

Developing Vector Space Models

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Developing vector space models



•Developing vector space models from scratch is not necessary (but a nice exercise to better understand them)

- •Use existing libraries (numpy, NLTK, gensim, glove-python, pytorch, TensorFlow)
 - More efficient
 - •Easy to use
 - Well documented
 - •Regular maintenance and updates

Data



word2vec: ~3 million 300-dimensional embeddings trained on part of the Google News dataset

(see https://code.google.com/archive/p/word2vec/)

GloVe: ~2.2 million 300-dimensional embeddings trained on Common Crawl (840 billion tokens)

(see https://nlp.stanford.edu/projects/glove/)

For obtaining the data, see

https://github.com/maximilianmozes/word_embeddings_ workshop_resources/blob/master/README.md

word2vec in Python



- •We use the **gensim** library
- •Available at https://radimrehurek.com/gensim/

Demo at

https://github.com/maximilianmozes/word_embeddings_worksh op_resources/blob/master/code/train_word2vec.py

GloVe in Python



- •We use the **glove-python** package
- •Available at https://github.com/maciejkula/glove-python

Demo at

https://github.com/maximilianmozes/word_embeddings_worksh op_resources/blob/master/code/train_glove.py

Visualising embeddings



- •We use the **tensorboardX** package
- •Available at https://github.com/lanpa/tensorboardX

Demo at

https://github.com/maximilianmozes/word_embeddings_worksh op_resources/blob/master/code/visualize.py

Other examples



•We have more examples of how to load pre-trained models at

https://github.com/maximilianmozes/word_embeddings_worksh op_resources/blob/master/code/examples.py

Other tips and tricks



- •Do not implement everything from scratch (use optimised packages)
- •Working with *millions* of vectors can be expensive
- Make use of matrix computations (numpy in Python)
- •Examples:
 - •Compute distance matrix for all words in vocabulary **ONCE** instead of computing distances over and over again
 - •Use existing functions to compute norms and distances
 - •scipy.spatial.distance.cosine for cosine distances
 - •numpy.linalg.norm for vector norms

GloVe in R



- •We recommend python for your work with word embeddings
- •Reasons: <u>speed</u> + libraries
- •However: you can also do it in R

- •Note: for all local approaches, memory is an issue (example: 840B, 300d model is a 5.65GB txt file)
- •For faster and more voluminous work: server or cloud

Note: tutorial for running RStudio on AWS in the workshop materials and on https://danielhammocks.uk/instructions-and-guides/ (thanks to Daniel Hammocks)

GloVe in R



- Initialising and using GloVe in R
- •Step-by-step guide and wrapper functions:
- •https://github.com/maximilianmozes/word_embeddings_workshop_resources/blob/master/code/glove_in_R.pdf (pdf and HTML notebook in the workshop materials)
- 1. Looks for pretrained models
- 2. Initialises GloVe models as DFM matrix in data.table (+ cleans up obsolete files from the R environment)
- 3. Uses any of the pre-trained GloVe models for neighbour calculation



Thank you for your attention

Any questions?

- Workshop website: https://maximilianmozes.github.io/word-embeddings-workshop/
- GitHub repo: https://github.com/maximilianmozes/word_embeddings_workshop_resources

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