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LANGUAGE EVOLUTION: THE EMPIRICAL TURN

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Ancestral State Reconstruction and Loanword Detection

Marisa Koellner and Johannes Dellert

University of Tuebingen

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Bibliography

References

IELex and Language Tree

- language sample (IELex): consists out of 207 concepts across 95 languages
- language tree: classifications from Ethnologue for living languages classifications from Glottolog for extinct languages
- cognate classes (IELex): represented at the leaves of the tree
- loanword judgements (IELex): binary annotation
 - 1 indicates loanwords judgement
 - 0 indicates either the absence of borrowing or incomplete data

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Ancestral State Reconstruction

Two versions of ancestral state reconstruction:

- the Sankoff algorithm for maximum parsimony, as implemented as part of the PAUP* software (assumes that there should be exactly one cognate class for each concept at each node, only allows multiple reconstructions if both variants lead to maximum parsimony)
- an alternative **threshold-based method** built on a recursively computed confidence measure (considers cognate classes separately, no bias against multiple reconstructed classes)

Bibliography

References

ASR: Sankoff Algorithm as in PAUP*



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ASR: Computing the Confidence Measure

- assign confidence cn(v, c) to each class c at each node v
- for attested languages, cn(v, c) := 1 or cn(v, c) := 0
- for non-leaves (i.e. reconstructed nodes), we recursively compute confidence values as follows (Ch(v) = children of v):

$$cn(v,c) := \max\left\{0, 1 - \frac{\sum_{\substack{v_i \in Ch(v) \\ |Ch(v)|}}{cn(v_i,c) + 0.5}}{\sum_{v_i \in Ch(v)} cn(v_i,c) + 0.5}\right\}$$

- intuition: close to 1 if average of child confidences is high, even closer to 1 if attested across many branches
- threshold: reconstruct class c for v whenever cn(v, c) > 0.4

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ASR: Our Reconstruction Method



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Loanword Detection



Internal Borrowing (concept mountain)



External Borrowing (concept mountain)



Cognate Classes: $t = t\bar{a}x$ $k = k\bar{u}$ $p = pu\bar{x}t\bar{a}$ g = gar/gorax = xox

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Semantic Evolution (concept *head*)



Cognate Classes: k = kopfh = head

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Evaluation

	detected loanwords	true loanwords	Precision	Recall
Confidence	1409	239/1100	16%	22%
Sankoff	4532	477/1100	10%	46%

- quite low overall performance
- loanword detection method highly depends on the quality of the hypothetical cognate classes at the internal nodes

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Evaluation

Limits of borrowing detection:

- $\bullet\,$ binary annotation $\rightarrow\,$ true loanwords might not be annotated
- $\bullet\,$ directionality $\rightarrow\,$ only target languages can be compared
- $\bullet~{\rm data}$ \rightarrow no gold standard including source languages

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Method Comparison

Performance of the methods shown on the concept mountain:

	detected loanwords	true loanwords
Confidence	2	2
Sankoff	39	3

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Method Comparison

Performance of the methods shown on the concept *spit*:

	detected lonwords	true loanwords
Confidence	1	1
Sankoff	8	0

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Additional Limits of loanword detection

- (1) external borrowing: the cognate class is not present in the tree
- (2) semantic evolution: the word changes its meaning over time
- (3) borrowing within a cognate class: the borrowing took place within one cognate class

Borrowing within a cognate class (concept mountain)



Further work

- (1) reconstruction: try more ancestral state reconstruction methods
- (2) data simulation: alternative evaluation of the model on much more data
- (3) more complex model of borrowing: detect more complex linguistic cases (e.g. within cognate classes)
- (4) collect expert loanword judgements: building a gold standard which includes source language
- (5) model directionality: getting a clearer picture of language contact

Thank you for your attention!

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