

Growing Trees from Big Data

Bayesian Phylogeny for Historical Linguistics

Gereon Kaiping

2017-07-18

1 Solutions to All Your Problems!

- Crunching Numbers
- Bayesian ...
- ... Phylogenetics

2 Examples

- Austronesian: Branches and times
- Bantu: Phylogeography
- Indo-European: Ancient written sources

3 And why actually not.

4 Conclusions

- Further Reading

Problem

Using the comparative method is hard and limited, because

- it is a lot of painstaking work,
- we don't know how to weigh the evidence,
- is a mix of hypothesis generation and validation,
- loan words and chance resemblances make our lives more difficult,
- cognates may have changed meanings (but how far?).

And then it doesn't even give us dates, just "not before" or "not after" if we are lucky.

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Solution

Tree reconstruction methods from **Bioinformatics**

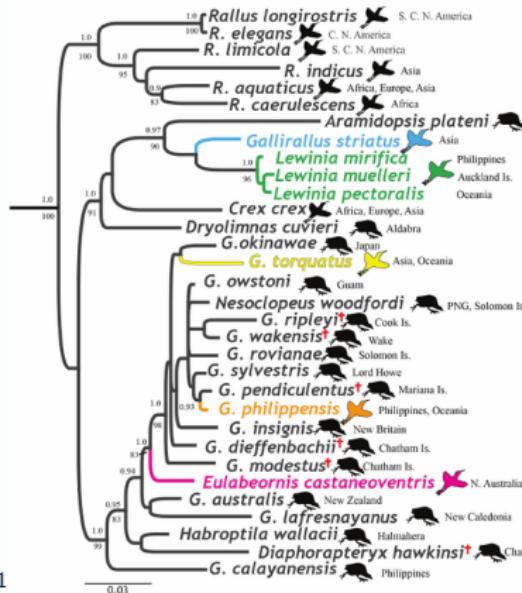
CCTCCACGCC AACGGAGCCT CATTCTTCTT
CCTCCACGCC AACAAAGCCT CATTCTT---
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CCTCCACGCC AACGGAGCCT CAGGTGTCTT
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¹Garcia-Ramirez et al. (2015)

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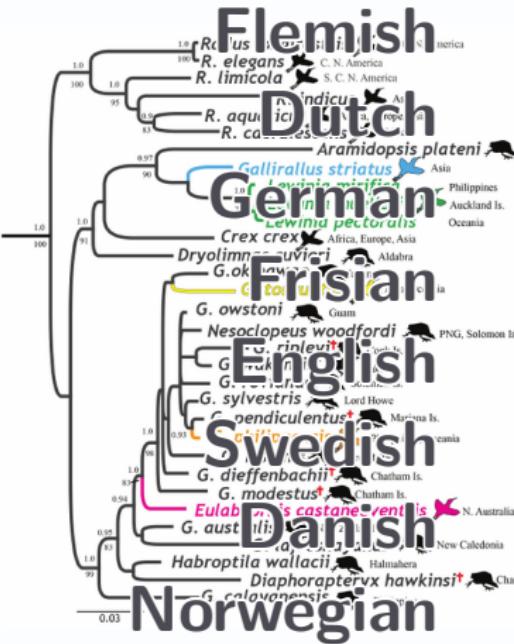
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Solution

Tree reconstruction methods from Bioinformatics

h	ʊ	n	-	ə	r	-	t	_	v	c	ʊ	-	t
h	ʊ	n	d	e	-	-	t	_	v	ɔ:	-	-	t
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h	e	n	d	-	ɹ	ə	d	_	w	ɜ:	-	-	d
h	ʊ	n	d	-	r	a	-	-	o:	-	-	-	d



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¹Garcia-Ramirez et al. (2015)

Evolution as a random process

Idea Evolution = a random process on a tree².

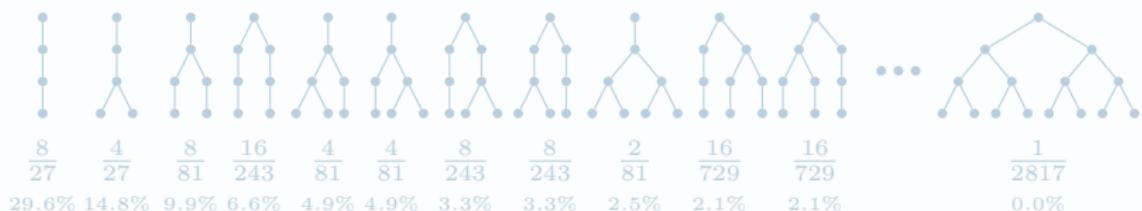
Generate a tree using dice rolls.



