

Phylogenetic comparative methods: Continuous characters

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Outline

1. Lecture on models of continuous evolution
2. Tutorial on fitting a model; λ
3. Tiny lecture on ancestral states
4. Tutorial on estimating ancestral states
5. Tiny lecture on Phylogenetic Generalized Least Squares
6. Tutorial on correlated evolution

Continuous characters in linguistics and anthropology

In biology, it is easy: body mass, brain size, metabolic rate, etc.

In linguistics, many well-studied comparative traits are discrete. However, consider:

- phonemic inventory size (number of vowels, consonants, tones, contrasts, etc.)
- lexicon size (number of numeral classifiers, genders, body part terms, color inventories, kinship terms, other specific semantic classes)
- morphology (number of cases, verbal and nominal affixes, length of maximally inflected verb or noun (in morphemes), amount of compounding)
- syntax (percentage of SOV word order in corpus, amount of inflection when not obligatory)
- psycholinguistics (age of acquisition of particular lexical items, reading speed)
- evolutionary anthropology (brain size, position of larynx)

Anthropology:

- subsistence (amount of time spent farming, herding, hunting, fishing, gathering)
- material culture (size of floor area in housing, amount of dowry)
- anthropometric characters (height, weaning age, age at circumcision, amount of formal schooling, age at first birth, length of post-partum sex taboo)

Components of a phylogenetic comparative analysis

- **data**: trees + trait data on tips
- **model**: description of evolutionary change
- **inference method**: Maximum Parsimony, Maximum Likelihood, Bayesian
- **parameter estimates**: node values, rates, etc.

Models of continuous character evolution

“I simply want to correct for phylogenetic relatedness, just tell me what to do”

phylogenetic regression (R: `pgls()` in `caper`)

phylogenetic ANOVA (R: `aov.phylo()` in `geiger`)

phylogenetic principal component analysis
(R: `phyl.pca()` in `phytools`)

phylogenetic t-test (R: `phyl.pairedttest()` in `phytools`)

Models of continuous character evolution

- Brownian motion
- Ornstein Uhlenbeck / stabilizing selection
- Spectational model of evolution / punctuated equilibrium
- Adaptive radiation / ecological niche-filling model

See Nunn (2011: 101-104) for an introduction and relevant references

Models of continuous character evolution

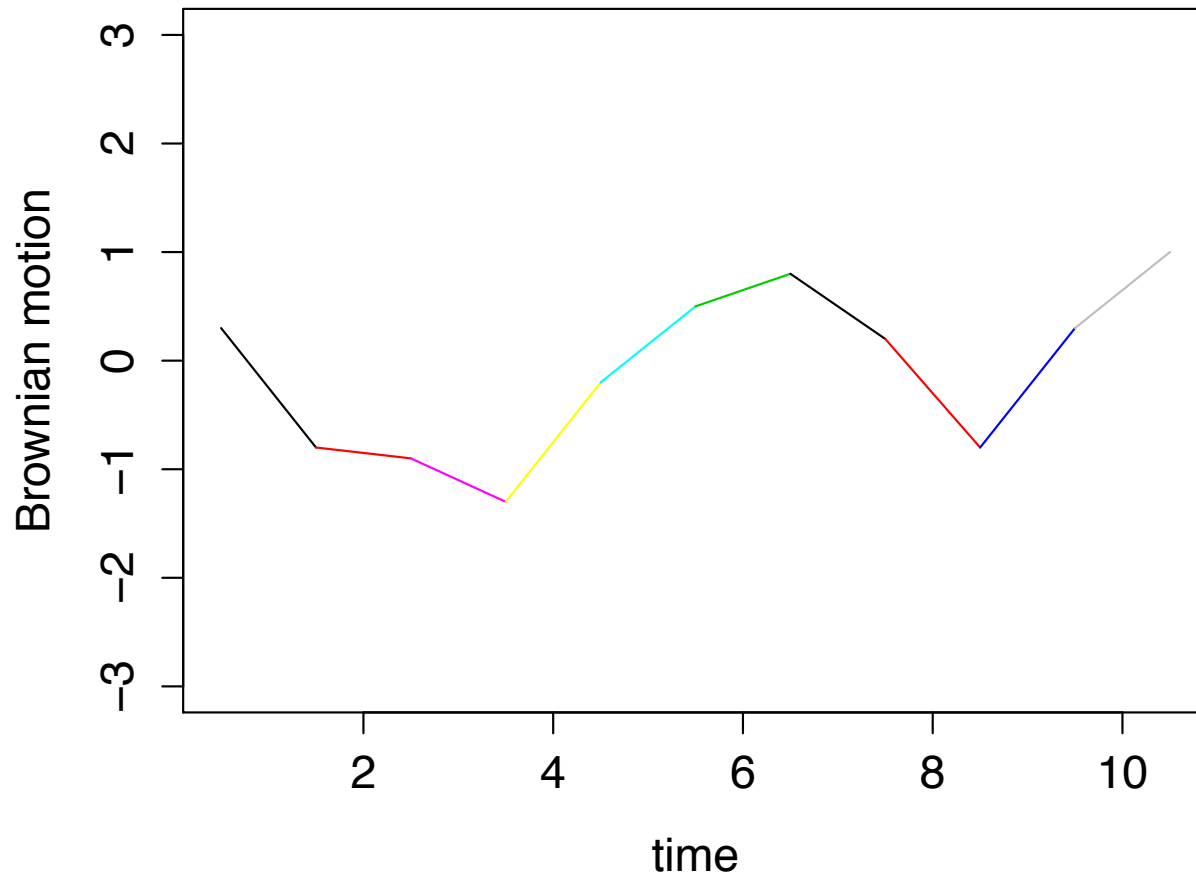
- Which evolutionary model is preferred?
- Are there differences across groups of taxa in which model is preferred?
- Is one direction of change more likely than the other?
- Do both directions of change happen at equal rates?
- Are there temporal evolutionary trends?
- Is evolutionary change bounded or unbounded?

Brownian motion

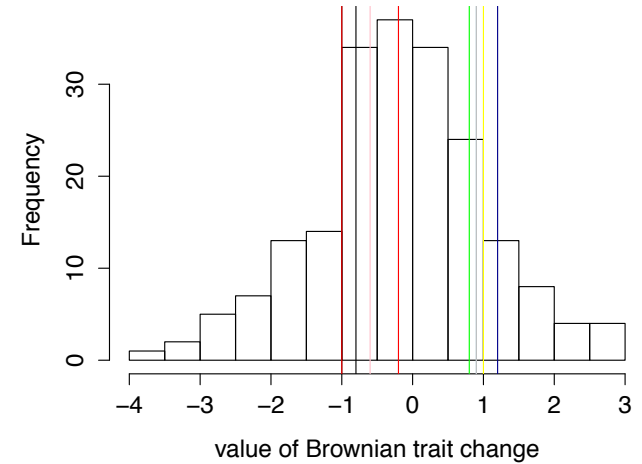
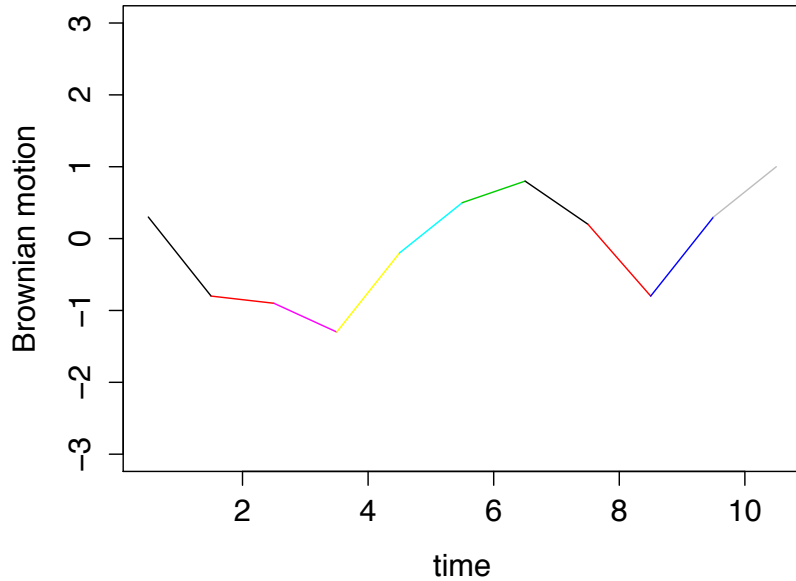
- can be considered a 'neutral' model of evolution
- a trait changes value at each of a huge number of very tiny time steps:
 1. Each change is independent of previous change on that branch and on other branches;
 2. Each change is randomly drawn from a normal distribution.
- For a collection of Brownian trait changes:
 1. the mean change is zero
 2. the variance of changes is constant
 3. the expected variance of trait change is proportional to branch length

