Extensions to 'classical' iSSF

Johannes Signer

January 2024





Topics

What to do, if we

- have different behavioral modes
- have irregular sampling rates
- have multiple animals
- want to account for non-linear relationships
- want to account for spatial autocorrelation in covariates
- want to include memory
- . .

Multiple individuals

Why should we care?

- Most telemetry studies have data from many animals.
- Often individual behave very different (and we can fully account for these differences in a model).
- We are often interested in population-level effects (i.e., how would an average animal behave).

How-to account for individual differences

- 1. Ignore individuals and fit data to all animals.
- 2. Fit an individual model for each individual.
- 3. Use a mixed-model strategy.

2. Fit an individual model for each individual.

- A somewhat naive approach could be, to fit to each individual animal the model of interests (e.g., a SSF or an iSSF).
- In a next step we can then "do statistics" with the coefficients of the individual model. For example, we could
 - calculate the mean and confidence intervals to obtain population level effects, or
 - use a linear models to relate coefficient values to other explanatory covariates.
- A difficulty is if we have extreme observations or some levels of a categorical covariate is not observed for all animals.

There are different programming strategies, how one could approach such a situation:

- a. Write customized code for each individual.
- b. Use some kind of looping structure (for example a for-loop).
- c. Use a nest-unnest approach, as we have seen previously (for example with the purr package).

An example of this approach was used in Signer et al. 2019

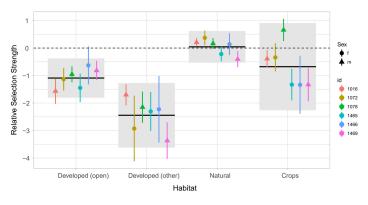


Figure 1: Source Signer et al. 2019

3. Use a mixed-model strategy.

- For HSF this is relatively straight forward. We can make use of well established tools that were developed for GLMMs.
- For iSSFs this is slightly more challenging. We have to use a likelihood equivalent reformulation of the iSSF as a poisson regression with random effects for each strata with a fixed large variance.

Random effects for HSFs

• Random effects were proposed for HSFs over 15 years ago²

 $^{^2}$ Gillies et al. "Application of random effects to the study of resource selection by animals." Journal of Animal Ecology 75.4 (2006): 887-898.

Random effects for HSFs

Random effects were proposed for HSFs over 15 years ago²

Application of random effects to the study of resource selection by animals CS Gillies, M Hebblewhite. SE Nelsen... - Journal of Animal ..., 2006 - Wiley Online Library) RSF models to those with random effects for the intercept, categorical ... of random effects in this empirical example, we simulated data for three common scenarios where random effects ... \$\frac{1}{2}\$ Save \$\sqrt{9}\$ Cite Cited by 78 Pealted articles All 22 versions

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 Majority of studies between 2016 and 2020 (80 %) only include random intercept and no random slope(s).

 $^{^2}$ Gillies et al. "Application of random effects to the study of resource selection by animals." Journal of Animal Ecology 75.4 (2006): 887-898.

Muff et al. 2020 had another look at this issue and extended this also to iSSF.



Accounting for individual-specific variation in habitat-selection studies: Efficient estimation of mixed-effects models using Bayesian or frequentist computation

Stefanie Muff X, Johannes Signer, John Fieberg X

First published: 27 August 2019 | https://doi.org/10.1111/1365-2656.13087 | Citations: 88

A case study for HSF/RSF

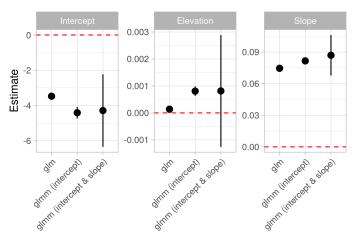
- Data on habitat selection of Mountain Goats³
- Generalized linear model with binomial response (GLM), random intercept (GLMM 1), and random intercept and slopes (GLMM 2).

 $^{^3}$ Lele & Keim, (2006) Weighted distributions and estimation of resource selection probability functions. Ecology 87, 3021–3028.

