

Continuous characters

- ML ancestral states for continuous characters under BM are equivalent to ancestral states estimated under weighted squared change parsimony
- Choose the ancestral states that minimize the amount of squared change on the tree

Maximum likelihood

- Find model parameters that maximize the likelihood of obtaining the data
- Calculate $P(a=x_i \mid \text{model, tree, data})$
- Can generate confidence intervals based on t-distribution (Schluter 1997)

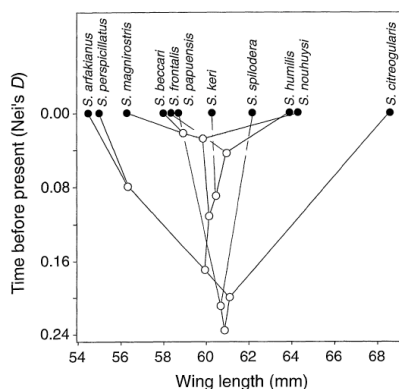
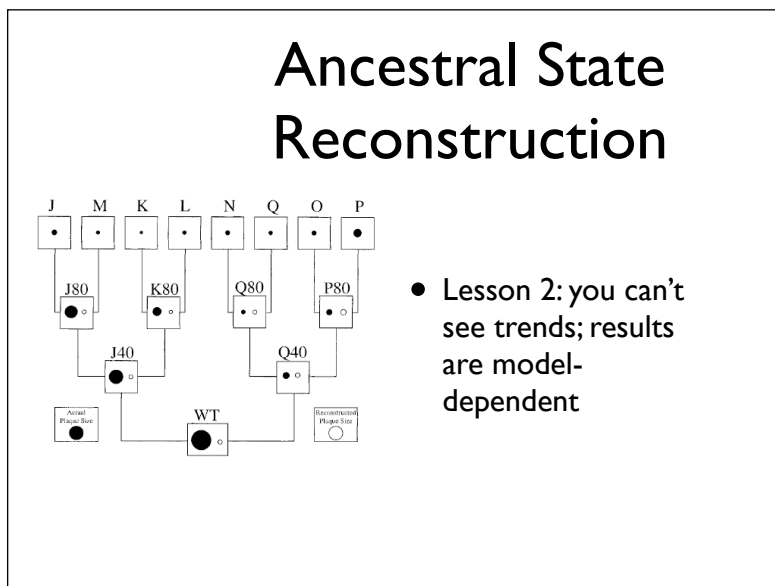
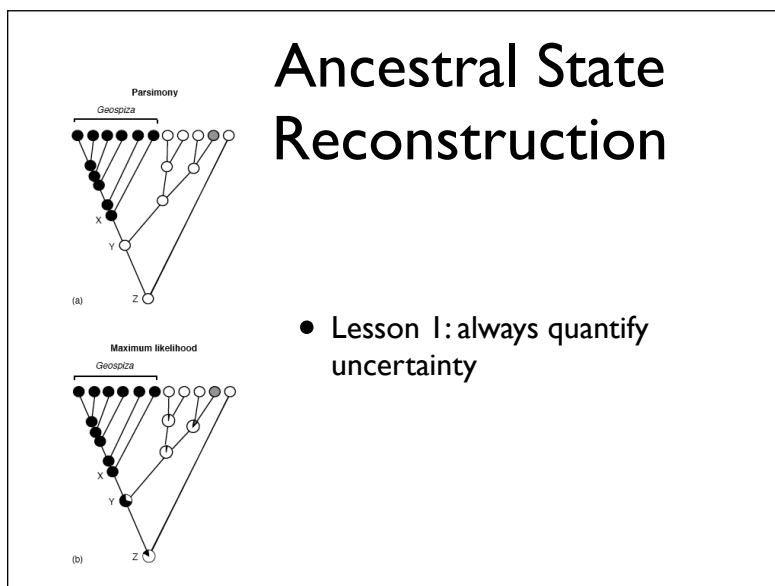
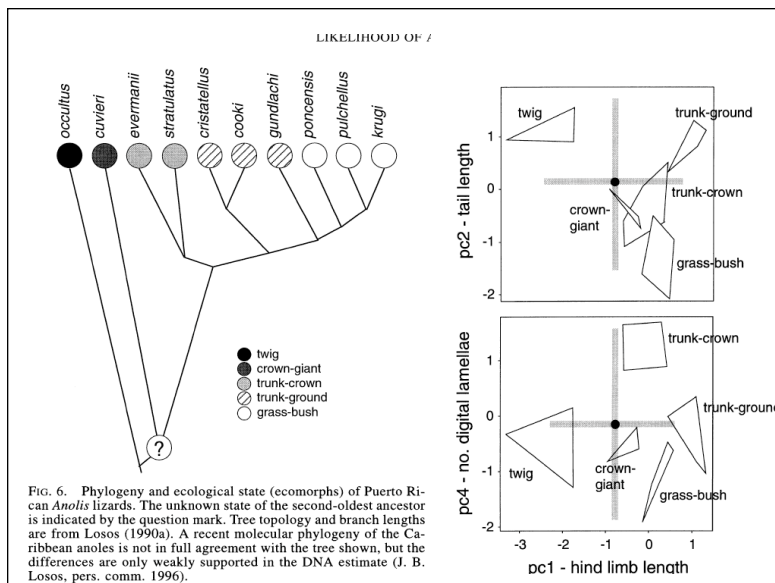