Brownian motion

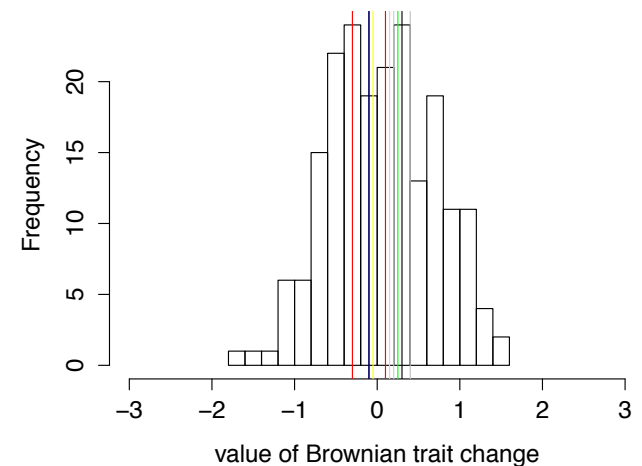
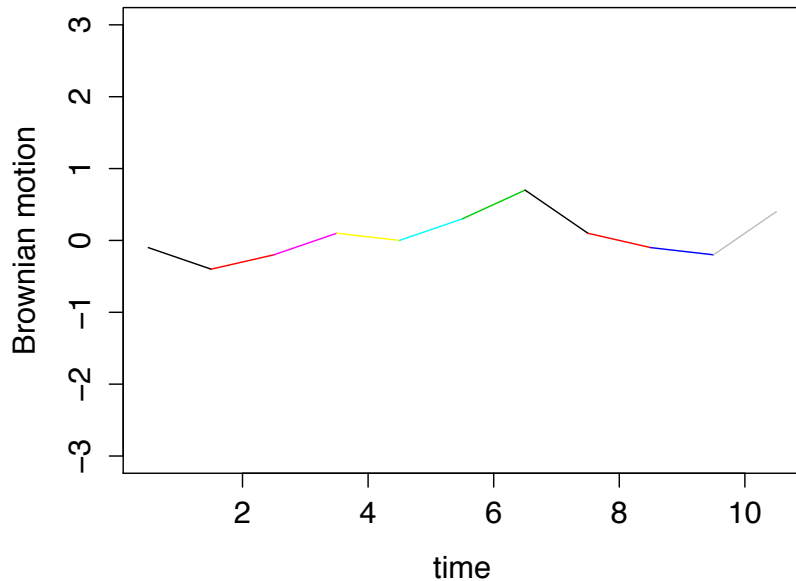
A single trait on a a single branch through trait space over time



