

Hierarchical Models in Population Ecology

What are they and why should we use them?

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Topics of Discussion

- > Introduction to hierarchical models
 - What is a *hierarchy*?
 - What is a *statistical model*?
 - What is a *hierarchical model*?
 - What is NOT a *hierarchical model*?
- > Hierarchical models in population ecology
 - Brief primer to population ecology
 - *Process-only* models
 - *Process + observation* model
 - *Hyper-parameter* models
- > Why should we use hierarchical models?
 - Scope and scale of inference
 - Correct accounting of variance
 - Borrowing strength
- > Areas of active development
 - Integrated population models
 - Spatial capture-recapture models
- > Hierarchical modeling resources

Introduction to Hierarchical Models

Introduction to Hierarchical Models
What is a hierarchy?

Definition: **hierarchy** (noun) – A series of *ordered* groupings of people or things within a system
Royle et al. 2013

Definition: **classification** (noun) – the arrangement of entities in a hierarchical series of nested classes, in which *similar or related classes at one hierarchical level are combined comprehensively into more inclusive classes at the next higher level*
Mayr and Bock 2002

Introduction to Hierarchical Models
What is a hierarchy?

➤ Hierarchies in population ecology:


ECOLOGICAL SCALES OF ORGANIZATION

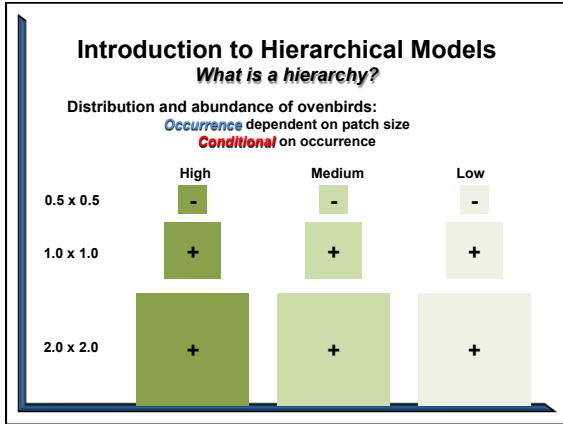
Metacommunity: distribution of communities
Community: distribution of metapopulations
Metapopulation: distribution of populations
Population: distribution of individuals

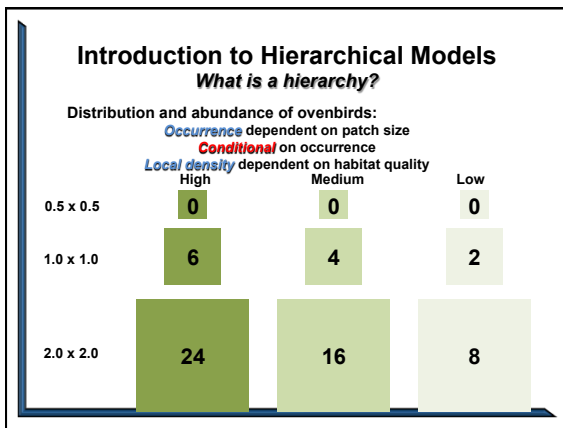
How different factors affect different hierarchical levels

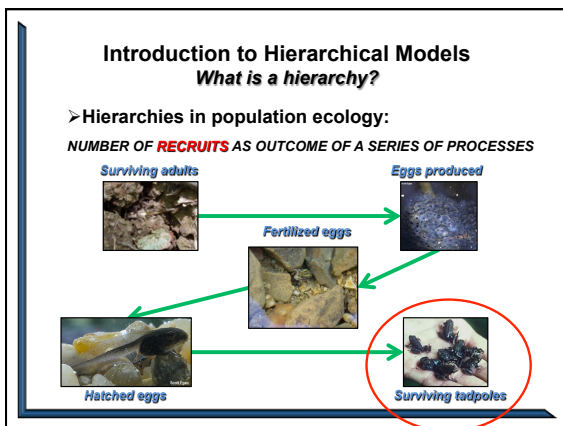
Introduction to Hierarchical Models
What is a hierarchy?

Distribution and abundance of ovenbirds:









Introduction to Hierarchical Models

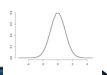
What is a statistical model?

