

How predictable is extinction?

Forecasting species survival at million-year timescales

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Foundational assertion of conservation paleobiology

By studying the **past**,
we can better predict the **future**.

What are we predicting?

Extinction is **hard** to predict, but is **important** to conservation decisions.

Predicting extinction

- ▶ A taxon with a **greater than average** global geographic range is likely to **survive for longer** than a taxon with **less than average** global geographic range.

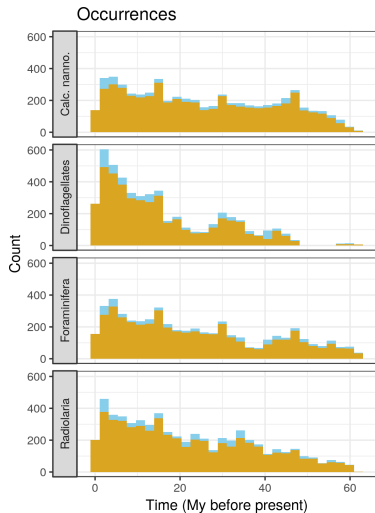
Predicting extinction

- ▶ A taxon with a **greater than average** global geographic range is likely to **survive for longer** than a taxon with **less than average** global geographic range.
- ▶ A taxon's global geographic range can change over time.

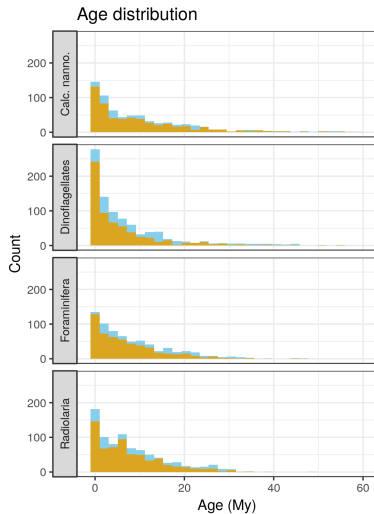
Predicting extinction

- ▶ A taxon with a **greater than average** global geographic range is likely to **survive for longer** than a taxon with **less than average** global geographic range.
- ▶ A taxon's global geographic range can change over time.
- ▶ What happens to extinction risk as a taxon changes geographic range? How is extinction risk impacted if that taxon's global geographic range has recently **increased** or **decreased**?

Data being analyzed



Occurrence type Last Standard

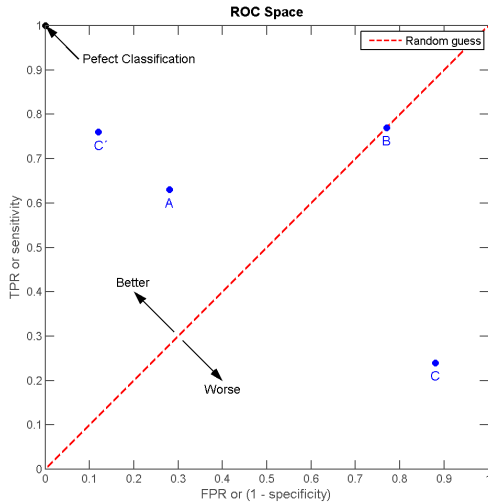


State Extant Extinct

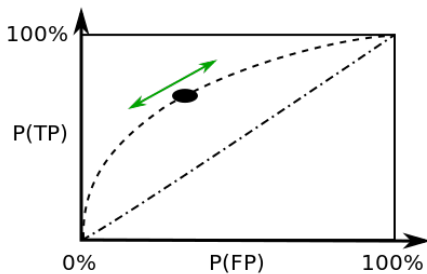
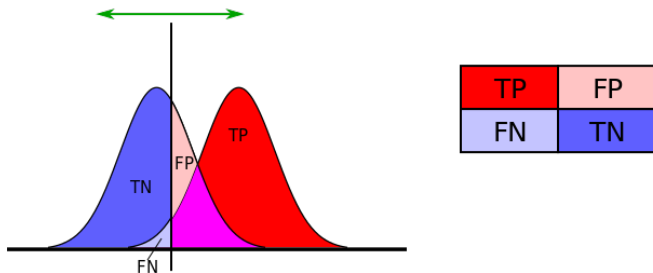
How we're analyzing the data

- ▶ Encoding the past
 - ▶ Change in geographic range between current observation and previous observation.
 - ▶ Average global temperature at time of previous observation (Mg/Ca isotope).
 - ▶ Age in millions of years at time of observation.
- ▶ Explore model adequacy using posterior predictive distribution.
- ▶ Estimate out-of-sample predictive performance using k -fold cross-validation.

Measuring performance: Receiver Operating Characteristic



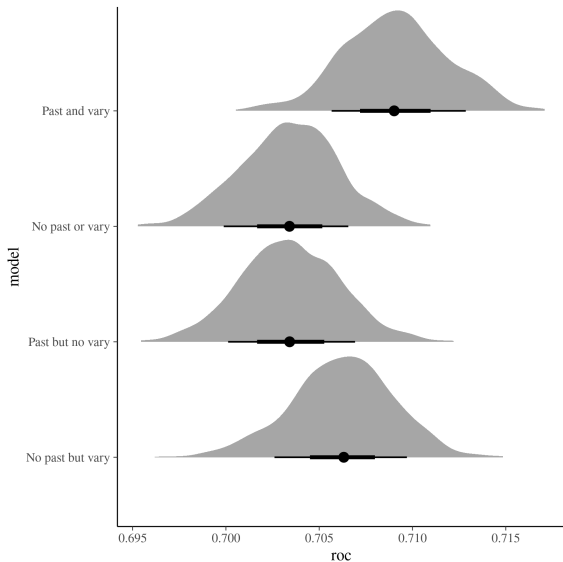
Measuring performance: Receiver Operating Characteristic



