92586 Computational Linguistics

Lesson 15. From document representations, towards sequences

Alberto Barrón-Cedeño

Alma Mater Studiorum-Università di Rologna

a.barron@unibo.it @_albarron_				
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Chapters 6 and 7 of Lane et al. (2019)				

Previously

► Training and loading (existing) embeddings

Doc2vec

Objective Computing a vectorial representation of a document.

Same idea as with word2vec: a NN to predict words

Input

- ► *k* context words (optional)
- ► A unique ID of the sentence/paragraph/document

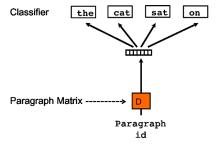
Output

- ▶ 1 target word
- ► The paragraph vector is unique among all documents
- ► The word vectors are shared among all documents
- ► The document vector is computed **on the fly**

Le and Mikolov (2014); (Lane et al., 2019, p. 215)

Doc2vec

Distributed Bag of Words version of Paragraph Vector (PV-DBOW) Similar to skip-gram



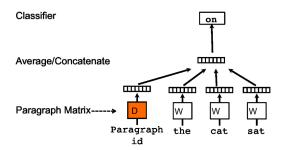
- ► Iteration: a text window and a random word from the text window are sampled, forming a classification task given the paragraph vector.
- ► No word vectors: faster + lower memory requirements

■ Let us see

Doc2vec

Distributed Memory Model of Paragraph Vectors (PV-DM)

Derived from CBOW



- ► Each column in the paragraph matrix is a vector representing one paragraph
- ► We can average or concatenate the word and paragraph vectors

References

Lane, H., C. Howard, and H. Hapkem

2019. Natural Language Processing in Action. Shelter Island,

NY: Manning Publication Co.

Le, Q. V. and T. Mikolov

2014. Distributed representations of sentences and documents.