Mapping Phonology to Semantics: A computational model of cross-lingual spoken word recognition

I. Zaitova, B.M. Abdullah and D. Klakow, 2022



Presented by Emma Angela Montecchiari

Paper Review

Human Language Technologies Università di Trento Cognitive Science Master's 09-02-2024 Closely related languages exhibit intelligibility = Metric of cross-language similarity

Inter-comprehension =

Ability of intelligible languages speakers of comprehending each other's speech without explicitly learning the second language

Key influential aspects:

- Lexical distance [cognates]
- Phonological distance (Levenshtein distance)

Spoken word recognition theories and models

Mapping between words acoustic-phonetic representation (*acoustic realization*) and semantic representation in memory (*lexical knowledge*)

General objective: Exploring mutual intelligibility

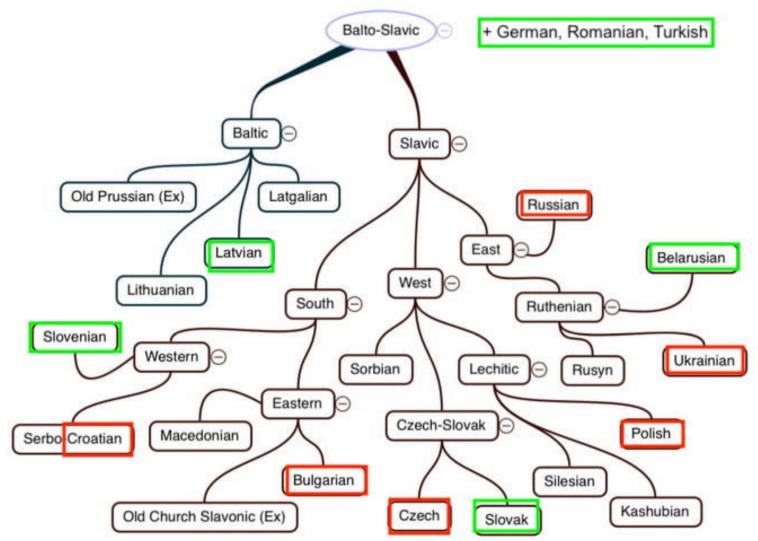
Method: Neural model of spoken-word recognition.

Procedure: Evaluating monolingual models on crosslingual performance

Cross-lingual performance: How are they able to understand spoken words meaning of related languages?

Monolingual Models trained on 6 Slavic languages

[exemplifying high mutual intelligibilty]



Cross-Lingual evaluation on Slavic and non-Slavic languages

Specific objectives:

a) How performance predicts languages mutual intelligibility?

B) Do results reflect languages genetic relation?

C) Identifying linguistic measures affecting performance.

6 Monolingual models — training on 6 Slavic languages

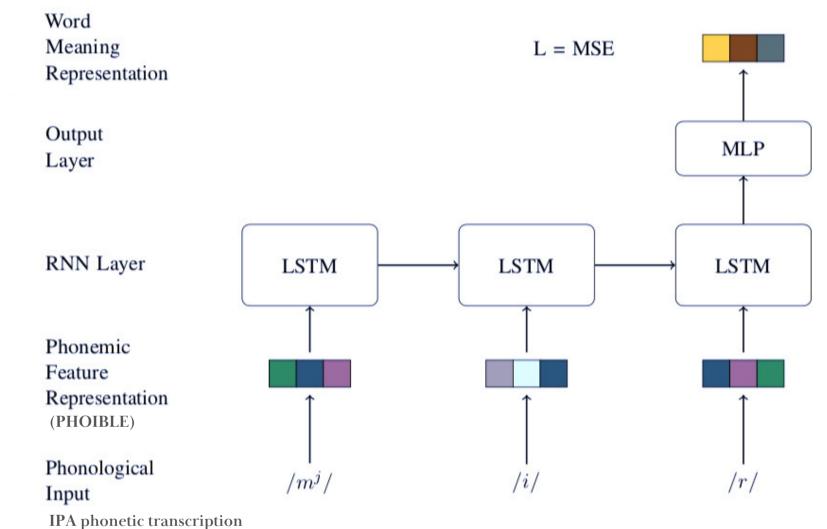
[exemplifying high mutual intelligibilty]