Brownian motion

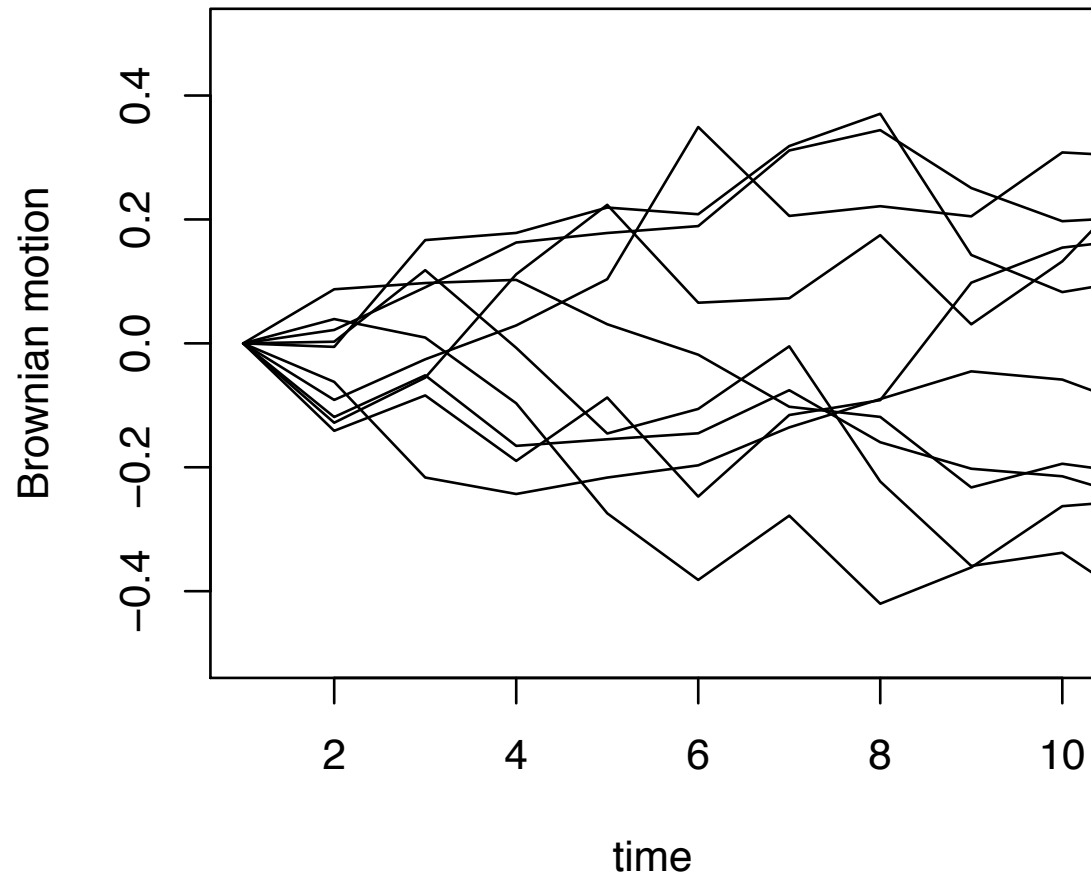


higher variance = higher rate of trait change σ^2

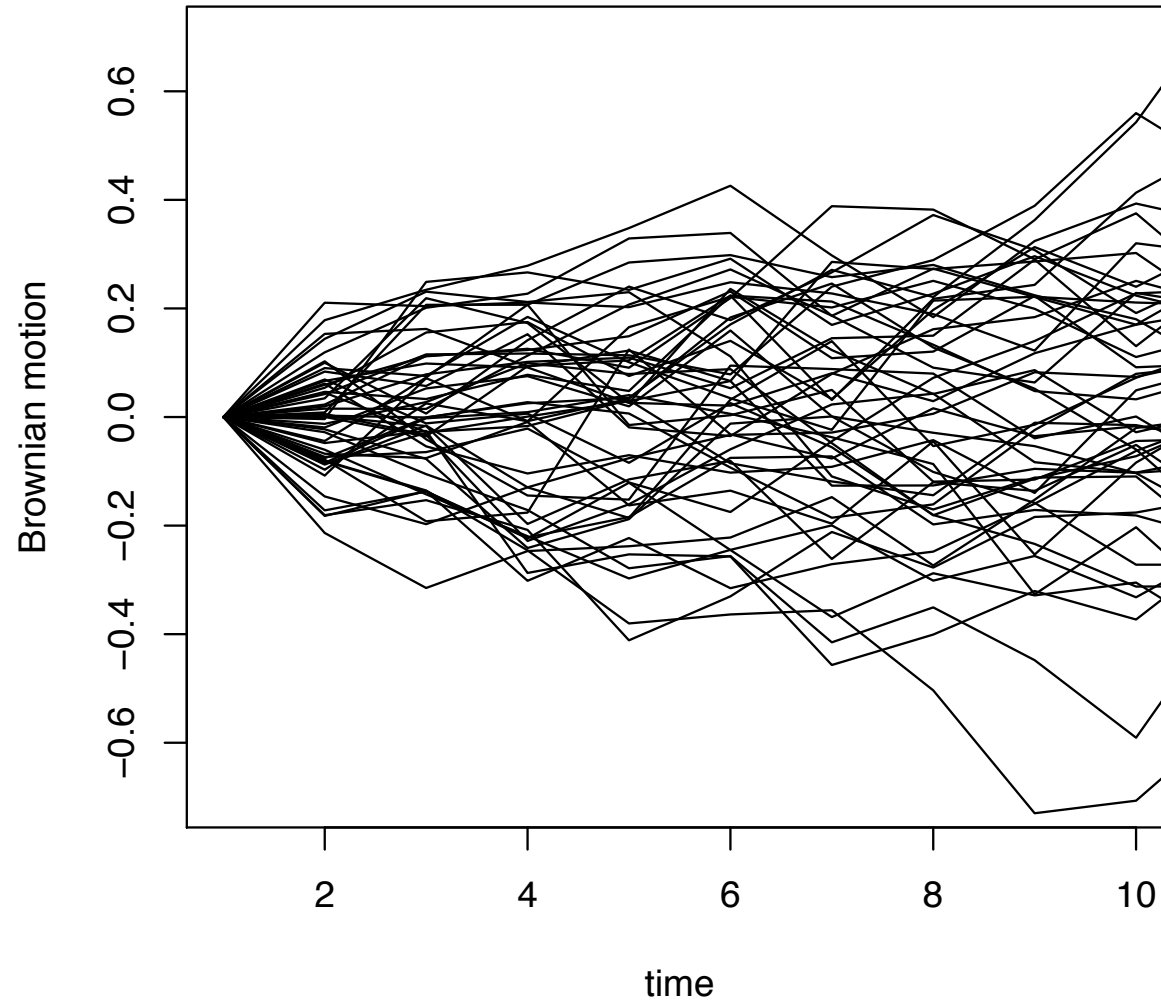


lower variance = lower rate of trait change σ^2

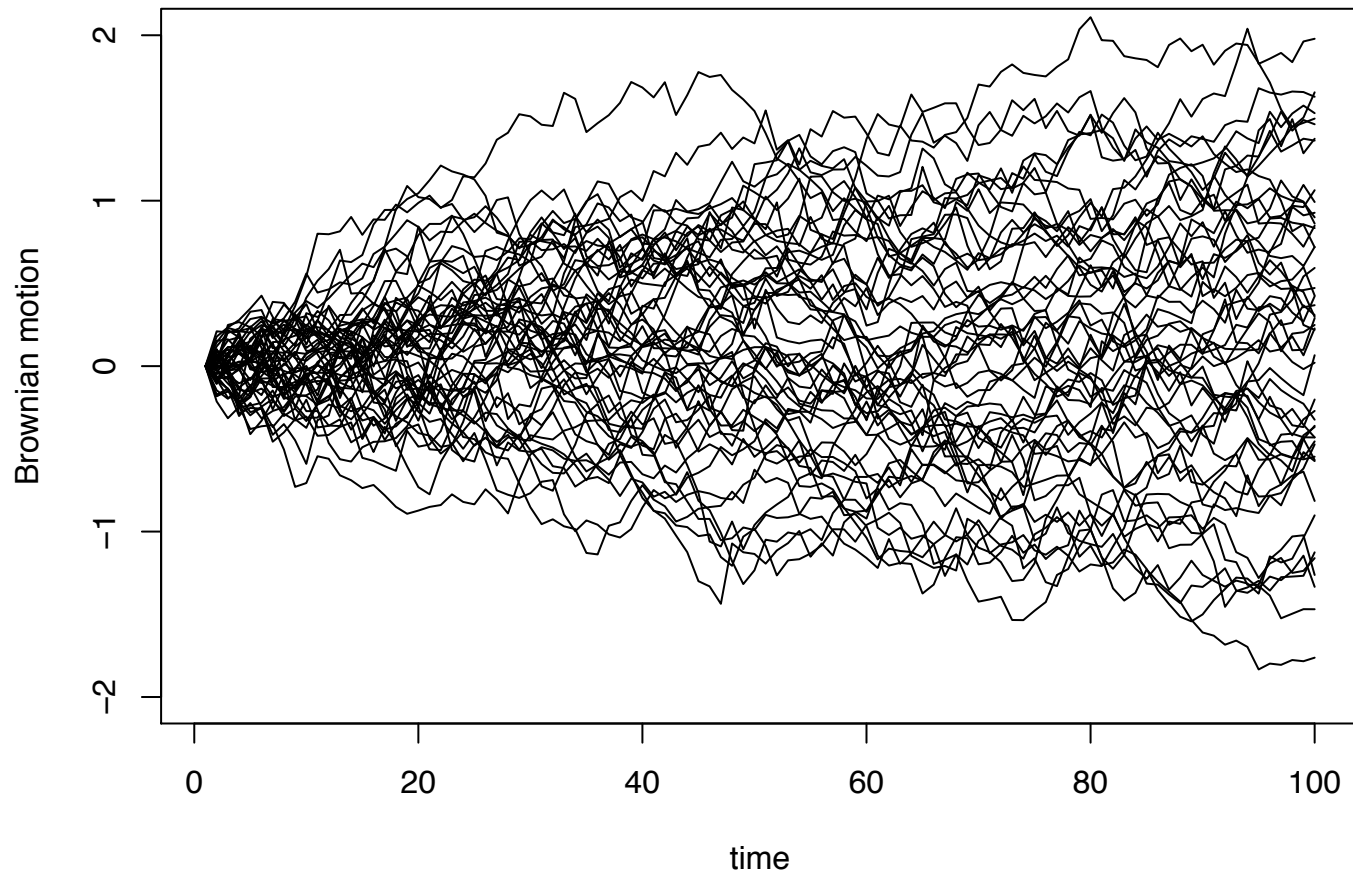
Brownian motion



Brownian motion

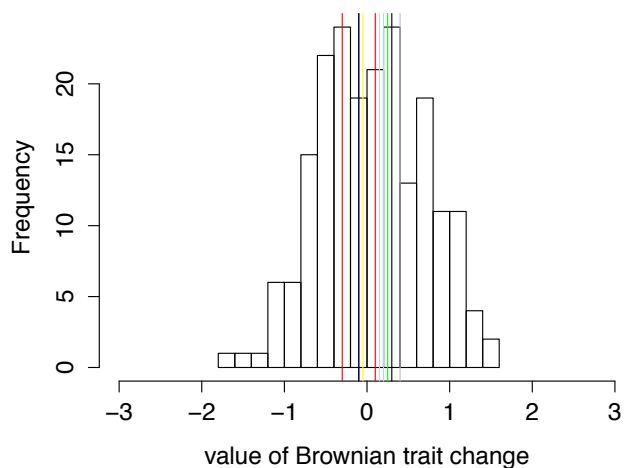
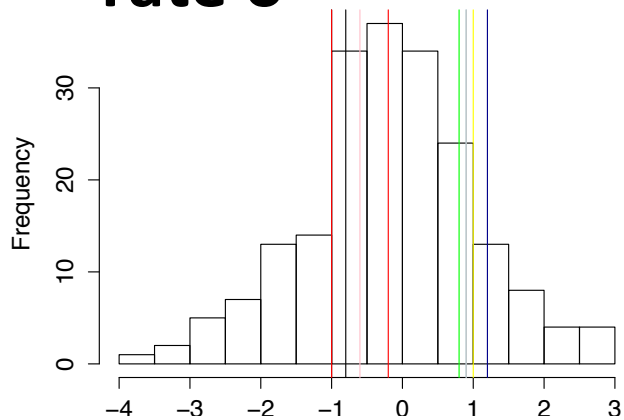


Brownian motion

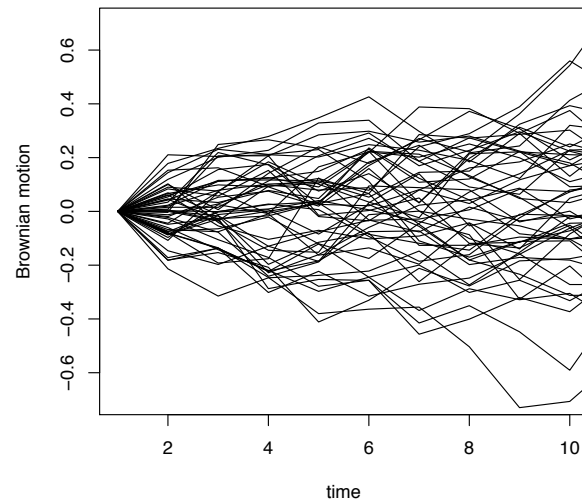


Brownian motion

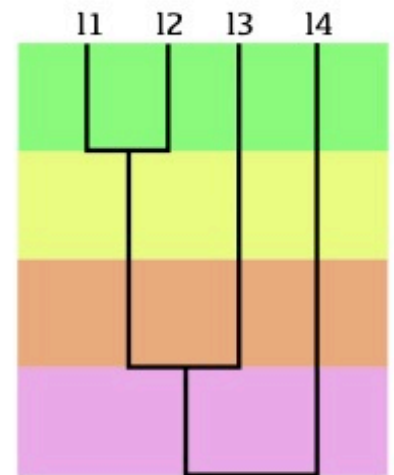
rate σ^2