Example

- Start with a single language, and proceed for 3 generations.
- In each generation and language, roll a dice: On 1, split the current language in 2.

Possible trees:



"Likelihood": $P(\text{Data} \mid \text{Model})$

²or a network or a population, as long as we can formalize it.

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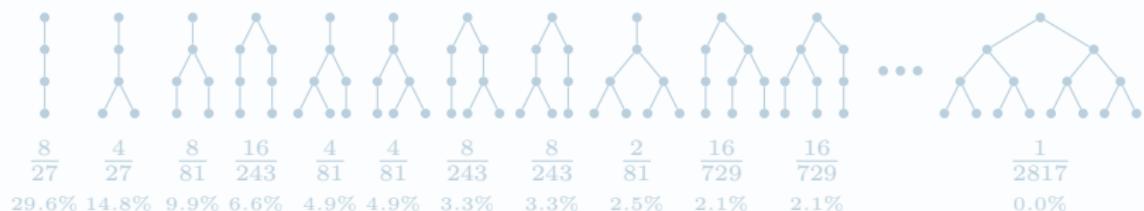
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Example

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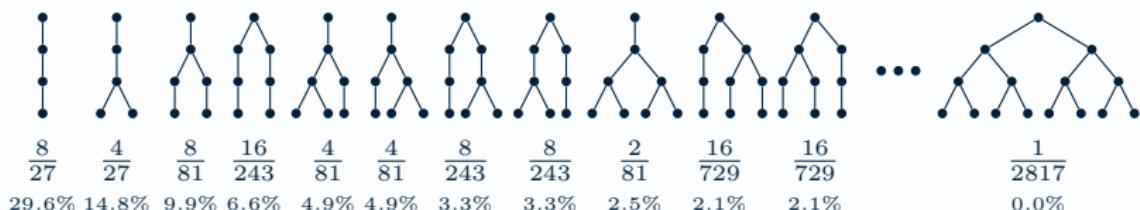
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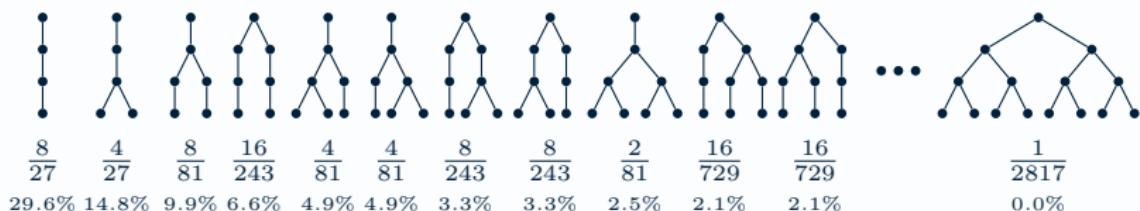
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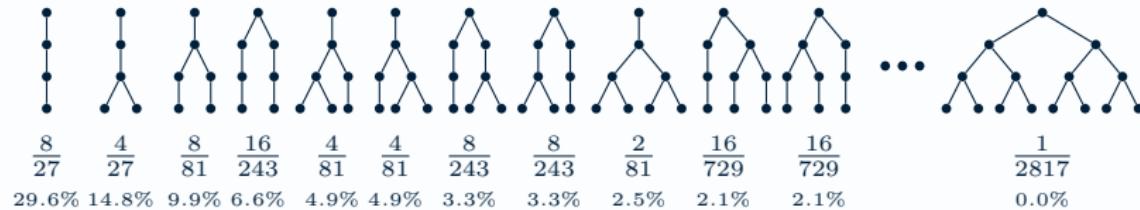


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Going back from data

Is given data compatible with this model?



"I generated a tree with three recent languages."

