

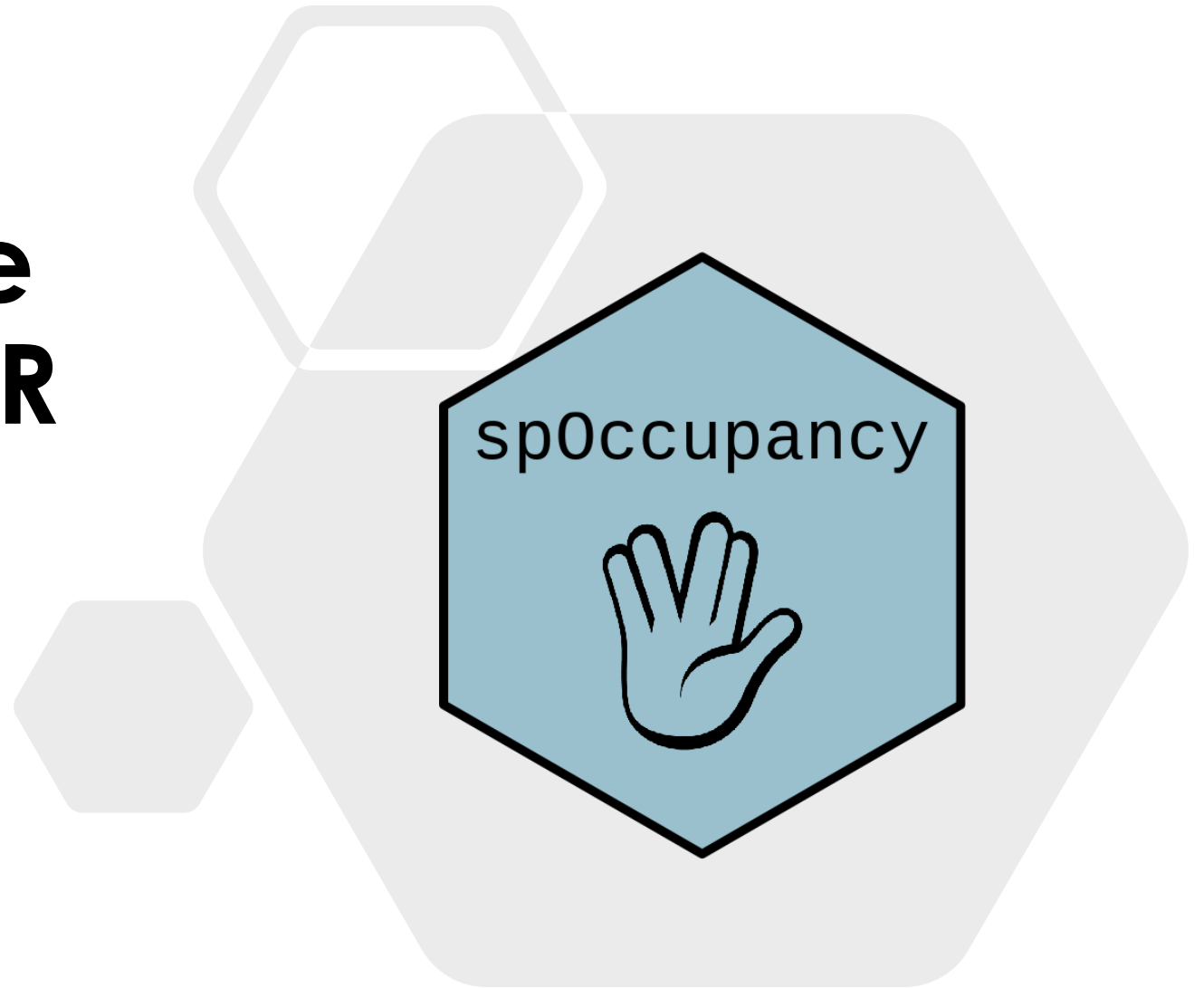
Spatial occupancy models with the spOccupancy R package

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ESA Statistical Methods
Seminar Series



Overview

- Overview of occupancy modeling and spatial modeling
- spOccupancy functionality
- spOccupancy syntax and example: single-species and multi-species spatial occupancy models
- Q&A

Motivation

- Species distribution modeling
 - Where do species occur and how does this change over time?
 - What drives species distributions?
- Two key complexities when modeling species distributions
 - Imperfect detection
 - Spatial autocorrelation

Imperfect Detection



How do we account for imperfect detection?

