GAČR EXPRO NEUREM³ Studying Representations

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unless otherwise stated

Deep NNs for Image Classification

It's deep if it has more than one stage of non-linear feature transformation



Caveat on Evaluation (1/2)

Consider word2vec "comprehensive" test set (Mikolov et al., 2013):

- 8.8k "semantic" and 10.6k "syntactic" questions,
- w2v "accuracy is quite good" (eyeballing)
 - The authors do mention that exact-match is "only about 60%").

Kocmi and Bojar (2016) carefully examined the test set:

- "Semantic" questions cover only 3 question types:
 - country \rightarrow city, country \rightarrow currency, masculine family member \rightarrow feminine
 - Vylomova et al. (2016) test many other relations, e.g. walk-run, dog-puppy, bark-dog, cook-eat.
- "Syntactic" questions constructed by combinations:
 - starting from only 313 distinct word pairs,
 - (leading to only 35 different pairs per question on average),
 - And of the 313 pairs, 286 are formed regularly.

Caveat on Evaluation (2/2)

	Test Set by	
Accuracy on "Synt Qs"	Mikolov et al.	Kocmi et al.
word2vec as released	62.5%	43.5%
word2vec on our data	42.5%	9.7%
SubGram on our data	42.3%	22.4%

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SubGram on our data	42.3%	22.4%
Nine rules	71.9%	66.4%

Caveat on Ultimate Evaluation

Kocmi and Bojar (2016):

- submitted to TSD on March 22, 2016.
- appeared in TSD in September 2016.
- ... cited by 4.

Bojanowski et al. (2017):

- submitted to arxiv on July 15, 2016.
- appeared in TACL 2017.

... cited by 1024.

Caveat on Ultimate Evaluation

Kocmi and Bojar (2016):

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• No code released, no fast code implemented at all.

Bojanowski et al. (2017):

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... cited by 1024.

• This is the FastText paper.

ÚFAL People in NEUREM³

- Ondřej Bojar
- Pavel Pecina
- Jindra Helcl (non-autoregressive MT, i.a.)
- Ivana Kvapilíková (unsupervised MT)
- Michal Auersperger (document representations)
- (Jindřich Libovický) (MT with images, i.a.)
- (Petra Galuščáková) (something with video?)

Expected Outcomes of NEUREM³

- Insight into what the representations look like (for ASR and NMT).
- Tools for diagnosing:
 - Which tasks are learned implicitly with the main one.
 - Why is the network making some particular types of errors.
 - Which generalizations has the network learned and which not.
- Methods for:
 - semi-supervised and unsupervised learning.
 - pre-training, reuse of model parts, combining larger models, model interfacing,
 - successful multi-task training, all esp. in the areas of ASR and NMT.

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- Methods for:
 - semi-supervised and unsupervised learning.
 - pre-training, reuse of model parts, combining larger models, model interfacing,
 - successful multi-task training, all esp. in the areas of ASR and NMT.
- Good papers, good papers, good papers...

References

Piotr Bojanowski, Edouard Grave, Armand Joulin, and Tomas Mikolov. 2017. Enriching word vectors with subword information. *Transactions of the Association for Computational Linguistics*, 5:135–146.

Tom Kocmi and Ondřej Bojar. 2016. SubGram: Extending Skip-gram Word Representation with Substrings. In Petr Sojka, Aleš Horák, Ivan Kopeček, and Karel Pala, editors, *Text, Speech, and Dialogue: 19th International Conference, TSD 2016*, number 9924 in Lecture Notes in Computer Science, pages 182–189, Cham / Heidelberg / New York / Dordrecht / London. Masaryk University, Springer International Publishing.

Tomas Mikolov, Kai Chen, Greg Corrado, and Jeffrey Dean. 2013. Efficient estimation of word representations in vector space. *CoRR*, abs/1301.3781.

Ekaterina Vylomova, Laura Rimell, Trevor Cohn, and Timothy Baldwin. 2016. Take and took, gaggle and goose, book and read: Evaluating the utility of vector differences for lexical relation learning. In *Proceedings of the 54th Annual Meeting of the Association for Computational Linguistics (Volume 1: Long Papers)*, pages 1671–1682, Berlin, Germany, August. Association for Computational Linguistics.