# Quantifying the Comparative Method

COMPUTATIONAL APPROACHES TO HISTORICAL LINGUISTICS

# The Comparative Method

The traditional workflow of historical and comparative linguistics

- Accounts for similarities that cannot be chance
  - Establishs genetic relationships among languages through commonly inherited forms

# The Comparative Method

A number of different steps (Trask 2000:64-67):

- 1. Establish genetic relationship *prima facie* 
  - Easy to do for closely related languages, such as Romance
- 2. Identify cognate sets through systematic correspondences of sounds in words of similar meaning
- 3. Set up proto-forms from the correspondence sets
  This allows us to reconstruct the proto-language and detect the sound changes that occurred from mother to daughter languages

### Comparative Method in Action

Fortson 2004: 131	Language	Word		
	Latin	centum		
	Greek	hekaton		
	Tocharian B	kante		
	Old Irish	cét		
	Middle Welsh	cant		
	Gothic	hund		
	Sanskrit	śatám		
	Avestan	satəm		
	Lithuanian	šim̃tas		
	Old Church Slavic	sŭto		

#### Comparative Method in Action

Latin			С	е	n	t	u	m
Greek	(h	e)	k	а		t	0	n
Tocharian			k	а	n	t	е	
В								
Old Irish			С	é		t		
Middle			С	а	n	t		
Welsh								
Gothic			h	u	n	d		
Sanskrit			Ś	а		t	á	m
Avestan			S	а		t	ə	m
Lith			Š	i	ñ	t	а	S
OCS			S	ŭ		t	0	

### Comparative Method

From these correspondence sets, we can reconstruct a proto-form: PIE  $*\hat{k}mtom$ 

This process requires expert knowledge of the languages involved

Easy with a limited data set

 How can we do something like the Austronesian language family (1200+ languages)?

# Computational Approaches

Quantitative and computational methods are being used more and more in historical linguistics

- More objective, transparent, and easily replicable (List & Moran 2013)
- Built from evolutionary phylogeny
- Concerned with the evolutionary history of species, genes, and morphological characteristics
- Compare to historical linguistics: investigates evolution of language, grammatical features, and words
- Data structure is similar—sequence of characters (DNA, etc.)

# **Computational Approaches**

Previous methods:

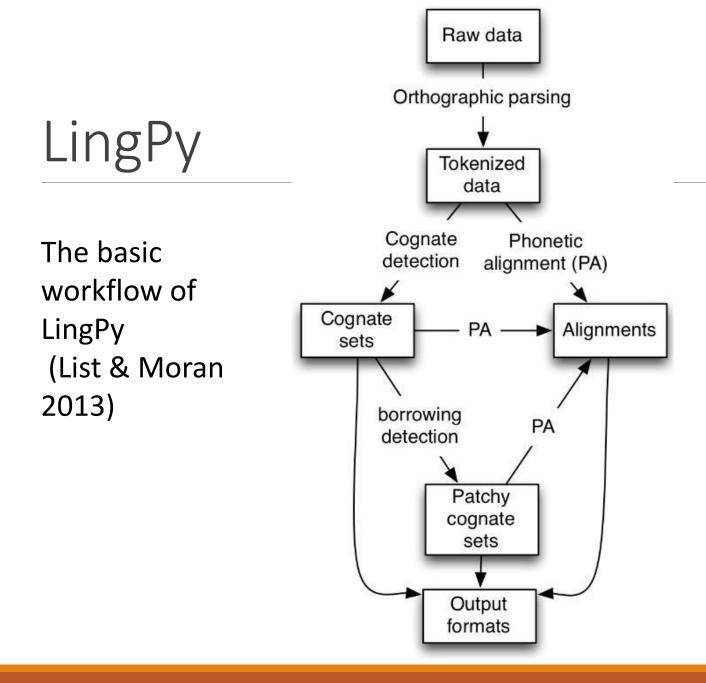
- Phonetic alignment algorithms (Kondrak 2000)
- Tests for genealogical relatedness (Kessler 2001)
- Phylogenetic reconstruction (Holman et al. 2001)
- Automatic cognate detection (Steiner et al. 2011)
- Automatic borrowing detection (Nelson-Sathi et al. 2011)
- Automatic proto-form reconstruction (Bouchard-Cote et al. 2013)