Common cross-language concepts sample of word forms selected from FastText (linguistic uniformity) (502 concepts)

Recurrent Neural Net (ADAM + MSE)

1 LSTM layer linear-tanh MLP Learning the mapping between (A) phonological sequences and (B) semantics embeddings (meaning representation)

FastText cbow

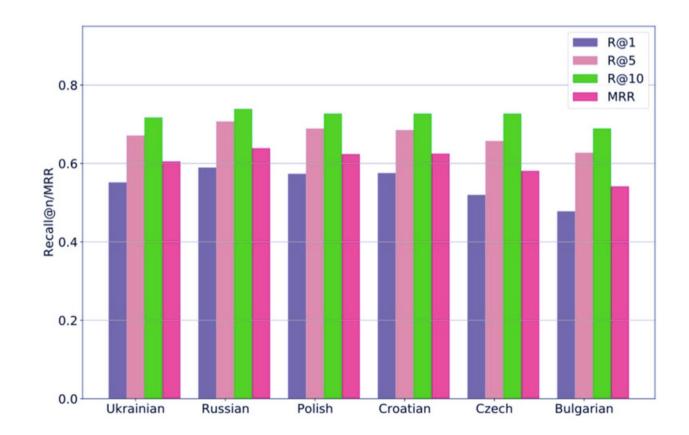


[502 common cross-lingual concepts words]

Procedure - Evaluation

<u>Evaluation</u>: cosine similarity between **embeddings Test set** retrieval - [output]-[target in training set]

Monolingual [training language=evaluation language]: [output]-[target in test set].



Cross-lingual evaluation [training (L1) language≠evaluation (L2) language]:

Cosine similarity between **embeddings** - [output]-[target in L2 test] + [L2 target]-[L1 target] + [L1 target]-[possible outputs in training set]

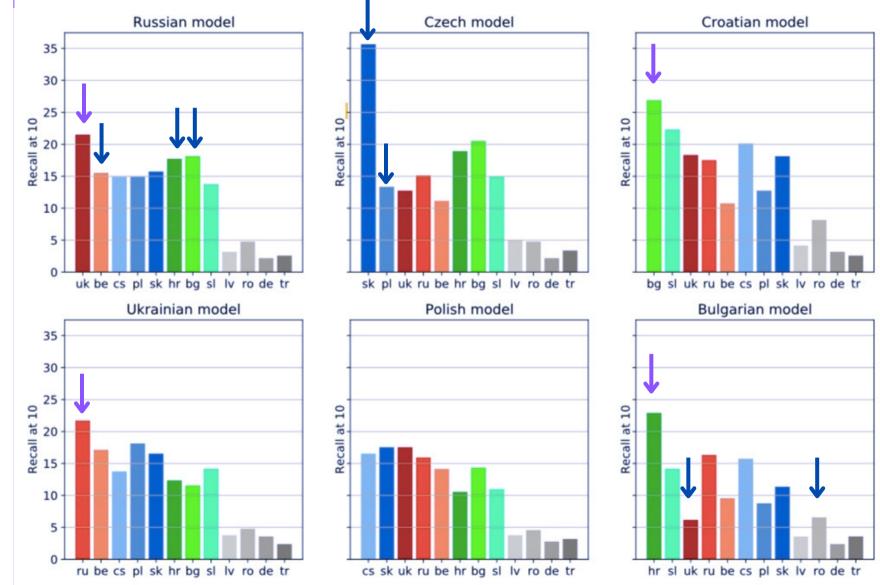
training: Russian -- e.g. word люди /ljudi/ 'people', testing: Czech concepts. evaluation: (1) compute the semantics representation of Czech lidé /lid@/ 'people' (2) estimate its similarity to test sequences in Russian with the target meaning representation being that of the Russian word люди /lj u d i/.