"Also, my first roll was one of ☀☀☀"

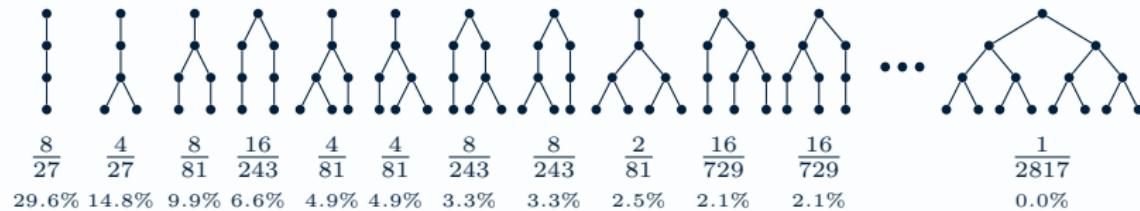
How compatible is this data with this or that model? Which models should I believe in (and how much)?

Probabilities = confidence of belief. Not: repeatable random experiment.

"Posterior probability": $P(\text{Model} \mid \text{Data})$

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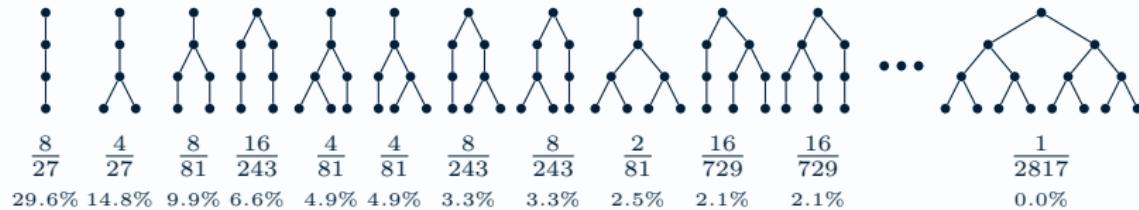
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Bayes' Theorem

$$P(\text{Model} \mid \text{Data}) \propto P(\text{Data} \mid \text{Model}) \times P(\text{Model})$$

“What did the language history look like?”

=

“What trees are compatible with the data and my idea of language change?”

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“Weighted by how ‘strange’ they are, how well does each tree explain my data?”



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Bayesian phylogenetic inference⁴ may look complicated, but it is

- model-based
- can incorporate prior knowledge
- outputs result uncertainty
- gives implicit weights from first principles

⁴Dunn (2015, 2009), Michael et al. (2015)

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Computational Phylogenetics

Markov Chain Monte Carlo (MCMC): “Roll dice to generate trees, but only keep the good ones”

Need:

- simple stochastic model(s) of language evolution, with parameters.
Example: *generalized binary model*



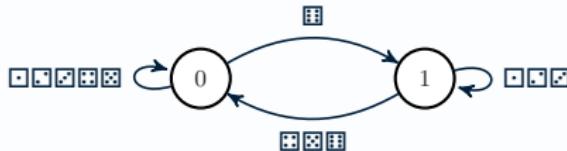
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- large dataset of model-compatible data

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Data for Computational Phylogenetics

- Swadesh lists: models based on semantic change (like Glottochronology – but much more flexible)

Language	<i>hand</i>	<i>two</i>
Dutch	hant	tve
English	hænd	tu
French	mẽ	dø
Indonesian	taŋan	dua

- Geography: various models
- Phonetic alignments: still in infancy
- Typological data: Some approaches, problems with universals/pathways
- Morphosyntax: ??????

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English	1	4
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Indonesian	3	5

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So much the theory.

- Austronesian: Branches and times – Gray, Drummond & Greenhill (2009)
- Bantu: Phylogeography – Currie et al. (2013)
- Indo-European: Ancient written sources – Chang et al. (2015)

Example 1: Austronesian

abvo.org ~ Austronesian ~ Trees ~ (Beta)

Austronesian Basic Vocabulary Database

Word: hand

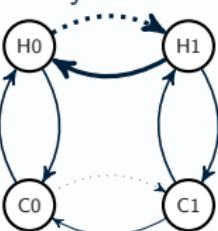
Entries for "hand":

