University of Copenhagen Faculty of Science



Limitations of cross-lingual learning from image search

Mareike Hartmann and Anders Søgaard **Department of Computer Science**

Lexicon Induction From Web Search Images

Bergsma and van Durme (2011) and Kiela et al. (2015) induce lexica for nouns from web search images:

Get images associated with a word from a web search engine by using the word as a query

Rank the German candidate translations of the English word **mug** based on similarities between images:





traurig:











Find a translation for the word by ranking candidates based on similarities between the images associated with the words

Can the approach be generalized to verbs and adjectives?

Data

- 3 wordlists of English words (Simlex, MEN, Bergsma) translated into 5 languages: German, French, Russian, Italian, Spanish
- 1406 nouns, 206 verbs, 159 adjectives
- For each query word, 50 images returned by Google Search engine

Similarity Computation

Image feature representations (4069-dimensional) extracted from a

| Results | | |
|---------|--|--|
| 0.8 | | |

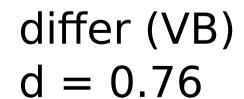
convolutional neural network pretrained on ImageNet

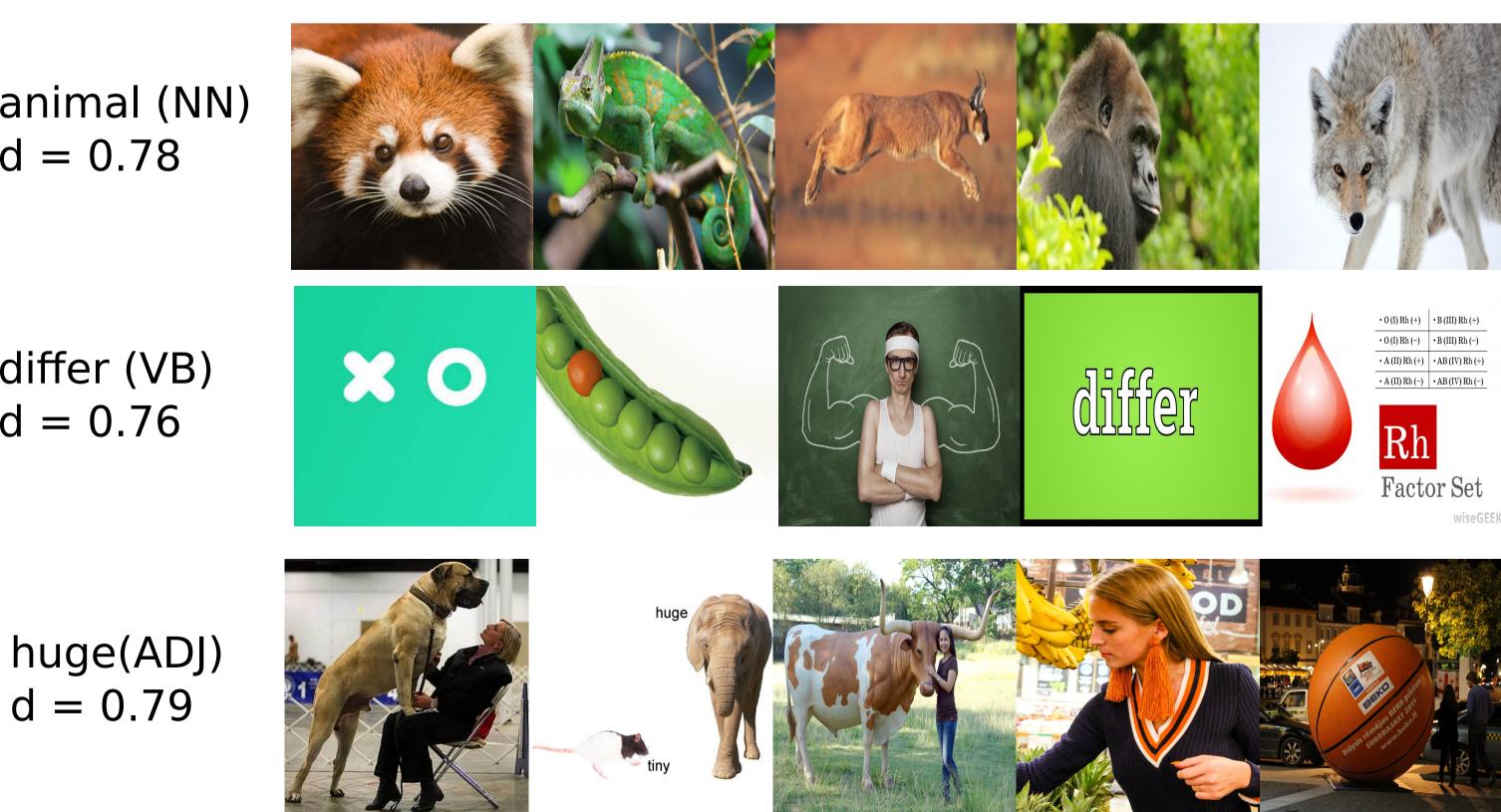
- Word translations are ranked according to: •
 - Cosine similarity between pairs of individual images
 - Cosine similarity between aggregated representations of image sets
 - Highest number of nearest neighbors