Definition: **statistical model** (noun) – A formal description of a number generating process comprised of a **deterministic** and a **stochastic** component, expressed algebraically, and based on **probability** distributions (i.e., parametric)

"Parametric statistical modeling means describing a caricature of the 'machine' that plausibly could have produced the numbers we observe"

Kery 2010

Linear regression example:
 $y_i = \alpha + \beta_1 X_i + \dots + \beta_k X_i + \epsilon_i$
 where $E[Y] = \alpha + \beta_1 X_i + \dots + \beta_k X_i$ ← Deterministic
 and $\epsilon_i \sim \text{Normal}(0, \sigma^2)$ ← Stochastic



Introduction to Hierarchical Models

The Hierarchical Model

Definition: **hierarchical model** (noun) – A series of [parametric] models, ordered by their **conditional** probability structure

Royte et al. 2013

aka: state-space, multi-level, random-effects, GLMM, nested


Example: **SPECIES OCCURRENCE MODEL**

$z_i \sim \text{Bernoulli}(\psi)$ State process

$y_i | z_i \sim \text{Bernoulli}(z_i \times p)$ Observation process

where y_i = observed presence at site i ,
 z_i = true occurrence status at site i
 and p = detection probability

Observation is **CONDITIONAL** on true state



Introduction to Hierarchical Models

NOT Hierarchical Models

➤ "Step-down" or "Stepwise" model selection

- The ad hoc process of holding model structure constant for some parameters, while investigating structures for others

Example: Cormack-Jolly-Seber model
 Model parameters: ϕ (apparent survival) and p (detection probability)

- Hold ϕ constant, test alternative structures for p
- Hold best structure for p constant, test ϕ

NOT RECOMMENDED

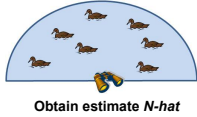
Doherty et al. 2012

Introduction to Hierarchical Models
NOT Hierarchical Models

➤ Multi-stage analyses (i.e., *statistics on statistics*)

- The process of using estimates from an initial analysis as input data for a secondary analysis

Example: Evaluate habitat effects on local abundance (N)



Obtain estimate $N\text{-hat}$


Introduction to Hierarchical Models
NOT Hierarchical Models

➤ Multi-stage analyses (i.e., *statistics on statistics*)

- The process of using estimates from an initial analysis as input data for a secondary analysis

Example: Evaluate habitat effects on abundance (N)

- 1) Estimate abundances from encounter data
- 2) Test for relationship between N estimates and habitat variables



A WELL KNOWN "NO" in STATISTICS

Linear regression of $N\text{-hat}$:
 $N\text{-hat} = \alpha + \beta_1 X_1 + \dots + \beta_k X_k + \epsilon$
 where X_1, \dots, X_k are habitat variables

Introduction to Hierarchical Models
NOT Hierarchical Models

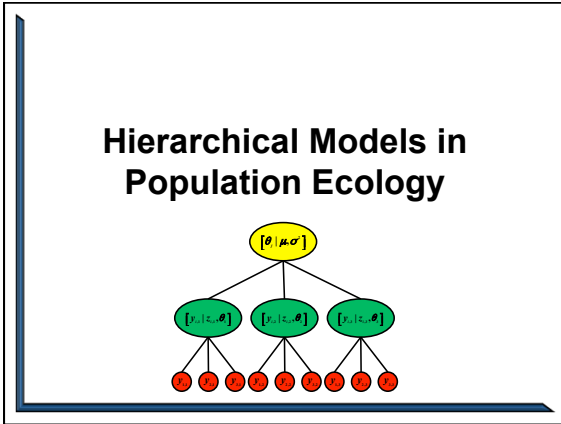
➤ Bayesian inference

- A statistical inference paradigm based on Bayes theorem that uses probability to describe all unknown quantities

Bayesian hierarchical modeling:

The fitting of hierarchical models using Bayesian methods

Hierarchical models can also be fit using frequentist methods



Hierarchical Models in Population Ecology

Population ecology

- > Abundance and distribution of individuals and species
- > Dynamics of populations, metapopulations, communities, etc.
- > Factors affecting abundance, distribution, and dynamics

Hierarchical Models in Population Ecology

How do we use hierarchical models in the study of population ecology?