ID	Language	Item	Annotate	Cognacy	Classification	Loan
Hand						
317959	Noroi (Bengt)	re-vorat				
318000	Noroi (Bengt)	re-vorat				
330644	Mohican Tsooping	kunay				
350918	Proto-Océan	*kunay				
327485	Osage-Missouri (1773)	in-e-nay				
215781	Proto-Mon-Khmer	*ŋ̚iŋ̚ai				
209395	Proto-Mon-Khmer	*ŋ̚iŋ̚ai				
209397	Chamorro	cos				
208506	Chamorro	cos				
177236	Hut	tap				
208624	Hut	tap				
208625	Moikorene, Cebuano	tae				
246690	Hang (Amping)	teritɔ̄s				
246691	Hang (Amping)	teritɔ̄s				
204647	Batak	teritɔ̄s				
247574	Bugis (Nalati)	teɔ̄s				
247575	Bugis (Makassar)	teɔ̄s				
247576	Bugis (Tana Toraja)	teɔ̄s				
308716	Satelite	tae				
205	Proto-Austronesian	*təŋ̚iŋ̚a				
184446	Proto-Austronesian (Lexic)	*kənay		1	Austronesian	
384949	Proto-Austronesian (Lexic)	*kənay		1, 79	Austronesian	
216	Malay - Cebuano (Bano)	tae		2	K-A-K-CUf	
109621	Malay - Cebuano (Bano)	tae		2	K-A-K-CUf	
205351	Malay - Cebuano (Bano)	tae?		2	K-A-K-CUf	
71434	Malay - Soglio (Pap)	tae?		2	K-A-N-Soglio	
71425	Malay - Soglio (Pap)	rapa?		2	K-A-N-Soglio	
71426	Malay - Soglio (Pap)	kaas		2	K-A-N-Soglio	
71427	Malay - Soglio (Pap)	ketem		2	K-A-N-Soglio	
71428	Malay - Soglio (Pap)	avet		2	K-A-N-Soglio	
205352	Malay - Soglio (Pap)	avet?		2	K-A-N-Soglio	
235	Sundanese (Sunda)	bae?		51	K-A-Sundanese	
207720	Sundanese (Pawon)	bae?		51	K-A-Sundanese	
207520	Sundanese (Tatah)	bae?		51	K-A-Sundanese	
207498	Sundanese (Husia)	bae?		51	K-A-Sundanese	
207353	Sundanese (Husia)	bae?		51	K-A-Sundanese	
71419	Burun Pit, Southern	inot		3	A-Burun	
71420	Burun Pit, Southern	lapid				
386000	Burun (Tabukabu LBS)	raŋ̚at?		1	A-Burun	
203720	Burun (Tabukabu LBS)	raŋ̚at?		1	A-Burun	
203946	Burun (Tabukabu)	raŋ̚a		1	A-Burun	
204355	Burun (Tabukabu)	raŋ̚a		1	A-Burun	
60822	Amis (Central)	kaŋ̚ay		13	K-A-SE	
203920	Amis (Central)	kaŋ̚ay		13	K-A-SE	
203941	Amis (Central)	kaŋ̚ay		13	K-A-SE	
203941	Amis (Central)	kaŋ̚ay?		1	K-A-SE	
203941	Amis (Central)	kaŋ̚ay?		1	K-A-SE	
210373	Borneo T'boli (T'boli)	chma		1	K-A-T'boli	
210395	Borneo T'boli (T'boli)	chma		1	K-A-T'boli	
210595	Borneo Ida	chma		1	K-A-T'boli	
63899	Kwantes Hill	lins?		1	K-A-N-Kwantes	
63900	Kwantes Hill	gralp		3	K-A-N-Kwantes	

- Austronesian Basic Vocabulary Database: several 1000 cognate classes for 210 meanings in 400 langs

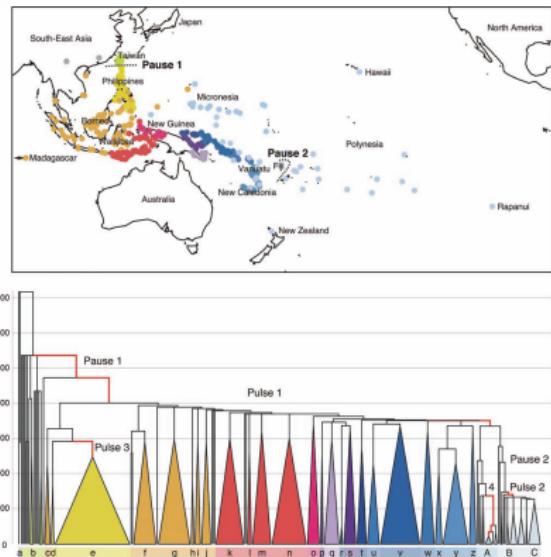
- Plus two “outgroup” langs, minus borrowings

