

Taxon occurrence as a function of both biological traits and environmental context

the changing North American species pool

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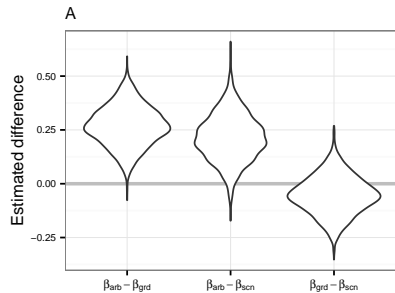
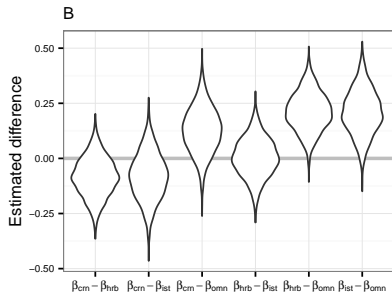
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Question

When are certain ecologies/ecotypes enriched or depleted?

Differences in extinction risk



(Smits, 2015, *PNAS*)

Species pool concept

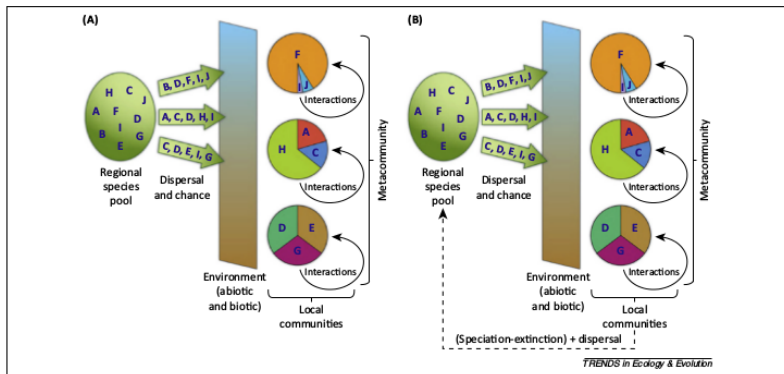
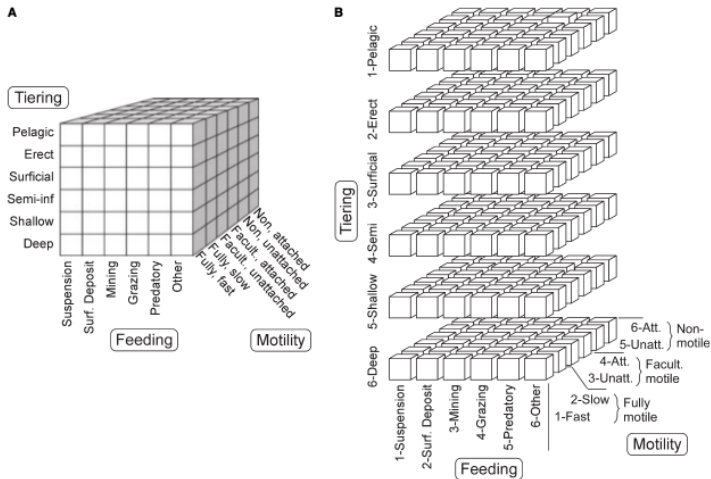


Figure 1. Two models of community assembly. (A) Local communities comprise a subset of species from the regional species pool that have passed through environmental filters. There is no feedback from the metacommunity (collection of local communities) to the regional species pool. Adapted from [5]. (B) Local communities are assembled as in (A), but speciation adds new species to the pool, extinction removes others, and dispersal allows the persistence of species that might otherwise go extinct.

(Mittelbach and Schemske, 2015, *TREE*)

Eco-cube and ecotypes



TEXT-FIG. 1. Ecospace as defined by the three axes of tiering, motility level and feeding strategy. A, the ecospace cube with categories on each axis labelled. B, the ecospace cube 'exploded', showing 216 'bins' or modes of life specified by the combination of the categories on each ecospace axis.

