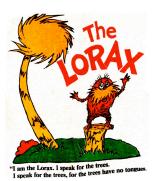
# Overview of Comparative Methods

- Learning from the trees
- General statistical framework
- Basic models of growing trees and changing characters





# General statistical framework

- We will focus on a model fitting approach
- Develop mathematical models
- Calculate likelihoods
- Compare the fit of a set of competing models

#### Core Models

- Birth-death model for diversification
- Mk model for discrete characters
- Brownian motion model for continuous characters

#### Markov Models

- Most of the models used in comparative biology satisfy the Markov property:
- The conditional probability distribution of future states of the process, given the present state and a constant number of past states, depends only upon the present state and the given states in the past, but not on any other past states

Expected species diversity under a birth-death model:

$$E[N_t] = N_o e^{(b-d)t}$$

 $N_t$  = species diversity after time t

 $N_o$  = starting species diversity

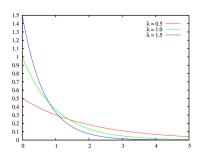
b = birth rate

d = death rate

t = time

## Waiting times

 Under a birth-death model, the time intervals between successive events are always drawn from exponential distributions



# Growing birth-death trees

- Same protocol, except waiting times are drawn from exponential distribution with λ = n (b + d), where n is the number of living species at that time
- Choose a random lineage
- Decide whether you have a speciation or extinction event
- Pr[speciation] = b/(b+d); Pr[extinction]=d/(b+d)



#### **Outline**



- Review of BD model
- Diversity of sister clades
- Tree balance
- Newer approaches

#### **ERM Model**

- BD is a special case of the Equal Rates Markov model (ERM)
- ERM Model assumptions
  - Markov model: next step depends only on current state
  - Equal rates: b and d are constant across lineages at any given time
  - b and/or d might change through time

#### Probability of a split at one node

$$\Pr[(k,n-k)] = \frac{\frac{2}{n-1}}{\frac{1}{n-1}} \quad k \neq n-k$$

\* Under any ERM model, including BD

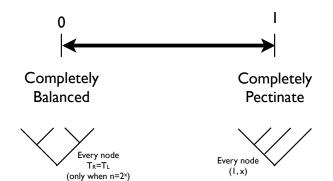
## I<sub>c</sub> (Colless 1982)

$$I_{\rm c} = rac{\sum\limits_{{
m (all interior nodes)}}\mid T_{
m R} - T_{
m L}\mid}{2} \,,$$

T<sub>R</sub>,T<sub>L</sub> = number of taxa subtended by "left" and "right" branches

Maximum possible value for  $I_c$  with a completely pectinate tree

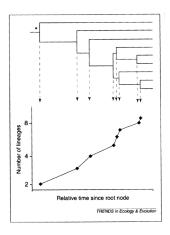
## I<sub>c</sub> (Colless 1982)



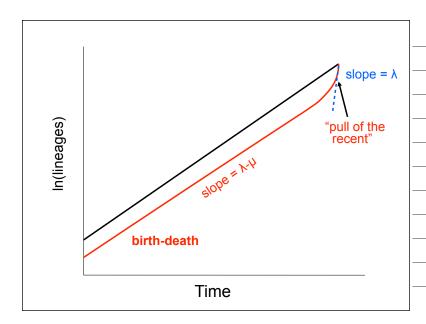
## Outline

- How do lineages accumulate under the birth-death model?
- Estimating speciation and extinction rates from trees
- Testing hypotheses about diversification

## Lineage-through-time



# 



$$Pr[t_1 = x] = 2\lambda e^{-2\lambda x}$$

$$Pr[t_2 = x] = 3\lambda e^{-3\lambda x}$$

$$Pr[t_3 = x] = 4\lambda e^{-4\lambda x}$$

$$Pr[t_4 = x] = e^{-5\lambda x}$$

likelihood = P[tree | 
$$\lambda$$
] =  $2\lambda e^{-2\lambda t1}$   $3\lambda e^{-3\lambda t2} 4\lambda e^{-4\lambda t3} e^{-5\lambda t4}$ 

In general,

likelihood = P[tree | 
$$\lambda$$
] =  $e^{-n\lambda x_n} \prod_{i=2}^{n-1} i\lambda e^{-i\lambda x_i} = (n-1)! \lambda^{n-2} e^{-\lambda s}$   

$$s = \sum_{i=2}^{n-1} ix_i$$

#### PB vs. BD

- Two simple ways to test for extinction in a tree
- Akaike Information Criterion (AIC)

AIC = 2 k - 2 ln L 
$$AICc = AIC + \frac{2k(k+1)}{n-k-1}$$
.