- Occupancy modeling
- Basic idea: perform multiple surveys (i.e., visits) at each site
- Multiple visits give information on detection probability
- Allows us to separately estimate occupancy probability from detection probability

Detection-nondetection matrix

Site	Survey 1	Survey 2	Survey 3	Survey 4
1	1	0	0	1
2	0	0	0	0
3	1	1	0	NA
4	1	NA	0	NA
5	0	1	1	1
6	0	0	0	1

Occupancy model: what is it?

- Two distinct sub-models
 - Model occupancy probability as a function of site-level covariates
 - Model detection probability as a function of site and/or survey-level covariates
 - Can only detect a species if it truly occupies a site
 - Detection probability is modeled "conditional" on true occupancy

Single-species occupancy model

Occupancy (ecological) sub-model

$j = 1, \dots, J$ (site)

$k = 1, \dots, K_j$ (replicate)

$$z_j \sim \text{Bernoulli}(\psi_j)$$
$$\text{logit}(\psi_j) = \beta_1 + \beta_2 \cdot X_{2,j} + \dots + \beta_r \cdot X_{r,j}$$

Detection (observation) sub-model

$$y_{j,k} \sim \text{Bernoulli}(p_{j,k} \cdot z_j)$$
$$\text{logit}(p_{j,k}) = \alpha_1 + \alpha_2 \cdot V_{2,j,k} + \dots + \alpha_r \cdot V_{r,j,k}$$