(Bambach *et al.*, 2007, Palaeontology)

The fourth-corner problem

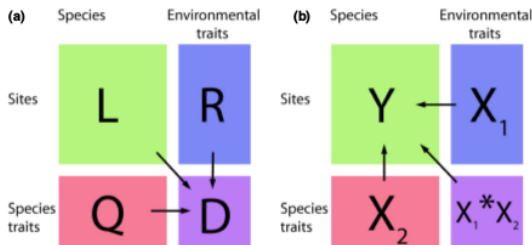
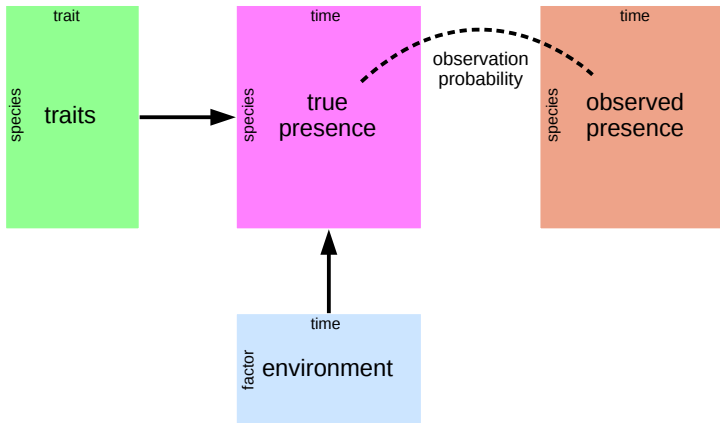


Fig. 1. Graphical representation of the fourth-corner problem and its solution. (a) The problem as posed by Legendre, Galzin & Harmelin-Vivien (1997), where the goal is to combine abundance (**L**), trait (**Q**) and environment (**R**) data in some way, to determine a matrix describing the trait–environment relationship (**D**). (b) The proposed model-based solution to the fourth-corner problem, where the goal is to predict abundance (**Y**) as a function of predictor variables for environment (**X₁**), species traits (**X₂**) and their interaction (**X₁*X₂**). The matrix of coefficients for the interaction between **X₁** and **X₂** is the fourth corner.

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

Paleo-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)

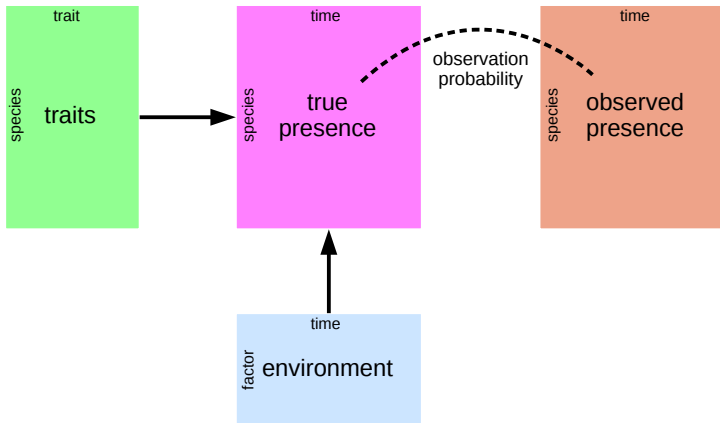
group-level (2 My time unit t)

- ▶ 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

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

Paleo-fourth corner model



Model and sampling statement definition

$$y_{i,t} \sim \text{Bernoulli}(\rho_t z_{i,t})$$

$$\text{logit}(\rho_t) \sim \mathcal{N}(\rho', \sigma_\rho)$$

$$z_{i,t} \sim \text{Bernoulli}(\theta_{i,t})$$

$$\text{logit}(\theta_{i,t}) = z_{i,t-1}(\alpha_t + X_i \beta_{t-}) + \left(\prod_{k=1}^{t-1} 1 - z_{i,k} \right) (\alpha_t + X_i \beta_{t-})$$

