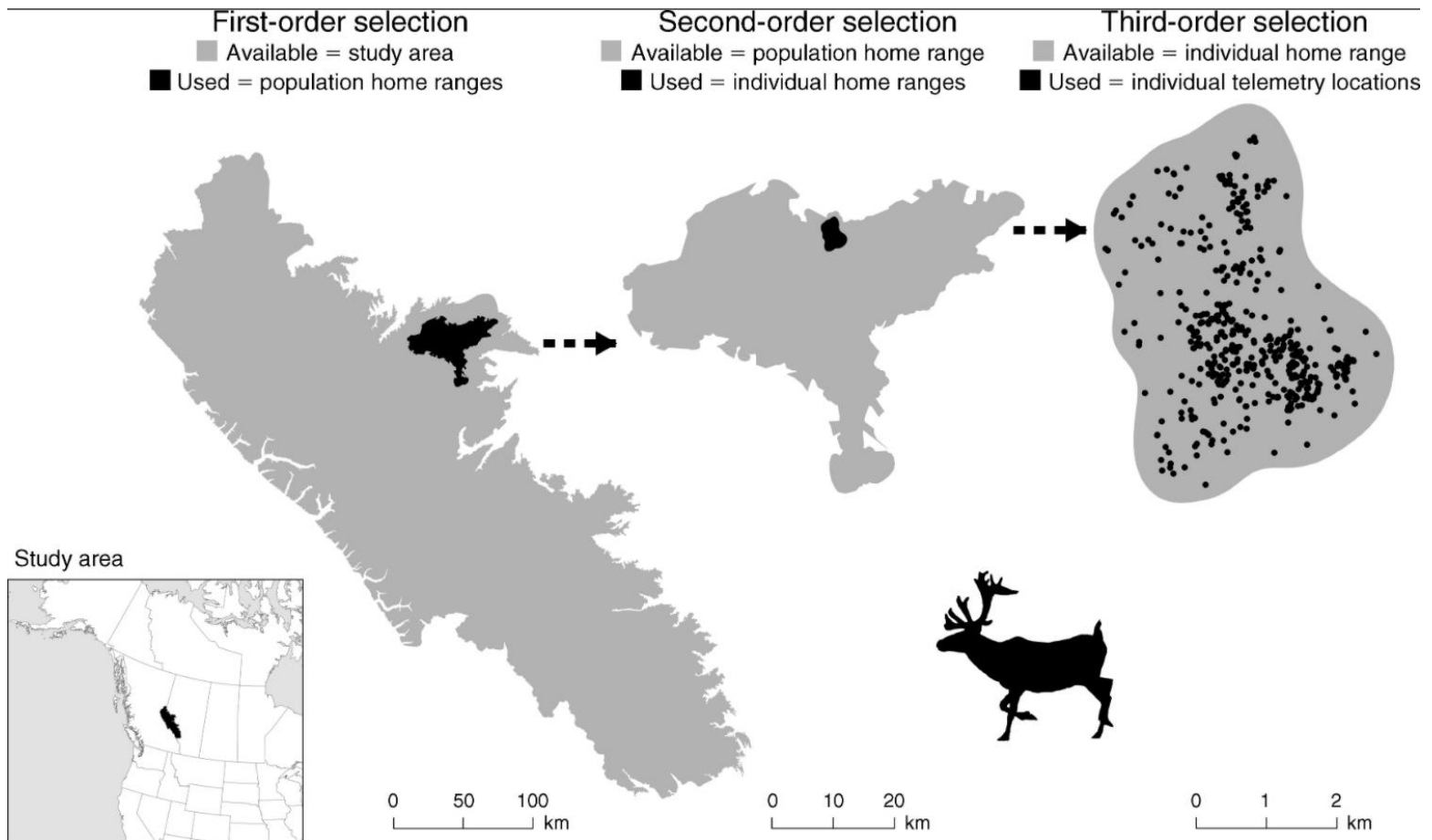
Positive response – Animal prefers habitat type

Negative response – Animal avoids habitat type



# Orders of Selection – Different scales of inference

Johnson, D. 1980. The comparison of usage and availability measurements for evaluating resource preference. *Ecology* 61:65-71.



