

Modelação Ecológica

AULA 23

4th December 2019



Kiirsti Owen  · 2 days ago · 3 min read



THE 25 DAYS OF CHRISTMAS: AN R ADVENT CALENDAR

Updated: 12 hours ago



[MARMAM] PhD opportunities at the University of St Andrews



MARMAM <marmam-bounces@lists.uvic.ca> on behalf of Sonja Heir
To 'marmam@lists.uvic.ca'


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Tue 11/26/2019 3:04 PM

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 ATT00001.txt
423 bytes

We have several exciting and funded* PhD opportunities available at the University of St Andrews, UK – please see the School of Biology website for details (PhD projects at the SOI/SMRU):

<https://synergy.st-andrews.ac.uk/research/phd-study/phd-study-projects/phd-study-soi-projects/>

Please note different deadlines & funding opportunities – some closing very soon!

01 December 2019 (*funded for students **worldwide**):

- **The seasonal ocean dynamics of the Amundsen Sea Embayment (using telemetry & oceanographic data from seal-born sensors)– Supervisor: Dr Lars Boehme**
- Impacts of fishing-induced changes in forage fish school structure on African penguin foraging – Supervisor: Prof Andy Brierley

13 December 2019 (*funded for UK/EU students):

- A lab on a chip: using nano-plasmonics tongues for building miniaturized ecosystem sensors (SUPER DTP)- Supervisor: Dr Lars Boehme
-

06 January 2020 (*funded for UK/EU students)

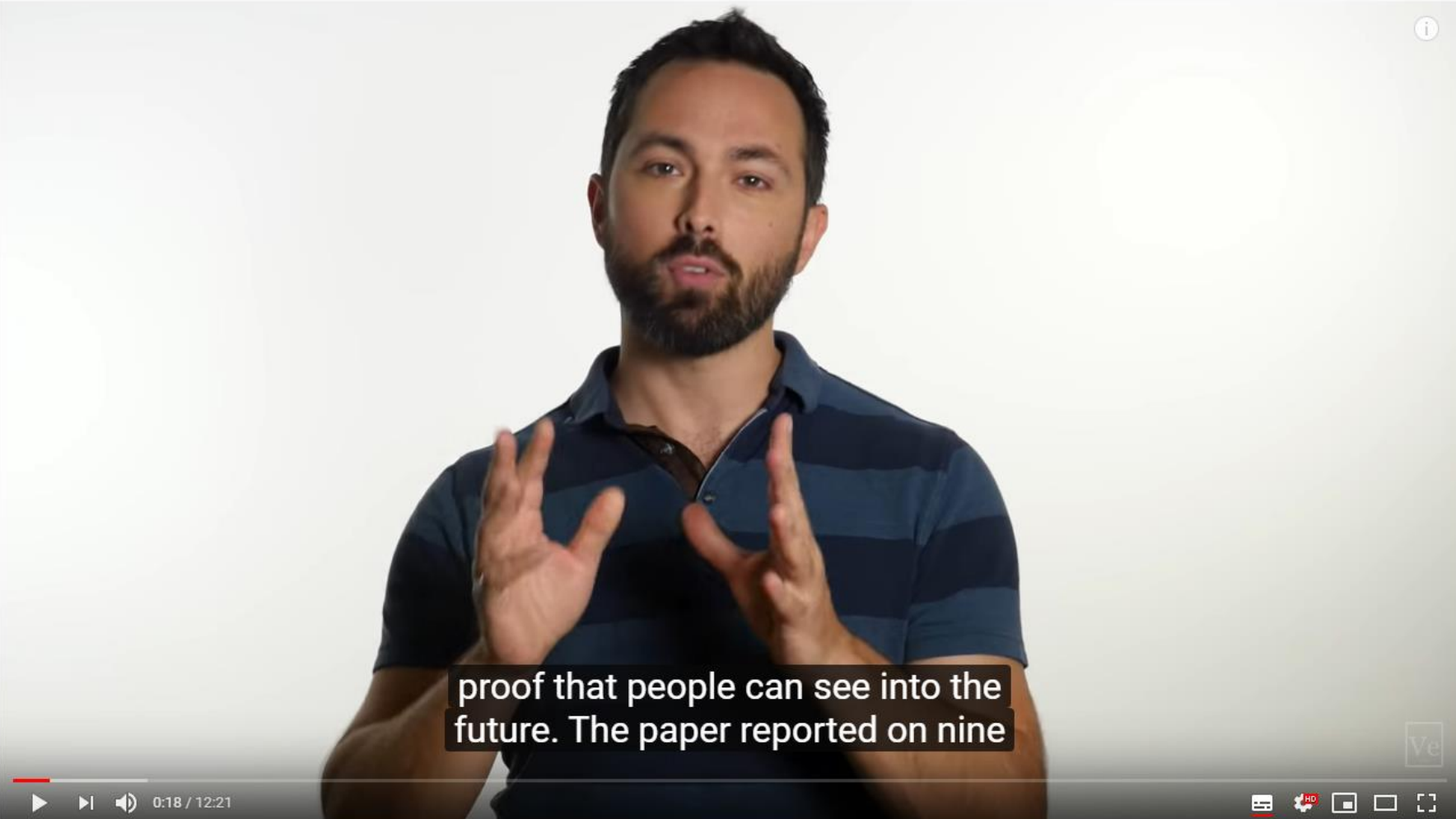
- Killer whale predation of harbour seals in the coastal waters of Scotland: investigating the ecological drivers and consequences of an apex predator-prey interaction – Supervisors: Dr Gordon Hastie & Dr Saana Isojunno

Posted on behalf of my colleagues.

Best wishes from Scotland,

Sonja

Dr Sonja Heinrich
Sea Mammal Research Unit
School of Biology
University of St Andrews
Scotland, UK



Is Most Published Research Wrong?

2,158,496 views • 11 Aug 2016

82K 1K SHARE SAVE ...

<https://www.youtube.com/watch?v=42QuXLucH3Q>

COMMENT · 27 NOVEMBER 2019

Climate tipping points – too risky to bet against

The growing threat of abrupt and irreversible climate changes must compel political and economic action on emissions.

Timothy M. Lenton , Johan Rockström, Owen Gaffney, Stefan Rahmstorf, Katherine Richardson, Will Steffen & Hans Joachim Schellnhuber





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Generalized Additive Models and Mixed-Effects in Agriculture

July 15, 2017

By Fabio Veronesi

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[This article was first published on R tutorial for Spatial Statistics, and kindly contributed to R-bloggers]. (You can report issue about the content on this page here)

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Introduction

In the previous post I explored the use of linear model in the forms most commonly used in agricultural research.

Clearly, when we are talking about linear models we are implicitly assuming that all relations between the dependent variable y and the predictors x are linear. In fact, in a linear model we could specify different shapes for the relation between y and x , for example by including polynomials (read for example: <https://datascienceplus.com/fitting-polynomial-regression-r/>). However, we can do that only in cases where we can clearly see a particular shape of the relation, for example quadratic. The problem is in many cases we can see from a scatterplot that we have a non-linear distribution of the points, but it is difficult to understand its form. Moreover, in a linear model the interpretation of polynomial coefficients become more difficult and this may decrease their usefulness.

An alternative approach is provided by Generalized Addictive Models, which allows us to fit models with non-linear smoothers without specifying a particular shape a priori.