$$\beta_{t,d} \sim \mathcal{N}(\mu_d, \sigma_d)$$

$$\alpha_t \sim \mathcal{N}(\mu + \phi_{p[t]} + U_t \gamma, \sigma_\mu)$$

$$\phi_p \sim \mathcal{N}(0, \sigma_\phi)$$

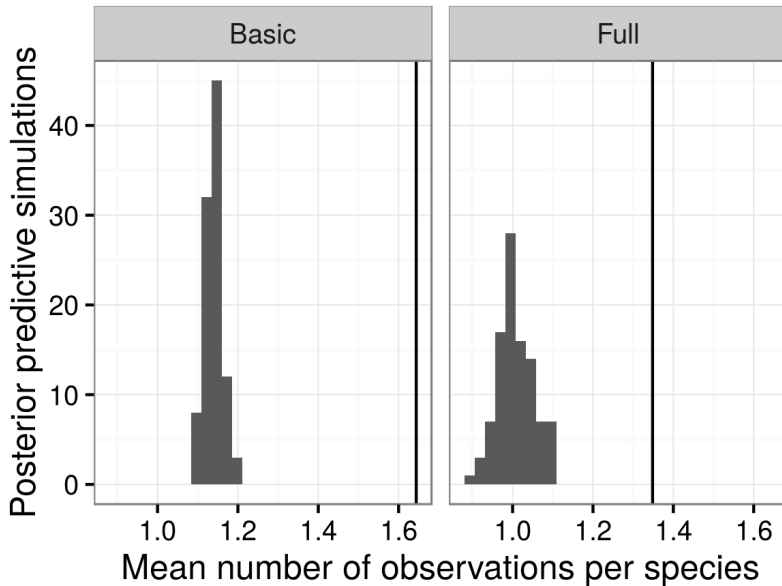
Note: Product term ensures taxon-loss is permanent. Implementation in Stan marginalizes over all possible (range-through) values of z instead of estimating the discrete parameters. I also use a noncentered parameterization of the hierarchical effects for better posterior sampling behavior. This presentation excludes final (hyper)priors.

Parameter estimation and inference

- ▶ full HMC/MCMC slow
- ▶ Automatic Differentiation Variational Inference (ADVI)
 - ▶ approximate Bayesian inference
 - ▶ assumes posterior is Gaussian, no correlation between parameters
 - ▶ true Bayesian posterior