- Binary covariation model



- Calibrations and variable replacement rates

Example 1: Austronesian



"The invention of the outrigger canoe and its sail may have enabled the Austronesians to move across this channel before spreading rapidly over the 7000 km from the Philippines to Polynesia (4). This is supported by linguistic reconstructions showing that the terminology associated with the outrigger canoe complex can only be traced back to Proto-Malayo-Polynesian and not Proto-Austronesian (41)."

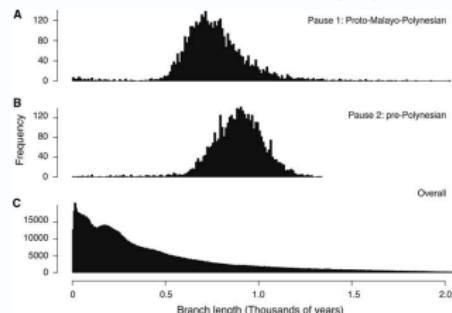


Fig. 3. Histograms of the branch length distributions. (A) The distribution of the Proto-Malayo-Polynesian pause, (B) the distribution of the pre-Polynesian pause, and (C) the overall branch-length distribution.

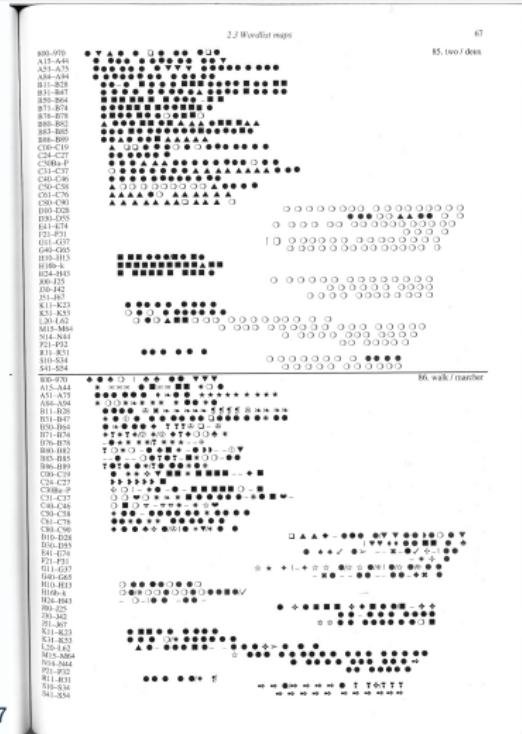
⁶Gray, Drummond & Greenhill (2009)

Example 1: Austronesian – Critique

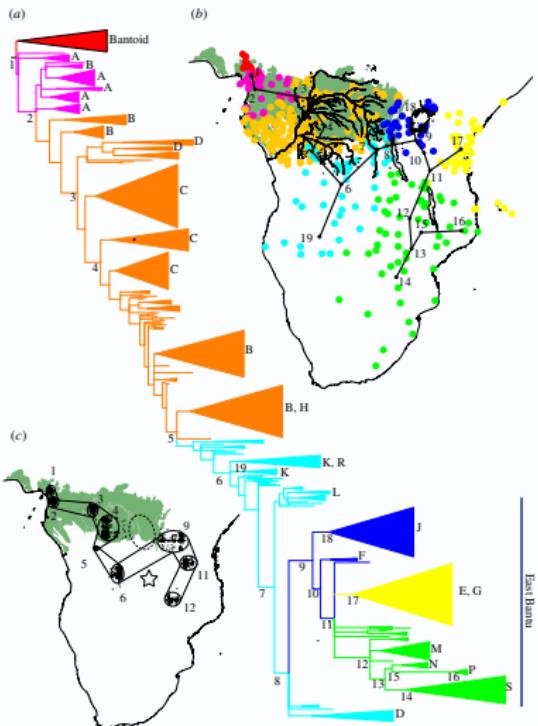
- Pauses and pulses appear with high posterior probability
- Prior? Do the results follow from data or original guess?
- Some subgroupings not linguistically supported – Data contains sociogeography
- How realistic is binary covarion?