time

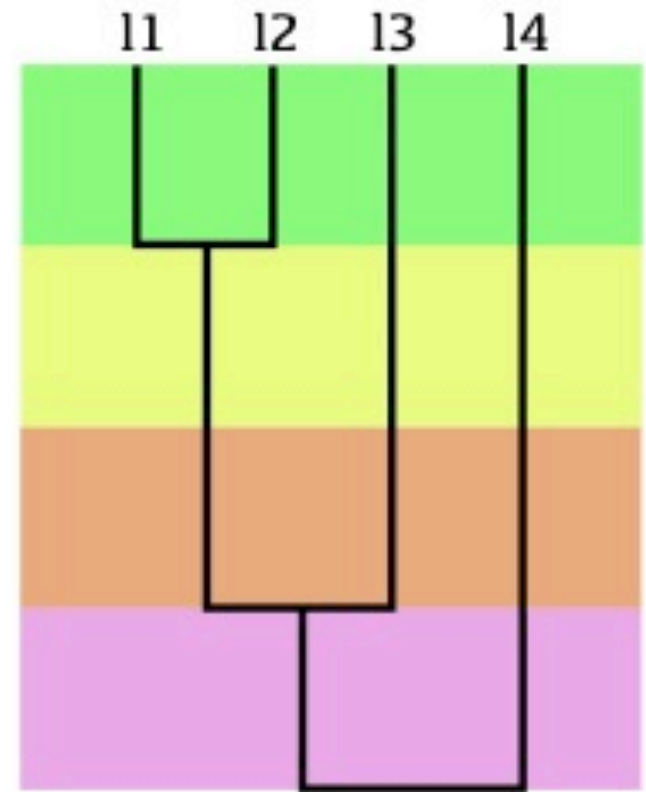


phylogenetic
structure



Phylogenetic variance-covariance matrix

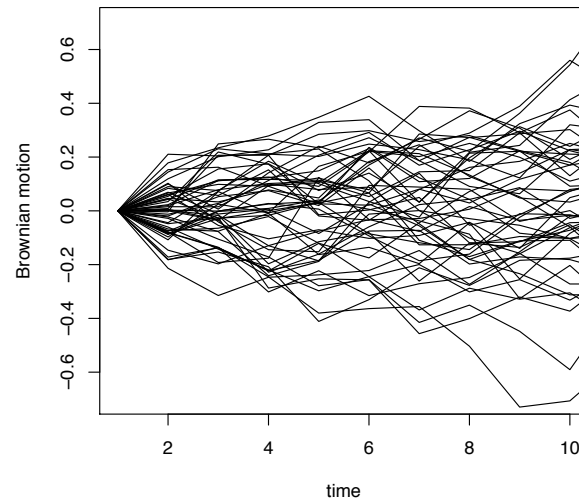
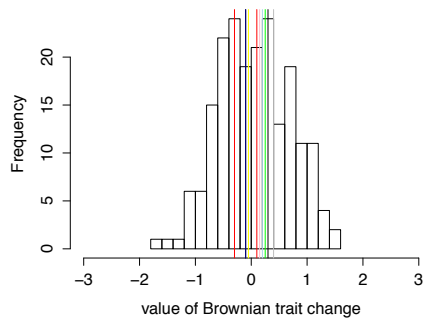
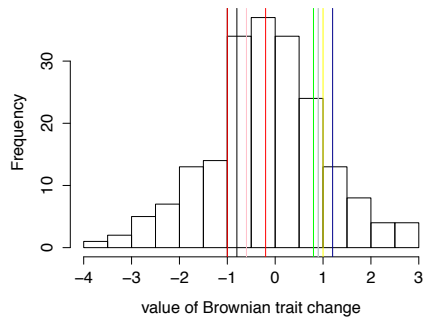
	l1	l2	l3	l4
l1	1+2+1			
l2	2+1	1+2+1		
l3	1	1	3+1	
l4	0	0	0	4



Brownian motion

rate σ^2

time



$$dX(t) = \sigma^2 dB(t)$$

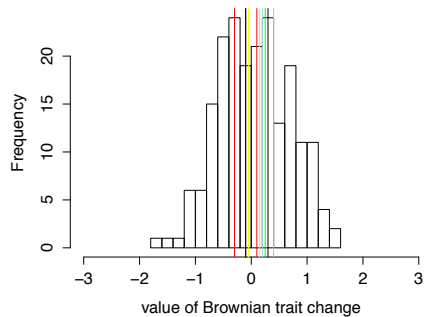
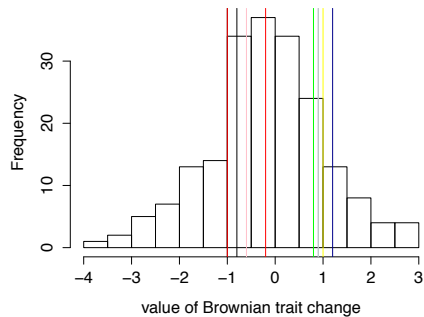
distribution d of
languages (X) by time t