<https://www.r-bloggers.com/generalized-addictive-models-and-mixed-effects-in-agriculture/>

Re: a suggestion in your 2009 book and a quick question on top



Highland Statistics Ltd <highstat@highstat.com>

To Tiago Marques

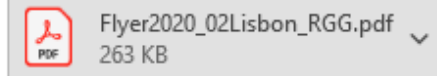
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Thu 11/28/2019 8:21 PM



On 28/11/2019 19:46, Tiago Marques wrote:

Hi Alain,

I hope all is well at your end. I guess you won't remember, but we have met a few years ago when I attended a course of yours in Lisboa. Any way, I am just writing with a suggestion for an improvement in the book, which I have been using as a resource in my Ecological Modelling classes.

In page 317, to obtain a required factor, you use the entire page with a quite complicated piece of code that you then try to simplify, but even the simplification is not straightforward. The result of all that code can be obtained with a single simple line of code:

```
Owls$NestNight=as.factor(paste0(Owls$Nest, ".", substr(Owls$FoodTreatment,1,3)))
```

which avoids distracting the reader from the fundamental.

Hope you find it helpful. I take the chance to ask, I must be confusing something, but once you use the above code there are now 54 clusters (27 nests times 2 nights) and the largest number of observations is 28, for the Oleyes.Sat nest night combination. This will be the variable to define the grouping id in the GEE. Where does the output from the summary of the geeglm function output "Number of clusters: 277 Maximum cluster size: 18" shown in page 319 comes from –which I get in my computer too. I seem to be missing something rather obvious, as I was expecting "Number of clusters: 54 Maximum cluster size: 28", but I am not sure what! Many thanks in advance for any feedback sent this way.

Cheers,


Tiago

Re: a suggestion in your 2009 book and a quick question on top



Highland Statistics Ltd <highstat@highstat.com>

To Tiago Marques

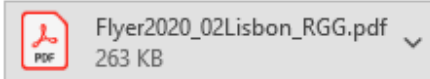
 Reply

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Thu 11/28/2019 8:21 PM



Dear Tiago,

Thank you for your email..and for your suggestions.

The thing with GEE is that I haven't used it since 2009. Since then I went more into GLMMs...and since the last 3 or 4 years more towards INLA. **So..I don't think I'm the most appropriate person to ask.** Maybe create a dummy data set and ask the programmers?

Maybe you could also have a look at the GEE chapter in the same book? That was something with badgers. Perhaps it is due to the coding of the ID? What happens if it is coded as a numerical variable? Shouldn't make a difference...but you never know.

Kind regards,

Alain

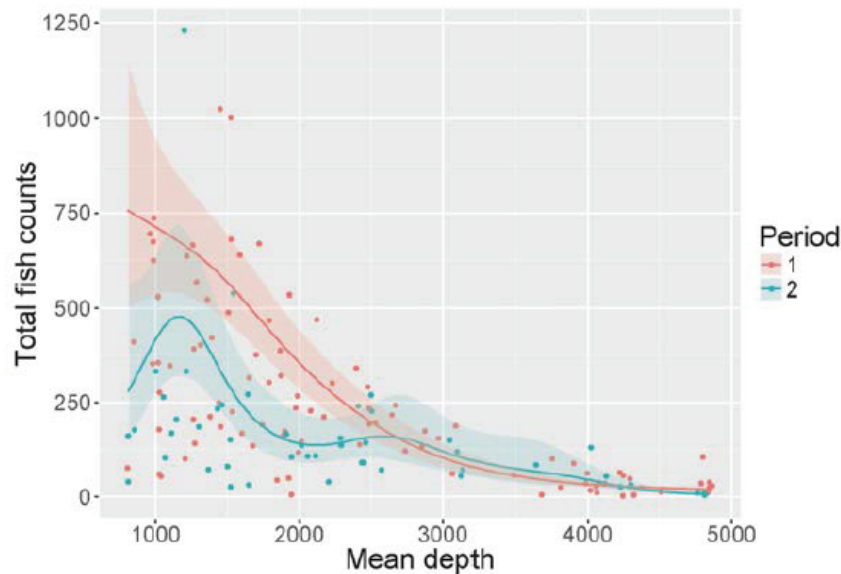
PS...I see that you also have contacts in Lisboa. Can you forward the attached flyer to any potential interested colleagues?

Data Exploration, Regression, GLM & GAM with introduction to R

Provided by: **Highland Statistics Ltd**

In cooperation with:

cE3c – eChanges, Faculty of Sciences, University of Lisbon, Portugal



We begin with an introduction to R and provide a protocol for data exploration to avoid common statistical problems. We will discuss how to detect outliers, deal with collinearity and transformations.

Date & Venue

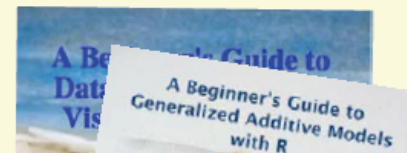
Date: 3 - 7 February 2020

Venue: PT Meeting Centre,
R. Bojador 47, Parque das
Nações, Lisbon, Portugal

Price: £500

Instructors: Dr. Alain Zuur
Dr. Elena Ieno

Authors of 11 books and
providers of over 150 courses



REGISTRATION

www.highstat.com

Dr Alain F Zuur
highstat@highstat.com
www.highstat.com

RE: question on geopack output



Tiago Marques
To Søren Højsgaard

Reply	Reply All	Forward	
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Wed 12/4/2019 11:06 AM

Fra: Tiago Marques <tiago.marques@st-andrews.ac.uk>
Sendt: 30. november 2019 20:04:21
Til: Søren Højsgaard
Emne: question on geopack output

Dear Dr. Søren Højsgaard,

I am using the geopack and it's function glmgee. While reproducing an analysis in the book by Zuur et al 2009, I realized there's a bit of the output that is not sensible to me (also shown in the book in fact). I attach here a short RMarkdown report that ends up with the question: the output lists as the number of clusters and the maximum cluster size something that is not what it should be – at least the way I see it. [Can you possibly explain what I might be doing wrong?](#)

All the best,

From: Søren Højsgaard <sorenh@math.aau.dk>
Sent: Monday, December 2, 2019 9:01 AM
To: Tiago Marques <tiago.marques@st-andrews.ac.uk>
Subject: Sv: question on geopack output

I can't possibly reproduce / comment on anything you send me without the data.

Tiago

You know: "A minimal reproducible example"...

RE: question on geopack output



Tiago Marques
To Søren Højsgaard

y	Reply All	Forward	
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Wed 12/4/2019 11:06 AM

Hi Soren,

Any feedback on this? I have a class this afternoon on GEE's and I was hoping I'd be able to tell the students something about what I have asked you. [Why are the number of groups \(and maximum number of observations per group\) in the GEE output not what I'd expect.](#)

[Am I doing something stupid, or is the software doing something funny?](#)

