An Empirical Investigation of Word Representations for Parsing the Web

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Overview

- Parsing is difficult for unrestricted web text (accuracy: 90% WSJ -> 80% web).
- Word representation features obtained from large unlabeled data may combat data sparseness.
- We observed that word clusters/embeddings help most in the case of predicted part-of-speech (POS) tags.

Dependency Parsing of Web Text

- Data: Google Web TreeBank from SANCL2012, containing 5 domains (Answers, Emails, Newsgroups, Reviews, Weblogs).
- Graph-based parser with arc-factored model.

Extra word representations features are added on top of baseline features.

Brown Clustering

- Hierarchical clustering algorithm based on class-based bigram language model.
- It has been shown to improve accuracy.

An example of Brown clustering



Collobert & Weston Embedding

- Word Embedding: word represented in a dense low dimensional real value vector form, often induced from a neural language model.
- We constructed features in following 2 ways;

2D visualization of word embeddings Australi**an** nerican British European Russian Japanese Germärench EU international central state domestic foreign local * Figure from Joseph Turian.

- I. <u>Convert an embedding into a bit-string</u>.
 - For each real-value in vector, we give a bit I if the value is positive and else give bit 0. Then we concatenate all bits.

* Figure from Koo et al. 2006. "Simple Semi-supervised Dependency Parsing".

We used short bit-string prefixes of the hierarchy, • combined with word forms or POS tags, as features.

2. <u>Cluster word embeddings</u>.

• We used repeated bisection algorithm to cluster embeddings, then use acquired cluster IDs as features, similar to Brown clustering.



Discussion

- Extra features improved results on experiments with predicted POS tag data sets, but not with gold POS tag data sets.
- Brown clustering features <u>outperforms</u> word embedding features.



Induce word embeddings on in-domain data sets.

Try different ways to construct features.