Ecologists push back

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FORUM

On misinterpreting the 'phylogenetic correction'

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Introduction

The last 10 years have seen important advances in methodology for taking phylogeny into account when analysing a comparative dataset. This commentary is about a class of interpretive procedures associated with these new statistical methods. We will call the interpretive logic 'phylogenetic correction' (PC for short), because this phrase summarizes the approach. Our essential message, however, is that a PC procedure is not in fact a 'correction', an adjustment to remove errors. Rather, it is a conceptual decision to give priority to one interpretation over another. Accordingly, it is an error to believe that PC is a methodology that must routinely be applied in all comparative analyses.

The present Forum was proposed by the Editor during the review process for a paper about comparative ecology of seed mass (Leishman et al. 1995; see p. 517). Accord-

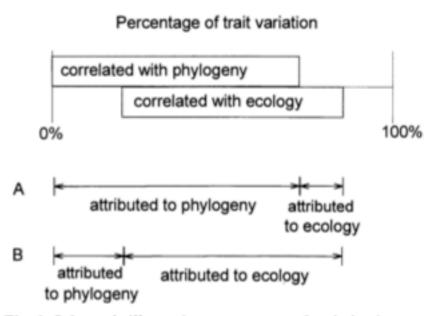
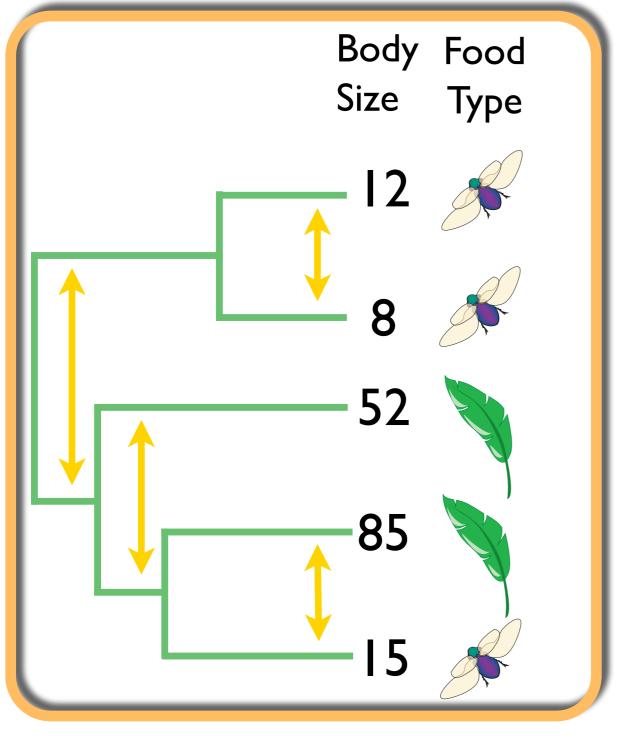


Fig. 1 Schematic illustrating percentages of variation between species in a trait such as seed mass correlated with phylogeny and ecology, and the two extremes (A and B) of the spectrum of possible attributions of variation. Extreme A is phylogenetic correction (PC).

Are we are attributing too much to evolution? (and not enough to ecology?)



Quantitative character associated with a particular selective "regime"

We want to know the correlation between morphology and ecology

Statistically remove the effects of phylogeny (using Brownian Motion)

All comparative analyses are constructed of 3 pieces

I. The comparative dataset (body size, plumage color, etc. for each species)

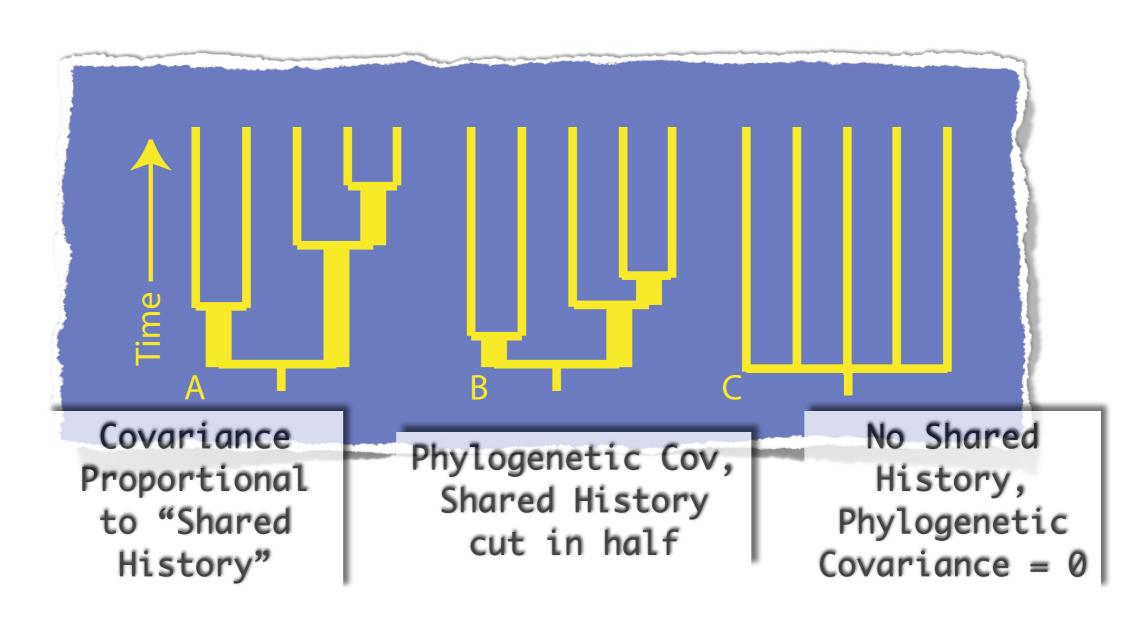
- 2. A particular pattern of shared ancestry (phylogeny and branch lengths)
- 3. A model of evolutionary change along each branch of the phylogeny

Together, these give you a predicted distribution for a phenotypic character among species

Most approaches use a BM (i.e., purely neutral) model

But... many datasets do not fit BM well.

A popular approach improves fit by "scaling" branch lengths



Why change the phylogeny?

1. The comparative dataset (body size, plumage color, etc. for each species)

2. A particular pattern of the ed ancestry (phylogeny and branch lengths)

3. A model of evolutionary change along each branch of the phylogeny

What about the Biology?