Brownian rate
parameter

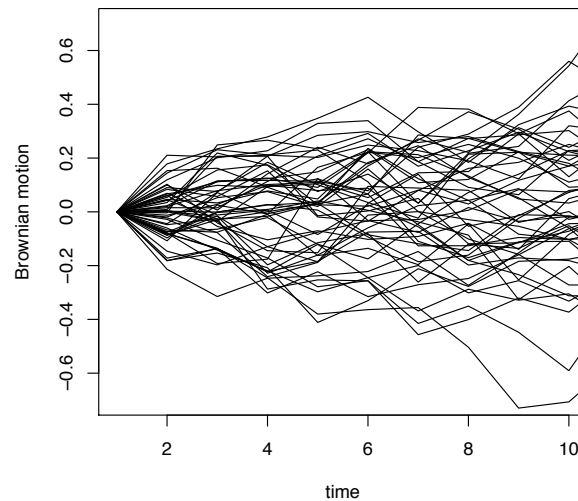
random normal
distribution by time t

Brownian motion

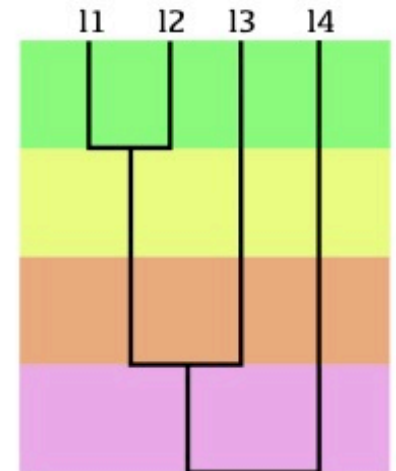
rate σ^2



time



phylogenetic
structure



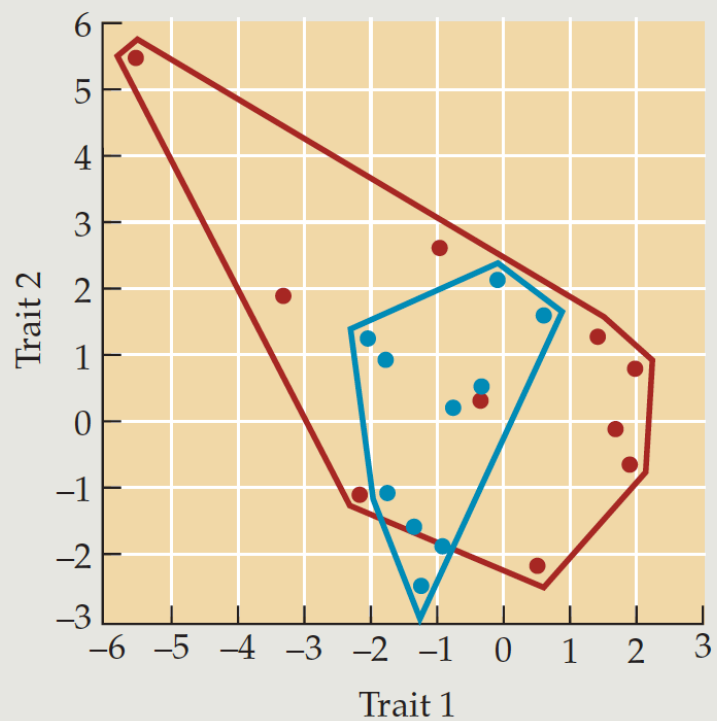
$$dX(t) = \sigma^2 \times$$

distribution d of
languages (X) by time t

Brownian rate
parameter

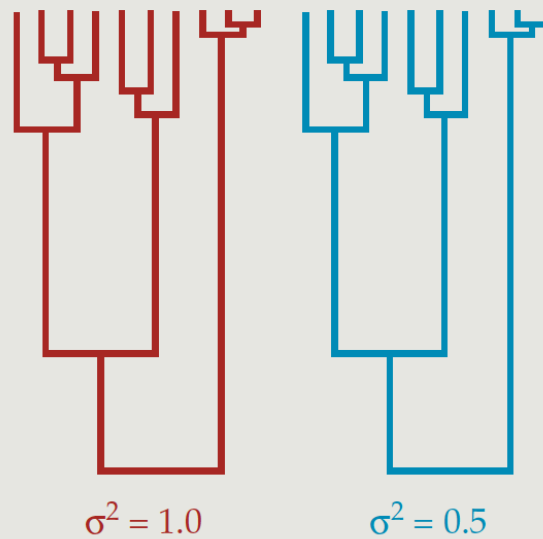
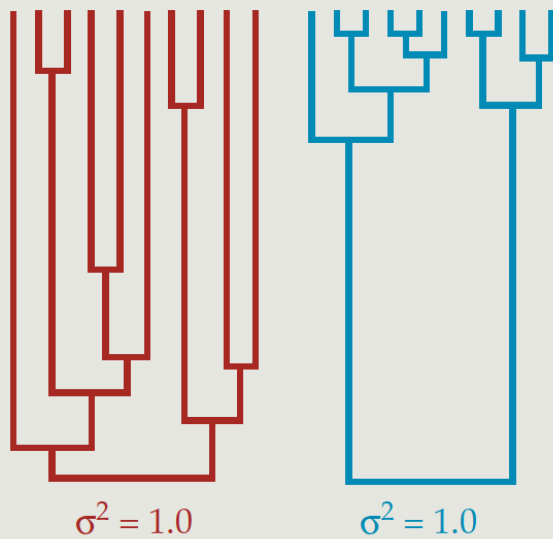
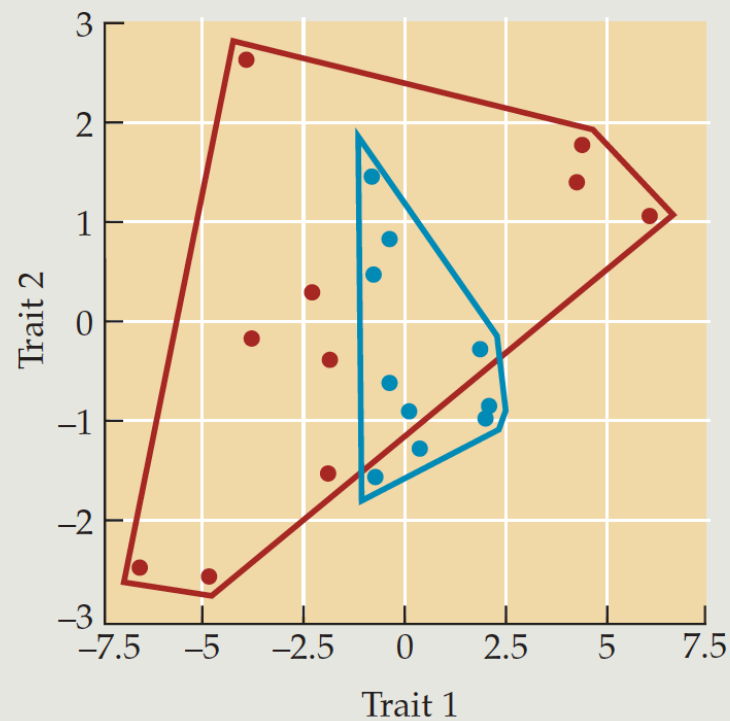
	l1	l2	l3	l4
l1	1+2+1			
l2	2+1	1+2+1		
l3	1	1	3+1	
l4	0	0	0	4

(A)



(B)

Losos and Mahler (2010)



Models of continuous character evolution

- Brownian motion
- Ornstein Uhlenbeck / stabilizing selection
- Spectational model of evolution / punctuated equilibrium
- Adaptive radiation / ecological niche-filling model

See Nunn (2011: 101-104) for an introduction and relevant references

Models of continuous character evolution

Nunn (2011: 103ff): the **Ornstein Uhlenbeck / stabilizing selection** model captures the force of selection on adaptive traits. In this model, the trait wanders around an optimum following Brownian motion. The model adds a parameter that tends to pull back the trait towards the optimum. The larger the restraining factor, the more reduced the expected variance of trait values.