Cheers

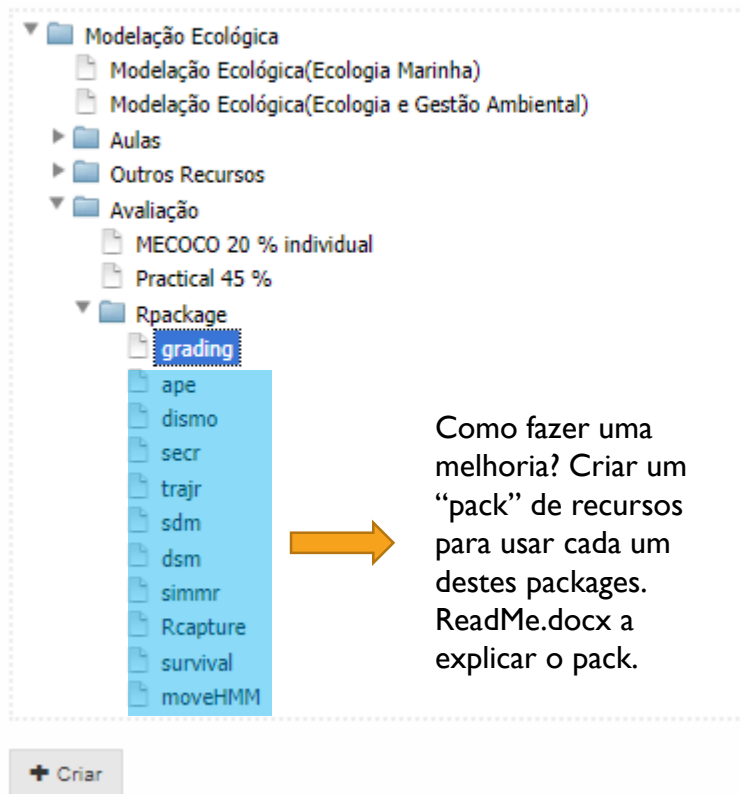
T

TODAY'S MENU

- Discussion on model averaging paper
 - Bespoke useful models in ecological statistics (capture recapture, occupancy, spatial point processes, distance sampling, hierarchical models)
 - Wrapping up GEEs - Modelling the owl data and the deer data
 - Modelling “your” grades data
 - Any other stuff you might want to discuss with me
-
- I will leave at 16:00, to compensate you from having spent an extra hour in class yesterday
 - Pure coincidence: I have to be at ISPA at 16:30 for a PhD defense on dolphin sound production

GRADING

Gestão de Páginas



- Modelação Ecológica
 - Modelação Ecológica(Ecologia Marinha)
 - Modelação Ecológica(Ecologia e Gestão Ambiental)
- Aulas
- Outros Recursos
- Avaliação
 - MECOCO 20 % individual
 - Practical 45 %
 - Rpackage
 - grading**
 - ape
 - dismo
 - secr
 - trajr
 - sdm
 - dsm
 - simmr
 - Rcapture
 - survival
 - moveHMM

+ Criar

Como fazer uma melhoria? Criar um “pack” de recursos para usar cada um destes packages. ReadMe.docx a explicar o pack.

grading

Página

Ficheiros 20

Permissões

Link

Título

grading

Conteúdo





Português (Portugal)

H B I S U x³ x₂ A

MODEL AVERAGING

Ecological Monographs, 0(0), 2018, pp. 1–20
© 2018 by the Ecological Society of America

Model averaging in ecology: a review of Bayesian, information-theoretic, and tactical approaches for predictive inference

CARSTEN F. DORMANN ^{1,22} JUSTIN M. CALABRESE ² GURUTZETA GUILLERA-ARROITA,³ ELENI MATECHOU,⁴ VOLKER BAHN,⁵
KAMIL BARTOŃ,⁶ COLIN M. BEALE,⁷ SIMONE CIUTI,^{1,8} JANE ELITH,³ KATHARINA GERSTNER,^{9,10} JÉRÔME GUELAT,¹¹ PETR KEIL,¹⁰
JOSÉ J. LAHOZ-MONFORT,³ LAURA J. POLLOCK,¹² BJÖRN REINEKING,^{13,14} DAVID R. ROBERTS ^{1,15} BORIS SCHRÖDER ^{16,17}
WILFRIED THUILLER,¹² DAVID I. WARTON,¹⁸ BRENDAN A. WINTLE,³ SIMON N. WOOD,¹⁹ RAFAEL O. WÜEST,^{12,20} AND FLORIAN HARTIG^{1,21}

Dealing with correlation
Random Effects, Mixed Models &
Generalized Estimating Equations

Wrap up –
Generalized Estimating Equations

LAST WEEK'S TASK - continued

Revisit two datasets from FT7b4ME 20 | 2019.pdf in “Aula 19”

7. Find a GLM that best fits the data “Owls.txt”, where you are trying to explain the begging behavior of owls offspring when the parents are absent from the nest. The variable “SiblingNegotiation” represents the number of calls produced by the chicks in the nest during a 30 second period, while “BroodSize” represents the size of the brood. More details about this data can be found in Zuur et al. 2009.

Account for variation over time in the same nest

8. The data “DeerEcervi.txt” contains the incidence of *E. cervi* parasites in deer pellets, and we have also the corresponding sex, length and farm the deer were on. How many farms were available? Ignore them for now, and model the presence/absence of parasites in pellets as a function of deer characteristics. This is a dataset also used by Zuur et al. 2009.

Account for variation across farms

```
> fmodelo <- formula(SiblingNegotiation~offset(LBroodSize)+FoodTreatment+ArrivalTime)
> mod1 <- geeglm(formula=fmodelo,data=owls,family = poisson,id = Nest, corstr = "ar1")
> summary(mod1)
```

```
Call:
geeglm(formula = fmodelo, family = poisson, data = owls, id = Nest,
        corstr = "ar1")
```

Coefficients:

	Estimate	Std.err	Wald	Pr(> W)	
(Intercept)	3.70322	0.66935	30.61	3.16e-08	***
FoodTreatmentSatiated	-0.56417	0.12254	21.20	4.15e-06	***
ArrivalTime	-0.12418	0.02691	21.30	3.93e-06	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Estimated Scale Parameters:

	Estimate	Std.err
(Intercept)	6.242	0.387

Correlation: Structure = ar1 Link = identity

Estimated Correlation Parameters:

	Estimate	Std.err
alpha	0.3854	0.0582

Number of clusters: 27 Maximum cluster size: 52

```
> sort(with(owls,tapply(BroodSize,Nest,length)))
```

Forel	Sevaz	Chevroux	GDLV	CorcellesFavres	Henniez
4	4	10	10	12	13
Gletterens	LesPlanches	Lully	Rueyes	Jeuss	Trey
15	17	17	17	19	19
ChEsard	Bochet	StAubin	Murist	Payerne	Franex
20	23	23	24	25	26
Seiry	Marnand	AutavauxTV	Lucens	Champmartin	Etrabloz
26	27	28	29	30	34
Yvonnand	Montet	Oleyes			
34	41	52			

```
> length(unique(owls$Nest))
```

```
[1] 27
```

```
> mod.by.nestnight <- geeglm(formula=fmodelo,data=owls,family = poisson,id=NestNight,corstr="ar1")
> summary(mod.by.nestnight)
```

Call:

```
geeglm(formula = fmodelo, family = poisson, data = owls, id = NestNight,
        corstr = "ar1")
```

Coefficients:

	Estimate	Std.err	Wald	Pr(> W)	
(Intercept)	3.593	0.668	28.9	7.6e-08	***
FoodTreatmentSatiated	-0.578	0.115	25.4	4.6e-07	***
ArrivalTime	-0.122	0.027	20.3	6.6e-06	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Estimated Scale Parameters:

	Estimate	Std.err
(Intercept)	6.64	0.524

Correlation: Structure = ar1 Link = identity

Estimated Correlation Parameters:

	Estimate	Std.err
alpha	0.517	0.0676

Number of clusters: 277 Maximum cluster size: 18



Just as in GLMs, non-normal responses and link functions get added to models with random effects

Review

Cell
PRESS

Generalized linear mixed models: a practical guide for ecology and evolution

Benjamin M. Bolker¹, Mollie E. Brooks¹, Connie J. Clark¹, Shane W. Geange², John R. Poulsen¹, M. Henry H. Stevens³ and Jada-Simone S. White¹

“... Despite the availability of accurate techniques for estimating GLMM parameters in simple cases, complex GLMMs are challenging to fit and statistical inference such as hypothesis testing remains difficult...”