# LingPy (List & Forkel 2016)

An open-source python library

Source code is readily available online (lingpy.org)

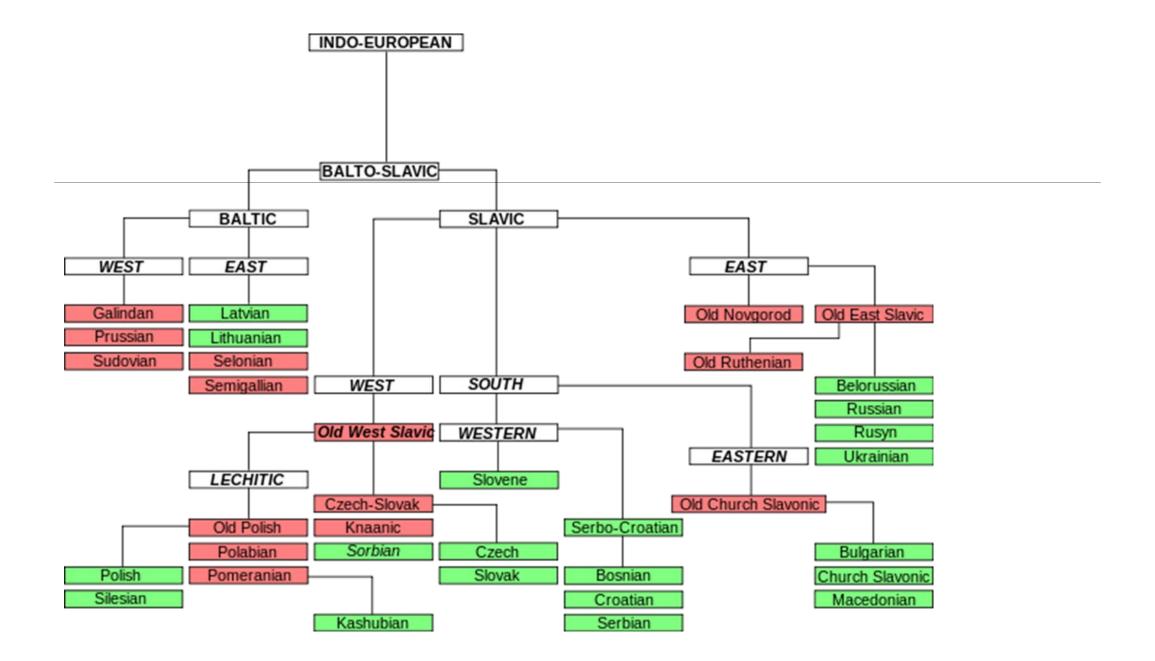
It implements many computational methods in a general workflow mimicking the Comparative Method



# Current Study

Apply the LingPy methods to Baltic and Slavic data

- Look at cognate judgements
- Establish a phylogenetic tree
- Create rough reconstructions
- Look at borrowing networks



# Balto-Slavic Languages

Branch of Indo-European

- The specific relation between them is controversial
  - A single branch, like Indo-Iranian?
  - Two separate branches?
- Large number of words shared exclusively by Baltic and Slavic (Trautmann 1923)
- no major isogloss that separates the two branches
- Relatively lately attested:
  - Slavic ca. 9<sup>th</sup> century
  - Baltic cs. 12<sup>th</sup> century

#### Data

Swadesh lists for 6 Slavic languages and 3 Baltic languages
 Bulgarian, Czech, Croatian, OCS, Polish, Russian

- Latvian, Lithuanian, Old Prussian
- 172 words in each list

Lists taken from the Indo-European Lexical Cognacy Database (IELex, http://ielex.mpi.nl/)