Models of continuous character evolution

Pagel (1999): κ (kappa) is a power function that raises the length of each of the branch segments on the tree to the power κ . As it approaches 0, all branch segments have a length = 1, this is evidence for punctuational evolution.

κ is used to analyze **spectational evolution** or **punctuated equilibrium**: “changes is assumed to occur rapidly at, or shortly after, speciation, and stasis follows until the next speciation event.”
Mooers (1999: 250). See also Nunn (2011: 119ff).

Models of continuous character evolution

Pagel (1999): δ (delta) is a power function that raises the length of the total path (root to tip) to the power δ . A value < 1 transforms branch lengths to be increasingly shorter towards the tips, suggesting that trait change occurs rapidly early in the tree and then slows down over time. Value > 1 describe an increasing rate of character evolution through time.

δ is used to analyze a model of adaptive radiation or ecological niche-filling, when early change is important and the rate of evolution slows down over time. Also known as early burst model. Changes in rate are captured by the initial rate of evolution and an additional slowdown parameter. See Nunn (2011: 121ff) for a discussion of approaches other than δ .

Models of continuous character evolution

Pagel (1999): λ (lambda) is a multiplication of only the internal branches by a constant λ , with a value = 0 resulting in no internal branches and thus a star phylogeny, a value = 1 suggesting the data evolves as Brownian motion predicts on the tree, and a value > 1 suggesting more covariance than predicted under Brownian motion.

λ is used to investigate phylogenetic signal, an assessment of whether closely related species, languages, or cultures resemble each others. It should always be investigated for any comparative analysis! See Nunn (2011: 110) for discussion of more methods.

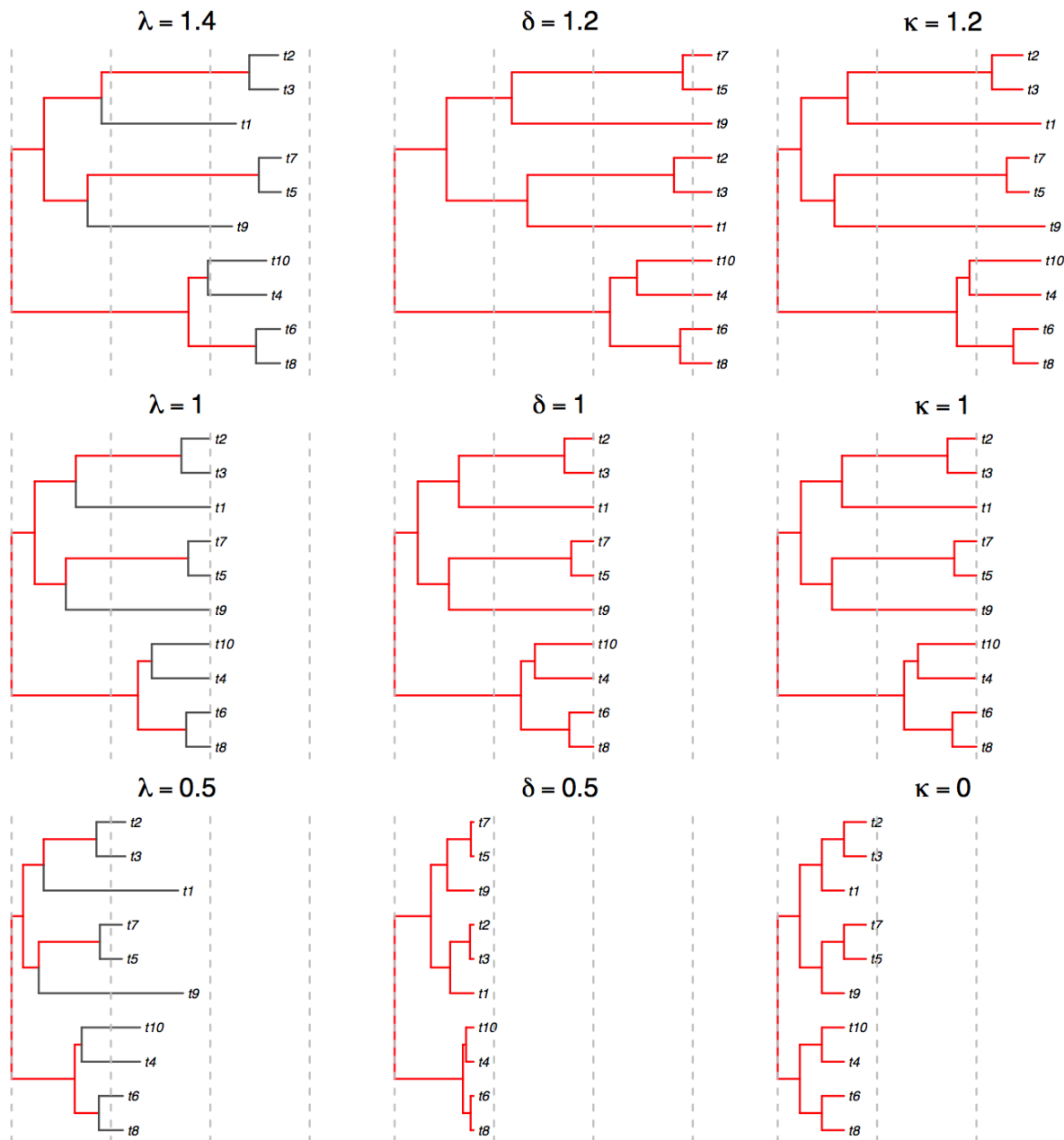


Figure 3: Examples of λ , δ and κ branch length transformations. The branches affected by a given transformation are shown in red.

Relevance of models

- Bentz et al. (2015): λ between .46 and 1 for lexical diversity measured in several parallel texts;
- Currie (2013): investigates the fit of λ and κ in a study of demographic factors of societal violence;
- Kamilar & Cooper (2013): review phylogenetic signal in 31 variables for primates;
- Verkerk (2013): finds high λ values for a variety of lexical/typological scores related to motion

Models of continuous character evolution in R

package **ape** can generate several phylogenetic error structures, i.e. variance-covariance matrices based on different evolutionary models:

`corBlomberg()` – fits Early-burst model AKA ACDC (accelarated/ decelerated) model according to Blomberg et al. (2003)

`corBrownian()` – Brownian model

`corGrafen()` – adjusted branch lengths according to Grafen (1989)

`corMartins()` – covariance structure according to Martins and Hansen (1997)

`corPagel()` – fits λ (lambda)

These are then added to `gls()` (Generalized Least Squares) in package `nlme`. More on GLS later.

Models of continuous character evolution in R

package **mvMORPH** and **geiger** in R implement continuous character evolution.

`fitContinuous()` in **geiger** can deal with:

- Brownian motion
- Ornstein Uhlenbeck / stabilizing selection
- Early-burst model / ACDC
- λ
- κ
- δ
- and more...