Fourth-order selection: Local habitat selection (individual steps)

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Transcending scale dependence in identifying habitat with resource selection functions

Nicholas J. DeCesare, Mark Hebblewhite, Fiona Schmiegelow, David Hervieux, Gregory J. McDermid, Lalenia Neufeld, Mark Bradley, Jesse Whittington, Kirby G. Smith, [Luigi E. Morgantini](#), Matthew Wheatley, Marco Musiani ... See fewer authors



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# Use v. Availability Study Design

## Logistic regression analysis

$$w(x, \beta) = \exp(\beta_1 x_1 + \beta_2 x_2 + \dots + \beta_n x_n)$$

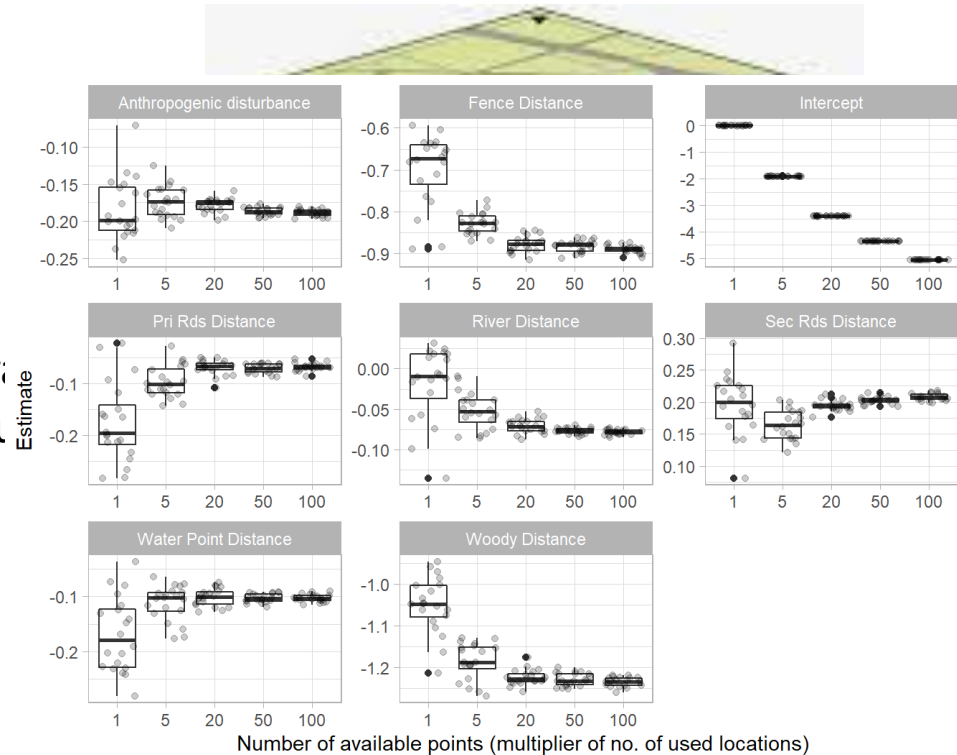
Habitat covariates:

Anthropogenic Footprint

Distance to landscape features (Roads, Rivers)

.....