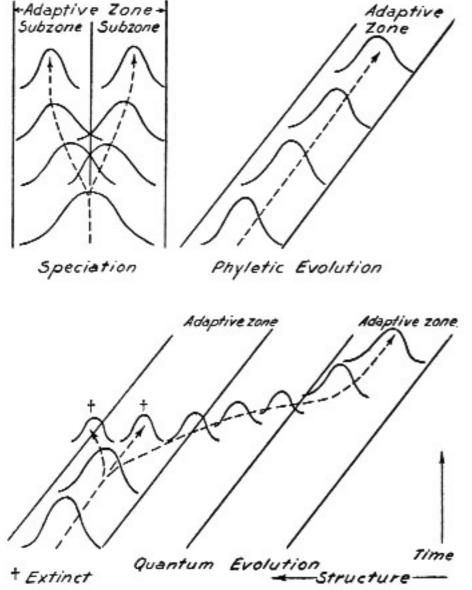
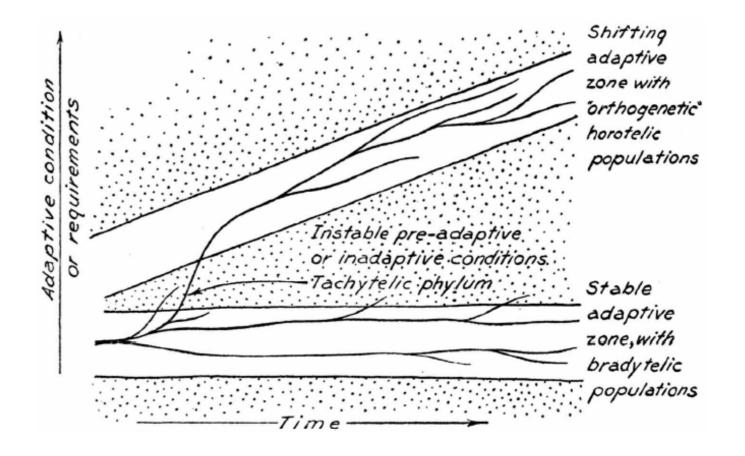


Fig. 31.—Diagrams of characteristic examples of the three major modes of evolution. In this and Figs. 32-33 the broken lines represent phylogeny and the frequency curves represent the populations in successive stages.



Simpson (1953) The major features of evolution

There is another option...

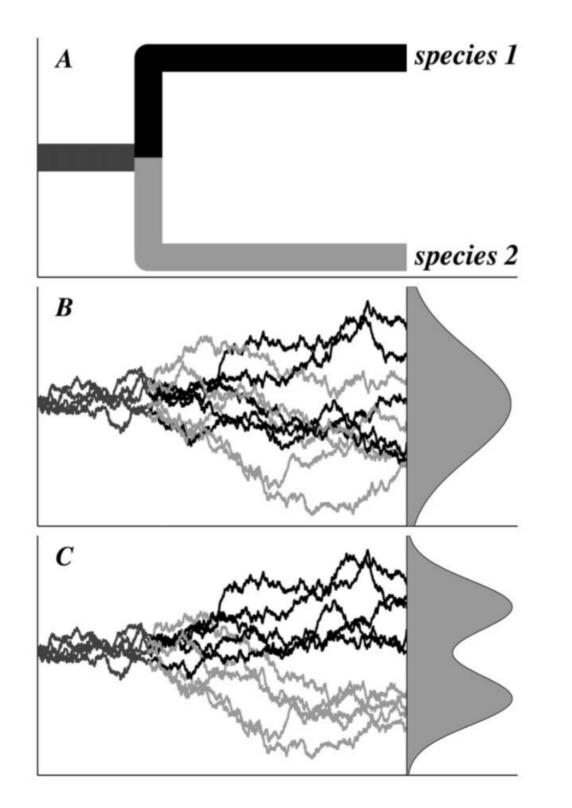
 The comparative dataset (body size, plumage color, etc. for each species)

2. A particular pattern of shared ancestry (phylogeny and branch lengths)

3. A model of evolutionary change along each branch of the phylogeny

Change the model of evolution !

How do we explain Patterns of Biodiversity?



Pattern of relationship

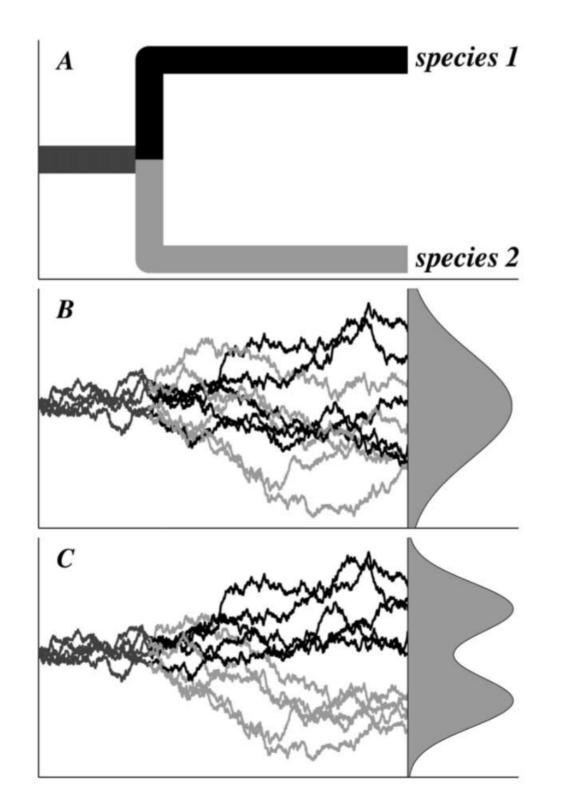
Are species phenotypes just subject to random evolution "drift"?

Or are other forces shaping the phenotype?

- Shifts in Selection
- Shifts in Environment
- Changes in Constraints

Figure from Butler & King 2004

How do we explain Patterns of Biodiversity?



How do we tease apart the factors?

- Statistical considerations
- Evolutionary or Tree Thinking

Figure from Butler & King 2004

Modeling adaptive evolution using OUCH*!

Marguerite Butler

University of Hawaii, Department of Zoology



Aaron King

University of Michigan, Ecology & Evolutionary Biology

(*Ornstein-Uhlenbeck for Comparative Hypotheses)

2. "Model the Evolutionary Process"

The phylogeny (pattern and timing of evolutionary diversification) as well as the data contains important information

Model the evolutionary process along each branch of the phylogeny

Brownian Motion

$$dX_i(t) = \sigma dB_i(t), \quad t_i^{j-1} \le t \le t_i^j.$$

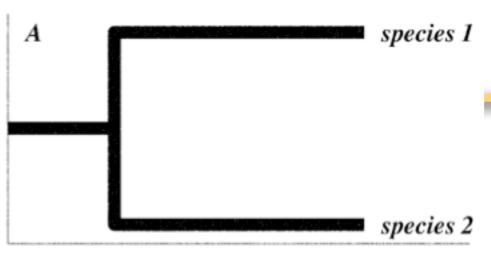
Orstein Uhlenbeck Process

$$dX_i(t) = \alpha \left(\beta_i^j - X_i(t)\right) dt + \sigma dB_i(t)$$

Vary the models to reflect biology Then compare to find the best model

Hansen (1997)

How different are BM and OU models?