Resources: online

- Bodega Applied Phylogenetics Workshop:
<http://treethinkers.org/tutorials/>
- Anthrotree: <http://www.anthrotree.info>
- Liam Revell's graduate level course "Methods in Phylogenetic and Macroevolutionary Analysis",
<http://www.phytools.org/mpma/>
- r-sig-phylo:
<http://www.mail-archive.com/r-sig-phylo@r-project.org/>
- BayesTraits manuals:
<http://www.evolution.reading.ac.uk/SoftwareMain.html>

Resources: (hand)books

- Lemey, Philippe, Salemi, Marco, & Vandamme, Anne-Mieke (Eds.). (2009) *The phylogenetic handbook: A practical approach to phylogenetic analysis and hypothesis testing*. Cambridge: Cambridge University Press.
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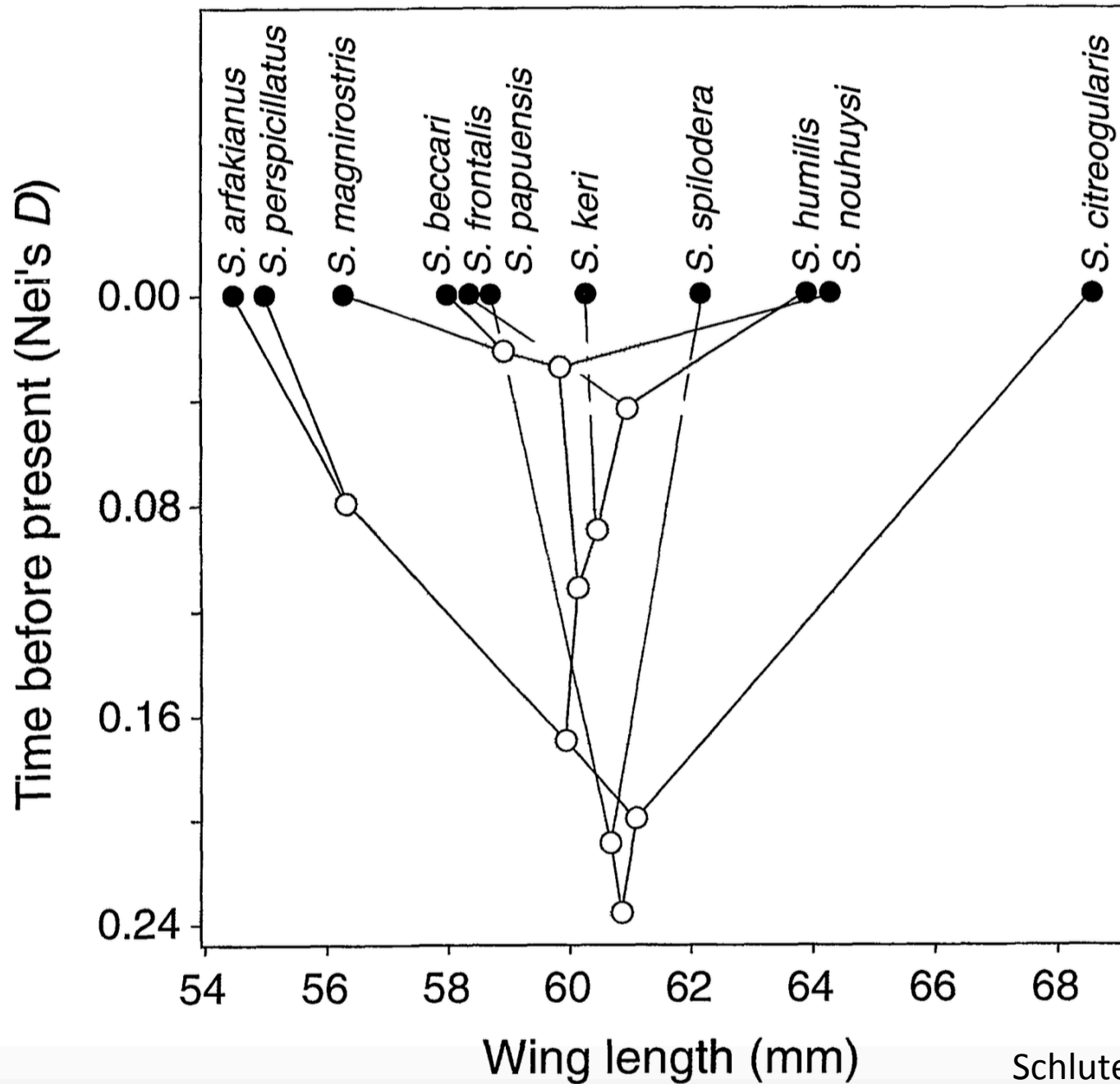
Tutorial 1

Phylogenetic signal; fitting lambda

Ancestral states

To be able to estimate what languages/cultures were like in the past is one of the main aims of phylogenetic comparative studies.

We can do this under Brownian motion, so: we aim to find the set of ancestral states that maximize the probability of our data and tree – aside from the rate parameter and the state of the root of the tree, we infer the behavior of each of the internal nodes of the tree.