- > Match structure of the statistical model to the structure of our conceptual model of ecological processes

Frog recruits revisited: # of **recruits (R)** into adult class

$E_i \sim \text{Poisson}(\lambda_i)$
 $FE_i | E_i \sim \text{Binomial}(E_i, f)$
 $T_i | FE_i \sim \text{Binomial}(FE_i, h)$
 $R_i | T_i \sim \text{Binomial}(T_i, m)$

Hierarchical Models in Population Ecology

How do we use hierarchical models in the study of population ecology?

> Match structure of the statistical model to the structure of our conceptual model of ecological processes

Frog recruits revisited: # of **recruits (R)** into adult class

$E_i \sim \text{Poisson}(\lambda_i)$	$\log(\lambda_i)$ - <i>eg. female body mass, pathologies, etc</i>
$FE_i E_i \sim \text{Binomial}(E_i, f)$	$\text{logit}(f)$ - <i>eg. male body mass, pathologies, etc</i>
$T_i FE_i \sim \text{Binomial}(FE_i, h)$	$\text{logit}(h)$ - <i>eg. predator density, temp, etc.,</i>
$R_i T_i \sim \text{Binomial}(T_i, m)$	$\text{logit}(m)$ - <i>eg. predator density, temp, etc.,</i>

Hierarchical Models in Population Ecology

How do we use hierarchical models in the study of population ecology?

> Incorporate conditional observation process into model structure to account for imperfect detection

Example: **CORMACK-JOLLY-SEBER MODEL**

$z_i \sim \text{Bernoulli}(\phi)$

$y_i | z_i \sim \text{Bernoulli}(z_i \times p)$

State process

Observation process

Observation is **CONDITIONAL** on true state

where z_i = true alive status for individual i ,
 y_i = observed alive status for individual i ,
and p = detection probability

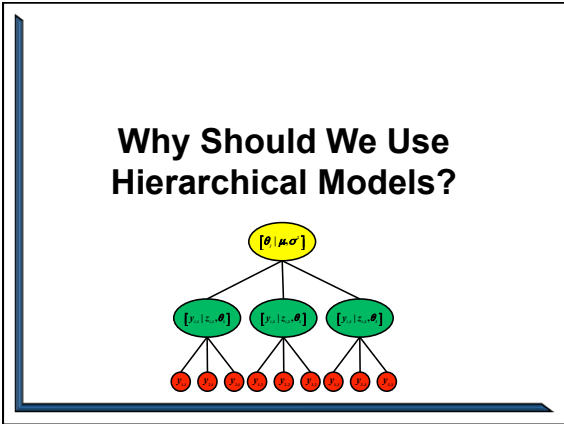
Hierarchical Models in Population Ecology

How do we use hierarchical models in the study of population ecology?

> Impose additional structure via hyper-parameters

Example: **CORMACK-JOLLY-SEBER MODEL**

<p><i>Individual covariates and random effects</i></p> <p>$\text{logit}(\phi) \sim \text{Normal}(\alpha + \beta x, \sigma^2)$</p> <p>$z_i \sim \text{Bernoulli}(\phi)$</p> <p>$y_i z_i \sim \text{Bernoulli}(z_i \times p)$</p> <p>Evolutionary processes on fitness</p>	<p><i>Temporal covariate and random effects</i></p> <p>$\text{logit}(\phi) \sim \text{Normal}(\alpha + \beta x, \sigma^2)$</p> <p>$z_i \sim \text{Bernoulli}(\phi)$</p> <p>$y_i z_i \sim \text{Bernoulli}(z_i \times p)$</p> <p>Environmental processes on fitness</p>
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Why Use Hierarchical Models?

Scope and Scale of Inference

- Extend inference beyond levels under study
 - Generalize to population from which sample units were drawn
 - ✓ *Need to know means and variances of global processes*
- Scale-dependent inference
 - Evaluate factors affecting different levels of ecological processes
 - ✓ *Distribution and abundance of ovenbirds*

Why Use Hierarchical Models?