Nuts and Bolts: 2 species example

Phenotype
$$\mathbf{X}(t) = \begin{bmatrix} X_1(t) \\ X_2(t) \end{bmatrix}$$

Bspecies 1 species 2 S Time

Log-likelihood equation

$$-2 \log \mathcal{L} = \{\mathbf{X}(T) - E[\mathbf{X}(T)]\}' \mathbf{V}^{-1}$$
$$\times \{\mathbf{X}(T) - E[\mathbf{X}(T)]\} + N \log (2\pi \det \mathbf{V})$$

BM

OU

$$E[\mathbf{X}(T)] = \begin{bmatrix} \theta_0 \\ \theta_0 \end{bmatrix}$$

$$E[X_1(T)] = \theta_0 e^{-\alpha T} + \theta_1 (1 - e^{-\alpha T})$$

= $\theta_0 W_{10} + \theta_1 W_{11}$

$$E[\mathbf{X}(T)] = \mathbf{W}\theta$$

$$\mathbf{V} = \sigma^2 \begin{bmatrix} T & s \\ s & T \end{bmatrix}$$

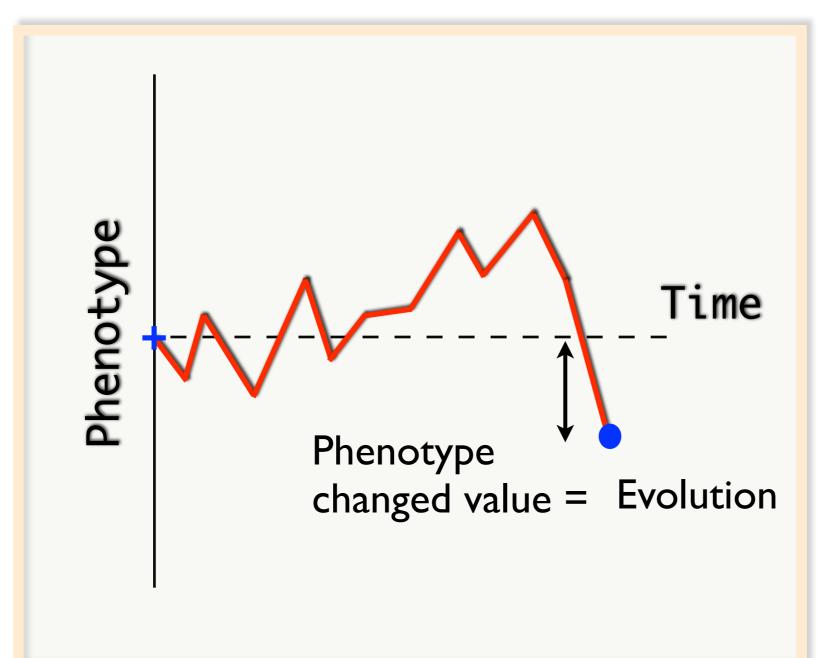
$$\mathbf{V} = \sigma^2 \begin{bmatrix} T & s \\ s & T \end{bmatrix} \qquad \mathbf{V} = \begin{bmatrix} 1 - e^{-2\alpha T} & e^{-2\alpha(T-s)}(1 - e^{-2\alpha s}) \\ e^{-2\alpha(T-s)}(1 - e^{-2\alpha s}) & 1 - e^{-2\alpha T} \end{bmatrix}$$

Hansen (1997), Butler & King (2004)

That's great... but what does it look like?

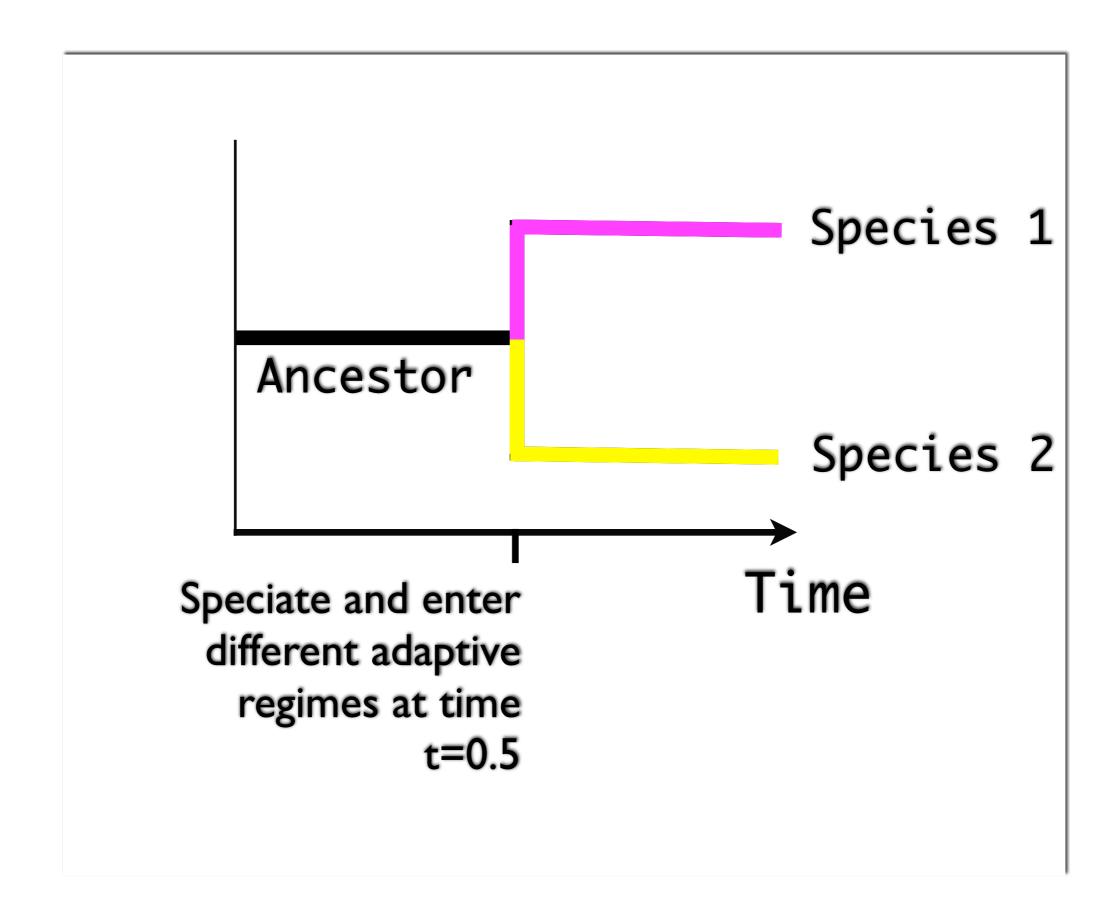
let's do some simulations!