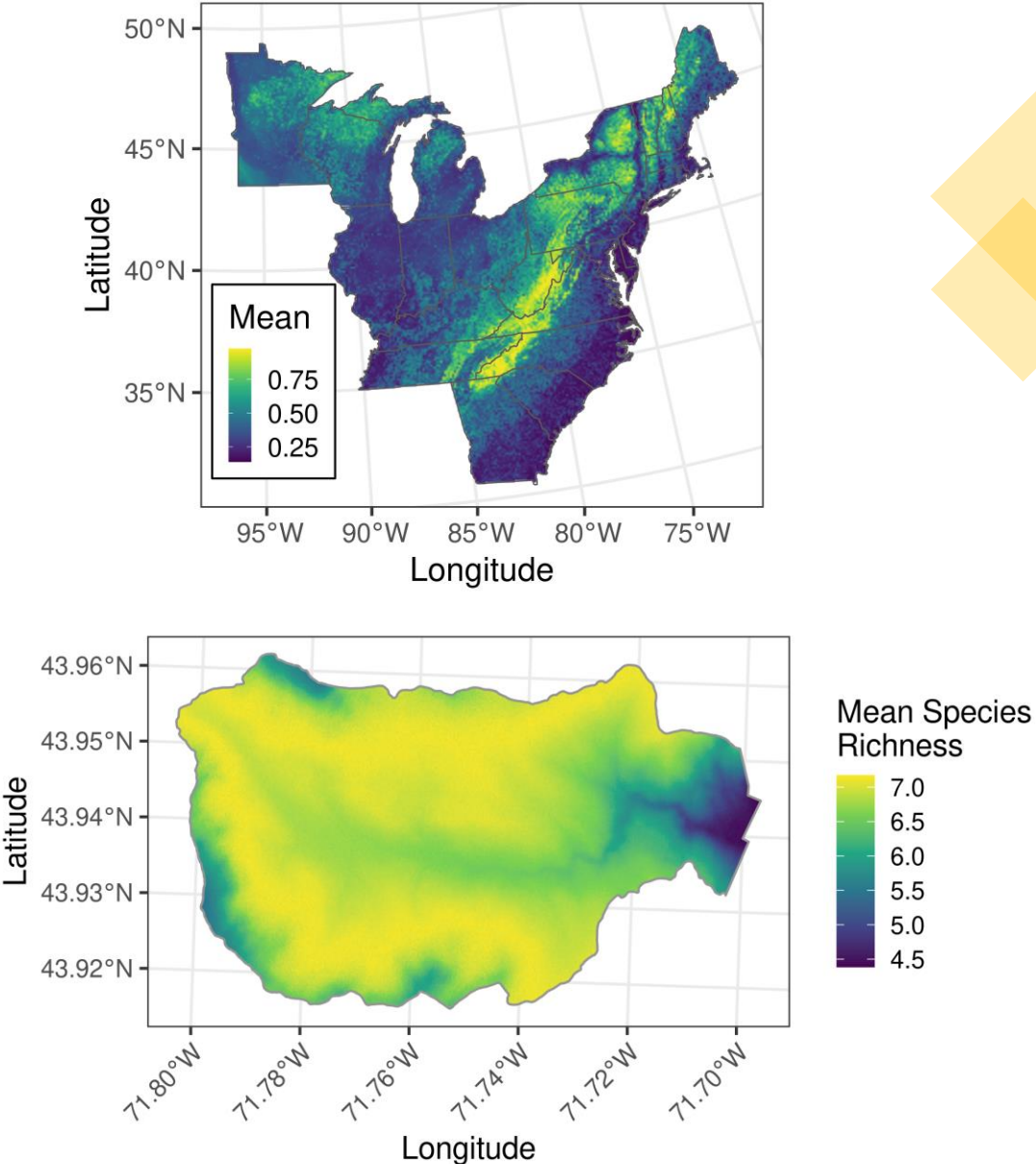
Multi-species occupancy models

- Two flavors in spOccupancy:
 - Dorazio-Royle multi-species (community) occupancy models
 - Spatial and non-spatial
 - Joint species distribution models with imperfect detection
 - Spatial and non-spatial
 - Account for species correlations using factor modeling ([Doser, Finley, Banerjee 2022](#))

					Site	Survey 1	Survey 2	Survey 3	Survey 4
				Site	Survey	Survey	Survey	Survey 4	
			Site	Survey	Survey	Survey	Survey 4		
		Site	Survey 1	Survey 2	Survey 3	Survey 4			
Site	Survey 1	Survey 2	Survey 3	Survey 4					
1	1	0	0	1	0	01		1	0
2	0	0	0	0	0	NA		NA	1
3	1	1	0	NA	NA	NA		1	0
4	1	NA	0	NA	0			1	
5	0	1	1	1	0				
6	0	0	0	1					

Spatial autocorrelation

- Things closer together in space tend to be more similar than things further apart
- What leads to spatial autocorrelation in species distributions?
 - Environmental drivers, habitat requirements
 - Biotic factors (dispersal, conspecific attraction)
- Initial approach: attempt to explain spatial variation in species distributions with covariates (e.g., forest cover, temperature, elevation)



Residual spatial autocorrelation

- Spatial correlation in data *after* including spatial covariates
- Often arises from missing/unavailable covariates
- Can lead to bias if unaddressed
- Account for using spatial random effects
 - Each site has a local adjustment in occupancy probability
 - The local adjustments are given a spatial structure
 - Estimated parameters: spatial variance and spatial decay
 - Caution: spatial confounding ([Hanks et al. 2015](#))

Single-species spatial occupancy model

$j = 1, \dots, J$ (site)

$k = 1, \dots, K_j$ (replicate)

Occupancy (ecological) sub-model

$$z_j \sim \text{Bernoulli}(\psi_j)$$

$$\text{logit}(\psi_j) = \beta_1 + \beta_2 \cdot X_{2,j} + \dots + \beta_r \cdot X_{r,j} + w_j$$

