# Analysing multivariate ecological data with Generalized Linear Latent Variable Models

Bert van der Veen Jenni Niku Sam Perrin









### Exercise material and package installation

https://github.com/BertvanderVeen/IRSAE2021GLLVMworkshop

#### Questions

In chat or "raise" your hand

♥ On twitter: #GLLVMs, @vdVeenB or @J\_\_Niku or @samperrinNTNU

On github: https://github.com/BertvanderVeen/IRSAE2021GLLVMworkshop/discussions

#### Installation

Ask in chat if you have issues

## Welcome! 😭





Affiliation

**Expertise** 

Bert van der Veen PhD candidate Norwegian institute of Bioeconomy research & Norwegian university of Science and Technology - Statistical ecology

- Ordination

- Species distribution modeling



Jenni Niku Postdoc University of Jyväskylä

- Statistical ecology

- Species distribution modeling



Sam Perrin PhD Norwegian university of Science and Technology Ducky AS

- Fresh water ecology
- Invasion ecology
- Species distribution modeling

### Program

#### **Topic**

- Basics of Multivariate analysis
- Latent variables
- Generalized Linear Latent Variable models

#### Who



#### Questions / Break

- gllvm R-package (Niku et al. 2019)
- How to: model-based ordination with GLLVMs



#### Questions / Break

- Some ecology with species associations



#### Finish

- Model-based quadratic ordination
- Model-based constrained ordination
- Wrap-up



#### What are these **GLLVMs**?

#### **Generalized Linear Latent Variable Models**

- Model-based multivariate analysis
- Mixed-effects model
- Joint Species Distribution Model Ovaskainen et al. 2017
- Unconstrained ordination Hui et al. 2015
- Constrained ordination
- "Factor analytic approach"

But in general, for the analysis of species (co-)occurrence patterns.

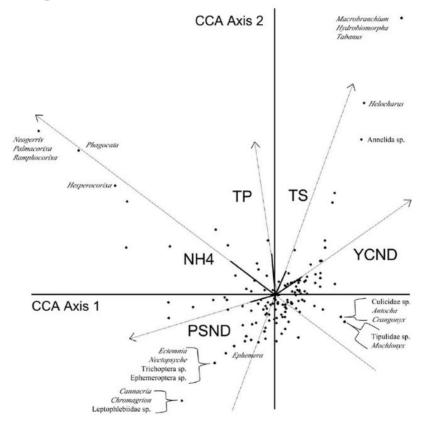
GLLVMs are flexible



### Classical multivariate analysis

E.g., PCA, CA, DCA, NMDS

With eigenvectors and using distance metrics



Maul et al. 2004

### Classical multivariate analysis

Here we go model-based!

Plant Ecol (2015) 216:669-682 DOI 10.1007/s11258-014-0366-3



#### Model-based thinking for community ecology

David I. Warton · Scott D. Foster · Glenn De'ath · Jakub Stoklosa · Piers K. Dunstan

Received: 29 January 2014/Accepted: 30 May 2014/Published online: 19 November 2014 © Springer Science+Business Media Dordrecht 2014

## Gathering data

We go out, register species at multiple sites



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## "Multivariate"

- What does multivariate mean?
- Multivariate: multiple **responses**
- E.g. counts of species at sites

	Species 1	Species 2	Species 3	Species 4	Species 5
Site 1	25	10	0	0	0
Site 2	0	2	0	0	0
Site 3	15	20	2	2	0
Site 4	2	6	0	1	0
Site 5	1	20	0	2	0

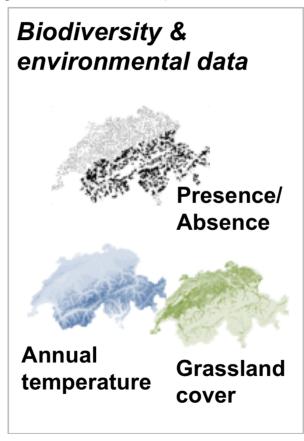
### e.g., camera trap data



Caravaggi et al. 2020

### "Multivariable"

Multiple **predictors** that represent the environment



https://damariszurell.github.io/SDM-Intro/



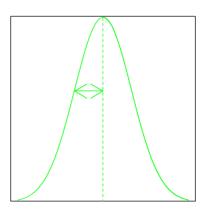
### To summarise

- Both data and method can be univariate or multivariate
- Multivariate data can be analysed with both multivariate and univariate methods (SDM, CA)
- Multivariable data can be used in multivariate or univariate analysis
  - Generally the same for all responses
  - (But, note that the model can of course set terms to zero)

