Filling the Gap: Decoding of Word Embeddings for Generation of Coherent New Words

Safa AlSaidi, Amandine Decker, Stephanie Monteiro

 $\mathsf{M2}-\mathsf{Software}\;\mathsf{Project}$



- State of the Project
- 2 Languages study
- 3 Results
- 4 Discussion
- Mhat to improve?
- 6 Future work

Reminder of our aim

• Apply decoder to the regression task (solving analogies)

$$A: B:: C: X \xrightarrow{X=?} A: B:: C: D$$

e.g. $star: stars:: cat: X \rightarrow cats$

- Current output: vectors (≠ word)
- Aim: transform these vectors into words

What we managed to do?

- Move all codes to PyTorch Lightning
- Research on morphology and variational auto-encoder
- Build the decoder based on word embeddings
- Train it on 11 Languages
- Test with different parameters
- Evaluate results with two metrics

- State of the Project
- 2 Languages study
- Results
- 4 Discussion
- 5 What to improve?
- 6 Future work

Language family

Language family	Languages		
Indo-European	German, Russian, Spanish		
Afro-Asiatic	Arabic, Maltese		
Uralic	Finnish, Hungarian		
Altaic	Turkish, Japanese		
Caucasian	Georgian		
Na-Dene	Navajo		

Figure: Classification according to language families

Morphological typology (1)

Figure: Classification according to the degree of internal complexity

Morphological typology (2)

Morphological type	Flectional	Agglutinating	
Characteristics	CumulationFusionInternal flection	 Morpheme ⇔1 meaning Clear-cut boundary Form not affected 	
Languages	German, Russian, Spanish, Arabic, Maltese	Finnish, Hungarian, Turkish, Japanese, Georgian, Navajo	

Figure: Classification according to the technique

Inflectional morphology

Affixes	Suffixes++	Suffixes+	=	Prefixes+	Prefixes++
Languages	German Russian Spanish Arabic Maltese Finnish Hungarian Turkish Japanese	Georgian			Navajo

Figure: Affixes used in inflectional morphology [Dryer, 2013]



Arabic, Maltese: templatic morphology (root-and-pattern strategy)

- State of the Project
- 2 Languages study
- Results
- 4 Discussion
- 5 What to improve?
- 6 Future work

Results with different parameters

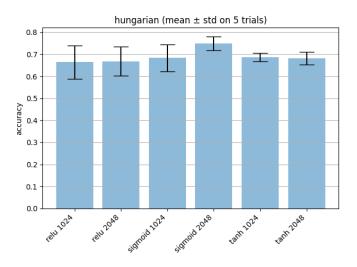


Figure: Mean accuracy (\pm standard deviation) on 5 trials for Hungarian

Results on all languages

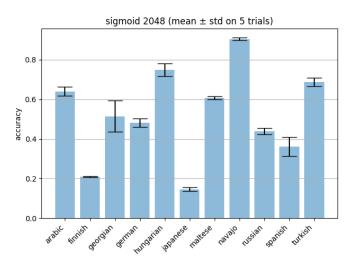


Figure: Mean accuracy (\pm standard deviation) on 5 trials with a sigmoid activation function and a hidden size of 2048

Results on all languages

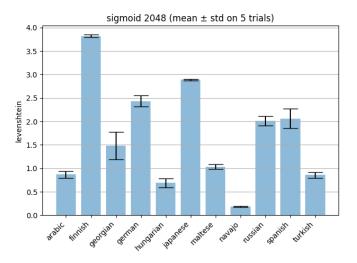


Figure: Mean levenshtein distance (\pm standard deviation) on 5 trials with a sigmoid activation function and a hidden size of 2048

- State of the Project
- 2 Languages study
- Results
- 4 Discussion
- 5 What to improve?
- 6 Future work

Leads to explain the results

- content of the embeddings
 - subwords ? (root-and-pattern strategy)
 - ► amount of different subwords
 - proximity of the subwords
- morphological features of the languages

- State of the Project
- 2 Languages study
- Results
- 4 Discussion
- **5** What to improve?
- 6 Future work

What to improve?

- Find a better evaluation metrics e.g.:
 - search for a new metrics that deals with word lengths
- Have a better understanding of the content of the embeddings:
 - subwords = morphemes ?
 - decoded words: real for some languages

- State of the Project
- 2 Languages study
- Results
- 4 Discussion
- 5 What to improve?
- 6 Future work

Future work

- 22 Nov Regression model + decoder / Variational auto-encoder
- 10 Dec Qualitative analysis / Multilingual model
- 14 Jan Application docker & webpage
- 3 Feb Report

شكراجزيلا Thank you Merci អរគុណ Obrigado

References I



Booij, G. E., Lehmann, C., Mugdan, J., and Skopeteas, S. (2008). *Morphologie: Ein internationales Handbuch zur Flexion und Wortbildung.* De Gruyter Mouton.



Dryer, M. S. (2013).

Prefixing vs. suffixing in inflectional morphology.

In Dryer, M. S. and Haspelmath, M., editors, *The World Atlas of Language Structures Online*. Max Planck Institute for Evolutionary Anthropology, Leipzig.



Eifring, H. and Theil, R. (2005).

Linguistic typology.

Linguistics for students of Asian and African languages.

Our decoder structure

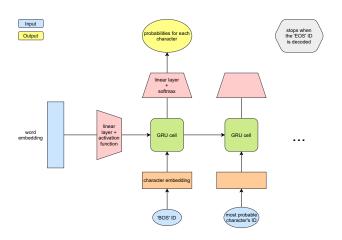


Figure: Our GRU based decoder

Inspired by this blogpost https://rajatvd.github.io/Generating-Words-From-Embeddings/

Results with different parameters (smaller hidden sizes)

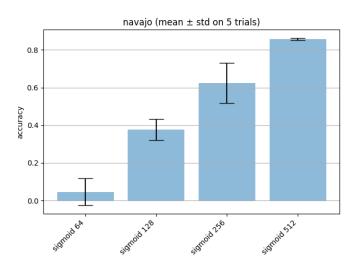


Figure: Mean accuracy (\pm standard deviation) on 5 trials for Navajo