$$w_j \sim \text{Normal}(0, \Sigma)$$

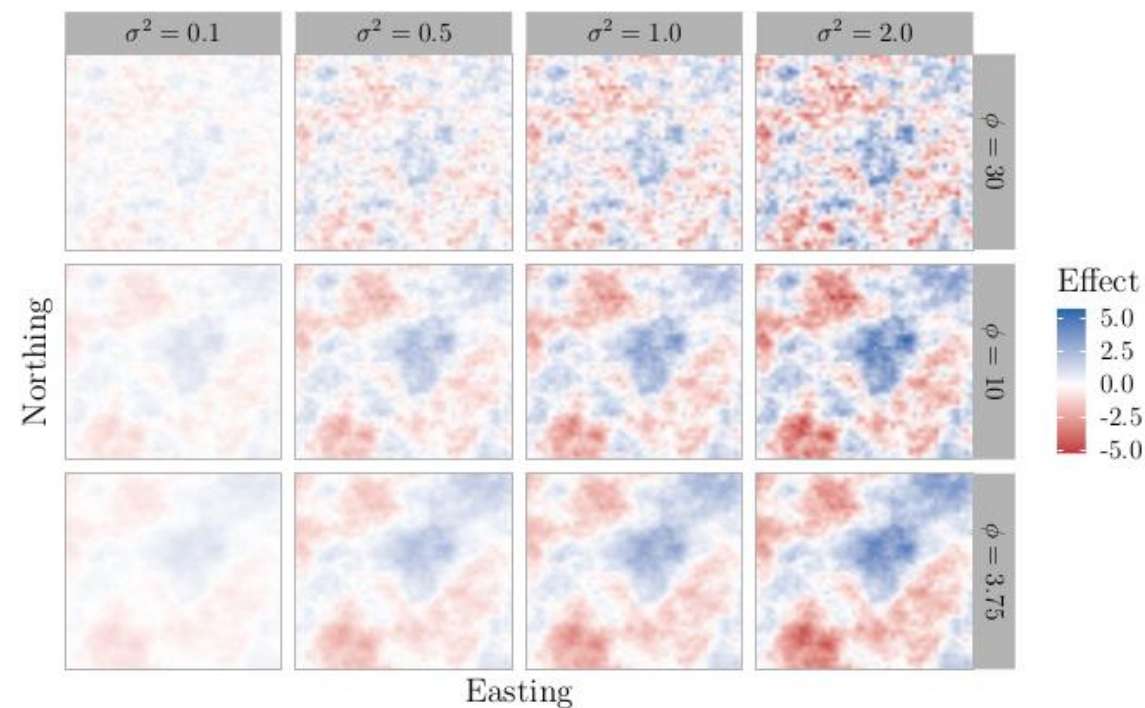
Detection (observation) sub-model

$$y_{j,k} \sim \text{Bernoulli}(p_{j,k} \cdot z_j)$$

$$\text{logit}(p_{j,k}) = \alpha_1 + \alpha_2 \cdot V_{2,j,k} + \dots + \alpha_r \cdot V_{r,j,k}$$

Gaussian processes

- "Gold standard" for modeling spatial data
- Spatial correlation function determines structure
- Downside: computationally intensive for big data
- Alternative: Nearest Neighbor Gaussian Processes ([Datta et al. 2016](#), [Doser et al. 2022](#))



spOccupancy



- Designed to fit a variety of Bayesian occupancy models
- Efficiently accommodates spatial autocorrelation
- Workflow completely in R using standard model syntax (no Bayesian programming languages necessary)
- Key functionality:
 - Single-species models
 - Multi-species models with options to account for species correlations
 - Data integration
 - Multi-season (spatio-temporal) models

Why Bayesian for occupancy modeling?

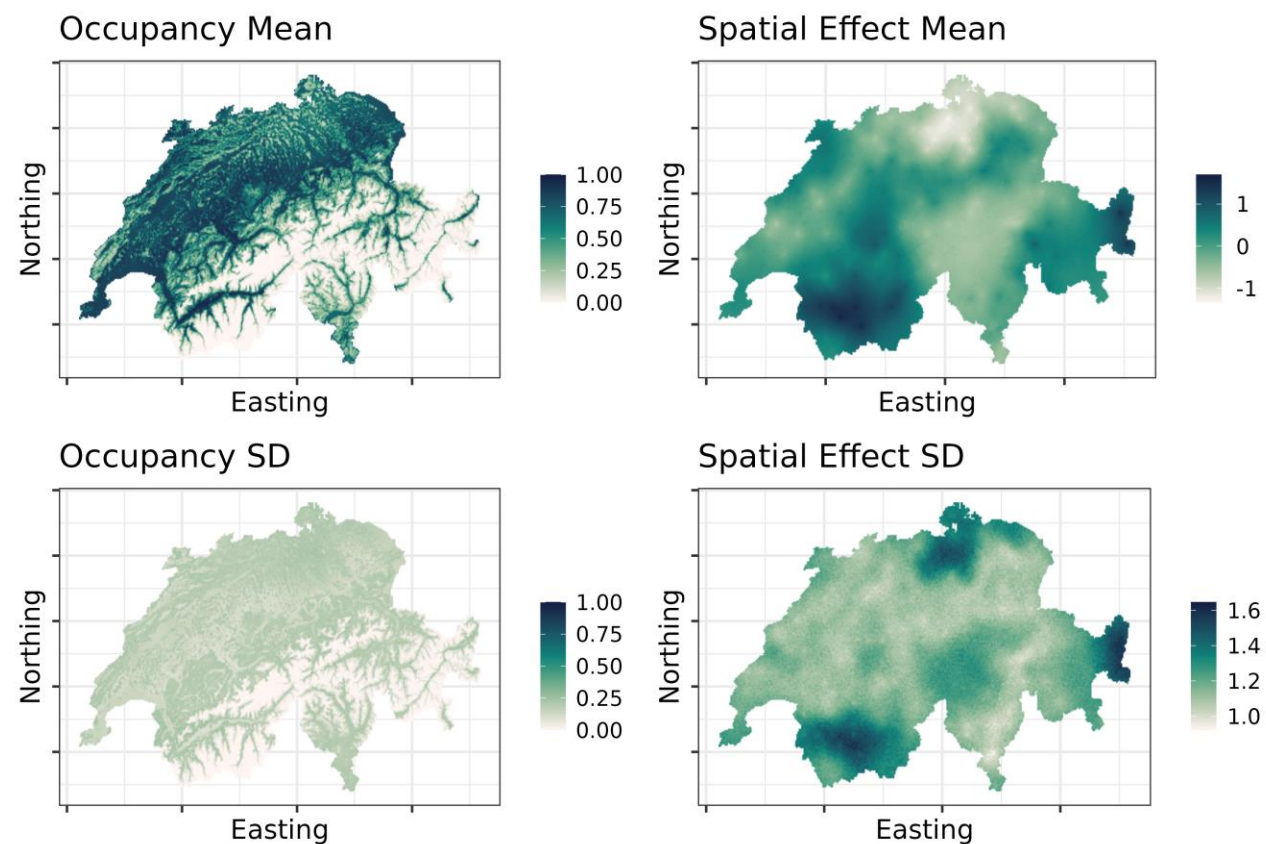
- Interpretation
- More flexible to accommodate spatial autocorrelation
- Easy to extend to multi-species frameworks/integrate multiple data sources
- Fully propagate uncertainty in all estimates (and derived quantities)