## Why analyse multivariate data?

- Interest in **co**-occurrence patterns
  - In contrast to only **occurrence** patterns (a species distribution)
- Why do species co-occur?
  - Similar environmental preferences
  - Similar history in the environment
  - Might result in Interactions
- Multiple species form a community

Abundance/occurrence

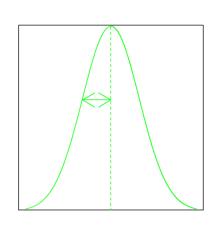


**Environment** 

## Why analyse multivariate data?

Occurrence pattern

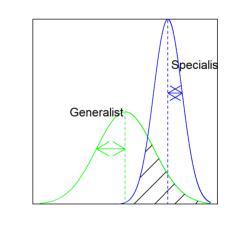
Abundance/occurrence



**Environment** 

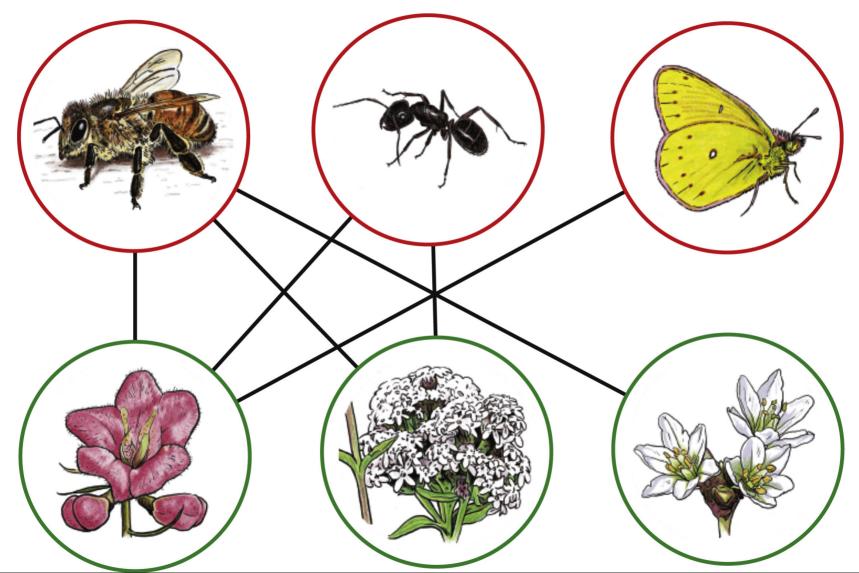
Co-occurrence pattern

Abundance/occurrence



**Environment** 

## But then for more species



## Examples

\_ \_ \_ .



### Joint modeling with latent variables

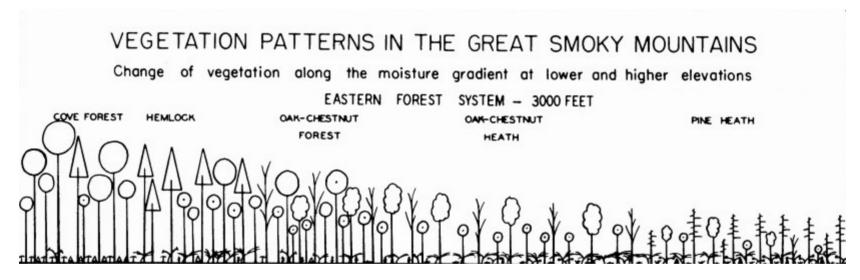
- Accounts for correlation between taxa
- "Borrows" information from other species for estimation
- Provides species associations
- Concept: fit a single model for all species
- I.e., a "Joint Species Distribution Model"
  - Faster
  - Less tedious
  - Explicitly model species co-occurrence
  - o Etc.

#### Key references

- Warton et al. 2015: "So many variabes: Joint modeling in community ecology"
- Blanchet et al. 2020: "Co-occurrence is not evidence of ecological interactions"
- Poggiato et al. 2021: "On the interpretations of joint modeling in community ecology"

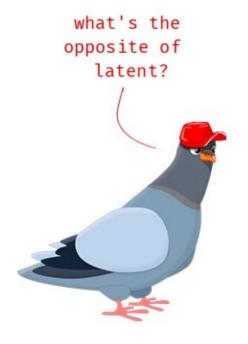


### Latent variables?



Whittaker 1967

#### Variables can be **observed** or **latent**



active, obvious, manifest, apparent, alive, clear, live, operative, working, open



If not measured, it is latent, like a random-effect.