Compiled from various etymological dictionaries

#### Data

Input data: Wordlist

 Tab-delimited text file organized into rows and columns with headers

ile Edit	Format \	/iew Help				
Word1		ien nep				_
DATA						
D	CONCEPT	IPA	DOCULECT	TOKENS		
ŧ						
540	I	as	Bulgarian	a s		
541	I	ja	Russian	ja		
.542	I	ja	Polish	ja		
.543	I	ja:	Czech	ja:		
.544	I	as	Old Prussian	a s		
545	I	es	Latvian	e s		
546	I	e∫	Lithuanian	e∫		
.547	I	jâ:	Serbo-Croatian		j â	:
548	Ι	azŭ	Old Church Slavi	ic	a z	Ŭ

# Implementing the data

Import the wordlist file

LingPy can manipulate the data

- Find specific entries for concepts
- Return entries for specific languages
- Add new entries

The IPA entries need to be tokenized and aligned

#### Cognate Judgements

After tokenization, cognate judgements can be determined

Follows the STARLING approach

Cognate words are assigned the same cognate ID

Accomplished through the LexStat method (List 2012)

• Other methods (Turchin, NED, and SCA) are also available in LingPy

### Results: Example Alignment

Language	Alignments					
Bulgarian	d		V	а		
Croatian	d		v	âː		
Czech	d		v	а		
Polish	d	-	V	а		
Russian	d	-	V	а		
OCS	d	ŭ	V	а		
Latvian	d	i	V	i		
Lithuania	d		V	i		
Old Prussian	d		W	ai		

Plays a crucial role in automatic approaches

Gets at the idea of sound correspondence sets

#### LexStat

Language-specific: no predefined scoring function

Uses an expanded version of Dolgopolsky's (1964) sound classes

Computes cognate distance scores through pairwise alignments, following Bouchard-Cote et al. (2013)

Close to the idea of sound correspondence sets

Words drawn from randomized sample

- Repeatedly aligned with each other
- Creates a distribution of sound transitions
- Compared to the actual distribution from aligned words in the wordlist

#### LexStat

Sequence conversion

Input converted to sound classes; sonority profiles determined

Scoring-scheme creation

• Language specific; created through a permutation method

Distance calculation

• Pairwise distance between all words are computed

Sequence clustering

- Sequences clustered into cognate sets whose average distance is beyond a certain threshold
- Flat cluster variant of the UPGMA algorithm

#### LexStat output

Balto-Slavic_lexstat.qlc - Notepad								
File Edit Format View Help								
# Wordlist								
# META								
@vowel	.s:TV_							
@json:	{"para	ms": {"cl	uster": "lexstat_	_upgma_0.	50", "c	scorer'		
# DATA								
ID	CONCE	PT IPA	DOCULECT	TOKENS	LEXSTA	TID		
#								
1540	I	as	Bulgarian	as	1			
1541	I	ja	Russian	jа	2			
1542	I	ja	Polish	jа	2			
1543	I	ja:	Czech	ja:	2			
1544	I	as	Old Prussian	as	1			
1545	I	es	Latvian	e s	1			
1546	I	e∫	Lithuanian	e∫	1			
1547	I	jâ:	Serbo-Croatian	jâ:	2			
1548	I	azŭ	Old Church Slav	vic	αzŭ	1		
#								

# Results: Cognate Judgements

Cognate words are assigned a CogID

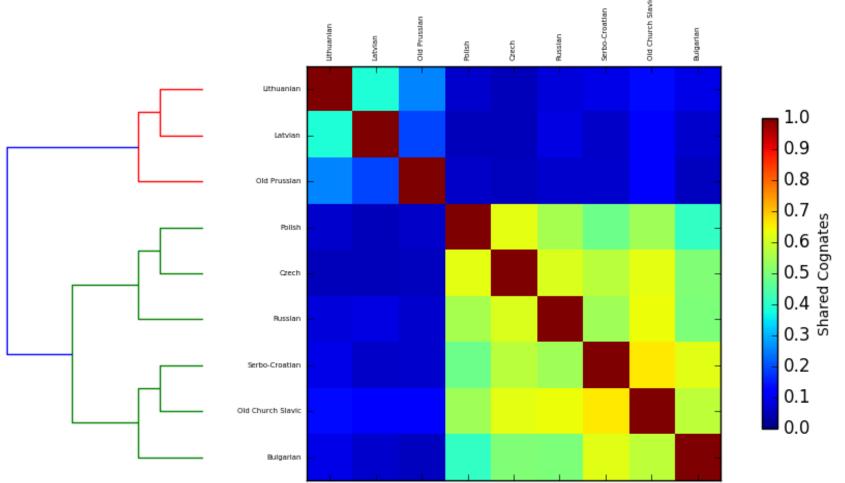
For example, every word for "two" has a CogID of 971