Correct accounting of variance

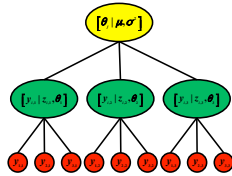
- Random effects allow partitioning of process and sampling variances
 - *Critical for population projection models used in population viability analyses*
- Avoids variance-accounting problems with multi-stage analyses
 - *Violation of constant sampling variance assumption*
- Allows modeling covariances among different parameters
 - *Temporal covariance between survival and recruitment*

Why Use Hierarchical Models?

Borrowing strength

- > Fixed effects can result in imprecise or extreme group-specific estimates for small samples
- > By constraining parameters by a common distribution (random effects), individual estimates are pulled toward the global mean (e.g., shrinkage)
- > Individual estimates “borrow strength from the ensemble”
- > Assumption of exchangeability must hold

Areas of Active Development



Areas of Active Development

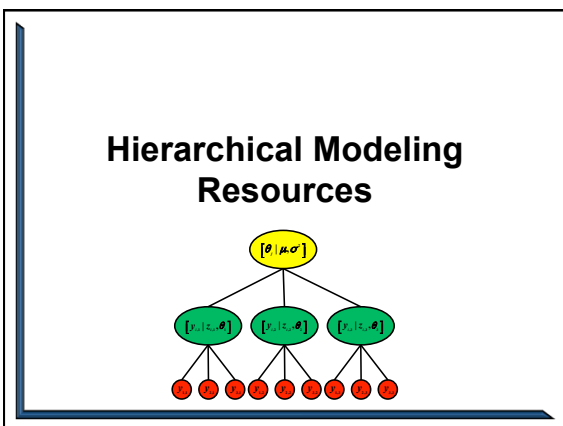
Integrated Population Models

- > Integrate data from multiple sources to model individual demographic processes
 - *Capture-recapture and known-fate data for survival*
- > Integrate data from multiple demographic processes to model population dynamics
 - *Capture-recapture, reproduction, known-fate, and band-return data*
- > Extend population models to metapopulation and community models
 - *Shared information among multiple populations or similar species*


Areas of Active Development

Spatial Capture-Recapture Models

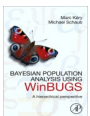
- Explicit modeling of territoriality
 - *Spatial interactions among individuals*
- Extending models to accommodate gregarious species
 - *Non-independence of individual activity centers*
- Development of explicit movement models
 - *Dispersal, transience, and migration*



Hierarchical Modeling Resources






Royle, J. A., and R. M. Dorazio. 2008. Hierarchical modeling and inference in ecology. The analysis of data from populations, metapopulations and communities. Academic Press, London, UK.



Kery, M., and M. Schaub. 2012. Bayesian population analysis using WinBUGS. A hierarchical perspective. Academic Press, Waltham, Massachusetts, USA.

Hierarchical Modeling Resources

Kery, M. 2010. Introduction to WinBUGS for ecologists. A Bayesian approach to regression, ANOVA, mixed models and related analysis. Academic Press, Burlington, Massachusetts, USA.

Link, W. A., and R. J. Barker. 2010. Bayesian inference with ecological applications. Academic Press, London, UK.

Royle, J. A., R. B. Chandler, R. Sollmann, and B. Gardner. 2013. Spatial capture-recapture. Academic Press, Waltham, Massachusetts, USA

LITERATURE CITED

- Doherty, P. F., G. C. White, and K. P. Burnham. 2012. Comparison of model building selection strategies. *Journal of Ornithology* 152:S317–S323.
- Kery, M. 2010. Introduction to WinBUGS for ecologists. A Bayesian approach to regression, ANOVA, mixed models and related analysis. Academic Press, Burlington, Massachusetts, USA.
- Mayr, E., and W. J. Bock. 2002. Classifications and other ordering systems. *Journal of Zoological Systematics and Evolutionary Research* 40:169–194.
- Royle, J. A., R. B. Chandler, R. Sollmann, and B. Gardner. 2013. Spatial capture-recapture. Academic Press, Waltham, Massachusetts, USA

PHOTO CREDITS

- http://www.fws.gov/uploadedImages/Region_3/NWRS/Zone_4/Illinois_River_Complex/Chautauqua/Sections/Seasons_Of_Wildlife/Waterfowl%20512x219.jpg
- http://www.uri.edu/cels/nrs/paton/LH_wood_frog.html
- http://www.uri.edu/cels/nrs/paton/photo_wofr2.htm