Choose the model with the lowest AIC score

Can also gauge support by the difference in AIC
by the difference in AIC
scores

$\Delta_{AIC}$	Interpretation
0-3	no support
3-7	weak
7-10	intermediate
>10	strong

#### Slowdown statistic

- Tests for longer branches near the tips of a phylogenetic tree
- Signature of an early, rapid radiation

$$\gamma = \frac{\left(\frac{1}{n-2}\sum_{i=2}^{n-1} \left(\sum_{k=2}^{i} k g_{k}\right)\right) - \left(\frac{T}{2}\right)}{T\sqrt{\frac{1}{12(n-2)}}}, \ T = \left(\sum_{j=2}^{n} j g_{j}\right)$$

#### Slowdown statistic

- Under pure-birth model,  $\gamma \sim N(0,1)$
- y<0 means diversification has slowed
- γ>0 difficult to interpret

$$\gamma = \frac{\left(\frac{1}{n-2}\sum_{i=2}^{s-1} \left(\sum_{k=2}^{i} kg_k\right)\right) - \left(\frac{T}{2}\right)}{T\sqrt{\frac{1}{12(n-2)}}}, T = \left(\sum_{j=2}^{s} jg_j\right)$$

# The problem... Slowdown Missing taxa Extinction Speedup

#### Outline - BM

- What is Brownian motion?
- When might characters evolve in a Brownian-like way?
- Simulating Brownian motion on trees
- Independent contrasts

## Three Facts Describe Brownian Motion

- Let W(t) be the value of the character at time t.Then:
  - E[W(t)] = W(0)
  - Successive steps are independent
  - $W(t) \sim N(W(0), \sigma^2 t)$

#### Parameters of BM

- Brownian motion models have two parameters:
  - $\Theta$ , the starting value;  $W(0) = \Theta$
  - $\sigma^2$ , the rate parameter

# Evolution might approximate BM...

- Genetic drift
- Random punctuated change
- Selection that is weak relative to the time interval considered
- Selection that changes randomly through time

## Simulating BM

- Simulating Brownian motion involves drawing values from normal distributions
- Variance of the distribution depends on  $\sigma^2$  and t
- Values along adjacent branches are added from the root to the tips of the tree

#### General form

- Tip data follow a multivariate normal distribution with mean vector Θ and variance-covariance matrix where
- $var(i) = \sigma^2(d_i)$ ;  $d_i$  = distance from root to tip i
- $cov(i,j) = \sigma^2(c_{i,j})$ ;  $c_{i,j}$  = shared path of tip i and j

TAB

THE FOUR CONTRASTS EXTRACTED FROM THE PHYLOGENY SHOWN IN FIGURE 9, EACH WITH ITS VARIANCE, ALL COMPUTED USING STEPS 1–4 IN THE TEXT

2	3	4 5	Contrast	Variance
$\langle v_2  v_3 \rangle$	V <sub>4</sub> \ /V <sub>5</sub>	$X_1 - X_2 \\ X_4 - X_5$	$ \begin{array}{c} \nu_1 + \nu_2 \\ \nu_4 + \nu_5 \end{array} $	
	\	y <sub>6</sub>	$X_3 - X_6 \ X_7 - X_8$	$v_3 + v_6' \\ v_7' + v_8'$
\	\	\	where	, ,

$$X_6 = \frac{v_4 X_5 + v_5 X_4}{v_4 + v_5}$$

$$v_6' = v_6 + v_4 v_3 (v_4 + v_5)$$

$$X_7 = \frac{v_2 X_1 + v_1 X_2}{v_1 + v_2}$$

$$v_7' = v_7 + v_1 v_2 (v_1 + v_2)$$

$$X_8 = \frac{v_6' X_3 + v_3 X_6}{v_3 + v_6}$$

 $v_8' \,=\, v_7' \,+\, v_3 \,\, v_6' / (v_3 \,+\, v_6')$ 

(Felsenstein 1981)

#### What are contrasts?

- Each standardized contrast is telling us something about the RATE of evolution
- The contrasts have a close relationship with  $\sigma^2$ , the rate parameter from BM
- The sum of the squared contrasts divided by n gives the ML estimate of  $\sigma^2$

# Using and Interpreting Contrasts

- Independent contrasts should be thought of as vectors
- They summarize information about the amount and direction of evolution at each node in the tree
- Standardized contrasts provide information about the **rate** of evolution

# Using and Interpreting Contrasts

$\hat{oldsymbol{\sigma}}^2$	=	$\sum S_i$
		n-1

- •This provides an unbiased estimate of evolutionary rate
- •The expected value of this estimate is equal to the actual rate parameter
- The maximum likelihood estimate of the rate parameter is biased

#### Character correlations

- Calculate independent contrasts for two characters, x and y
- Carry out a regression analysis of y on x with **no intercept** (force regression line through the origin)
- P < 0.05, then reject the null hypothesis of no evolutionary correlation

#### **Outline**

- Calculating the likelihood for a single character evolving under a BM model
- Alternative models for continuous character evolution
- Multivariate character evolution

## Likelihood for continuous characters on trees

- Given phylogeny, measurements of character y for each tip (y<sub>i</sub>)
- Choose a rate parameter  $\sigma^2$  and mean  $\Theta$
- Calculate the phylogenetic variancecovariance matrix for the tree V

## Likelihood for continuous characters on trees

- $y_i \sim MVN(\boldsymbol{\Theta}, \sigma^2 \boldsymbol{V})$
- Determine the probability of drawing the vector of  $y_i$  from the MVN distribution with mean  $\Theta$  and vcv  $\sigma^2 V$