“...GLMMs are surprisingly challenging to use even for statisticians. Although several software packages can handle GLMMs (Table 1), few ecologists and evolutionary biologists are aware of the range of options or of the possible pitfalls....”

<https://biologyforfun.wordpress.com/2014/03/12/generalized-linear-mixed-models-in-ecology-and-in-r/>

[Ecol Evol.](#) 2013 Sep; 3(9): 3141–3151.

Published online 2013 Aug 2. doi: [10.1002/ece3.707](https://doi.org/10.1002/ece3.707)

PMCID: [PMC3790557](#)

PMID: [24102000](#)

Generalized additive mixed models for disentangling long-term trends, local anomalies, and seasonality in fruit tree phenology

[Leo Polansky](#) and [Martha M Robbins](#)

► [Author information](#) ► [Article notes](#) ► [Copyright and License](#)

This article has been [cited by](#) other articles in PMC.

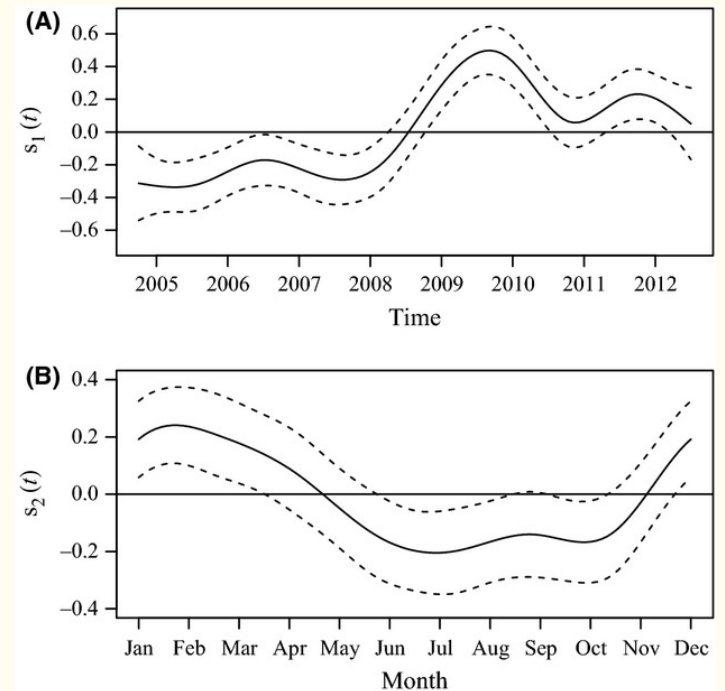


Figure 3

Smooth functions (solid lines) with 95% credible region delineated by the dashed lines estimated from the forest aggregated data for the (A) interannual smooth $s_1(t)$ and the (B) intraannual smooth function $s_2(t)$, both plotted on the linear scale.

Hierarchical generalized additive models in ecology: an introduction with mgcv

Bioinformatics tool

Ecology

Statistics

Data Science

Spatial and Geographic Information Science

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researchEric J. Pedersen^{1,2}, David L. Miller^{3,4}, Gavin L. Simpson^{5,6}, Noam Ross⁷

Published May 27, 2019

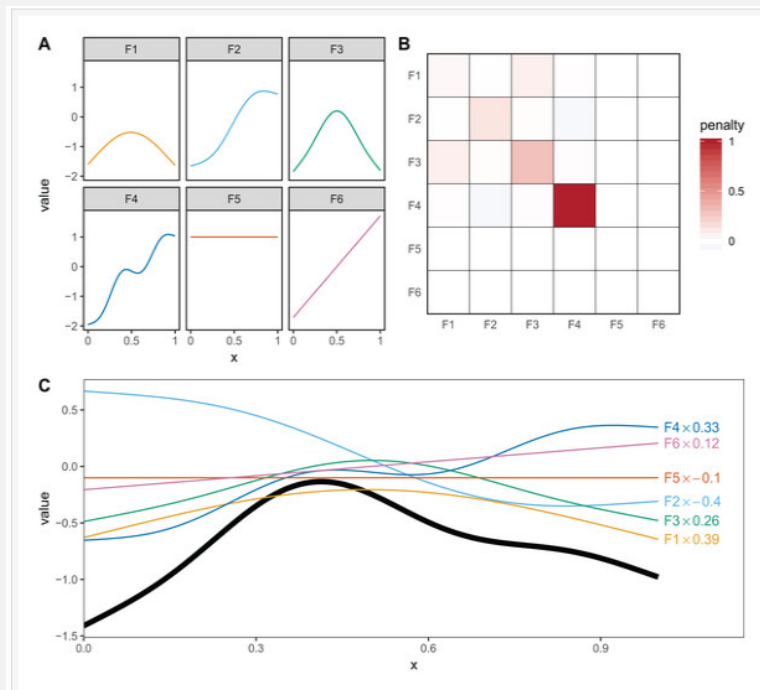
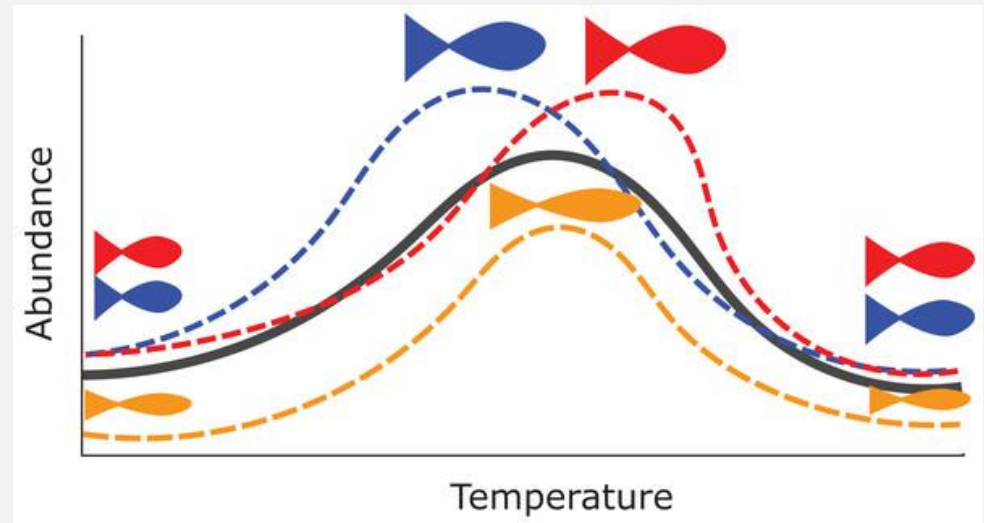


Figure 3: (A) Examples of the basis functions associated with a six basis function thin plate regression spline (TPRS, $m = 2$), calculated for data, x , spread evenly between $x = 0$ and $x = 1$.