FIG. 8. Wing lengths of Australian and New Guinea scrubwrens (*Sericornis*) and of their ancestors. The tree is the same as in Figure 3. Shaded lines span the support limits for each ancestor state. Marginal distributions were computed separately for each ancestor one at a time.



Summary

- We can test for character correlations, or other patterns of interest, with discrete characters
- It is possible to estimate ancestral character states and incorporate uncertainty

Comparative Methods and Quantitative Genetics

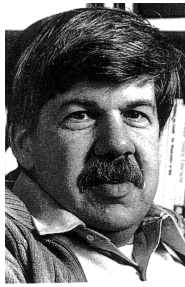
April 14, 2009

Quantitative Genetics and Macroevolution

- Comparative biology has the opportunity to test long-term prediction of quantitative genetic models
- How much change would we expect if we extrapolate from present-day trends?
- How much change do we actually see in the fossil record?

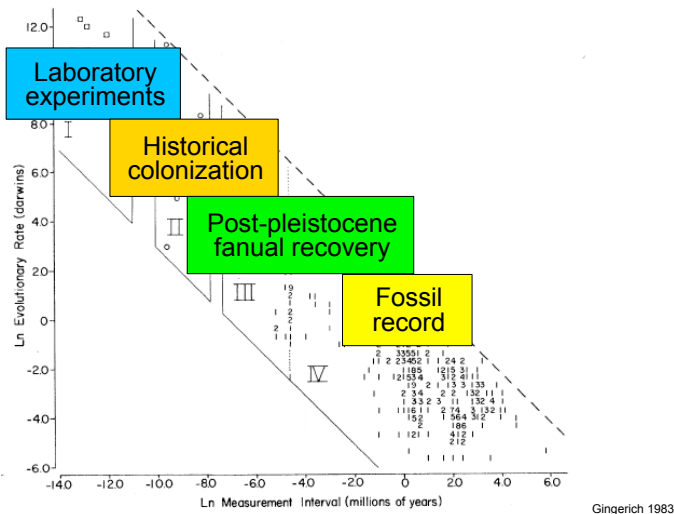
Roots of the Problem

- Gould pondered the connection between micro- and macroevolution
- Identified the problem of stasis: why do we see so little change in so many groups, even over long time periods?

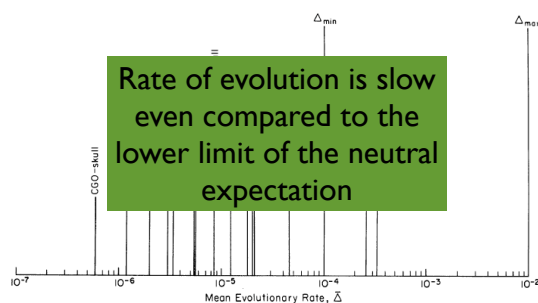


PROFESSOR STEPHEN JAY GOULD

A General Pattern in Macroevolution



Lynch (1990) Am. Nat.



What is the point?

- We can learn things by comparing results from comparative biology to microevolutionary models
- This helps bridge the gap between micro- and macroevolution

VOL. 169, NO. 2 THE AMERICAN NATURALIST FEBRUARY 2007

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Resolving the Paradox of Stasis: Models with Stabilizing Selection Explain Evolutionary Divergence on All Timescales

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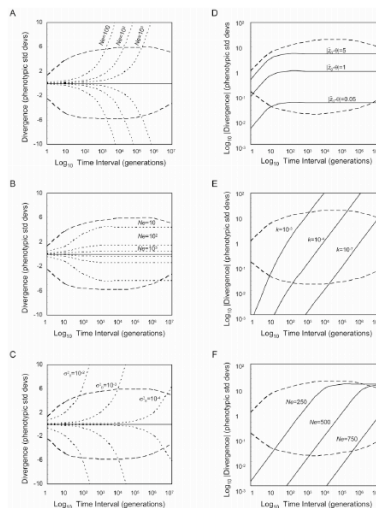


Figure 6. Example fits of central (A), white noise (B), Brownian motion of the optimum (C), displaced optimum (D), moving optimum (E), and peak shift (F) models to the divergence data. Heavy dashed lines in each panel represent the 95% confidence ellipse for the divergence data of Gage et al. (2001). If model predictions fell within this ellipse, we considered the fit to the data acceptable. To illustrate how each model represents

Two New Models

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- What is Felsenstein up to now?



