Improving Semantic Parsing Using Statistical Word Sense Disambiguation Ritwik Bose, Siddharth Vashishtha & James Allen

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Objective

Improve sense decisions made by logical semantic parsing using the outputs of a statistical WSD system while honoring semantic role restrictions.

Hints can be provided to the parser by curating inputs to the chart or by actively reordering constituents as they are built.

• **Pre-Hinting:** The chart is populated with the top candidates from SupWSD





Strategies

We consider the case study of improving a semantic parser, the TRIPS Parser [1], which is a best-first bottom-up chart-parser with a hand-built, lexicalized context-free grammar. It consists a TRIPS ontology which is a hand-crafted single-inheritance hierarchy where nodes specialize or override hierarchical features and argument templates from their parent. We improve the sense disambiguation of this logical semantic parser by integrating advice from a statistical Word Sense Disambiguation (WSD) system, SupWSD [3].

TRIPS Ontology and Lexicon



Figure 1: Entry for ONT : : REACH in the TRIPS Ontology

SupWSD

• Progressive Hinting: Constituents are reordered to prefer the best outputs from SupWSD
• Combined Hinting: All of the above



Figure 3: Progressive hinting strategy

Results

We report three metrics to evaluate the performance of the semantic parser – (i) accuracy (exact sense agreement) (ii) Mean Wu-Palmer similarity (Wu and Palmer 1994) and (iii) Mean accuracy over semantic factors (computed as Wu-Palmer over the factorized ontology).

Metric SupWSD Plain Pre Prog Comb

We use a supervised-system SupWSD [3] which provides a **probability distribution** over **WordNet senses** for a given sentence. The system uses an **SVM classifier** to predict the word senses and its **features** include various linguistics properties such as POS tags, syntactic relations, local collocation, word-embeddings, and information about surrounding words. We use the version which is trained on **SEMCOR**.

Hinting



Accuracy	66.19	39.42 50.33 42.45 53.19
WuP	84.22	73.97 79.12 75.75 80.53
Sem-fac	70.26	49.33 60.76 53.38 63.37

Table 1: Results from evaluation on SemEval2013 [2].

 Pre-hinting produces the greatest individual benefit Missing sense mappings are a problem.

 ● Plain → Progressive has almost the same improvement as Pre → Combined Missing senses and mis-ranked senses are likely separate issues

Future Work

• Universal Decompositional Semantics provides super-sense vectors

• Abstract Meaning Representation/Unscoped Logical Forms can provide sense bracketing and argument structures as further restrictions on chart entries.

• Discourse analysis can produce better global sense decisions

References

[1] James F Allen, Mary Swift, and Will De Beaumont. Deep semantic analysis of text. In *Proceedings of the 2008 Conference on Semantics in Text Processing*, pages 343–354. Association for Computational Linguistics, 2008.

[2] Roberto Navigli, David Jurgens, and Daniele Vannella. Semeval-2013 task 12: Multilingual word

Figure 2: Hinting Pipeline: A hint for the word **deals** is produced by SupWSD, generating a wordnet sense. The WordNet sense is mapped to a TRIPS type and marked up with a syntax template and features and added to the chart.

sense disambiguation. In Second Joint Conference on Lexical and Computational Semantics (* SEM), Volume 2: Proceedings of the Seventh International Workshop on Semantic Evaluation (SemEval 2013), volume 2, pages 222–231, 2013.

[3] Simone Papandrea, Alessandro Raganato, and Claudio Delli Bovi. Supwsd: A flexible toolkit for supervised word sense disambiguation. In *Proceedings of the 2017 Conference on Empirical Methods in Natural Language Processing: System Demonstrations*, pages 103–108, 2017.