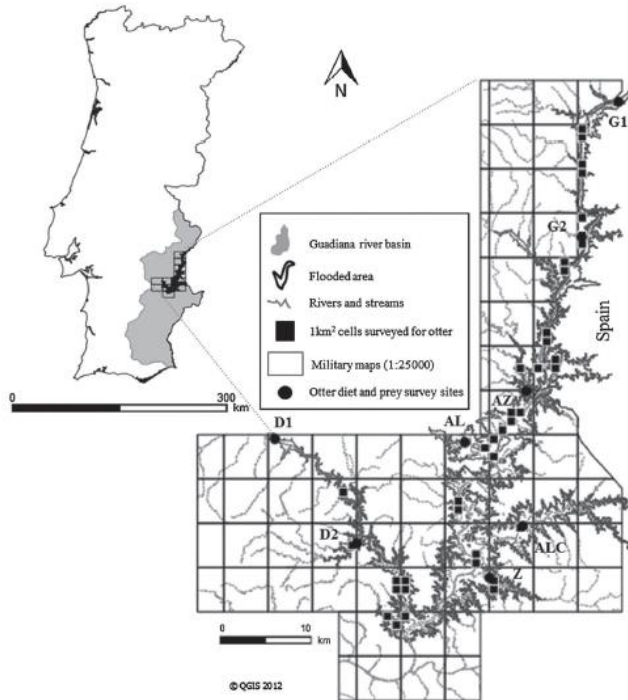


Figure 1. Location of the Alqueva Reservoir in southern Portugal, showing the 25 km² and the 1 km² (black squares) otter survey grid (circles) where otter diet and prey were assessed. Two sites (G1 and G2) were located in the main Guadiana River. Other sites were located in the Guadiana River: the Azeite stream (AZ), Alamo stream (AL), Degebe stream (D1 and D2), Alcarrache stream (ALC) and Zel...

To account for multiple surveys of the same location, a generalized additive mixed model (GAMM) regression framework was used (Wood, 2006). Presence/absence of otter signs was modelled by smoothing of trimer values, with the smoothness chosen by using the default generalized cross-validation procedure in the R mgcv library (Wood, 2006; R Development Core Team, 2011). Grid was included as a random effect and the residuals within sites were assumed to follow a first-order autoregressive model.

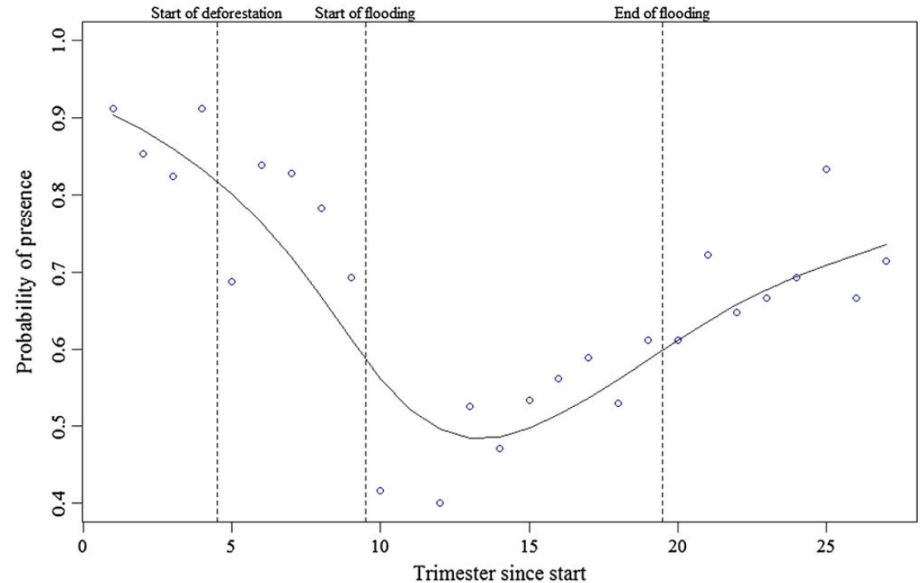
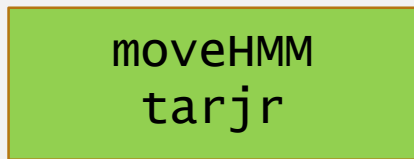


Figure 3. Probability of otter presence as a function of trimer in the flooded area of the Alqueva Reservoir. Data are represented by points and fitted model by a black line.

The theory R package work – a summary



Estimating abundance :
capture recapture vs. distance sampling



Animal movement:
modelling vs. descriptive



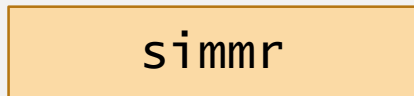
Survival analysis: just a very special GLM



Species distribution models, also just very special regression models



See also next class with Prof. Susana França



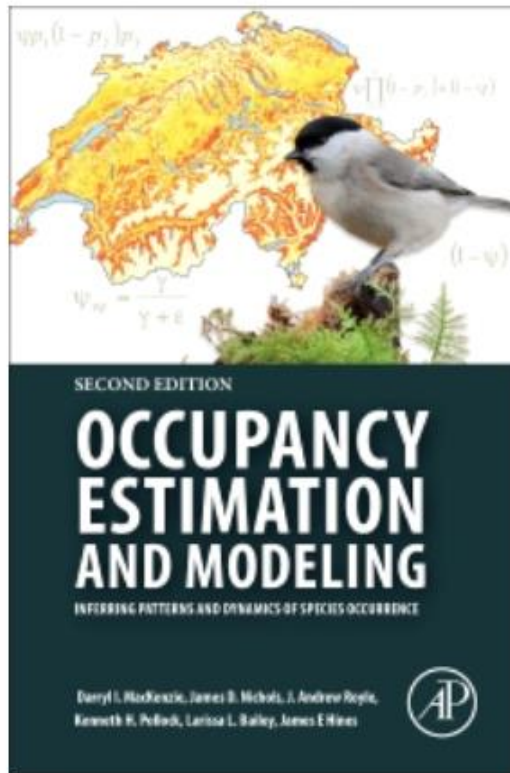
Stable Isotope Mixing Models – really, just a fancy regression with mixture models (also present in `moveHMM`)



Genetic data – just another clustering method

SPECIFIC YET STILL VERY
LARGE CLASSES OF
MODELS FOR
ECOLOGICAL DATA
(SO LARGE THAT EACH
OF THESE COULD BE A
FULL SEMESTER
COURSE!)

OCCUPANCY MODELLING



Occupancy Estimation and Modeling

2nd Edition

Inferring Patterns and Dynamics of Species Occurrence

Material blatantly stolen from

Using occupancy models to understand snow leopard distribution.