### Ecological gradient analysis

"Gradient analysis is a research approach for the study of spatial patterns of species." Whittaker 1967

Our sites describe the environment. Multiple environmental gradients can form a **complex** gradient.

	Predictor 1	Predictor 2	Predictor 3	Predictor 4	Predictor 5
Site 1	2.3321	3.0445	0.0000	3.0445	4.4543
Site 2	3.0493	3.2581	1.7918	1.0986	4.5643
Site 3	2.5572	3.5835	0.0000	2.3979	4.6052
Site 4	2.6741	4.5109	0.0000	2.3979	4.6151
Site 5	3.0155	2.3979	0.0000	0.0000	4.6151

### **Ecological gradients**

- 1) Ecological gradient: gradual change in the environment
  - e.g. temperature
- 2) Complex gradient: change in several ecological gradients
  - e.g., soil moisture and acidity on an elevation gradient
  - e.g., a gradual urban to rural change in the landscape
  - Can be represented as a single factor, covariate, predictor, latent variable, ordination axis

So a latent variable is an ecological gradient or ordination axis representing one, or multiple, missing predictors.

### Latent variables

"Few major complex ecological gradients normally account for most of the variation in species composition." Halvorsen 2012

#### In essence:

Community structure is generally low-dimensional.

#### At this point you might think:

- Community ecology has been doing it for a hundred years!
- e.g. Forbes (1907)
- Ordination:
  - Principal Component Analysis Pearson 1901
  - Correspondence analysis Hirschfeld 1934
  - o NMDS Kruskal 1964a,b
- Niche overlap

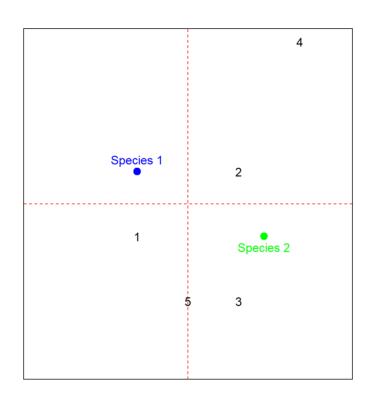
### Analysing multivariate data: ordination

- Termed by David Goodall in 1954: "An essay in the use of factor analysis"
- Applied factor analysis For the analysis of data on a plant community
- Reducing dimension of data
- ordering species or samples along an ecological gradient
- classically e.g.,
  - Principal Component Analysis (PCA; prcomp())
  - Correspondence Analysis (CA; cca() in vegan)
  - Multidimensional scaling (PCoA; cmdscale(), NMDS; metaMDS() in vegan)
  - Factor analysis: Precursor to GLLVMs (FA; factanal())
- Treats latent variables as fixed-effects
- So a multivariate Generalized Linear Model (kind of), i.e., a joint model

## Ordination: visual inspection

- Most common tool is the biplot Gabriel 1971
- Distance between species indicates dissimilarity
- Distance between sites indicates dissimilarity

Dimension 2



**Dimension 1** 

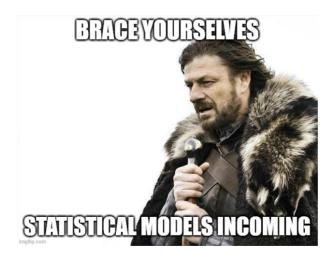
### Classical methods have some issues...

- Ordination axis (ecological gradient) treated as fixed (parameter)
- Horseshoe or arch effect (PCA, CA)
- Difficult (near to impossible) to check any assumptions
- Mean-variance relationships Warton and Hui 2017

In general, not very flexible.