Evolution in Cartoon

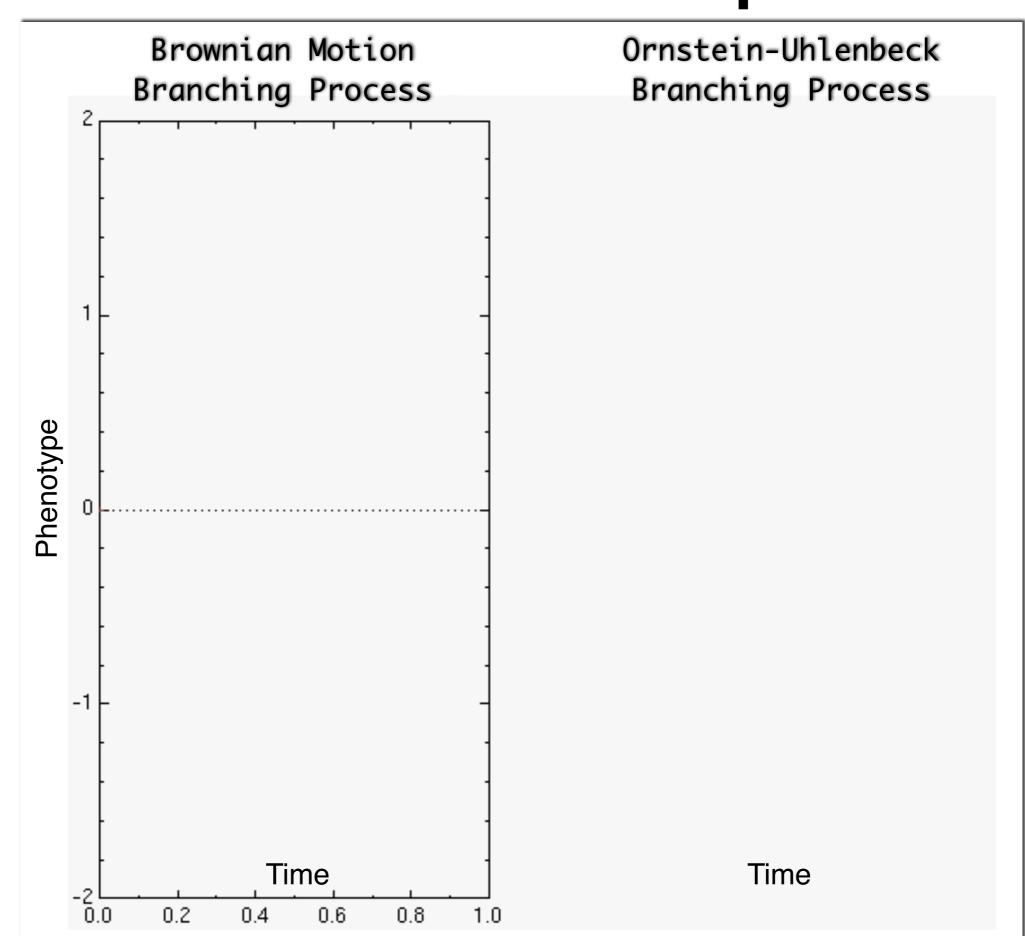


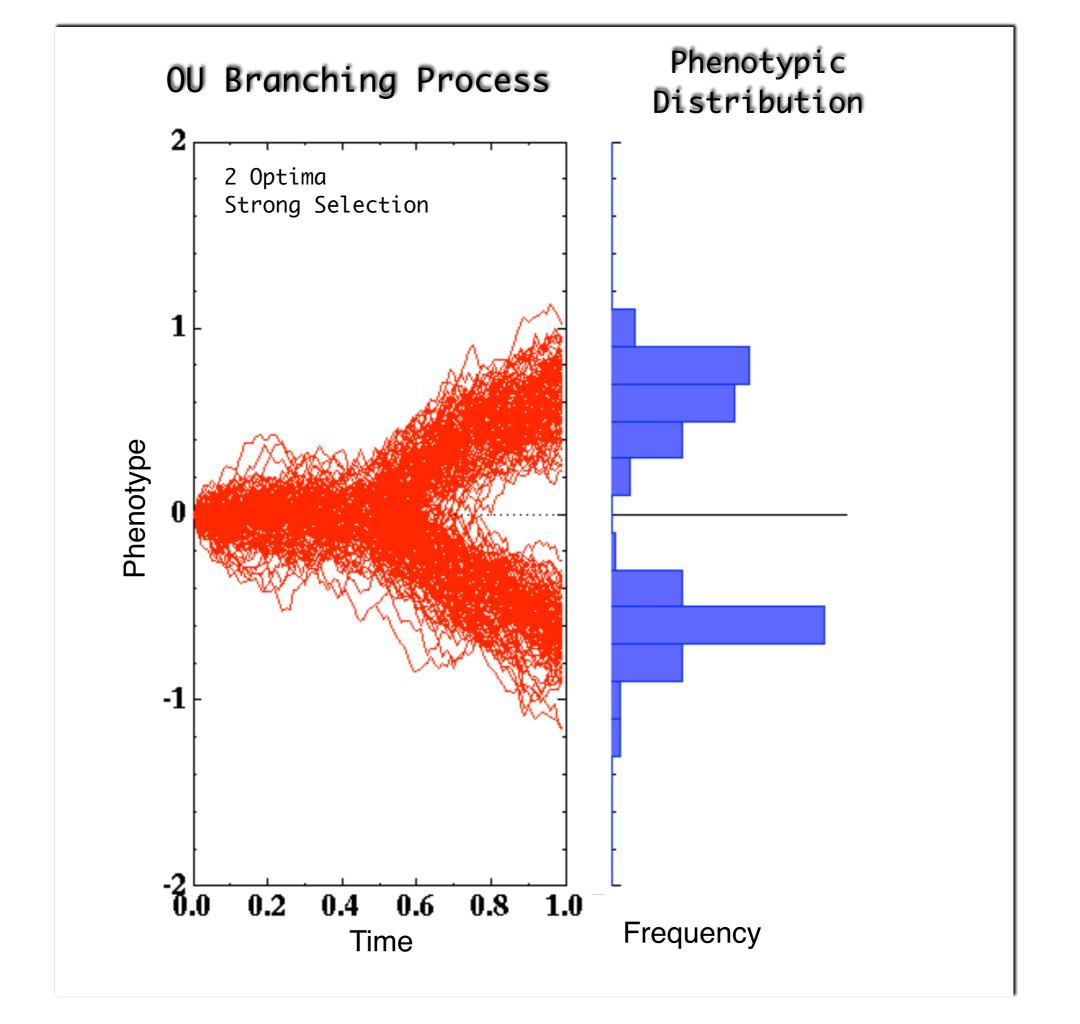
- At each infintessimal increment of time, there is a probability of change in the phenotype.
- The probability is drawn from a normal distribution, thus the BM is "Gaussian."
- The state at any given time is only dependent on the previous state.