Darryl I. MacKenzie

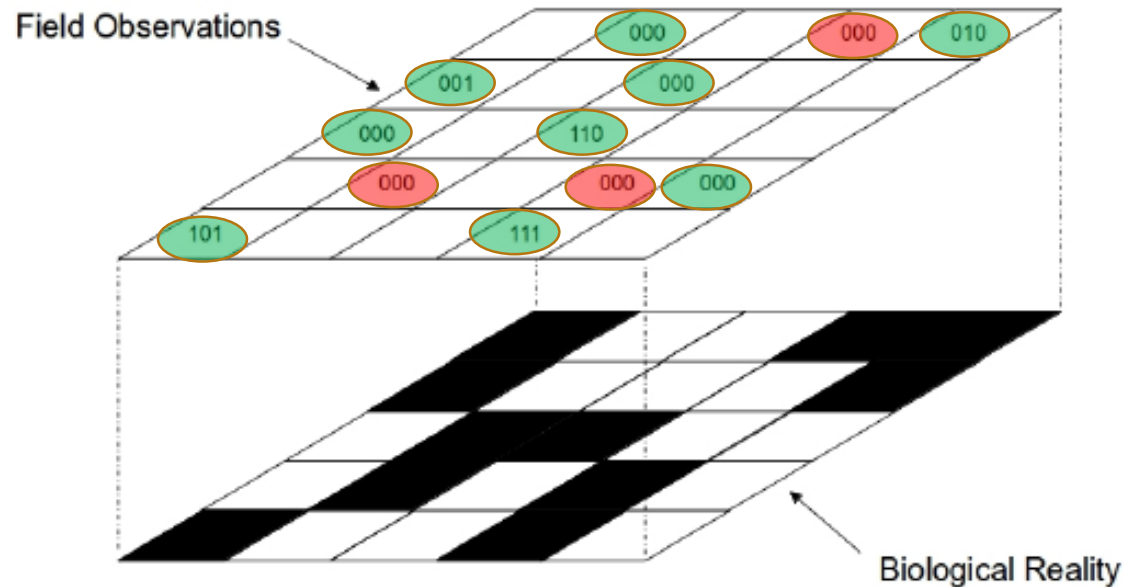


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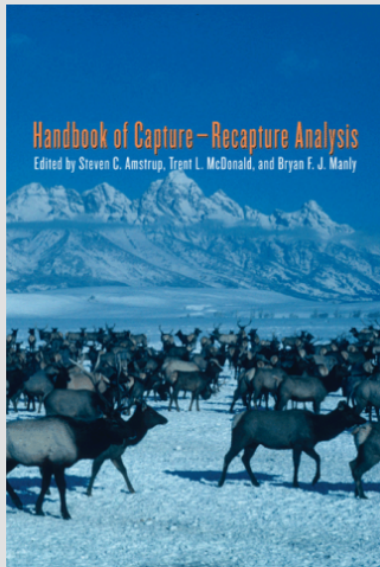
https://www.researchgate.net/publication/328417995_Using_occupancy_models_to_understand_snow_leopard_distribution



- Field observations are a combination of biological and observation processes.
- Important to account for observation process.
 - i.e., sampling and detection

- Many popular methods of analysis ignore imperfect detection.
 - e.g., logistic regression, MaxENT, random forests.
- Model where snow leopards were *found*.
- Occupancy models explicitly account for detection probability.
 - separate the observation and biological processes
- Require repeated surveys of each survey unit.
- Often have that information.
 - e.g., multiple camera-trap days, multiple segments along a transect.

CAPTURE RECAPTURE MODELS



Handbook of Capture-Recapture Analysis

Edited by Steven C. Amstrup, Trent L. McDonald & Bryan F. J. Manly

Editions

Paperback | 2005 | © 2006 | **\$90.00** | **£70.00** | ISBN: 9780691089683 | 336 pp. | 6 x 9 1/4 | 20 halftones, 6 line illus, 74 tables.
E-book | ISBN: 9781400837717

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A mais simples questão que se coloca quando pensamos numa espécie é: Quantos animais de uma determinada espécie existem?

Outras questões poderão ser, “o que comem?”, “que habitats preferem?”, “o que podemos fazer para ajudar na sua conservação, caso seja preciso?”, “quais as principais ameaças?”.

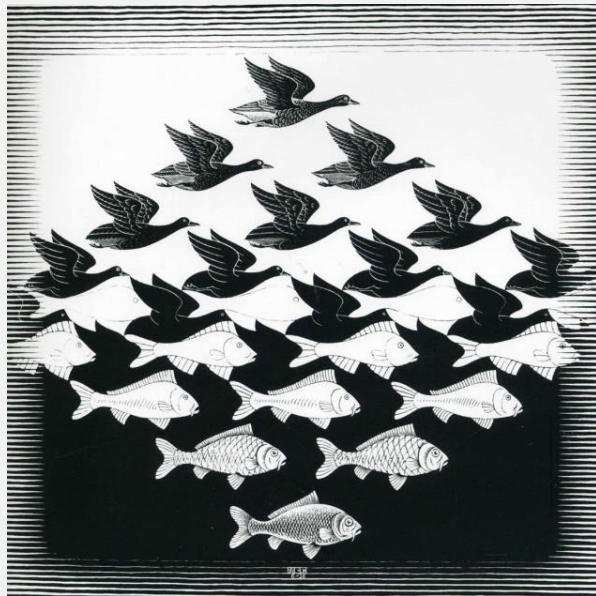
Para responder a estas perguntas é necessário usar modelos, pois a realidade é demasiado complexa para ser compreendida como um todo.



Como estimar a dimensão de uma população?

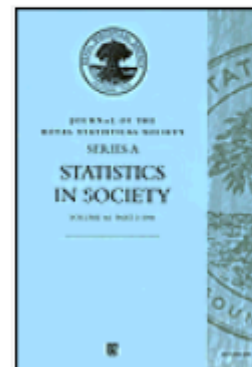
Uma área de ativa investigação em estatística é a da estimação do tamanho de populações naturais, que terá começado com Laplace quando este usou as ideias subjacentes aos métodos de captura-recaptura para estimar o tamanho da população de França, em 1783.

Em 1896, Carl G. J. Petersen foi o primeiro a empregar este método no estudo do fluxo migratório de peixes no mar Báltico. Em 1930, Frederick Lincoln utilizou-o para estimar o tamanho da população de patos selvagens da América do Norte. Nas décadas de 30 e 40 observou-se um consistente desenvolvimento teórico e aplicado do método. São frequentes estudos de captura-recaptura para estimar e monitorizar o tamanho das populações de várias espécies, tanto em populações fechadas como abertas.



Who Captures the Marks for the Petersen Estimator?

I. B. J. Goudie and M. Goudie



Journal of the Royal Statistical Society. Series A (Statistics in Society)
Vol. 170, No. 3 (2007), pp. 825-839 (15 pages)

Published by: [Wiley](http://www.wiley.com) for the [Royal Statistical Society](http://www.rstat.org).

O método utilizado por Laplace consistia em recolher uma primeira amostra da população (captura) de dimensão n_1 , “marcando” todos os indivíduos recolhidos e devolvendo-os de seguida à população. Posteriormente, recolhe-se uma segunda amostra (recaptura) de dimensão n_2 e conta-se o número de indivíduos marcados, n_3 . Como a segunda amostra é aleatória, a proporção de marcados na segunda amostra deve ser igual à proporção de marcados na população, donde se conclui que a dimensão estimada da população, N , é dada por:

$$\frac{n_3}{n_2} = \frac{n_1}{N} \Leftrightarrow N = \frac{n_1 \times n_2}{n_3}$$



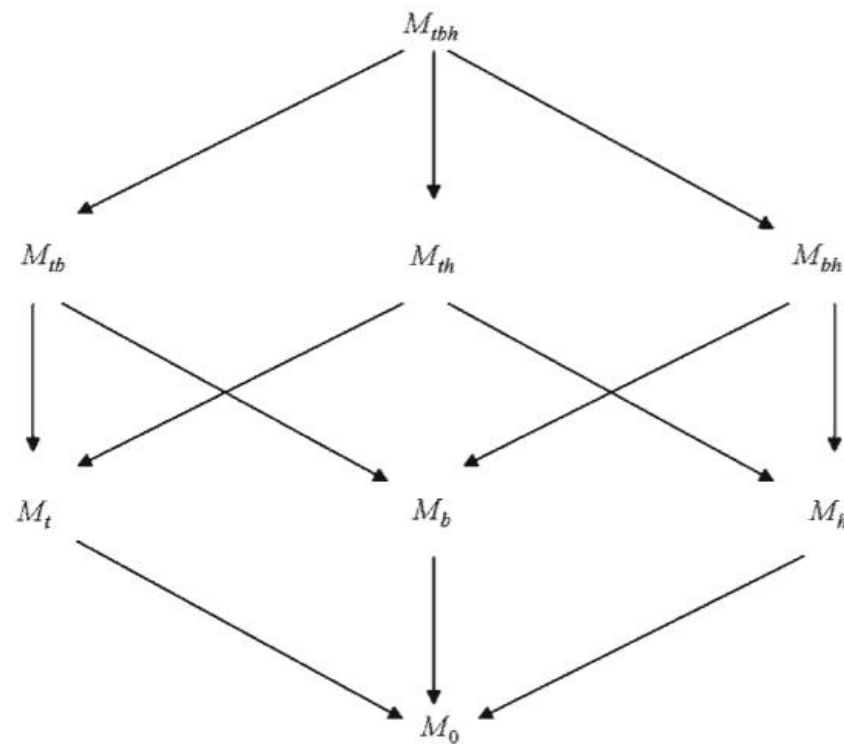


Fig. 1 The suite of eight models proposed by Otis et al. (1978) to allow for time variation (t), behavioral response to initial capture (b), and individual heterogeneity (h) of encounter probabilities. Models pointed to by arrows are nested within the model above

