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A ColWordNet API





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A CoolWordNet API*







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Outline

- Motivation
- ColWordNet
- ► The API

Motivation

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WordNet as a lexical resource for language learning and AI

Motivation: WordNet is a useful resource in NLP

- "The list of papers citing WordNet seems endless" (Hovy, Navigli and Ponzetto, AI 2013)
- It is a useful lexical resource for many tasks at different spectrums of LTs.
- One area where there is clear room for improvement is on *lexical combinations* of words: collocations.

WordNet could be extended with collocational information

- In Espinosa-Anke et al. (Coling 2016) we describe and evaluate ColWordNet (CWN).
 - Previous work on collocation acquisition focuses on compiling collocation lists (Church and Hanks 1989, Kilgarriff 2006) - No semantic classification.
 - We tackled fine-grained collocation classification, drawing upon the lexical relation between the base and the collocate:
 - 'perform': take [an] exam, make [a] decision, pose [a] question
 - 'put an end': solve [a] problem, break the silence
 - Linguistic motivation based on the Meaning Text Theory (Melćuk, 1987)

00 0 Σ CWN: HOW IT W

A brief description of our method for creating CWN













Bases (lacobacci et al., 2015)



Mapping train data sense-level embeddings models Collocates (Mancini et al., 2016)

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Train a transformation matrix between bases and collocates

Run the learned transformation to encode collocation relations between synsets



Mapping train data sense-level embeddings models (vectors)

Train data manual compilation and disambiguation Run the learned transformation to encode collocation relations between synsets

Train a

transformatio

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The API

A preliminary approach to enabling CWN querying

It generates a reliable mapping

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It works equally well for all LFs

Dependent on the *amount* and *quality* of training data. New issues related to semantics that we cannot answer yet.

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Multilingual potential

Since we used BabelNet as pivot, all assets provided by BabelNet can be explicitly leveraged.

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It is fast

It's actually not very fast. We've experimented with additional transformation appraches, but the difference in speed and performance has not been evaluated yet.

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Its algorithmic nature makes it possible to set specific thresholds for each lexical function.

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It is finished

This is a very preliminary prototype, so by no means this is a finished project.

Thank you

Any questions?

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