BM and OU models make different predictions



BM and OU models make different predictions





What about the Biology?

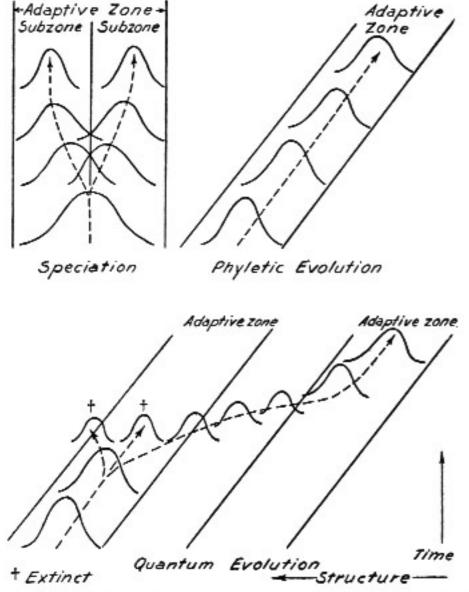
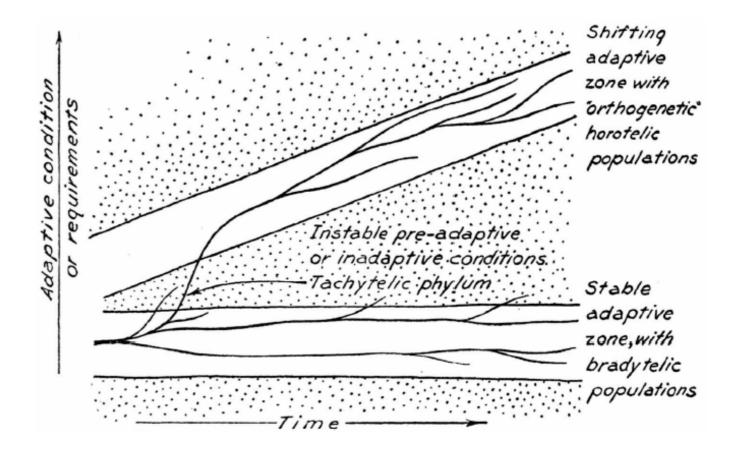
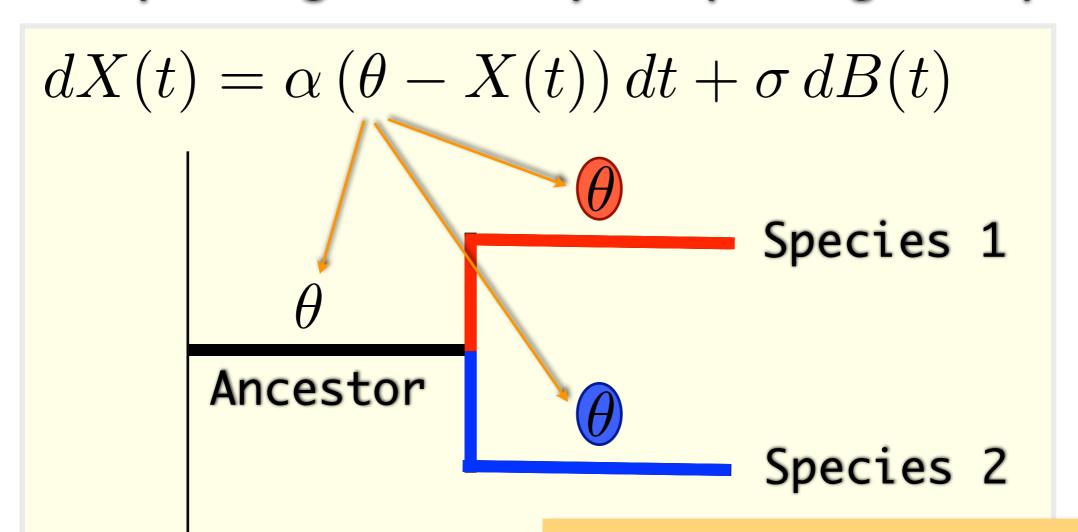


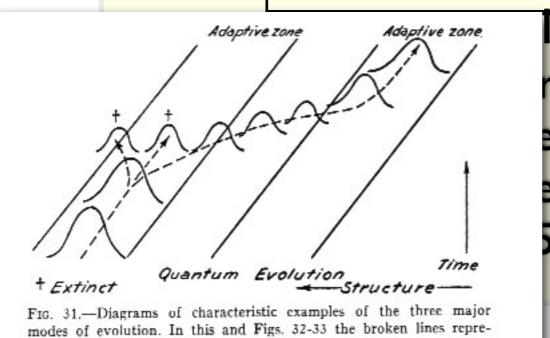
Fig. 31.—Diagrams of characteristic examples of the three major modes of evolution. In this and Figs. 32-33 the broken lines represent phylogeny and the frequency curves represent the populations in successive stages.



Simpson (1953) The major features of evolution

Shift in adaptive regimes: as easy as a painting new optimum!





sent phylogeny and the frequency curves represent the populations in successive stages.