Not foolproof

- Some missed cognates: 'I', 'full', etc.
  - Actual cognates can be misidentified because of sound classifications, alignments, etc.

Can display the percentage of cognates shared by languages in a heat map

#### Percentage Shared Cognates



#### **Consensus Reconstruction**

From this, we can create "quick and dirty" reconstructions

- Consensus strings are calculated from all alignments
- Selects the most frequent characters
- Typically around 2 edit operations from expert reconstructions

#### Results: Reconstructions

Examples:

For 'two', we get \*dva • Cf. PSI \*dъva, PBSI \*duwō

For 'day', we get \*dein-Cf. PBSI \*dein-/\*din-

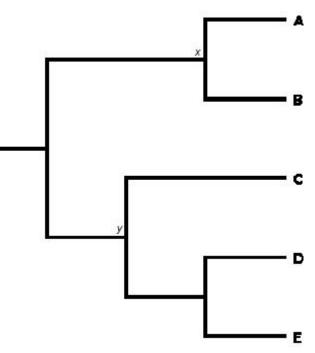
For 'stone', we get Slavic \*kamen-, Baltic \*akmens
Cf. PSI \*kamen~kamy, PB \*akmo

For 'good', we get Slavic \*dɔbr, Baltic \*labs • Cf. PSI \*dobrъ, PB \*labas

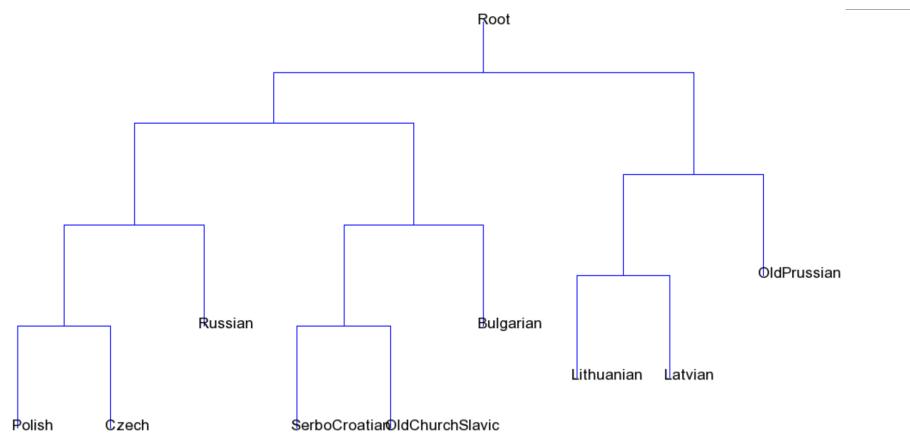
# Phylogenetic Trees

Also from this, we can create a simple bifurcating tree for the languages

- Use either Neighbor-joining or UPGMA
- Distance matrices=number of shared cognates
- Outputs simple Newick tree format
  - ((A,(B,C,),(D,E));



# Results: Phylogenetic Tree



# **Borrowing Detection**

Evolution of language is both a vertical and horizontal process

- Vertical=inheritance
- Horizontal=borrowing

Follows the method of Nelson-Sathi et al. (2011)

 Apply phylogenetic networks to recover frequency of hidden borrowings

# **Borrowing Detection**

Minimal Lateral Network (MLN)

- Networks=mathematical structures used to model pairwise relations between entities
  - Entities=vertices
  - Edges=interactions between vertices
- Applies the technique of gain-loss mapping to presence-absence patterns of cognate sets
- Searches for cognate sets incompatible with a reference tree typology
  - Points to borrowing