# Analytic Solution for MLE

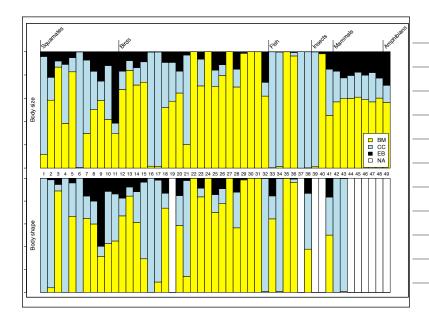
$$\hat{\sigma}^2 = \frac{(\mathbf{x} - \hat{a}\mathbf{1})'\mathbf{C}^{-1}(\mathbf{x} - \hat{a}\mathbf{1})}{n}$$

$$\hat{\mathbf{a}} = [(\mathbf{1}'\mathbf{C}^{-1}\mathbf{1})^{-1}(\mathbf{1}'\mathbf{C}^{-1}\mathbf{X})]'$$

x = vector of trait values, n = number of species,C = coancestry matrix (shared path lengths)

# Three Models for Phenotypic Evolution

- Single Rate (SR)
- Early Burst (EB)
- Constant Constraints (CC)



# What if you have more than one character?

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

#### Outline

- Example of the problem
- The Mk Model
- Alternatives to Mk

## The Mk Model

- Simple model for the evolution of discrete variables
- Constant instantaneous probability of changing between any two character states out of k possibilities
- Described by rate matrix Q:

$$Q = \alpha \begin{bmatrix} 1 - k & 1 & \dots & 1 \\ 1 & 1 - k & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 - k \end{bmatrix}$$

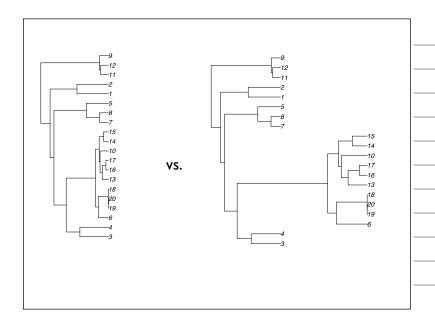
## In general:

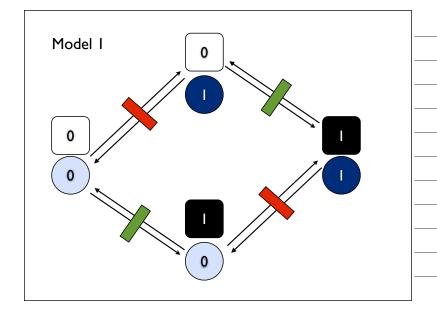
$$L_{(\tau)} = \sum_{r=1}^{k^{n-1}} \Pr(R_r \big| \tau)_{\mid}$$

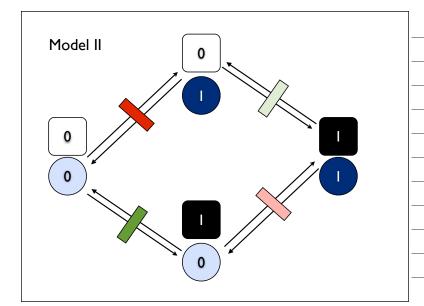
## Alternatives to Mk



$$Q = \begin{bmatrix} -\alpha & \alpha \\ \beta & -\beta \end{bmatrix}$$



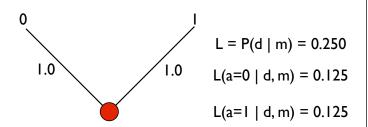




## Reconstructing Ancestral Character States

- Goal: to make inferences about the states of characters in ancestral lineages
- Usually make these inferences at nodes in the tree
- These inferences can be made using maximum likelihood

# Discrete Ancestral State Reconstruction

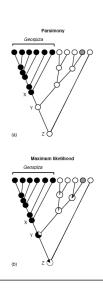


Model: Mk with  $q_{01} = q_{10} = 8$ 

### Continuous Characters

- Can also reconstruct ancestral characters using likelihood
- Assume characters are evolving under a Brownian motion model

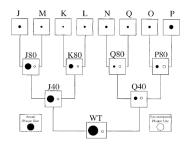
$$\Pr[y = x] = \frac{1}{\sigma\sqrt{2\pi t}} \exp\left(-\frac{(x-\theta)^2}{2\sigma^2 t}\right)$$



# Ancestral State Reconstruction

Lesson I: always quantify uncertainty

# Ancestral State Reconstruction



 Lesson 2: you can't see trends; results are modeldependent

#### Two New Models



## General approach

- Model evolution of liabilities li
- 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

### **Advantages**

- Can estimate both within- and amongspecies 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

# ML test of multiple rates

- Compare models where some parts of the tree have distinct speciation and/or extinction rates
- If you have likelihoods and the number of parameters, you can compare these models

# The BiSSE Model Speciation rate Extinction rate State 0 $\lambda_0$ $\mu_0$ State I $\lambda_1$ $\mu_1$

## Biogeography

- Goals of biogeographic studies
- Classic approaches
- Newer model-fitting methods for biogeography