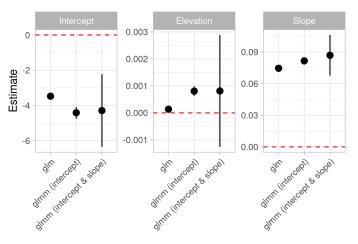
Let us fit three models to tracking data from wild goats:

```
# This is a naive approach (ignoring different animals)
m1 <- glmmTMB(STATUS ~ ELEVATION + SLOPE,
              data = goats, family = binomial())
# This is the random intercept model
m2 <- glmmTMB(STATUS ~ ELEVATION + SLOPE + (1 | ID),
              data = goats, family = binomial())
# This is a random slope and intercept model
m3 <- glmmTMB(STATUS ~ ELEVATION + SLOPE +
                (ELEVATION + SLOPE | ID),
              data = goats, family = binomial())
```

Comparing the model coefficients:



Comparing the model coefficients:



For RSF use random intercept and random slope(s) 4 .

Accounting for animal-specific variation (SSF)

Conditional logistic regression with random effects is more difficult

- The conditional logistic regression is a special case of the multinomial model.
- The multinomial model is likelihood-equivalent to the Poisson model.
- Thus we can rewrite the conditional logistic regression as a Poisson regression.

SSF as poisson model

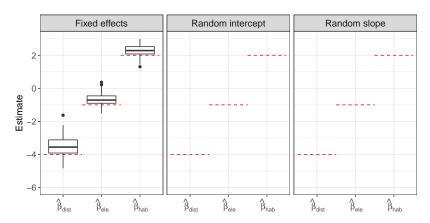
Reformulation as Poisson model⁵ ⁶

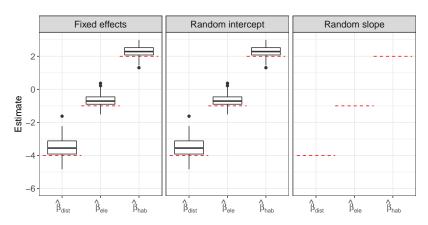
$$\mathsf{E}(y_{nti}) = \mu_{nti} = \exp(\alpha_{nt} + \boldsymbol{\beta}^{\top} \boldsymbol{x}_{nti} + \boldsymbol{u}^{\top} \boldsymbol{z}_{nti}) \;, \quad y_{nti} \sim \mathsf{Po}(\mu_{nti})$$

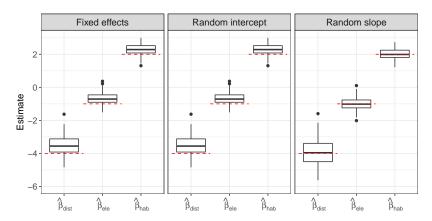
- $\alpha_{nt} \sim N(0, \sigma_{\alpha}^2)$ are the stratum specific intercepts with σ_{α}^2 being fixed at a very large value.
- $\boldsymbol{\beta}^{\top} \boldsymbol{x}_{nti}$ are the selection coefficients and the design matrix, respectively.
- $u^{\top} z_{nti}$ specify the random effect structure.

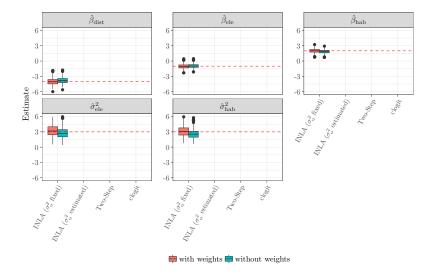
Simulation study from Muff et al. 2020

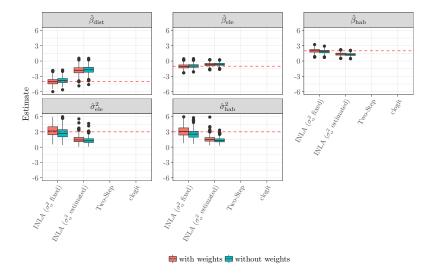
- Simulation of movement for 20 animals with animal-specific selection coefficients.
- For RSFs sample random points within the availability domain