spOccupancy workflow

1. Data simulation/prep
2. Model fitting
3. Model validation
4. Model comparison
5. Posterior summaries
6. Prediction

Example dataset: Swiss MHB Survey

- Data from the Switzerland Breeding Bird Survey in 2014
- 266 survey locations distributed throughout Switzerland
- Single species (European Goldfinch) and multi-species models (20 species)



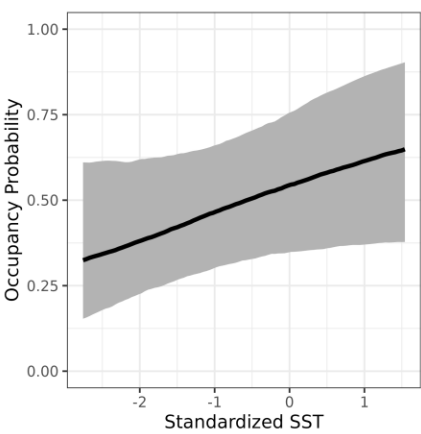
Additional examples

Single-species and multi-species occupancy models



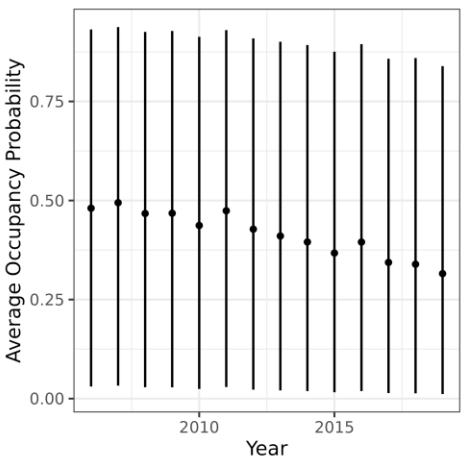
Amphibian community from Ribeiro Jr et al (2018) Eco Apps