### Model-based thinking

- Concept: apply regression concepts to multivariate analysis Warton et al. 2015
  - Explicit statistical models
  - Residual diagnostics
  - Model selection
  - et cetera



### Specifying a multivariate statistical model

- $\beta_{0i}$  intercept per species
- ullet  $x_{ik}$  site-specific predictors
- β<sub>j</sub> species-specific slopes

$$g(E(y_{ij}|\boldsymbol{x}_i)) = \boldsymbol{\beta}_{0j} + \boldsymbol{x}_i^{\top} \boldsymbol{\beta}_j$$
 (1)

- Stacked SDM (fit with glm(.) function)
- No random-effects

### A Multivariate Mixed-effects model

ullet Add residual for  $i=1\dots n$  sites and  $j=1\dots p$  species

$$g(\mathrm{E}(y_{ij}|oldsymbol{x}_i)) = oldsymbol{eta}_{0j} + oldsymbol{x}_i^ op oldsymbol{eta}_j + oldsymbol{u}_{ij}, \qquad oldsymbol{u}_i \sim \mathcal{N}(0, \Sigma)$$

- Structure  $u_{ij}$  with  $\Sigma$  by species
- ullet  $\Sigma$  are species covariances or associations
- A "joint species distribution model" Pollock et al. 2014
- Can be fit using standard mixed-effects modeling software.

#### In lme4:

glmer(abundance~species+x:species+
(0+species|sites),family="poisson",data=data)

- ullet Too many parameters in  $\Sigma$
- Number of parameters increases rapidly
- Often not feasible to fit

### Model-based ordination to the rescue!

- Ordination = dimension reduction
- Represent the latent complex ecological gradient
- A model like in regression
- Represent species associations with latent variables
- So JSDM = ordination? Yes! (for GLLVMs)
- "Model-based approaches to unconstrained ordination" Hui et al. 2015

## All the benefits from regression and ordination!, e.g.:

- Procrustus analysis
- Biplots
- Model-selection
- Residual diagnostics
- Appropriate mean-variance relationships
- Hypothesis testing
- No distance metrics



#### Generalized Linear Latent Variable Models

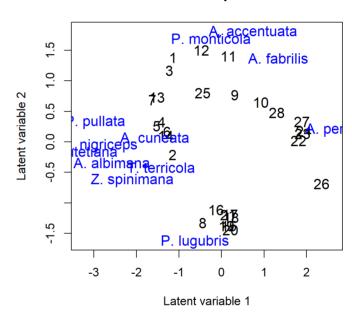
- GLLVM for short
- ullet Add factor analytic structure to  $\Sigma$
- Ordination = dimension reduction
- $m{u}_{ij} = m{\epsilon}_i^ op m{ heta}_j$   $\circ$  i.e.  $m{\epsilon}_i \sim \mathcal{N}(0, m{ heta}_j m{ heta}_j^ op)$
- Faster and fewer parameters:
  - Number of parameter doesn't grow so fast
  - $\circ$  More latent variables, better estimation of  $\Sigma$
- So we are estimating residual covariances!

$$\Sigma = \begin{bmatrix} \theta_{11} & 0 & 0 \\ \theta_{12} & \theta_{22} & 0 \\ \vdots & \ddots & \vdots \\ \theta_{1i} & \cdots & \theta_{di} \end{bmatrix} \begin{bmatrix} \theta_{11} & \theta_{12} & \cdots & \theta_{1j} \\ 0 & \theta_{22} & \ddots & \vdots \\ 0 & 0 & \cdots & \theta_{dj} \end{bmatrix}$$
(3)

#### Generalized Linear Latent Variable Models

- Still a mixed-effects model
- ullet d latent variables treated as random-effect
- Produces ordination
  - $\circ$  "site scores" :  $\epsilon_i$
  - $\circ$  "species scores" or "loadings":  $oldsymbol{ heta}_i$
- Species and sites far apart are dissimilar
- E.g., because species prefer different environments

#### Ordination of spider data



$$g(\mathrm{E}(y_{ij}|oldsymbol{x}_i,oldsymbol{\epsilon}_i)) = oldsymbol{eta}_{0j} + oldsymbol{x}_i^{ op} oldsymbol{eta}_j + oldsymbol{\epsilon}_i^{ op} oldsymbol{ heta}_j, \qquad oldsymbol{\epsilon}_i \sim \mathcal{N}(0,oldsymbol{I})$$

### gllvm R-package

## Methods in Ecology and Evolution





gllvm: Fast analysis of multivariate abundance data with generalized linear latent variable models in R

Jenni Niku 🔀, Francis K. C. Hui, Sara Taskinen, David I. Warton

First published: 21 September 2019 | https://doi.org/10.1111/2041-210X.13303 | Citations: 4

### Break / Questions

- **У** On twitter: **#GLLVMs**, **@vdVeenB** or **@J\_\_Niku** or **@samperrinNTNU**
- On github: https://github.com/BertvanderVeen/IRSAE2021GLLVMworkshop/discussions

Or in the chat.