Example 2: Bantu

- 2908 cognate classes for 90 meanings in 542 varieties of Bantu/Bantoid, with geographical point-coordinates
- Binary covariation with 6 (empirical) rate categories
- Brownian motion ancestral state reconstruction of latitudes and longitudes on 500 best trees
- Branch-dependent speed of movement and lexical change
- Other statistical analyses



Example 2: Bantu



"These debates have implications regarding the origin and spread of important cultural innovations, such as metallurgy and cattle-keeping."

"[...] explicit mapping of ancestral locations to make inferences about the specific route taken during the dispersal of Bantu languages. The results clearly support the 'pathway through the rainforest' scenario for the expansion of Bantu through much of sub-Saharan Africa. There is no support in these analyses for an early, deep split between East and West Bantu languages and a movement by one branch north of the rainforest."

Example 2: Bantu – Critique

- Several robustness checks of parameters
- Prior? Geography without lexical data?
- How good is Brownian motion as model for language spread?
Language shift and post-split contact might affect geographic inference.
(Though the fundamental results look robust.)

Example 3: Indo-European – Data and Prologue

a 066 HAND

b

066 73 Ossetic
066 59 Gujarati

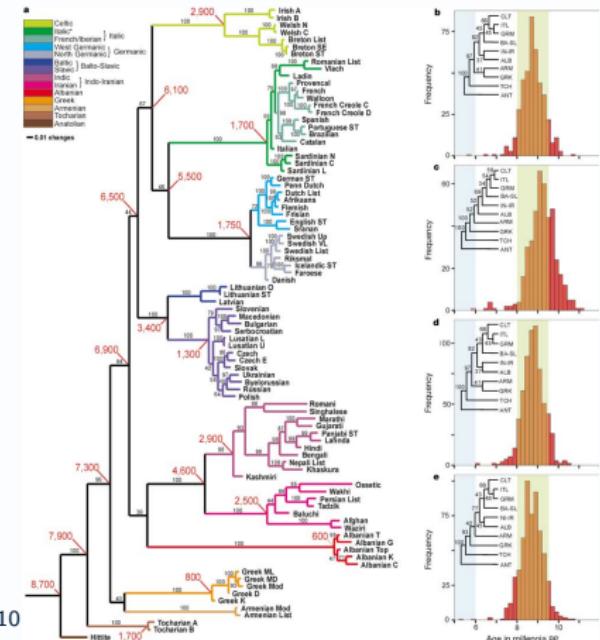
001

K"YX
NATH

c

066 17 Sardinian N	MANU
066 18 Sardinian L	MANU
066 09 Vlach	MYNE
066 22 Brazilian	MAO
066 21 Portuguese ST	MAO
066 15 French Creole C	LAME
066 13 French	MAIN
066 16 French Creole D	LAME
066 14 Walloon	MIN
066 12 Provencal	MAIN
066 20 Spanish	MANO
066 23 Catalan	MA
066 10 Italian	MANO
066 19 Sardinian C	MANU
066 11 Ladin	MAUN
066 08 Rumanian List	MINA