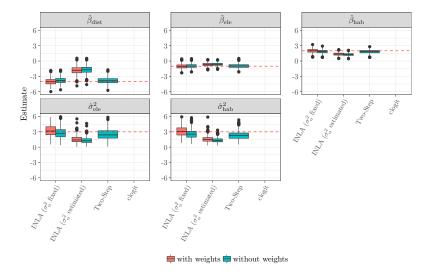


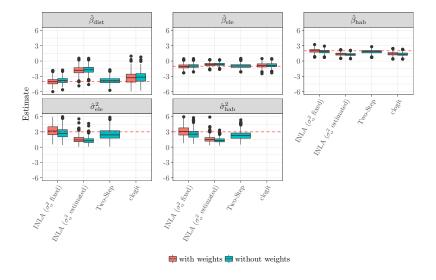












Software to fit these models

- HSF/RSF:
 - Any standard software package that can fit GLMMs is suitable.
- iSSF:
 - Frequentist: In R the package glmmTMB can be use, because it allows to fix the variance of random effects.
 - Muff et al. 2020 primarily used a Bayesian approach (INLA), as it straightforward to fix the variance.

An example from Roshier et al. 2021

- This work I did together with David Roshier from Australian Wildlife Conservancy.
- I was only involved in the statistical modeling.

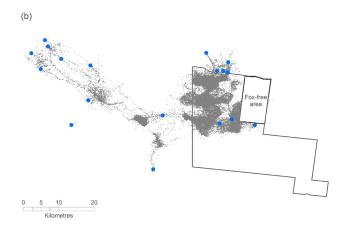


Visitation of artificial watering points by the red fox (Vulpes vulpes) in semiarid Australia

David A. Roshier^{1,2} Johannes Signer³ Andrew Carter^{1,4}

Data

- Telemetry data for 22 individual foxes at 20 min sampling rate
- Location of water sources (in blue)



Analysis

- 1. Revisitation rates with the recurse package¹
- 2. aKDE home ranges
- 3. integrated step selection analysis

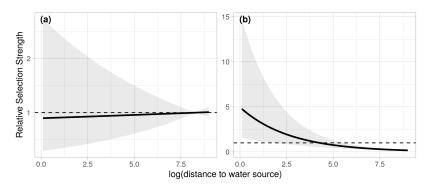
¹Bracis, C., Bildstein, K. L., & Mueller, T. (2018). Revisitation analysis uncovers spatio-temporal patterns in animal movement data. Ecography, 41(11), 1801-1811.

Questions: iSSA

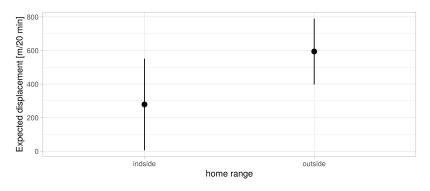
- Do foxes select for pixels closer to water sources?
- Does the selection depends on whether or not foxes are inside their home range?
- Do foxes move faster/slower when inside/outside their home range?

Habitat selection

Habitat selection of foxes inside (a) and outside (b) of their home range.



Expected displacement (i.e., how far do we expect a 'typical' fox to travel within a 20 minutes).



Different behavioral modes

We often want to account for different behavioral modes in habitat selection and movement:



RESEARCH ARTICLE © Open Access © 🚱

Accounting for behaviour in fine-scale habitat selection: A case study highlighting methodological intricacies

Different behavioral modes

We often want to account for different behavioral modes in habitat selection and movement:



Behavior-specific habitat selection by African lions may promote their persistence in a human-dominated landscape

Justin P. Suraci M., Laurence G. Frank, Alayne Oriol-Cotterill, Steve Ekwanga, Terrie M. Williams, Christopher C. Wilmers

First published: 03 February 2019 | https://doi.org/10.1002/ecy.2644 | Citations: 64

Different behavioral modes

We often want to account for different behavioral modes in habitat selection and movement:









Behavioural state-dependent habitat selection and implications for animal translocations

Simona Picardi XI, Peter Coates, Jesse Kolar, Shawn O'Neil, Steven Mathews, David Dahlgren

First published: 12 November 2021 | https://doi.org/10.1111/1365-2664.14080 | Citations: 10

An integrated approach



September 2017

A multi-state conditional logistic regression model for the analysis of animal movement