	Russian		Czech		Bulgarian	
Concept	Orth	IPA	Orth	IPA	Orth	IPA
EAR	yxo	/u x a/	ucho	/u x o/	yxo	/u x ɔ/
NOSE	нос	/n o s/	nos	/n o s/	нос	/n ɔ s/

Performance metrics:

Performance Results

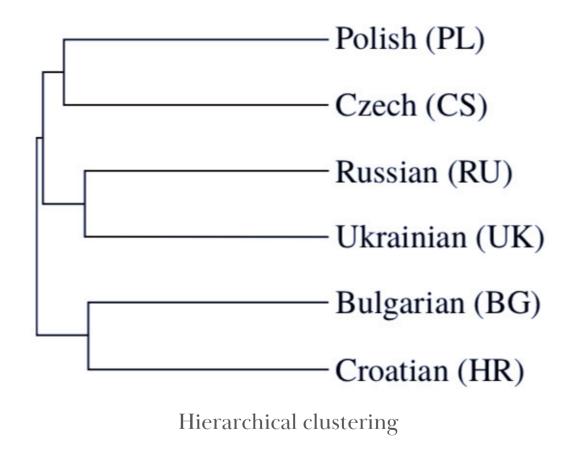
• Average recall at 1, 5 and 10 & Mean Reciprocal Rank (MRR)



Ukrainian – uk, Russian – ru, Belarusian – be, Czech – cs, Polish – pl, Slovak – sk, Croatian – hr, Bulgarian – bg, Slovene – sl, Latvian – lv, Romanian – ro, German – de, Turkish – tr

Linguistic predictors:

Levenshtein distance and PWLD & Hierarchical clustering among Recall@10



Additional Positive Overcomes:

- Long-short term memory network (LSTM) and multi-layer perceptron (MLP) are proven for cognitive validity in predicting human behaviour and cognitive features
- Adherence to multiple-trace theory, embracing continuity and coupling between speech perception, production, and memory.
- Ruled system discrete phonological representation, tackling the acoustic-phonetic invariance challenge

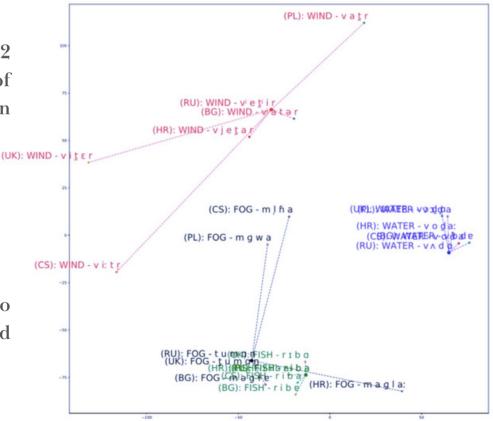
Additional Negative Overcomes:

• PWLD, has a lower correlation with retrieval metrics than LD

	R@10	MRR	cos sim	LD	PWLD
R10		0.98***	0.5***	-0.74***	-0.57***
MRR			0.5***	-0.75***	-0.56***
cos sim				-0.29*	-0.44***
LD					0.8***
PWLD					

• t-SNE clustering results raise 2 concerns about the alignment of similarly sounding words in different languages.

 Lack of accessibility to implementation code and training/testing data



Overcomes

Remarks:

- Claim for better explanation of semantic space computations (more examples);
- Claim for a visual representation of Slavic languages family tree.

Suggestions

- Underlie the quality of the model's architecture to enhance validity;
- Enhance a theoretical/historical framework in respect to the used model.

Suggestions for future directions:

- Bilingual training set towards an observation whether multilingualism enhance inter-comprehension;
- Explore a behavioural paradigm for model testing (multilingual modelling tasks as such hinting word meaning for L2 languages inputs, computing word similarity assessments);
- Asymmetrical intelligibility (Spanish and Portuguese)

Thank you for your attention!