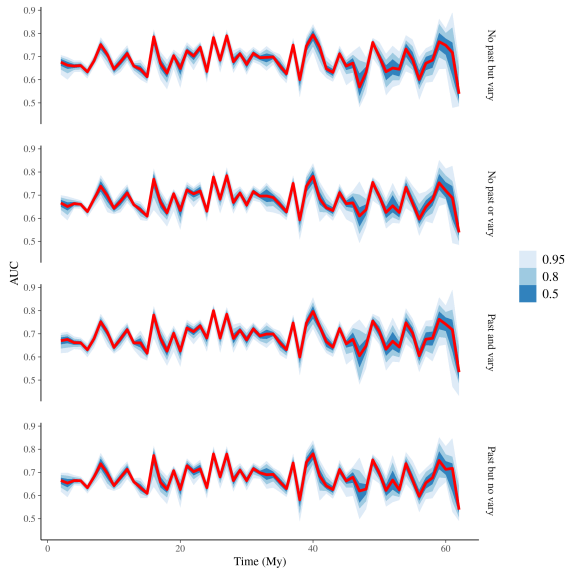
(wikimedia)

A conceptual model for predicting extinction

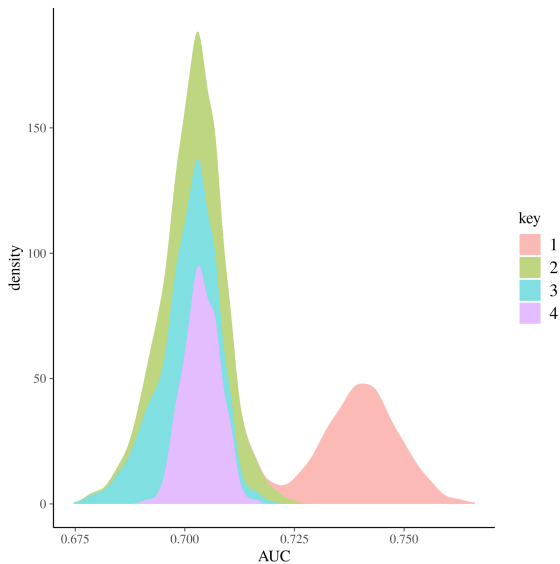
In-sample predictive performance, full dataset



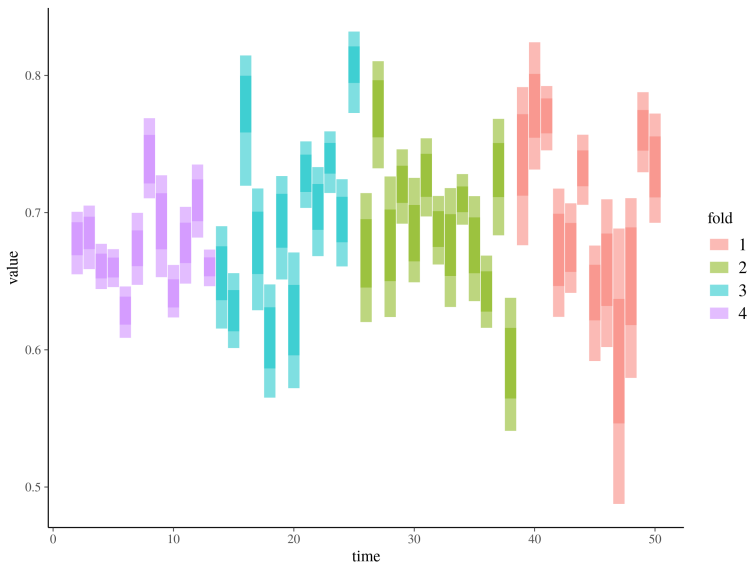
In-sample predictive performance, by time



Cross-validation results, full dataset



Cross-validation results, by time



Summary

Conclusions

Acknowledgements

A statistical model for predicting extinction

$$t_i \sim \text{Bernoulli}(\Theta)$$

$$\Theta_i = \text{logit}^{-1}(X_i B_{w[i],p[i]} + A_{d[i],p[i]})$$

$$B_{w,p} \sim \text{MVN}(\alpha_p, \Sigma_B)$$

$$\alpha_p \sim \text{MVN}(\mu, \Sigma_\alpha)$$

$$A_{w,p} \sim \text{MVN}(\delta_p, \Sigma_A)$$

$$\delta_p \sim \text{N}(0, \sigma_\delta)$$

$$\mu_d \sim \begin{cases} N(0, 10) & \text{if } d = 1 \\ N(-1, 3) & \text{if } d = 2 \\ N(0, 3) & \text{if } d > 2 \end{cases} \Sigma_B = \text{diag}(\tau_B) \Omega_B \text{diag}(\tau_B)$$

$$\Sigma_\alpha = \text{diag}(\tau_\alpha) \Omega_\alpha \text{diag}(\tau_\alpha)$$

$$\Sigma_A = \text{diag}(\tau_A) \Omega_A \text{diag}(\tau_A)$$

$$\tau_B \sim C^+(5)$$

$$\tau_\alpha \sim C^+(5)$$

Effects of age on extinction risk, by phylum

