## Remodeling the fossil record

analysis of emergent evolutionary and ecological patterns

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## Macroevolution and macroecology

### Structured data and modelling emergent patterns

#### Patterns in extinction

Background extinction and expected differences in species survival

Interplay between extinction intensity and extinction selectivity

## Patterns in functional diversity

Mammal species pool functional composition

### Conclusions and commentary

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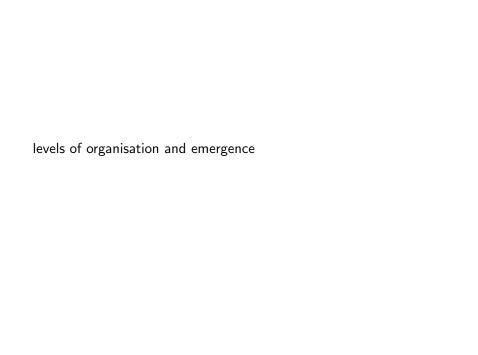
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## Definition

- Macroevolution
- Macroecology

trait: identifiable property of an organism	
species trait: idetifiable property of the entire species	

functional trait: class of traits describing interaction with

environment

species selection			

Rabosky and McCune: due to heritability of speciation/extinction

operationalize through traits while paying attention to definition of

rates

fitness

species fitness

derived definitions depend on definition of extinction

logic: if more fit, more likely to be present

Cooper 84 fundamental defintion: expected time till extinction

(lines-of-descent)

### Extinction

fundamentally emergent phenomenon; all members must die for collective to die.

## Law of Constant Extinction

Extinction risk, in a given adaptive zone, is taxon-age independent.

(Van Valen 1973 Evol. Theory)

## Survival of the unspecialized

When related phyla die out ... more specialized phyla tend to become extinct before less specialized. This phenomenon is also far from universal, but it is so common that it does deserve recognition as a rule or principle in evolutionary studies: the rule of the survival of the relatively unspecialized.

(Simpson, 1944, Tempo and Mode of Evolution, p. 143)

## Species pool concept

(Mittelbach and Schemske, 2015, TREE)



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## Inference

 $Learning\ from\ incomplete\ information.$ 

statistical model as inference device
engineering approach to analysis: building blocks used to create the device



## Models of macroevolution

birth-death for diversity

Brownian motion for trait

## Species distribution models

aboitic associations – traditional, maxent biotic associations – cats/maxent assembly fourth-corner – where we are

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## Question

Why do taxa go extinct at different rates?

## Motivating questions

- ► How do mammal species traits affect extinction risk?
  - ► How do shared time of origination or evolutionary history relate to extinction risk?
- ▶ How do my findings compare to current risk factors?
  - ► Is species extinction risk age-independent?

## Relationship between range size and extinction risk

(Harnik and Simpson 2013 Proc B)

# Hypotheses of effects of dietary category

# Hypotheses of effects of locomotor category

# Hypotheses of effects of locomotor category

# Survival model diagram

# Pattern of species survival under two models

# Effect of dietary category on extinction risk

# Effect of locomotor category on extinction risk

## Difference in risk between origination cohorts

## Three sources of variance

#### Conclusions

- ► Survival of the unspecialized as time-invariant generalization.
- Decrease in extinction risk with time.
  - ▶ Both cohort/temporal and phylogenetic effect.
- Some incongruence with risk factors in the Recent.
  - e.g. effect of body size, trophic category, phylogenetic clustering.

## Observation

At K/Pg mass extinction, biological traits (except geographic range) have no effect on bivalve taxonomic survival.

(Jablonski, 1986, Science)

## Questions and analysis

- ► How do the effect of emergent traits on duration (extinction selectivity) vary with expected duration (extinction intensity)?
- ► **Approach:** hierarchical Bayesian survival model; effect estimates vary with origination cohort; correlation btw effects modeled.