Extr:  
Valu



Simulation analysis required to estimate  
required availability sample



# Use v. Availability Study Design

Logistic regression analysis

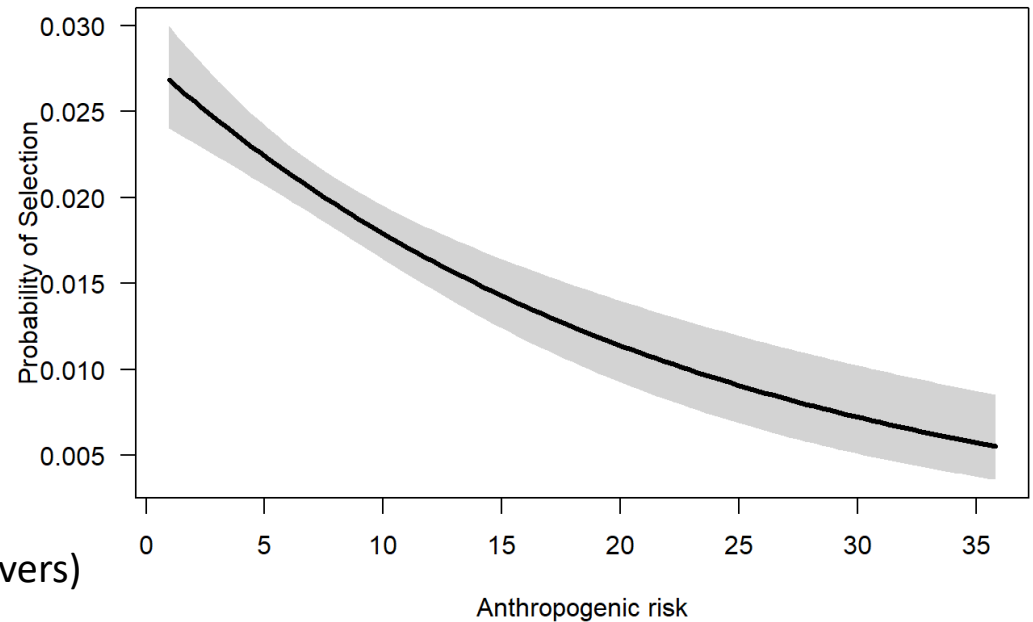
$$w(x, \beta) = \exp(\beta_1 x_1 + \beta_2 x_2 + \dots + \beta_n x_n)$$

Habitat covariates:

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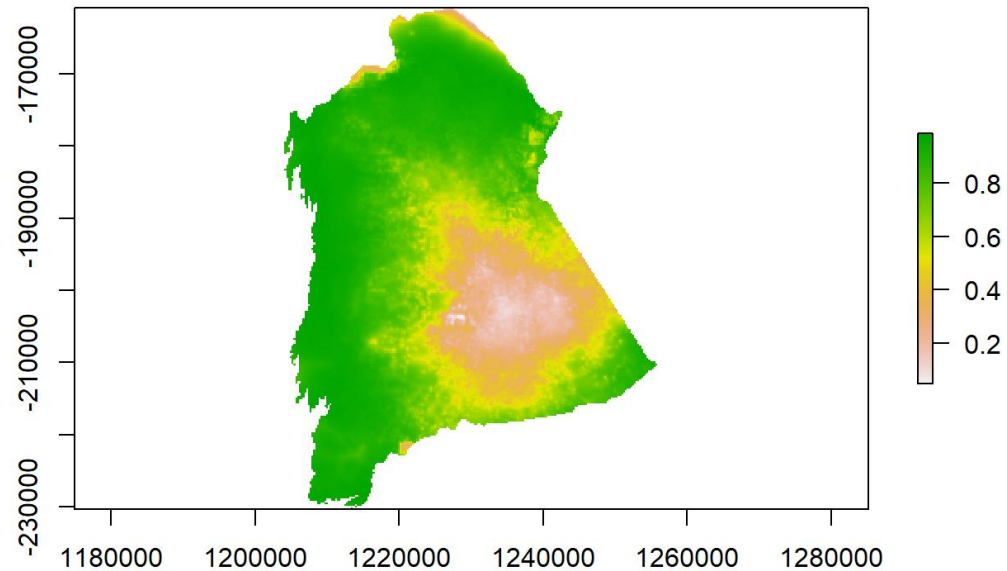
Distance to landscape features (Roads, Rivers)

.....



# Resource Selection Function Analysis

Final Prediction



For more information:

## **A 'How-to' Guide for Interpreting Parameters in Habitat-Selection Analyses**

 John Fieberg,  Johannes Signer, Brian Smith,  Tal Avgar

**doi:** <https://doi.org/10.1101/2020.11.12.379834>

This article is a preprint and has not been certified by peer review [what does this mean?].



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