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

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



- Reminder
- 2 Our approach
- Regression task
- 4 Results & discussion
- Software
- 6 Future work

### Reminder of our aim

Regression task based on transfer

$$A: B:: C: X \xrightarrow{X=?} A: B:: C: D$$
  
e.g.  $dog: dogs:: chat: X \rightarrow chats$ 

- Input: A and B in language 1, C in language 2
- Output: D in language 2
- Same transformation for A, B and C, D

### First trial results

	hungarian, german	turkish, finnish	hungarian, finnish
Cosine similarity	58.9	39.5	18.9
Euclidean distance	57.7	39.1	16.8

Table: Accuracy for the regression task on the three (source, target) language pairs.

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## 3 approaches

- Comparable data
- Omnilingual model
- Sigmorphon 2019

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#### New results

Table: Accuracy (in %) of 3 runs of the regression model.

Language	$\begin{array}{l} \textbf{ANNr (previous)} \\ (\text{mean} \pm \text{std.}) \end{array}$	actual
Arabic	<b>77.97</b> $\pm$ 16.03	$61.13 \pm 0.83$
Finnish	$37.78 \pm 9.28$	<b>77.56</b> $\pm$ 1.78
Georgian	$94.66 \pm 1.13$	$86.40 \pm 0.62$
German	$86.38 \pm 0.45$	$86.93 \pm 0.78$
Hungarian	$53.83 \pm 3.12$	$78.98 \pm 0.50$
Maltese	$75.00 \pm 5.08$	$79.66 \pm 1.11$
Navajo	$31.74 \pm 0.90$	$45.88 \pm 0.24$
Russian	$75.15 \pm 0.44$	$70.53 \pm 0.37$
Spanish	$86.27 \pm 0.71$	$91.12 \pm 1.06$
Turkish	$61.95 \pm 10.86$	$80.34 \pm 0.79$
Japanese	<b>61.60</b> ± 1.33	<b>37.54</b> ± 37.33

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## Bilingual analogies

In our dataset: WORD<sub>1</sub> FEATURES WORD<sub>2</sub>

An analogy:  $WORD_{1,A}:WORD_{2,A}::WORD_{1,B}:WORD_{1,B}$ where  $FEATURES_A = FEATURES_B$ 

Bilingual analogies: LANGUAGE<sub>A</sub>  $\neq$  LANGUAGE<sub>B</sub>

 $\rightarrow$  keep only the subset of *shared features* 

### Shared features

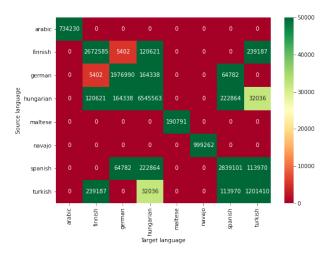


Figure: Number of possible analogies for each pair of languages

# Comparison between monolingual and bilingual results

	Finnish	German	Hungarian	Spanish	Turkish
Finnish	/	$43.96{\pm}1.48$	$80.93{\pm}1.94$	/	$82.00 \pm 1.90$
German	92.63±0.10	/	$68.17 \pm 3.12$	$68.17 \pm 3.12$	/
Hungarian	43.07±0.48	$85.92 \pm 0.83$	/	$85.92 \pm 0.83$	$40.92{\pm}2.46$
Spanish	/	$93.97 \pm 0.25$	$93.97 \pm 0.25$	/	$94.05 \pm 0.31$
Turkish	65.89±1.59	/	$71.76 \pm 0.92$	$93.18 \pm 1.90$	/

Table: Monolingual analogies: Accuracy ( $\pm$ std) on 3 runs

	Finnish	German	Hungarian	Spanish	Turkish
finnish	/	81.88	35.88	/	30.19
german	80.31	/	30.41	35.10	/
hungarian	48.83±3.19	$78.41 \pm 1.59$	/	91.62	33.93
spanish	/	17.63	83.26	/	40.63
turkish	45.81±0.17	/	16.17	70.27	/

Table: Bilingual analogies: Accuracy ( $\pm$ std) on 3 runs

# Omnilingual model

Languages which share features with at least one other language: Finnish, German, Hungarian, Turkish, Spanish

	Finnish	German	Hungarian	Spanish	Turkish
Finnish	60.30±1.26	$3.08 \pm 0.86$	$31.78 \pm 1.79$	/	$52.62 \pm 1.84$
German	$3.08\pm0.86$	$63.27 \pm 0.68$	$57.71 \pm 0.48$	$62.47{\pm}2.41$	/
Hungarian	31.78±1.79	$57.71 \pm 0.48$	$71.12 \pm 1.04$	$62.89{\pm}1.86$	$24.73 \pm 1.26$
Spanish	/	$62.47{\pm}2.41$	$62.89{\pm}1.86$	$66.82{\pm}1.34$	$62.20{\pm}6.57$
Turkish	52.62±1.84	/	$24.73 \pm 1.26$	$62.20{\pm}6.57$	$49.73 \pm 0.82$

Table: Accuracy (±std) on 5 runs

# Next time: Sigmorphon 2019 [McCarthy et al., 2019]

88 languages: 8/10 from Sigmorphon 2016 [Cotterell et al., 2016]

 $\rightarrow$  Arabic, Finnish, German, Hungarian, Russian, Spanish, Turkish, Maltese (Georgian and Navajo missing)

Aim: apply trained models to the new dataset

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### What we want

- Solving analogies: monolingual and bilingual
- Use the omnilingual model

### What it looks like

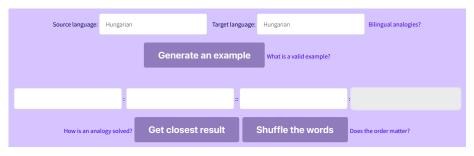


Figure: Preview of our software

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#### Future work

- Run final experiments
- Improve and adapt our software
- Continue writing the report

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### References I



Cotterell, R., Kirov, C., Sylak-Glassman, J., Yarowsky, D., Eisner, J., and Hulden, M. (2016).

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