## Intensity and selectivity

### Brachiopods

(ComputerHotline, wikimedia CC BY 2.5; Dwergenpaartje, wikimedia CC BY-SA 3.0)

### Post-Cambrian Paleozoic brachiopod genera and covariates

- ▶ time range approx. 488-252 Mya.
- stage as time unit; duration measured in stages (2-5 My each)
- multiple emergent traits analyzed; estimates vary by origination cohort
  - geographic range
  - body size
  - environmental preference (v, v²)
- gap statistic as measure of sampling (Foote and Raup 1996 Paleobio)
   imputed for taxa with short durations

### New measure of taxon's environmental affinity

(# epicontinental / total # occurrences) is what quantile of the distribution of all other background occurrences Beta( $\alpha$ ,  $\beta$ ).

- $\alpha$  is the # epicontinental background occurrences (+ 1).
  - $\blacktriangleright$   $\beta$  is the # open ocean background (+ 1).

### Measure of sampling and imputed values

Sampling is measured as the gap statistic r: (number of bins with an occurrence - 2) / (duration in bins - 2)

Can only be estimated for taxa with duration of three or more. Have to impute (e.g. fill-in) the values for all other taxa  $r^*$ .

$$s \sim \mathsf{Beta}(\phi, \lambda)$$
  
 $\phi = \mathsf{logit}^{-1}(W\gamma)$   
 $s^* \sim \mathsf{Beta}(\phi^*, \lambda)$   
 $\phi^* = \mathsf{logit}^{-1}(W^*\gamma)$ 

Note: Beta distribution parameterized in terms of mean  $\phi$  and total count  $\lambda$ . Also, this presentation excludes final (hyper)priors.

### Sampling statement for the joint posterior probability

$$\mu_{intensity} \sim \mathcal{N}(0,5)$$
 $\mu_{range} \sim \mathcal{N}(-1,1)$ 
 $\mu_{envpref} \sim \mathcal{N}(0,1)$ 
 $\mu_{envpref} \sim \mathcal{N}(0,1)$ 
 $\mu_{envpref} \sim \mathcal{N}(0,1)$ 
 $\mu_{envcurve} \sim \mathcal{N}(1,1)$ 
 $\mu_{envcurve} \sim \mathcal{N}(1,1)$ 
 $\mu_{size} \sim \mathcal{N}(0,1)$ 
 $\mu_{size} \sim \mathcal{N}(0,1)$ 

Note: Calculation of log probability of right and left censored observations is modified from the above

### Hierarchical survival model

# Model adequacy

### Variation in trait effects between cohorts

# Overall effect of environmental preference

### Change in effect of environment between cohorts

### Change in effect of environment between cohorts

### Correlation of effects between cohorts

#### Effect summary

- ► Effect of geographic range consistent with prior expectations; low variance.
  - ▶ No effect of body size; low variance.
  - ► Epicontinental environmental preference slightly favored on averaged; high variance.
- Strong support for survival of unspecialized as generalization wrt environmental preference; medium variance.

### Macroevolutionary process

- ► Magnitude of effect of geographic range and environmental preference increase with extinction intensity.
  - ► As extinction risk decreases, the differences between taxa matter less.

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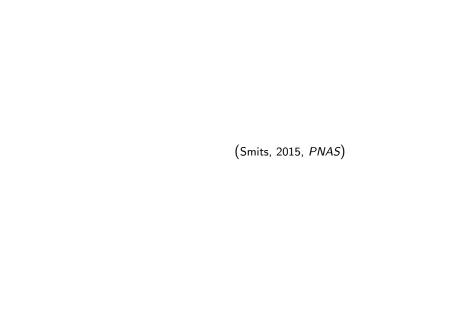
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### Eco-cube and ecotypes

(Bambach et al., 2007, Palaeontology)

### Fourth-corner modelling problem

(Brown et al., 2014, Methods Ecol. Evol.)