Integrated occupancy model



Bottlenose dolphin data from Lauret et al. (2021) Ecology

Multi-season occupancy model

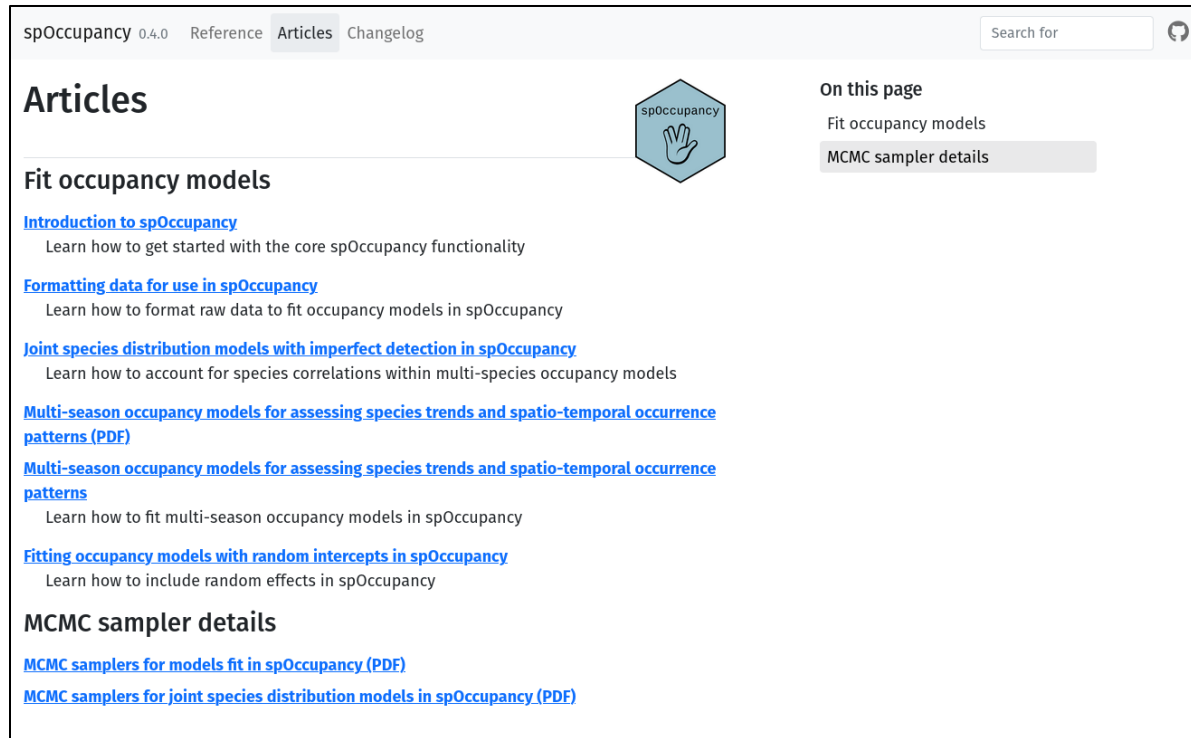


Eastern wood pewee data from Doser et al. (2021) Eco Apps

Coming soon...

- Spatially-varying coefficient occupancy models
- Multi-species spatio-temporal occupancy models
- Multi-species integrated occupancy models

spOccupancy




The screenshot shows the spOccupancy website interface. At the top, there are navigation links: "spOccupancy 0.4.0", "Reference", "Articles" (which is highlighted), and "Changelog". A search bar is located on the right. Below the navigation, the "Articles" section is displayed. It includes a sub-header "Fit occupancy models" and a list of articles with brief descriptions: "Introduction to spOccupancy", "Formatting data for use in spOccupancy", "Joint species distribution models with imperfect detection in spOccupancy", "Multi-season occupancy models for assessing species trends and spatio-temporal occurrence patterns (PDF)", and "Multi-season occupancy models for assessing species trends and spatio-temporal occurrence patterns". Below this, there is a section for "MCMC sampler details" with links to "MCMC samplers for models fit in spOccupancy (PDF)" and "MCMC samplers for joint species distribution models in spOccupancy (PDF)". A sidebar on the right contains a "spOccupancy" logo and a section titled "On this page" with links to "Fit occupancy models" and "MCMC sampler details".

- Package website
 - <https://www.jeffdoser.com/files/spoccupancy-web/>
- GitHub development page
 - <https://github.com/doserjef/spOccupancy/>
- [MEE intro paper](#)
- [arXiv preprint](#)
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The cover image shows the front page of a scientific paper. At the top, it says "Received: 21 December 2021 | Accepted: 20 April 2022" and "DOI: 10.1111/2041-210X.13897". Below this, the journal name "Methods in Ecology and Evolution" is displayed with the "BRITISH ECOLOGICAL SOCIETY" logo. The paper title "spOccupancy: An R package for single-species, multi-species, and integrated spatial occupancy models" is prominently featured. At the bottom, the authors are listed: "Jeffrey W. Doser^{1,2} | Andrew O. Finley^{1,2} | Marc Kéry³ | Elise F. Zipkin^{2,4}".

Joint species distribution models with imperfect detection for
high-dimensional spatial data 

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Acknowledgements



Elise Zipkin



Marc Kéry



Sudipto Banerjee



Thank you!