Aurélien Nicosia, Thierry Duchesne, Louis-Paul Rivest, Daniel Fortin

Ann. Appl. Stat. 11(3): 1537-1560 (September 2017). DOI: 10.1214/17-AOAS1045

ABOUT	FIRST PAGE	CITED BY	REFERENCES	SUPPLEMENTAL CONTENT	
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An integrated approach

Methodology | Open access | Published: 03 June 2023

Flexible hidden Markov models for behaviourdependent habitat selection

N. J. Klappstein [™], L. Thomas & T. Michelot

Movement Ecology 11, Article number: 30 (2023) | Cite this article

2447 Accesses 2 Citations 52 Altmetric Metrics

An integrated approach

How to account for behavioural states in step-selection analysis: a model comparison

J. Pohle¹*, J. Signer², J. A. Eccard³, M. Dammhahn⁴ and U. E. Schlägel¹

¹Institute of Biochemistry and Biology, University of Potsdam, Potsdam, Germany ²Wildlife Sciences; Faculty of Forest Sciences and Forest Ecology; University of Goettingen, Göttingen, Germany

³Animal Ecology, University of Potsdam, Potsdam, Germany

 $^{\rm 4}$ Behavioural Biology, University of Münster, Münster, Germany

Abstract

The idea is to simultaneously fit a Markov-switching variant of the conditional logistic regression:

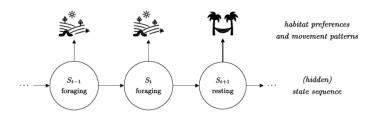
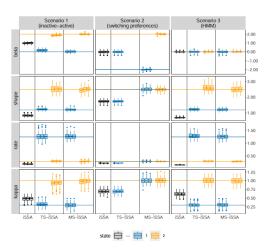


Figure 2: From Pohle et al. 2023

A simulation study was done by Pohle et al. 2023



Irregular sampling rates

In a recent preprint Hofmann et al. compared 4 approaches:





New Results Follow this preprint Methods for Implementing Integrated Step-Selection Functions with **Incomplete Data** David D. Hofmann, D Gabriele Cozzi, D John Fieberg doi: https://doi.org/10.1101/2023.11.08.566194 This article is a preprint and has not been certified by peer review [what does this mean?]

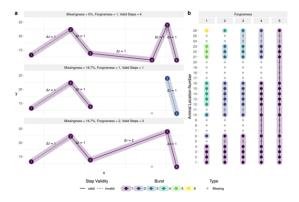


- 1. Imputation using the crawl package.
- 2. Naive: Following Munden et al. 2021
- 3. Dynamic+Model: Sample random steps from different tentative distributions.
- 4. Multistep: Using multiples of initial step length

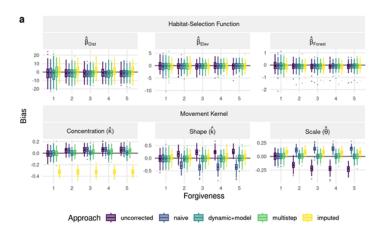
David D. Hofmann, G Gabriele Cozzi, G John Fieberg doi: https://doi.org/10.1101/2023.11.08.566194

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

Forgiveness: The maximum step-duration, measured in multiples of the regular step-duration, a modeler is willing to include in the step-selection analysis.

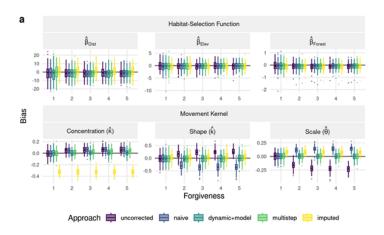


They then did a simulation study and compared the four approaches:



Account for non-linear relationships

They then did a simulation study and compared the four approaches:



Account for non-linear relationships



Key resources/publications

- Muff, S., Signer, J., & Fieberg, J. (2020). Accounting for individual-specific variation in habitat-selection studies: Efficient estimation of mixed-effects models using Bayesian or frequentist computation. Journal of Animal Ecology, 89(1), 80-92.
- Schielzeth, H., & Forstmeier, W. (2009). Conclusions beyond support: overconfident estimates in mixed models. Behavioral ecology, 20(2), 416-420.