### Paleontological fourth-corner model

#### Covariates of interest

# individual-level (species i at time unit t)

- log-odds of occurrence probability at time t
- effect of locomotor type
  - arboreal, digitigrade, plantigrade, unguligrade, fossorial, scansorial
- effect of dietary type
  - carnivore, herbivore, insectivore, omnivore
- effect body size (rescaled log body mass)

#### group-level (2 My time unit t)

- overall mean of log-odds of occurrence probability
- temperature record based on Mg/Ca estimates
  - mean and interquartile range of rescaled value
- plant community phase following Graham 2011

#### Model of taxon occurrence

- response is p/a of genus in NA at time t
  - Bernoulli variable
  - probability is (observation prob) times ("true" presence)
- observation probability is effect of sampling/fossil record
- the latent discrete "true" presence modeled as a multi-level logistic regression
  - individual- and group-level covariates

### Paleontological fourth-corner model

### Model and sampling statement definition

```
y_{i,t} \sim \text{Bernoulli}(p_{i,t}z_{i,t})
                                                                                                                                      \Sigma^{\phi} = \operatorname{diag}(\tau^{\phi})\Omega^{\phi}\operatorname{diag}(\tau^{\phi})
p_{i,t} = \operatorname{logit}^{-1}(\alpha_0 + \alpha_1 m_i + r_t)
                                                                                                                                      \Sigma^{\pi} = \operatorname{diag}(\tau^{\pi})\Omega^{\pi}\operatorname{diag}(\tau^{\pi})
    r_t \sim \mathcal{N}(0, \sigma)
                                                                                                                                          \rho \sim \mathsf{U}(0,1)
  \alpha_0 \sim \mathcal{N}(0,1)
                                                                                                                                       b_1^{\phi} \sim \mathcal{N}(0,1)
  \alpha_1 \sim \mathcal{N}(1,1)
                                                                                                                                       b_1^{\pi} \sim \mathcal{N}(0,1)
    \sigma \sim \mathcal{N}^+(1)
                                                                                                                                       b_2^{\phi} \sim \mathcal{N}(-1,1)
z_{i,1} \sim \mathsf{Bernoulli}(\phi_{i,1})
                                                                                                                                       b_2^{\pi} \sim \mathcal{N}(-1,1)
z_{i,t} \sim \mathsf{Bernoulli}\left(z_{i,t-1}\pi_{i,t} + \sum_{i=1}^{t} (1-z_{i,x})\phi_{i,t}\right)
                                                                                                                                      \gamma^{\phi} \sim \mathcal{N}(0,1)
                                                                                                                                      \gamma^{\pi} \sim \mathcal{N}(0, 1)
\phi_{i,t} = \text{logit}^{-1}(a_{t,i[i]}^{\phi} + b_1^{\phi}m_i + b_2^{\phi}m_i^2)
                                                                                                                                       	au^{\phi} \sim \mathcal{N}^+(1)
                                                                                                                                       	au^\pi \sim \mathcal{N}^+(1)
\pi_{i,t} = \text{logit}^{-1}(a_{t,i[i]}^{\pi} + b_1^{\pi}m_i + b_2^{\pi}m_i^2)
                                                                                                                                      \Omega^{\phi} \sim \mathsf{LKJ}(2)
  a^{\phi} \sim \text{MVN}(U\gamma^{\phi}, \Sigma^{\phi})
                                                                                                                                      \Omega^{\pi} \sim LKJ(2).
  a^{\pi} \sim \mathsf{MVN}(U\gamma^{\pi}, \Sigma^{\pi})
```

Note: Product term ensures taxon-loss is permanent. Implementation in Stan marginalizes over all possible (range-through) values of z instead of estimating the

#### Parameter estimation and inference

- ▶ full HMC/MCMC slow
- Automatic Differentiation Variational Inference (ADVI)
  - approximate Bayesian inference
  - assumes posterior is Gaussian
  - true but approximate
     Bayesian posterior

### Posterior predictive performance

### Effect of mass on log-odds of observation

### Effect of mass on log-odds of occurrence

### Probability of ecotype origination

### Probability of ecotype survival

Group-level effects (plant phase, climate)

### Total species pool diversity and diversification

# Ecotype-specific diversity

# Ecotype-specific origination

# Ecotype-specific extinction

#### Concerns and conclusions

- basic and full models have similar results until Neogene
- posterior predictive simulations disimilar to observed; poor model adequacy
  - previous work has never evaluated model adequacy
  - second-order Markov process?
  - full posterior inference?
- decreasing ability to discern arboreal taxa over time (absence/increased rarity)
- increase in scansorial taxa over time
- increase in herbivorous taxa over time
- plant phase has small, idiosyncratic effects

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