OUCH does this by writing a model for each branch of the phylogeny, according to the selective regimes we hypothesize

Shift in adaptive regimes: as easy as a painting new optimum!

$$dX(t) = \alpha \left(\theta - X(t)\right) dt + \sigma dB(t)$$

$$\alpha \left(\theta - X(t)\right) dt + \sigma dB(t)$$
 Species 1
$$\alpha \left(\theta - X(t)\right) dt + \sigma dB(t)$$
 Ancestor
$$\alpha \left(\theta - X(t)\right) dt + \sigma dB(t)$$
 Species 2

Speciate and enter different adaptive regimes at time t=0.5

OUCH does this by writing a model for each branch of the phylogeny, according to the selective regimes we hypothesize

Thus, with:

a set of interspecific data, a phylogeny, and a little biological insight,

we can explore alternative evolutionary scenarios

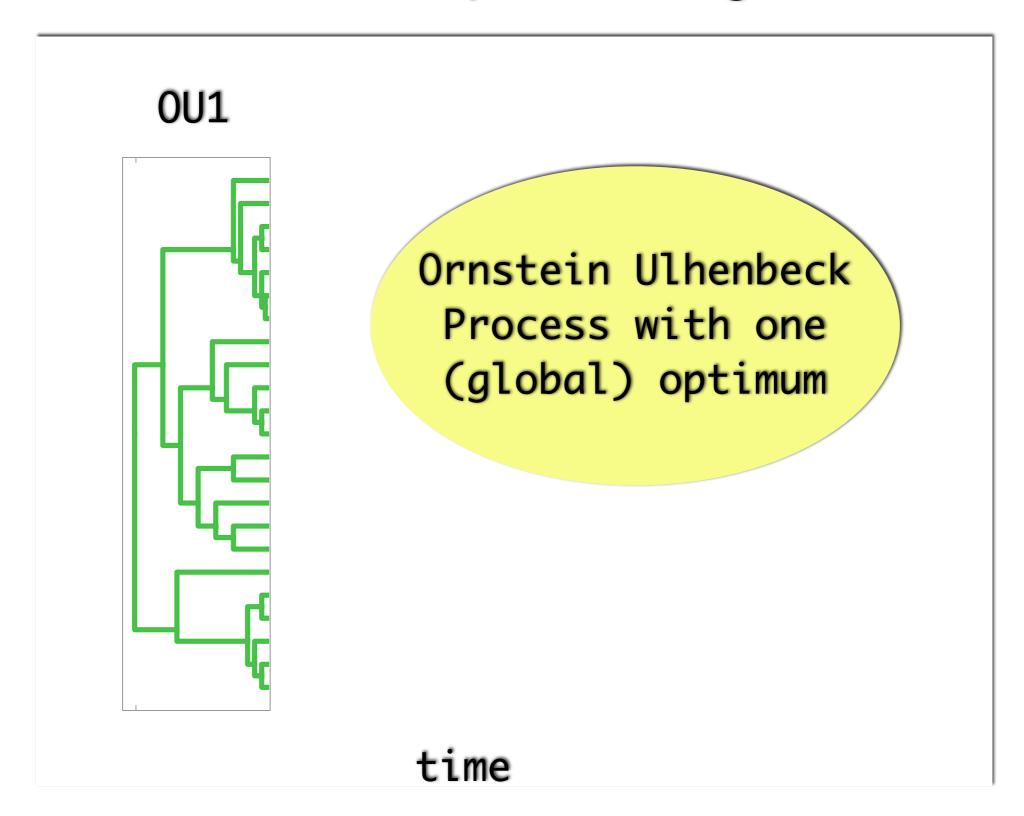
and potentially make a statement about how characters evolved!

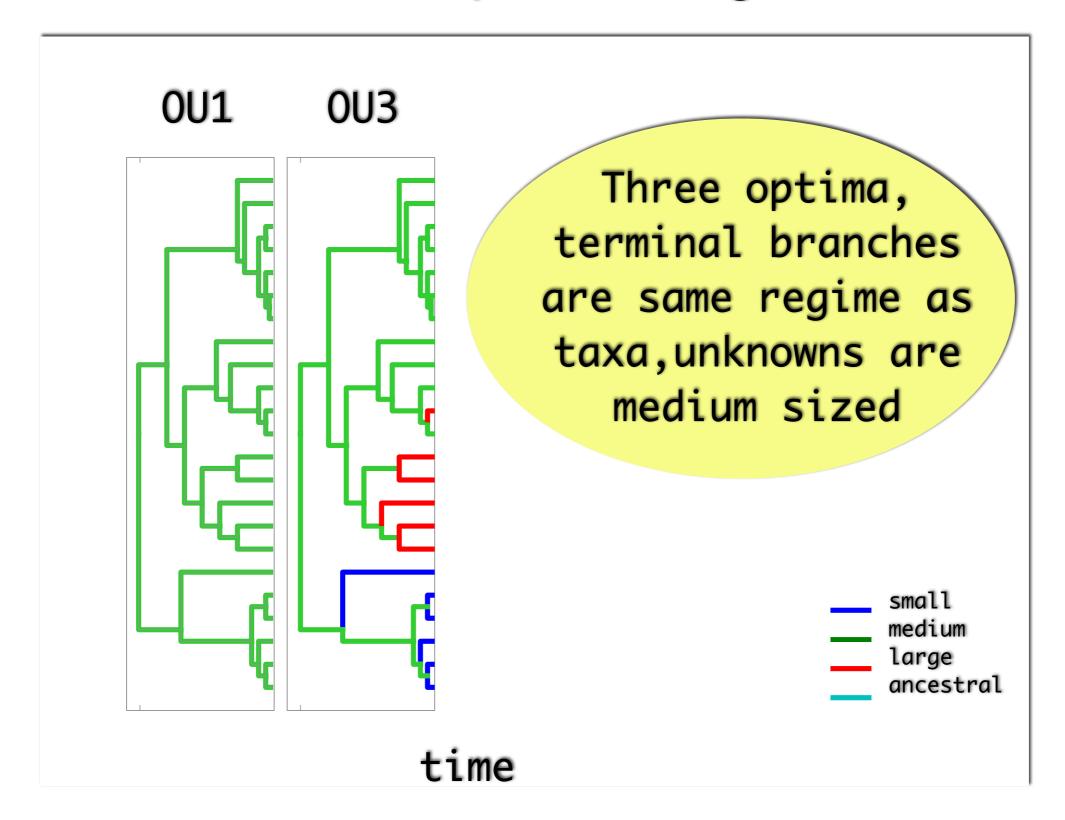
Islands in the Lesser Antilles have only 1 or 2 species of anoles.