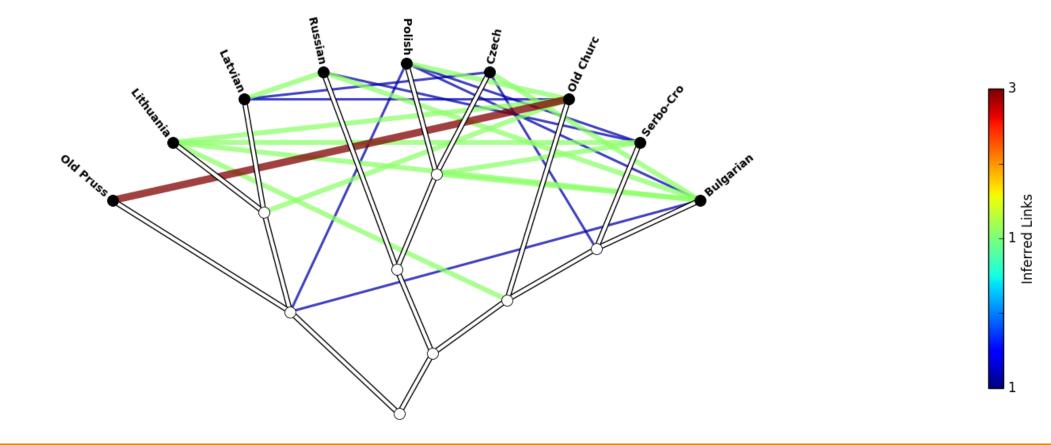
# Results: Borrowing Detection

Use MLN to capture the inferred horizontal relationships

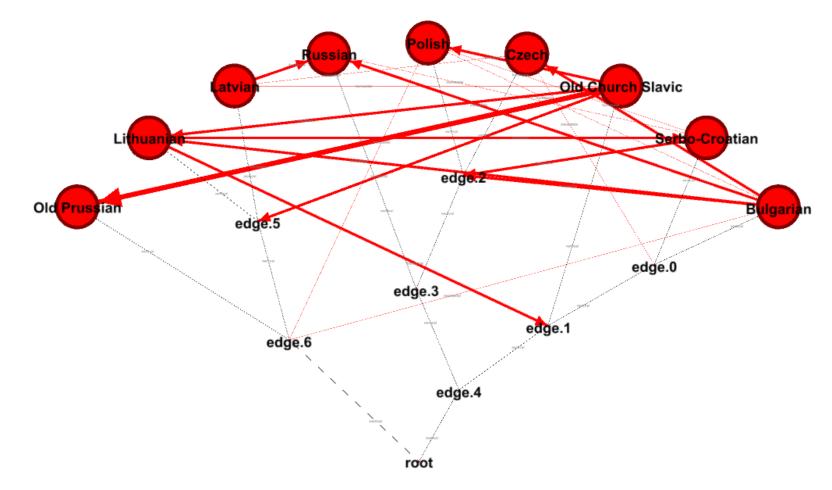
- Example: Old Prussian nage 'foot'
  - Cf. Lith. koja, Lat. kãja; Rus. noga, OCS noga

Plot the results against our reference tree

#### Results: Borrowing Detection



#### Results: Borrowing Detection w/Direction



## Conclusions:

Useful, but not infallible

Best combined with expert knowledge

Needs refinement in cognate judgements and reconstructions

Baltic and Slavic:

- Still uncertain about their exact relationship
  - Need to examine it further within a wider Indo-European context
- Extensive borrowing into Baltic from Slavic
- Latvian and Lithuanian are more closely related than Old Prussian

# Further Study

Cognate judgements

- Low B-Cubed scores (Bouchard-Côté et al. 2013)
- Expand on sound classes that are used to establish the cognate sets
- Implement expert judgements

IPA transcription

- This still has to be done by hand
- Letter-to-phoneme conversion as Machine Translation (Rama & Gali 2009)

# Further Study

Track the development of individual words through the language network

- Both inheritance and borrowing
- Examine intermediate stages of words

#### Implement more data

- More languages
- Longer wordlists
- Examine Balto-Slavic within a wider Indo-European context



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