Closed population estimation models and their extensions in Program MARK

March 2008 · Environmental and Ecological Statistics 15(1):89-99

DOI: 10.1007/s10651-007-0030-3

Table 1 A three-occasion example, where only eight possible encounter histories are possible, is presented to illustrate the construction of encounter history probabilities considering only time and behavior variation, where p_i is initial capture probability on occasion i , and c_i is the recapture probability on occasion i (with no c_1)

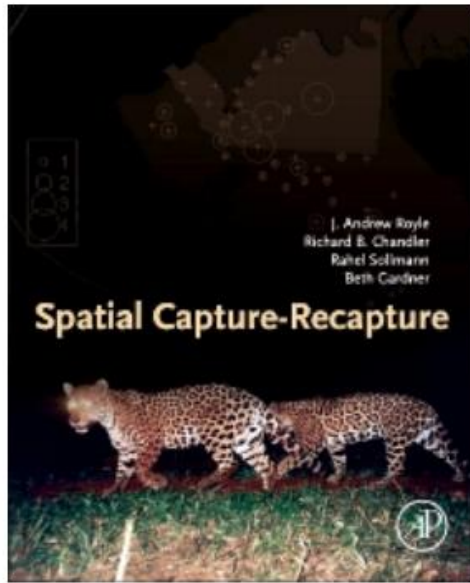
Encounter history (i)	Animals observed (X_i)	Probability (P_i)
100	X_{100}	$p_1(1 - c_2)(1 - c_3)$
010	X_{010}	$(1 - p_1) p_2(1 - c_3)$
001	X_{001}	$(1 - p_1)(1 - p_2) p_3$
110	X_{110}	$p_1 c_2(1 - c_3)$
101	X_{101}	$p_1(1 - c_2) c_3$
011	X_{011}	$(1 - p_1) p_2 c_3$
111	X_{111}	$p_1 c_2 c_3$
000	X_{000}	$(1 - p_1)(1 - p_2)(1 - p_3)$

These are never seen!

Unmodelled heterogeneity is a plague for conventional capture recapture !

And N is typically ill defined...

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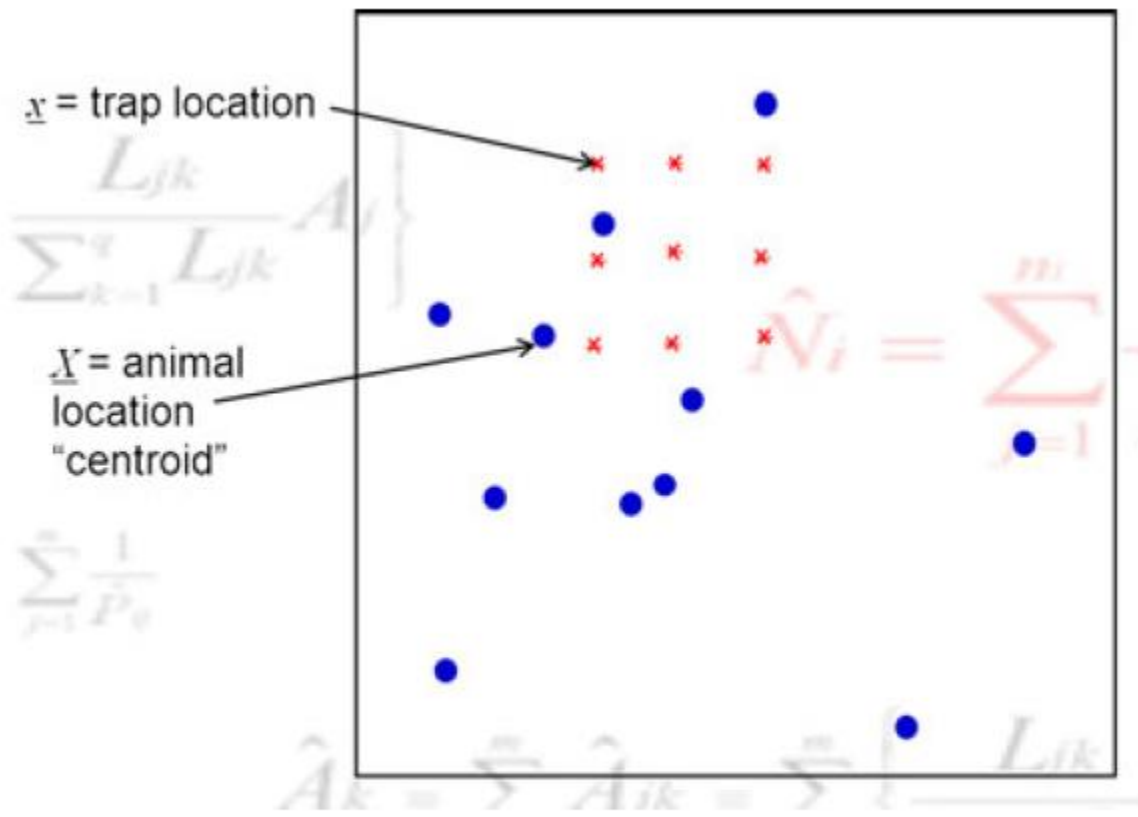


Fig. 1 Schematic representation of a spatial trapping grid relative to centroids of animal location

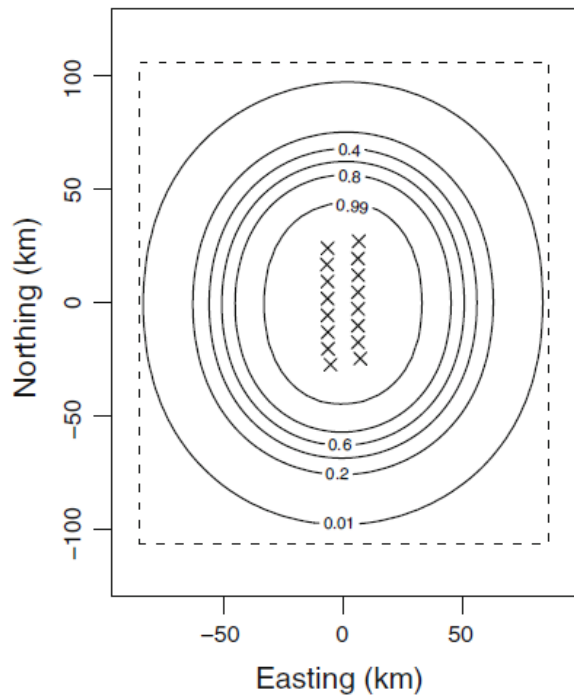


Fig. 1 Layout of BSURE case study hydrophones (*crosses*), *solid contour lines* showing probability of detecting a sound from that location with one or more hydrophones (denoted $p(X; \theta)$ in the text) estimated from a likelihood-based analysis with the half-normal detection function model, and the *dashed rectangle* showing the 80-km buffer used in that analysis