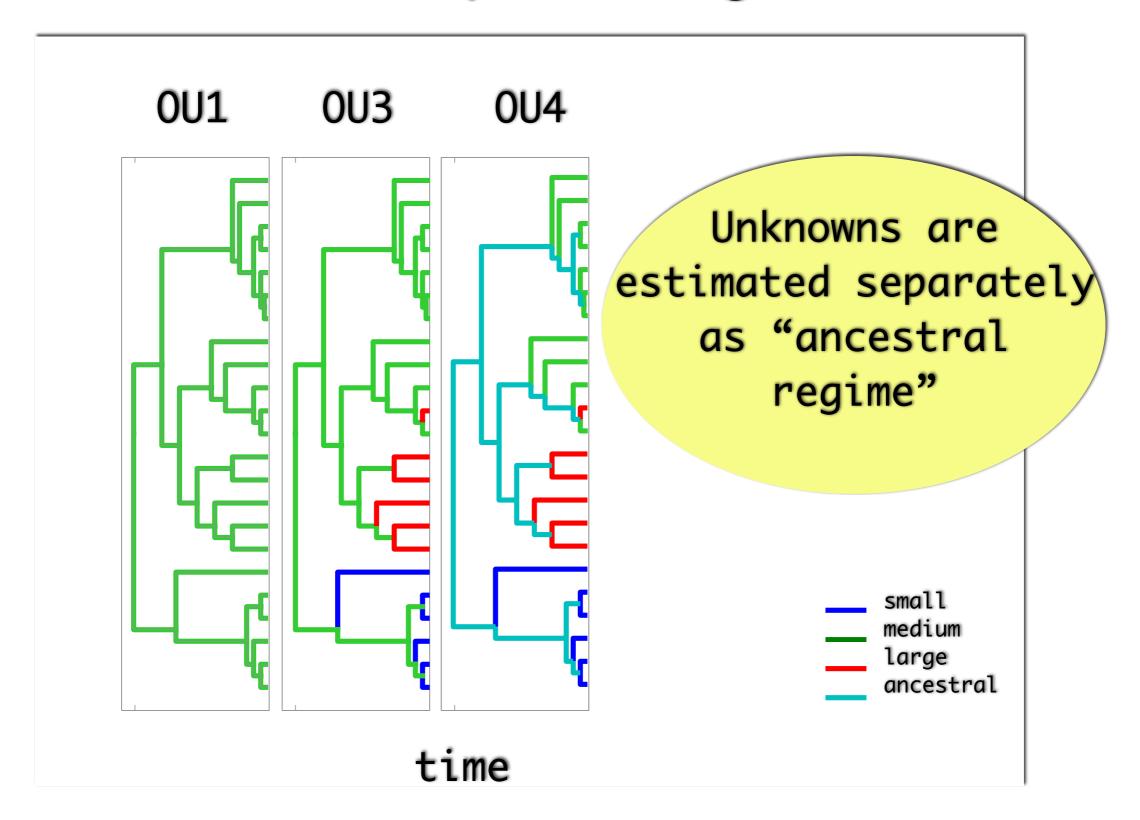
On the 2 species islands, anoles differ substantially in size.

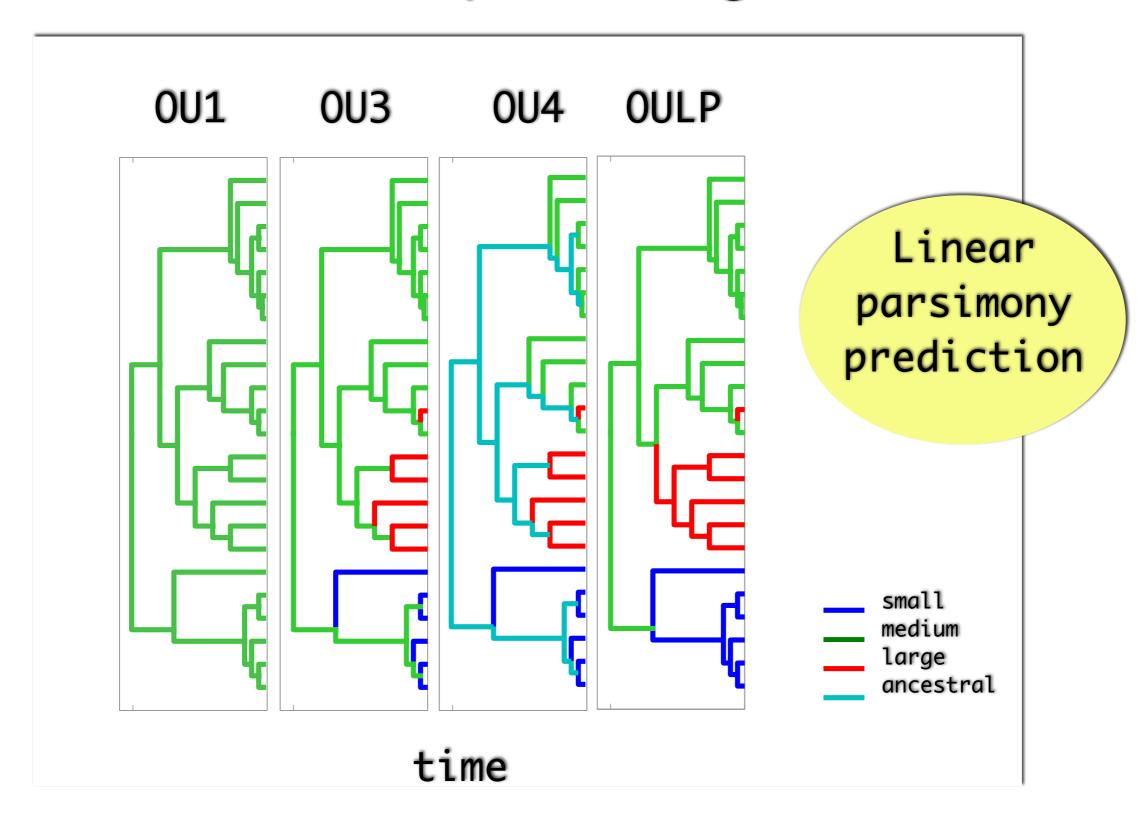
On the single species islands, anoles are "intermediate" in size.