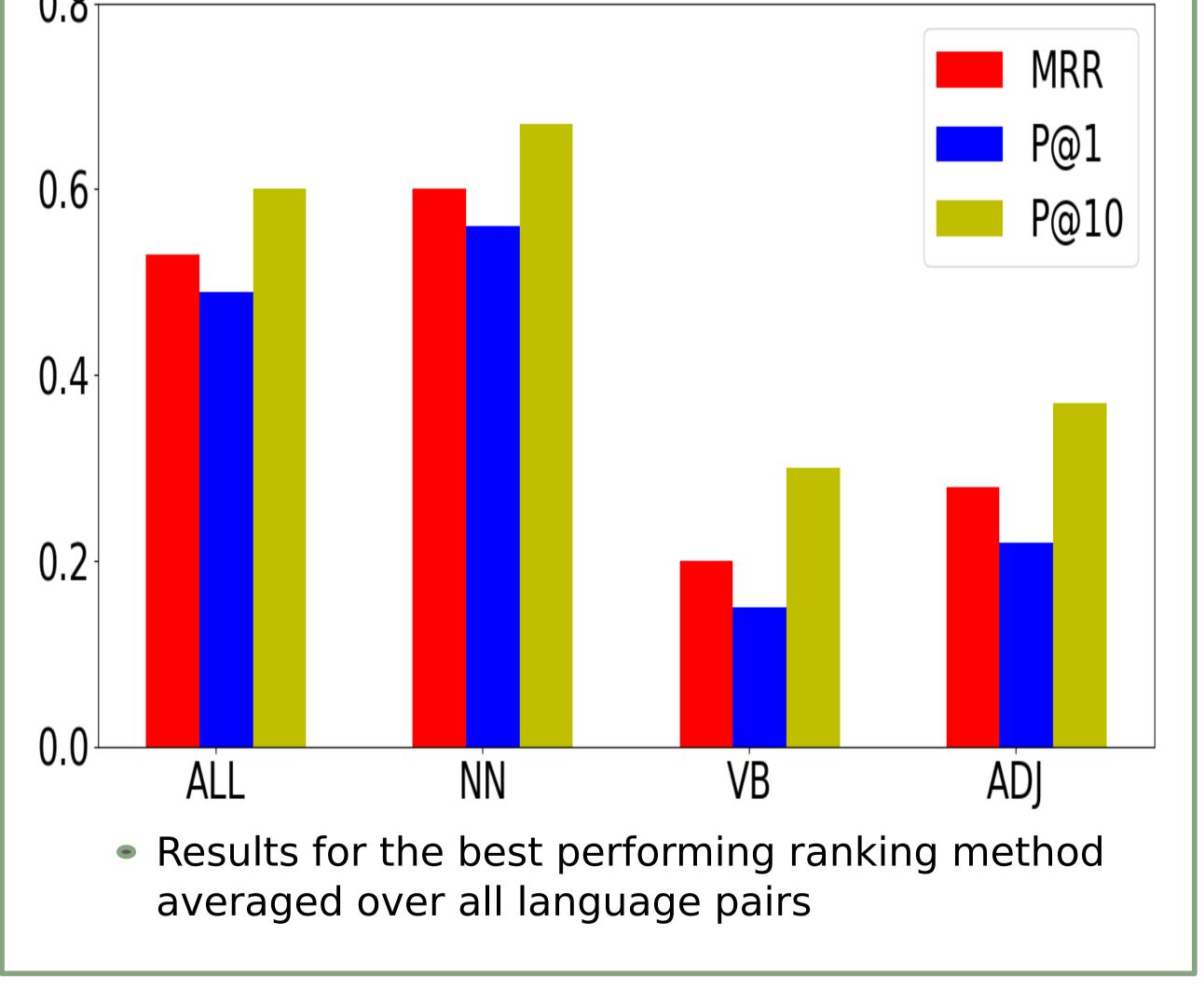
Analysis

Image dispersion is the average cosine similarity between all image pairs in an image set. Adjectives and verbs have higher dispersion values than nouns: nouns: d = 0.60, adjectives: d = 0.66, verbs: d = 0.68

animal (NN) d = 0.78







Conclusion

- Adjectives and verbs have a higher number of wordsenses according to WordNet than nouns: nouns: n = 5.08, adjectives: n = 6.88, verbs: n = 9.18
- In many cases, the search engine does not capture the intended POS of the query word:

vanish (VB) d = 0.43





- The approach does not scale to adjectives and verbs
- One explanation is that adjectives and verbs are more difficult to visualize in an iconic way (higher dispersion) values).

Ideas for improvements:

Use words in context as queries or collect images based on natural language captions rather than isolated semantic tags

Train the feature extractor on a different resource than ImageNet categories with its concrete object categories