

The Integration of Syntax and Semantic Plausibility in a Wide-Coverage Model of Human Sentence Processing



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Abstract

We present the SynSem-Integration model of difficulty in human sentence processing. It integrates a probabilistic wide-coverage grammar-based model with a separate model for verb-argument thematic role assignment and thematic fit prediction which accounts for semantic plausibility effects.

Motivation

Important properties of the human sentence processor are

- Sensitivity to prior linguistic experience
- Immediate incremental interpretation
- Robust and accurate processing of unseen input (wide coverage)
- Influence of semantic plausibility

Different existing models account for different properties; none covers all.

The SynSem-Integration Model

The Syntax-Semantics Integration model combines a probabilistic model of syntactic processing with a general model of semantic plausibility. Both models have wide coverage and are automatically induced.

Both models rank the proposed syntactic analyses:

The syntax model by syntactic probability estimates, the semantics model by plausibility estimates for the verb-argument pairs in each structure.

A globally preferred analysis is determined by interpolating the two models' predictions. This analysis is assumed to be adopted by readers.

Predicting Difficulty

We associate difficulty in human sentence processing with two events:

- A conflict in syntactic and semantic preferences for the highest-
- ranked structure, e.g., during an ambiguous region
- A revision in the interpretation of the globally preferred analysis, e.g., at disambiguation

Conflict Revision The patient cured by the therapy had invented it himself.

Evaluation

The SynSem-Integration model's predictions were tested against patterns of processing difficulty found for the main clause/reduced relative (MC/RR), NP/S, NP/0 and PP attachment phenomena (two studies per phenomenon).

The model significantly predicts the observed patterns of human difficulty, while a syntax-only baseline (equivalent to a lexicalized grammar-based model) fails.

The Semantic Model

The semantic model approximates world knowledge by exploiting the link between plausibility and word co-occurrence in a corpus annotated with thematic roles. Padó et al., 2006

Given an arbitrary verb-argument pair, it predicts a preferred role relation and its plausibility. Plausibility is equated to probability of encountering the pair in the respective relation in the FrameNet corpus.

Smoothing sparse data to achieve wide coverage:

- Semantic generalization: Pooling observations of words from the same semantic class

- Nouns: WordNet synsets
- Verbs: classes automatically induced from FrameNet
- Re-estimation smoothing: Good-Turing smoothing

The predictions of this widecoverage semantic model are significantly correlated to human plausibility judgments

Data Set	N	Spearman's ρ
McRae et al., 98	64	0.415, **
Pado et al., 06	414	0.522, ***



- Rely on frequency information induced from large corpora

Constraint-Based Models McRae et al., 1998; Narayanan and Jurafsky, 2002

- Combine information from different sources

Grammar-Based Models Crocker and Brants, 2000; Levy, 2005

- Constraints support pre-defined analyses, most active one wins
- Do not show broad coverage (due to hand-selection of constraints)



Comparison to Existing Models

- Shared with constraint-based architectures:
 - Combination of preferences from different sources
 - Prediction of difficulty if preferences conflict
- But: Wide-coverage model, no need to hand-select constraints Shared with grammar-based architectures:
- Probabilistic ranking of generated analyses
 - Models automatically induced from large corpora
 - But: Integration of plausibility

Phenomenon	Ν	Model	Spearman's ρ			
A11	26	SynSem	0.700,***		MC/ BB	McRae et al., JML 98 MacDonald, I CP 94
	50	Baseline	-0.223,ns		NP/	Garnsey et al., JML 97 Pickering & Traxler, JEP:LMC 98 Pickering & Traxler, JEP:LMC 98 Pickering et al., JML 00
	44	SynSem	0.792,***	\mathbb{N}	S NP/	
MC/nn	14	Baseline	0.199,ns	$\left \right\rangle$	0	
NP/S	12	SynSem	0.688, *	$ \rangle$	PP	Rayner et al., J.V Learn V Beh 83 Taraban & McClelland, JML 88
		Baseline	-0.165,ns			

The Syntax Model

The syntax model incrementally constructs structural analyses of the input and ranks them by their syntactic probability.

We use an incremental top-down parser (Roark 2001). Its grammar and lexicon are derived from the syntactically-annotated Penn TreeBank corpus.

Performance on the standard parser test set shows that the model assigns accurate analyses to unseen input.

1001001	ooverage	F-Score
PTB 23	100%	86.29

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