Threshold models

- Problem with analyzing quantitative characters:
- Number of parameters increase too fast with multivariate data sets
- Biologically unrealistic patterns; easy to reverse states even after long periods of time

Sewell Wright

- Introduced threshold model for guinea pig toes in 1934

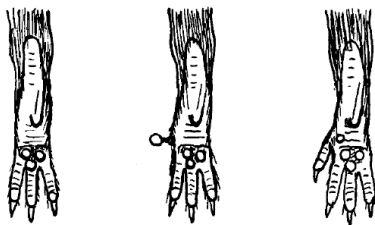


FIGURE 1.—Sketches of left hind feet of guinea pigs. The normal 3-toed condition is shown at the left. At the right is a foot with well-developed little toe and a corresponding plantar tubercle. Between is a grade of imperfect development of the little toe.

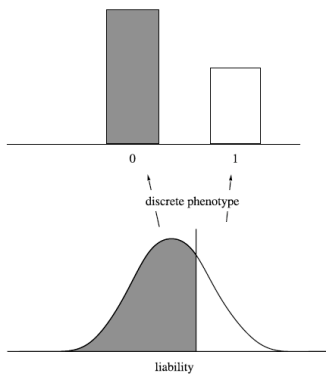


Figure 1. The threshold model of quantitative genetics, showing the continuous distribution of the underlying liability characters, and the resulting discrete distribution of the observed phenotype.

p characters, n species:

$$L = \text{Prob}(y; \mu, C)$$

$$= \int_{x_{11} \in R_{11}} \int_{x_{12} \in R_{12}} \cdots \int_{x_{1p} \in R_{1p}} \int_{x_{21} \in R_{21}} \cdots \int_{x_{np} \in R_{np}} \phi(x; \mu, C), \quad (2.2)$$

Hard to calculate:
Felsenstein suggests MCMC

General approach

- Model evolution of liabilities l_i
- These evolve under a Brownian motion model with some variance-covariance matrix, but cannot be observed
- Use likelihood or Bayesian MCMC to fit model parameters

Advantages

- Fewer parameters
 - Discrete Markov, 10 variables, >1000 parameters
 - Threshold model, $p(p+1)/2 = 55$ params
- Characters that just changed are more likely to change back.
- Allows for polymorphism

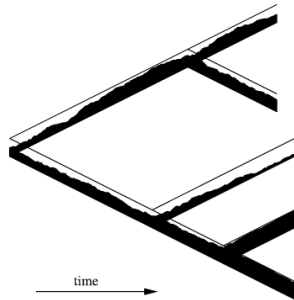


Figure 2. A simulation of a threshold model with one character, evolving by Brownian motion of the liability on a simple five-species tree. Each lineage is shaded to indicate the proportion of individuals in state 1 (shown by black shading). The upper two species end up with polymorphism, the middle species nearly all with state 0, and the lower ones nearly all with state 1.

Testing for correlations with the threshold model

- If liabilities are correlated then the two characters are evolving in a correlated manner
- Can test this using model-fitting approach

Two New Models

- Threshold models
- Intra- and interspecific contrasts

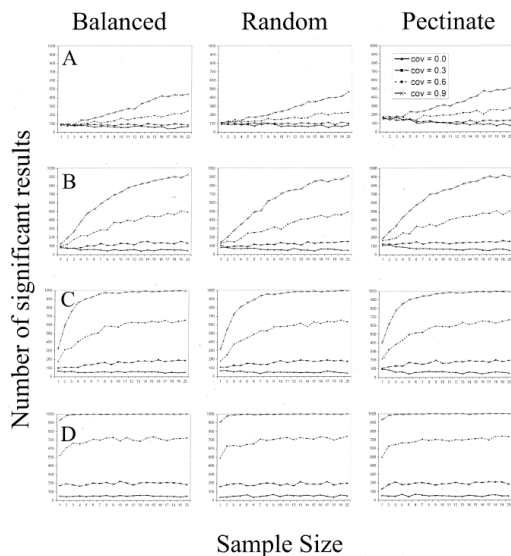


Contrasts Revisited

- There is no clean line separating phylogenetic trees from intraspecific genealogies
- Comparative methods almost all take species values as fixed and estimated perfectly
- But traits vary in the natural world, and we estimate them with error