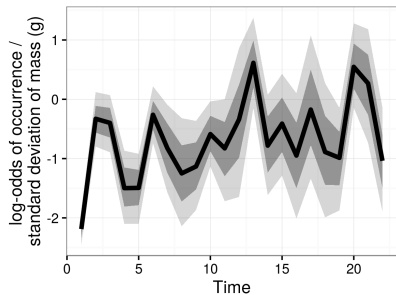


Posterior predictive performance

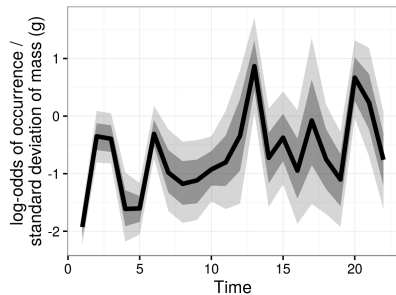


Effect of mass on log-odds of occurrence

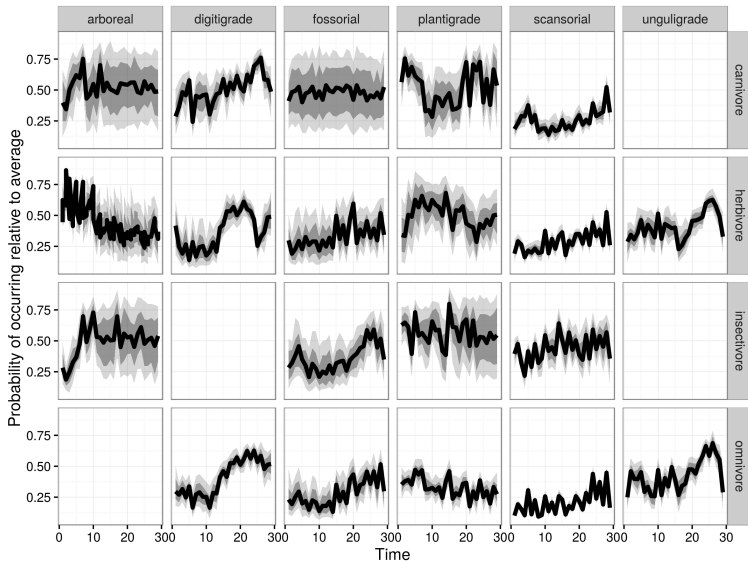
Basic model



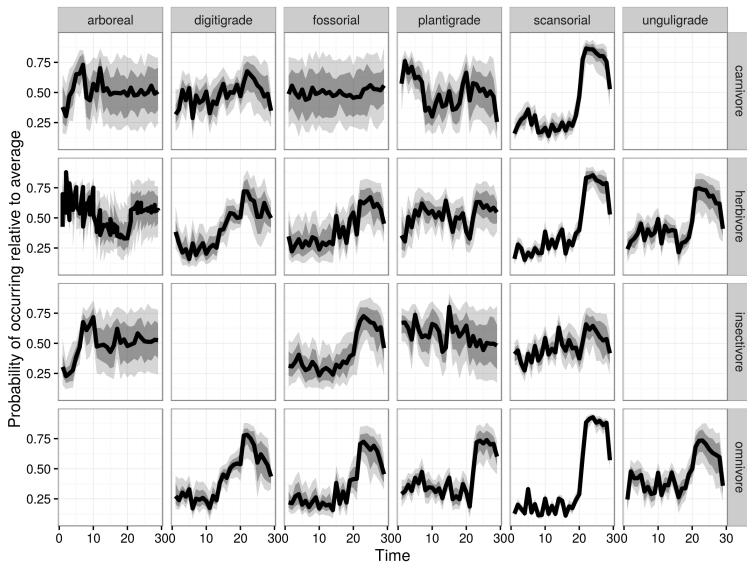
Full model



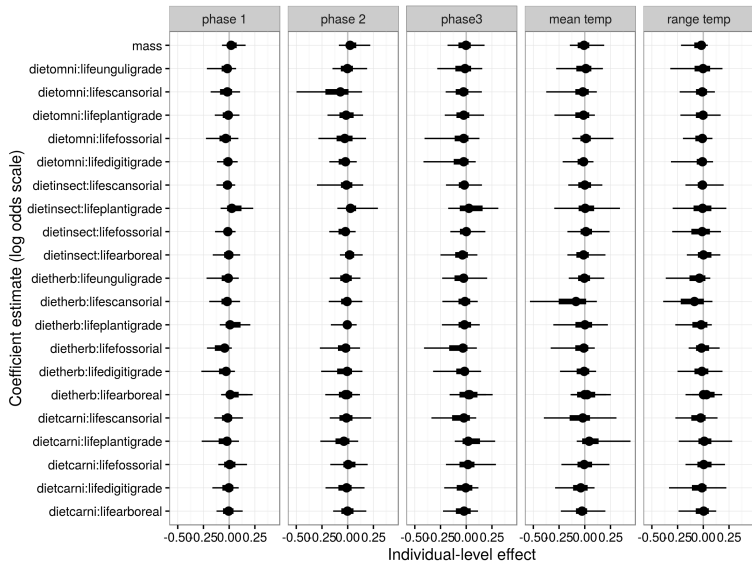
Probability occurrence is of ecotype (basic model)



Probability occurrence is of ecotype (full model)



Group-level effects (climate, plant phase)



Concerns and conclusions

Acknowledgements

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 - ▶ Kenneth D. Angielczyk,
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