002

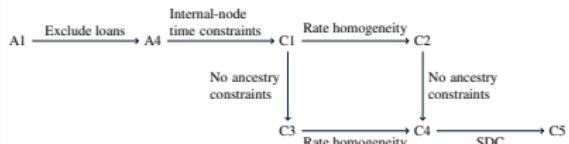


Example 3: Indo-European

IELex hand login

ID	Language	Source Form	Phonological Form	Notes	Cognate Class
11.4	Proto-Indo-European	*mon-u-			E
11.4	Proto-Indo-European	*gʰ̥es-r(o)-			C
11.4	Proto-Indo-European	*gʰ̥es-t(o)-			
80	Hittite	keššar			C
133	Luvian	iššaris			C
134	Lycian	izre			C
81	Tocharian A	tsar			C
82	Tocharian B	ṣar			C
88	Albanian	dorë		A singularised neut. plural PAIb ...	C
143	Standard Albanian	dorë			C
2	Albanian Sicily	dorë		A singularised neut. plural PAIb ...	C
4	Albanian Corinth	dorë		A singularised neut. plural PAIb ...	C
3	Albanian Gheg	dorë		A singularised neut. plural PAIb ...	C
6	Albanian Tsk	dorë		A singularised neut. plural PAIb ...	C
173	Mycenaean Greek	ke-º	kʰer-	Attested as an element in ...	C
110	Ancient Greek	χείρ	kʰé:r	G.sg. χειρός	C
152	Tsakonian	χερά			C
32	Greek	χερί	'çeri		C
31	Greek Lesbos	CHERI			C
129	Classical Armenian	ծեռն	jeñn		C
8	Armenian Eastern	ծեռց	džerkʰ		C
7	Armenian Western	ծեռց	ts'erkʰ		C
11	128 Avestan	zastō			C

11 Dunn (2015)

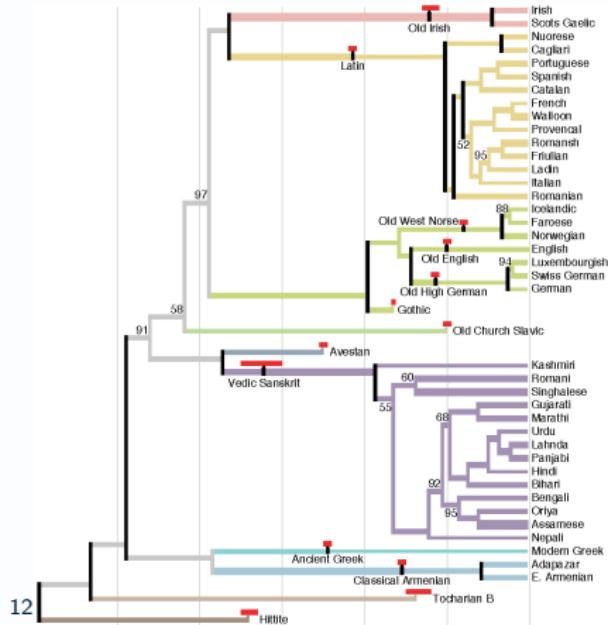


Starting from a replication of previous work (Bouckaert et al. 2012), improve

- data
- methodology
- tree prior
- post-processing

comparing each step.

Example 3: Indo-European



“Here we present a phylogenetic analysis in which ancestry constraints permit more accurate inference of rates of change, based on observed changes between ancient or medieval languages and their modern descendants, and we show that the result strongly supports the steppe hypothesis.”

“Because previous statistical phylogenetic research supported the Anatolian hypothesis, linguists who find that hypothesis implausible for other reasons may dismiss statistical analyses that purport to determine ancestral chronology. [...] statistical phylogenetic analysis can yield reliable information about pre-historic chronology, at least where all of the available data is taken into consideration.”

¹²Chang et al. (2015)

Example 3: Indo-European – Critique

- Very explicit about methodology (small steps, driver files available)
- Careful description of data coding
- Would someone have been this careful if the original results *had* matched the linguists' expectations?
- Ancestral constraints are very strong, and somewhat artificial in the model.

The discussion goes on¹³

¹³Verkerk (2017)

Bayesian phylogenetics will not solve all problems

- The papers show problems with Bayesian phylogenetics in practice
- Fundamental problems of Bayesian phylogenetics

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Problems with specific papers

Some papers

- disregard prior knowledge
- use models that don't fit their data
- don't show their priors

Not even in a better world.

State of the art models

- can only build trees, no language contact
- only support already cognate-coded data (baby steps towards phonetic data and typology), no distinction between innovation and retention
- are not realistic / not calibrated / have biases
- can't actually decide high-level relationships

I think we will always

- need domain experts
- have problems modelling morphosyntax
- have qualitative data that are hard to integrate

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Conclusions

- It is useful to talk about probabilities of events in the past
- Computer models can help make sense of large data sets
- The computer only tests consistency or helps build intuition, it does not replace expertise
- Very few language-appropriate models so far
- Building a *good* inference is hard!
- Mathematical models can handle and combine new types of data for new types of results

If you disagree with results, *what parameters or choices do you disagree with?*

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