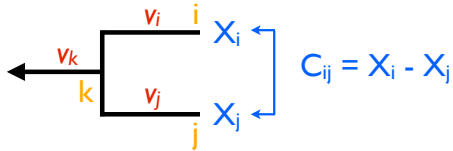
Contrasts Revisited

- Why is there a problem?
- We assume that species means are estimated without error in most comparative methods
- This can lead to statistical problems



Calculating Contrasts*

1. Find two tips on the phylogeny that are adjacent (say nodes i and j) and have a common ancestor, say node k
2. Compute the contrast $X_i - X_j$. This has expectation zero and variance proportional to $v_i + v_j$



Also inflates rates

$$v_{ij} = v_i + v_j$$

Contrasts

$$v_{ij} = v_i + v_j + \frac{\sigma_i^2}{n_i} + \frac{\sigma_j^2}{n_j}$$

Actual

This leads to inflated values for independent contrasts, especially towards the tips of the tree

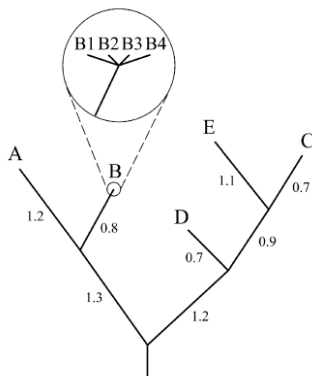
Felsenstein's model

- Character evolution on the tree is described by an evolutionary variance-covariance matrix R
- Remember:

- All characters for all species are drawn from a multivariate normal distribution
- $V = R \otimes C$
- \otimes is the *Kronecker product*

Felsenstein's model

- Characters also vary within species according to a (distinct) variance-covariance matrix W
- Assumed to be the same for each species in the tree



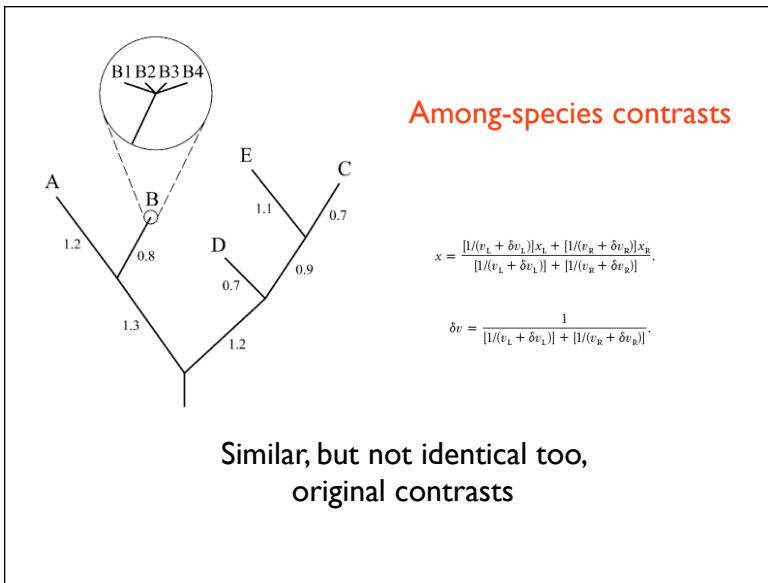
Within-species contrasts

$$c_1 = f_1(x_{B1} - x_{B2}),$$

$$c_2 = f_2 \left[x_{B3} - \left(\frac{1}{2} x_{B1} + \frac{1}{2} x_{B2} \right) \right],$$

$$c_3 = f_3 \left[x_{B4} - \left(\frac{1}{3} x_{B1} + \frac{1}{3} x_{B2} + \frac{1}{3} x_{B3} \right) \right]$$

$$f_m = \sqrt{\frac{m}{m+1}}$$



Advantages

- Can estimate both within- and among-species variances and covariances
- Could potentially connect this even better to data collected within species
- Can try to reconstruct selection vectors necessary to get pattern

Disadvantages

- Not really independent contrasts in the same sense as originals
- Cannot obtain a set of standardized independent contrasts
- Requires iterative method, fancy ML estimation of the two covariance matrices

Two New Models

- Threshold models
- Intra- and interspecific contrasts