Model Comparison

BM)	00(1)	00(3)	0U(4)	OU(LP)
-36.31	-34.66	-40.06	-47.22	-49.69
-32.31	-26.66	-28.06	-33.22	-37.69
	0	0.56	0.95	0.99
	1	0.44	0.05318	0.00955
	-36.31	-36.31 -34.66 -32.31 -26.66 0	-36.31 -34.66 -40.06 -32.31 -26.66 -28.06 0 0.56	-36.31 -34.66 -40.06 -47.22 -32.31 -26.66 -28.06 -33.22 0 0.56 0.95

Model Comparison

	BM	00(1)	00(3)	0U(4)	OU(LP)
-2x Log Likelihood	-36.31	-34.66	-40.06	-47.22	-49.69
AIC	-32.31	-26.66	-28.06	-33.22	-37.69
LR		0	0.56	0.95	0.99
Pvalue		1	0.44	0.05318	0.00955

Bad Models Strange Parameter Estimates

	BM)	0U(1)	0U(3)	0U(4)	OU(LP)
alpha		0	0.32	14.67	2.49
sigma	0.2	0.21	0.2	0.47	0.22
ancestor	2.83	2.95	3.99)	0.86
optima 1)	-1.4	2.58	2.75
optima 2			0.18	3.11	3.24
optima 3			2.71	3.3	3.56
optima 4				2.83	

Best Model

	BM)	OU(1)	00(3)	0U(4)	OU(LP)
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Best Model

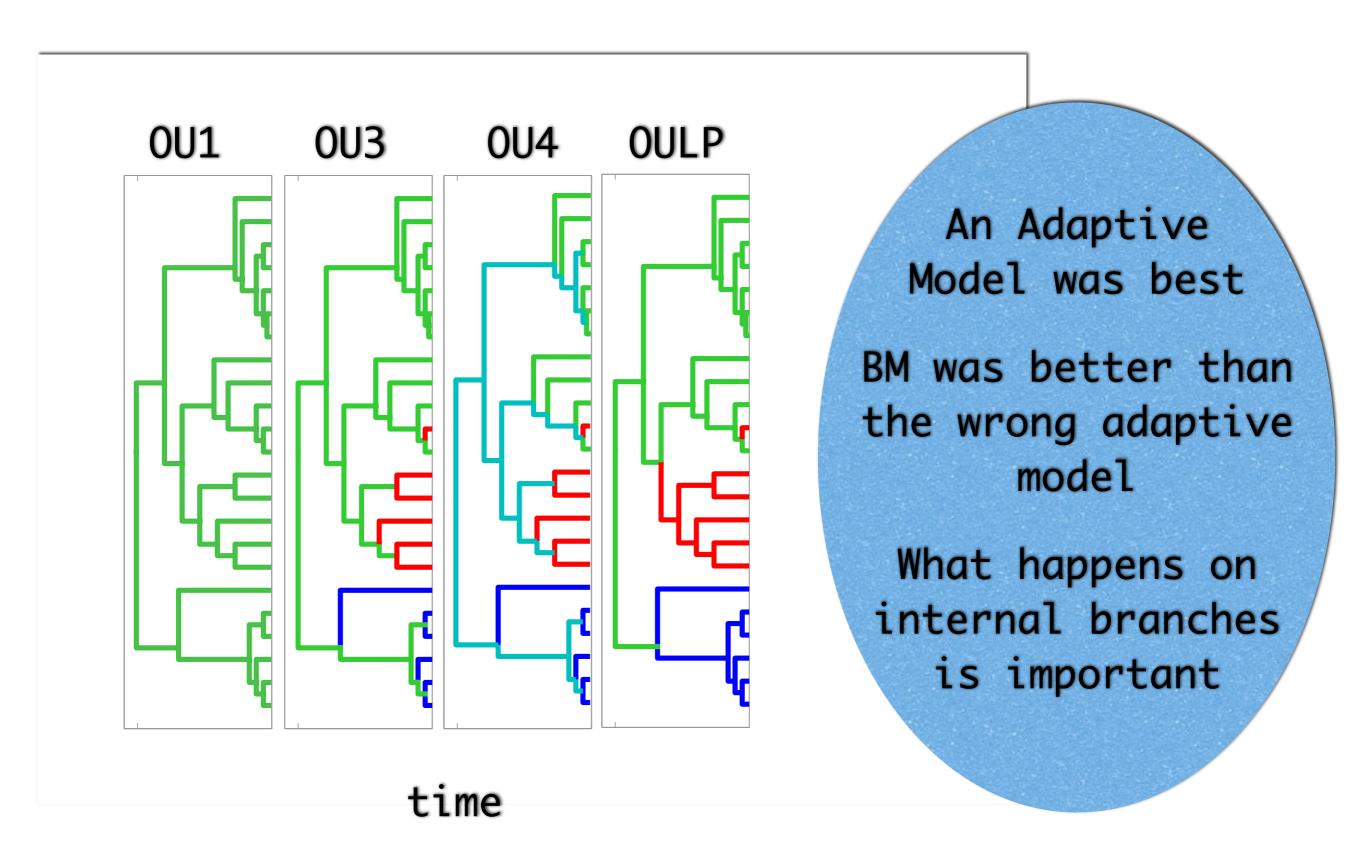
$\mathbb{E}\left[X_i(T) \mid X_i(0) = \theta_0\right]$		OU(LP)
$= e^{-\alpha T} \theta_0 + \sum_{i=0}^{k(i)} e^{-\alpha T} \left(e^{\alpha t_i^j} - e^{\alpha t_i^{j-1}} \right) \beta_i^j$	alpha	2.49
j=1	sigma	0.22
	ancestor	0.86
Ancestral Optima State	optima 1	2.75
(at root)	optima 2	3.24
	optima 3	3.56

Best Model

$\mathbb{E}\left[X_i(T) \mid X_i(0) = \theta_0\right]$		OU(LP)
$= e^{-\alpha T} \theta_0 + \sum_{i=0}^{k(i)} e^{-\alpha T} \left(e^{\alpha t_i^j} - e^{\alpha t_i^{j-1}} \right) \beta_i^j$	alpha	2.49
j=1	sigma	0.22
	ancestor	0.86
Ancestral State 92% Optima	optima 1	2.75
(at root)	optima 2	3.24
	optima 3	3.56

Butler & King 2004

Univariate Conclusions



Parametric Bootstrap indicates "high" and "low" SSD optima

associated with habitat type

	0U(7)	95% CI
alpha	1.44	(0.24, 4.5)
sigma	0.09	(0.04, 0.11)
ancestor	0.23	(0.03, 0.27)
optima tg	0.327	(0.26, 0.97)
optima tc	0.134	(0.00, 0.27)
optima gb	-0.013	(-0.28, 0.13)
optima cg	-0.06	(-0.39, 0.09)
optima tw	-0.074	(-0.42, 0.06)
optima tr	-0.051	(-0.39, 0.12)
optima uk	0.468	(0.34, 1.25)

Butler & King 2004

Final Thoughts (Rantings):

Biologists think about adaptive evolution tion, so WHY use a

tral?

BE EXPLICIT about your models!! It's important.

This should be a conclusion, not an assumption.

We gain nothing by using a null model that is really uninformative and then rejecting it!

We gain much more insight by using models that approximate our best ideas and comparing them.

"Correcting for phylogeny" is not a useful way to think about comparative analysis.