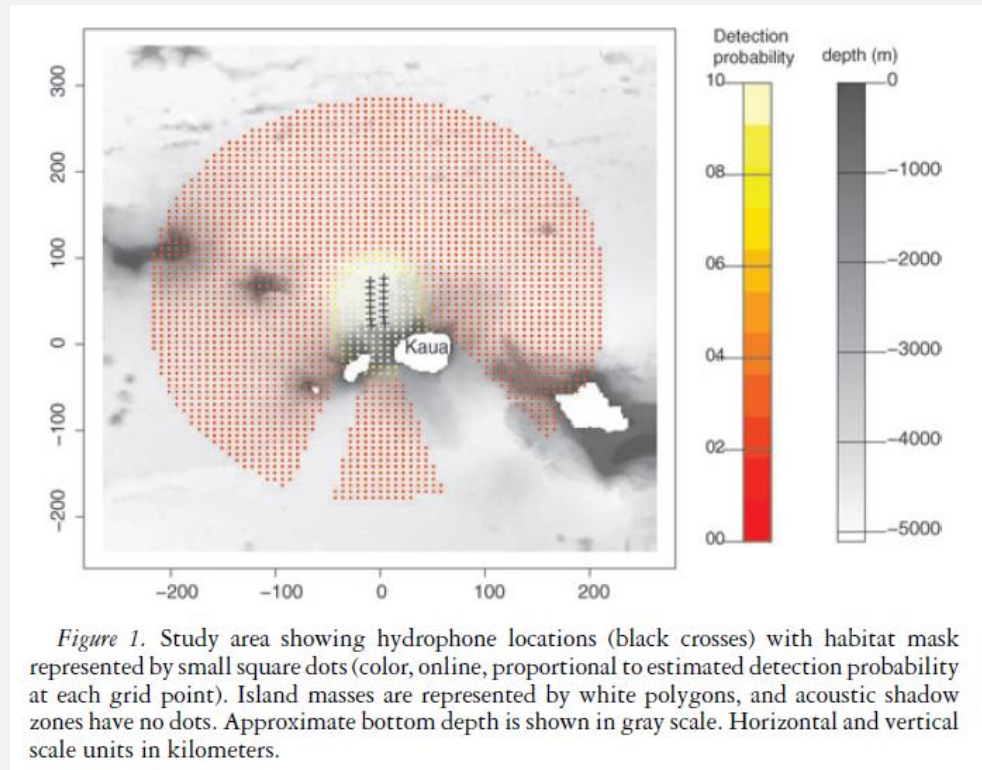


Figure 1. Study area showing hydrophone locations (black crosses) with habitat mask represented by small square dots (color, online, proportional to estimated detection probability at each grid point). Island masses are represented by white polygons, and acoustic shadow zones have no dots. Approximate bottom depth is shown in gray scale. Horizontal and vertical scale units in kilometers.

Marques, T.A.; Thomas, L.; Martin, S.W.; Mellinger, D. K.; Jarvis, S.; Morrissey, R. P.; Ciminello, C.-A. & DiMarzio, N. 2012 Spatially explicit capture recapture methods to estimate minke whale abundance from data collected at bottom mounted hydrophones *Journal of Ornithology* **152**: 445-455

Martin, S.W.; Marques, T.A.; Thomas, L.; Morrissey, R. P.; Jarvis, S.; DiMarzio, N.; Moretti, D. & Mellinger, D. K. 2012 Estimating minke whale (*Balaenoptera acutorostrata*) boing sound density using passive acoustic sensors *Marine Mammal Science* **29**: 142-158

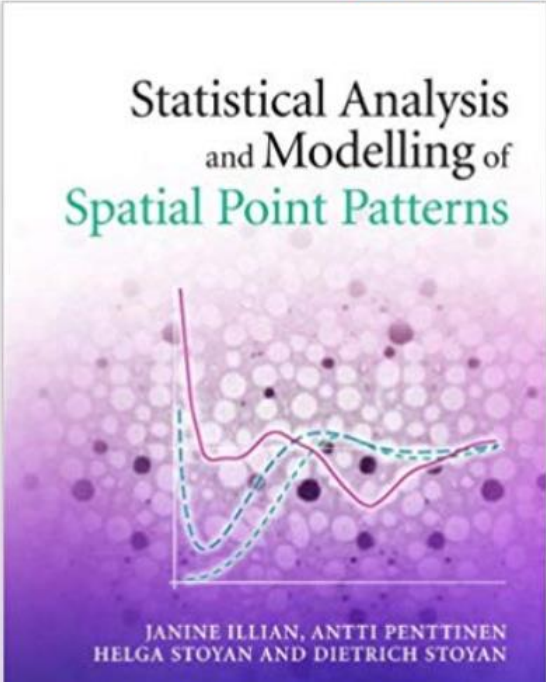
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Fig. 2. A point pattern, a window, and a pixel image.

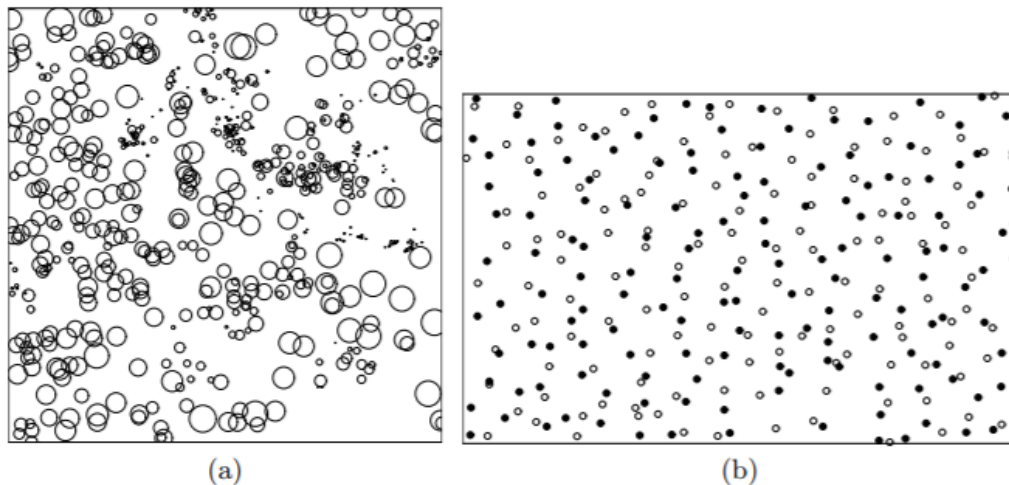


Fig. 3. Examples of marked point patterns. (a) continuous marks. Mark values (tree diameters) represented as radii of circles. The Longleaf Pines dataset, available as `longleaf`. (b) categorical marks. Mark values (cell types) represented as different graphical symbols. Hughes' amacrine cell dataset, available as `amacrine`.

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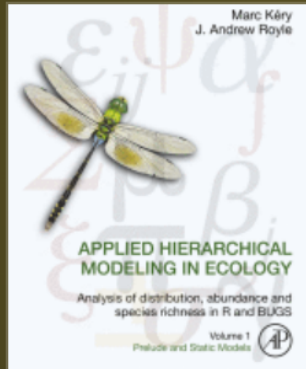
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keep that in mind...
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First published: 29 November 2013 | <https://doi.org/10.1111/geb.12138> | Cited by: 82

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