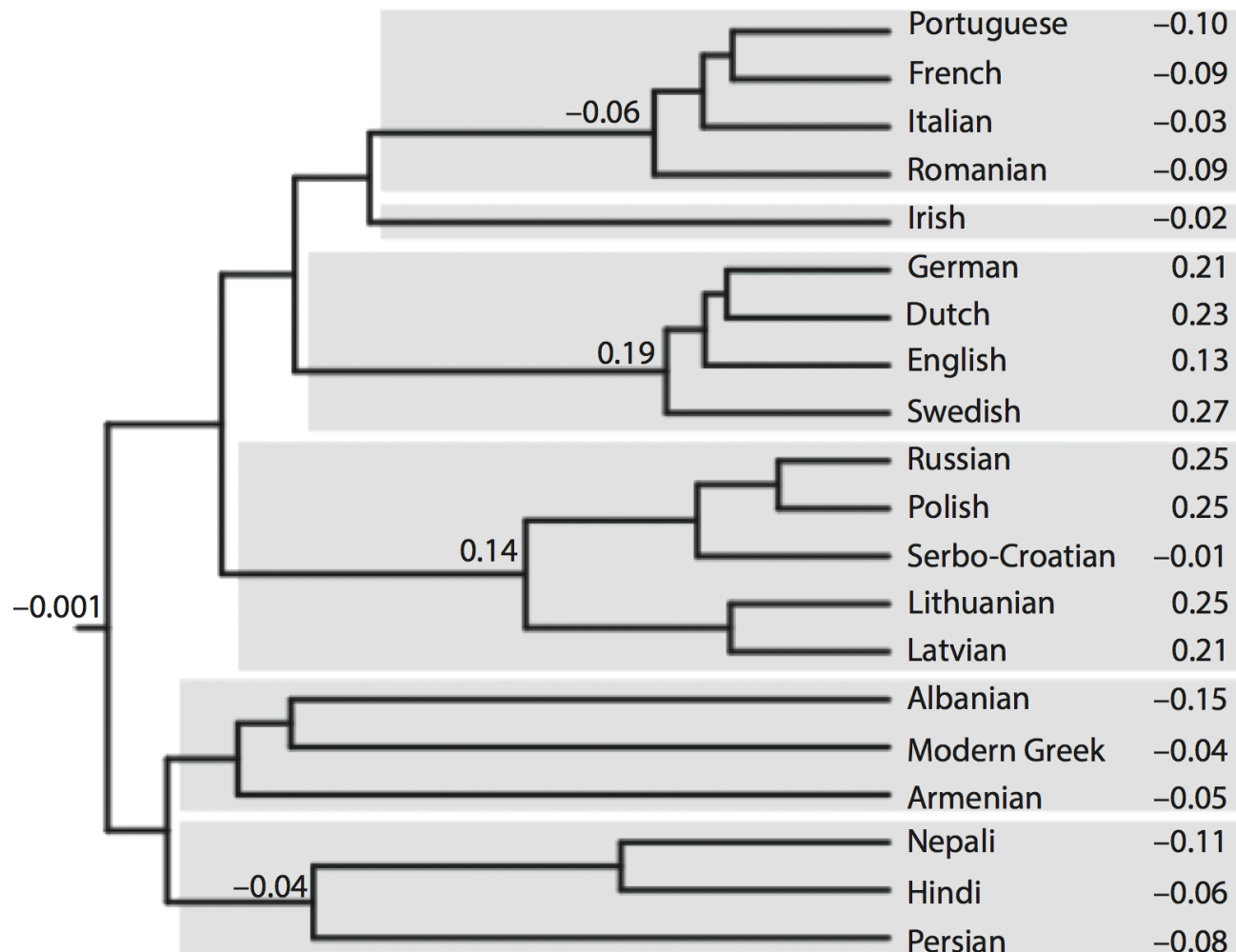


Figure 5. Scores on the 1st PC placed on the maximum clade credibility tree based on Bouckaert et al. (2012). Ancestral state estimates of the score on the 1st PC are added for Proto-Indo-European and the Germanic, Romance, Balto-Slavic and Indo-Iranian subgroups

Ancestral states

There are a number of different approaches to fit a Brownian model for continuous ancestral state estimation, the most common of which are implemented in R's **ape**, function `ace()`:

Maximum Likelihood (Schluter et al. 1997)

Least Squares (Felsenstein 1985)

Generalized Least Squares (Martins and Hansen 1997)

If you want to infer ancestral states with Bayesian methods, use `Continuous` in `BayesTraits` (Pagel and Meade n.d.)

Tutorial 2

Ancestral state estimation

Table 11.1 Demographic correlates of sex ratio at birth, using CONTINUOUS.

Variables correlated	LR	p	λ	r(given λ)	
SRB \times SRC	24.21	0.000	0.69	0.70	***
SRB \times SRA	0.45	0.341	0.59	0.11	
SRB \times lnTFR	3.33	0.010	0.73	-0.30	**
SRB \times lnIMR	1.99	0.045	0.79	-0.23	*
LnTFR \times lnIMR	20.42	0.000	0.70	0.66	***
SRB \times lnMMR	2.60	0.023	0.70	-0.26	*

Key: SRB = sex ratio at birth, SRC = sex ratio in children between 0–14, SRA = sex ratio in adults between 15–65, lnTFR = natural log of the Total Fertility Rate, lnIMR = natural log of the mortality rate of those under one year of age, lnMMR = natural log of the maternal mortality ratio (mothers death at or around childbirth per 100,000 births). LR = likelihood ratio, λ is a scaling parameter, r(given λ) is the correlation coefficient when the scaling parameter is applied (see CONTINUOUS manual: www.ams.rdg.ac.uk/zoology/pagel).

Ordinary Least Squares

$$y = X\beta + \varepsilon$$

vector of
response values

matrix with
predictor values

vector of
regression
coefficients

matrix with errors
(residuals)

Assumption of OLS: Errors must be
independent and identically distributed

	I1	I2	I3	I4
I1	1			
I2	0	1		
I3	0	0	1	
I4	0	0	0	1

(Phylogenetic) Generalized Least Squares

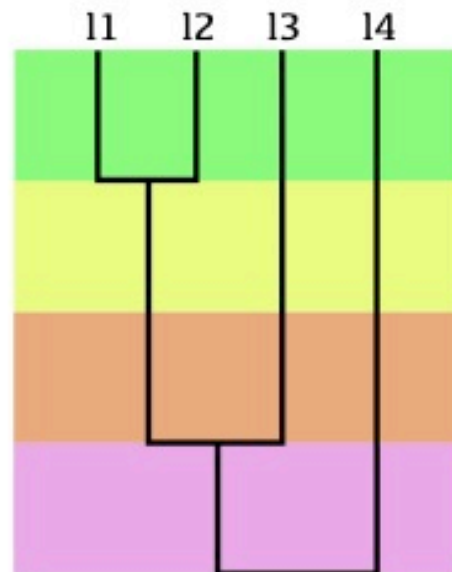
$$y = X\beta + \varepsilon$$

vector of
response values

matrix with
predictor values

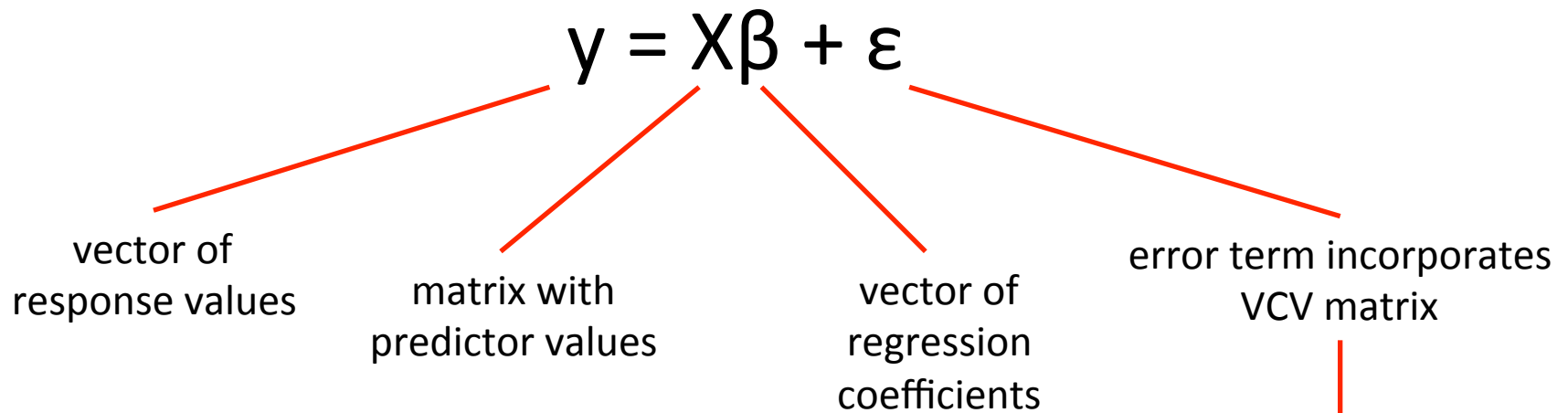
vector of
regression
coefficients

error term incorporates
VCV matrix

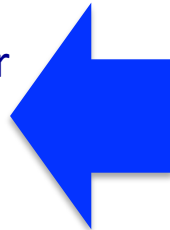


	l1	l2	l3	l4
l1	1+2+1			
l2	2+1	1+2+1		
l3	1	1	3+1	
l4	0	0	0	4

(Phylogenetic) Generalized Least Squares



!!! Evolutionary models other than Brownian motion can be substituted here !!!



	I1	I2	I3	I4
I1	1+2+1			
I2	2+1	1+2+1		
I3	1	1	3+1	
I4	0	0	0	4

Mixed Models...?

This has been a single level model; what about multi level models as presented by Heidi?

R has a package **MCMCglmm** (MCMC Generalised Linear Mixed Models) that can deal with that – but I have never worked with it 😊.

(Phylogenetic) Generalized Least Squares

In R: `pgls()` in package `caper`; there is also a package `pGLS` that I am not familiar with but seems promising

Tutorial 3

Correlated evolution with PGLS

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Credits

- Luke Mahler's continuous presentation
- Nunn (2011)
- Fiona Jordan 😊