

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

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M2 — Software Project



- 1 Reminder
- 2 Autoencoder
- 3 Results
- 4 Decoder
- 5 How to proceed?
- 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 ( $\neq$  word)
- Aim: transform these vectors into words

# What we managed to do?

- ❶ Build the autoencoder
- ❷ Applied it on the classification task
- ❸ Obtained a primary set of results
- ❹ Started analyzing what was faulty in our decoder

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# Autoencoder

## ① Encoder

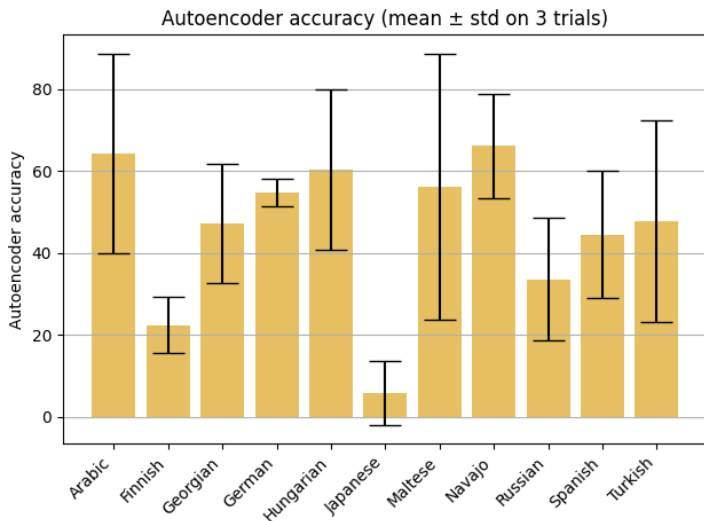
- ▶ CNN based model
- ▶ filters of sizes 2 to 6 → encode subwords ?

## ② Decoder

- ▶ 1 layer GRU
- ▶ hidden layer input = word embedding
- ▶ cells input = last decoded character's ID (start with 'BOS' ID)

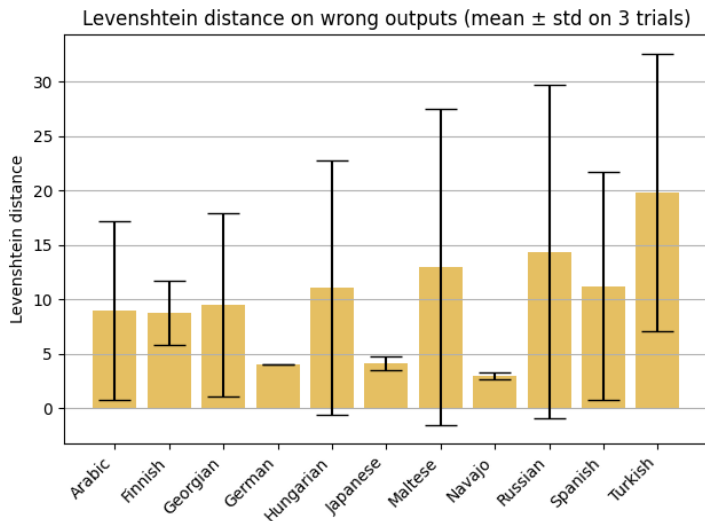
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# Autoencoder accuracy (input word = output word)

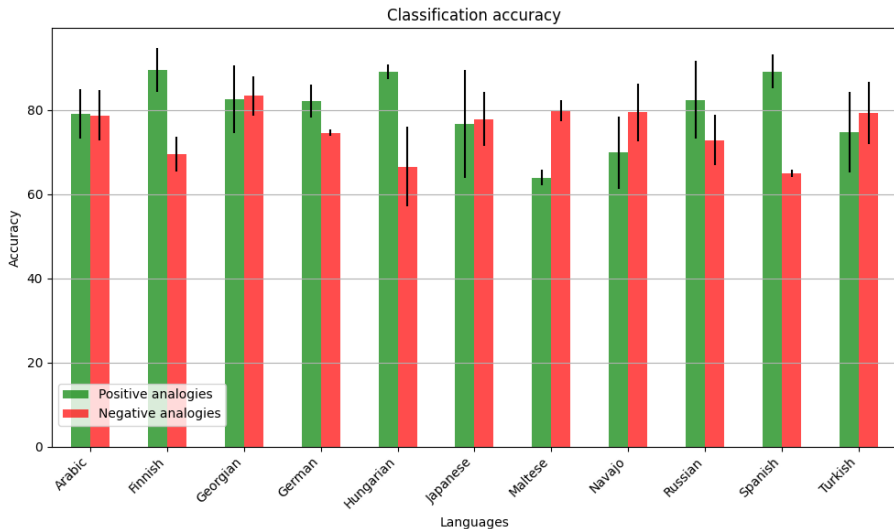




# Levenshtein distance when output word $\neq$ input word



# Accuracy for the classification task



# Discussion

Some modifications we tried:

- bidirectional: outputs empty words
- concatenate the embeddings to the input cells (= last character ID embedding + word embedding): random outputs of max\_len

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# What went wrong with the decoder

- ❶ Built files per language with only incorrect words
- ❷ Evaluated manually by fluent/native speakers
- ❸ Result: obtained some leads that would help us investigate how the model produces the words

# Comments and leads to investigate

## Summary of the comments:

- prefixes, inflections, and suffixes: exist in the languages;
- incorrect words: variability among the languages (real words: German around 65%, Russian none, Spanish 15%, Arabic 10%)

## Leads to investigate:

- how the model decodes the central part of a word
- the ratio of words that exist and don't exist per language
- force the model to work with the embeddings (seem to prioritise intern coherence)

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# How we plan to proceed

- ➊ Proceed with two languages: Arabic and German
- ➋ Query a dictionary to know the ratio of Exist/DNE words



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# Future work

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

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Thank you

Merci

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Obrigado

# Our decoder structure

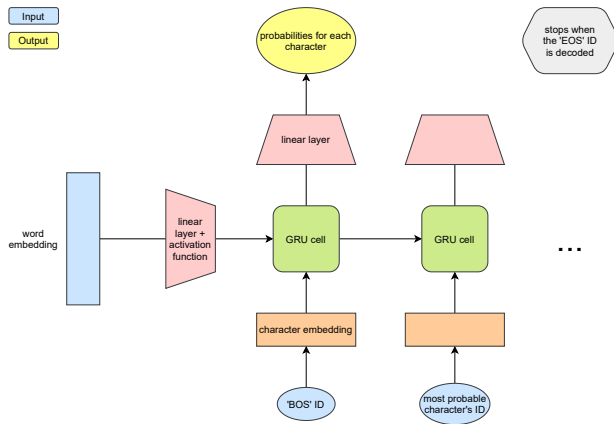


Figure: Our GRU based decoder

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