ACS Syntax and Semantics of Natural Language

Lecture 7: CCG Supertagging



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- The first stage in the CCG parsing pipeline is to assign CCG lexical categories to the words in the sentence
- This process is known as *supertagging*, since the labels being assigned are detailed syntactic structures
- Srinivas and Joshi (1998) also called this process *almost parsing*, since the detailed labels mean that, once the supertagging has been performed, there is less work for the parser to do
 - Srinivas and Joshi used Lexicalised Tree Adjoining Grammar (LTAG), in which the 'labels' are LTAG elementary trees, but the principle is the same

- We can apply any sequence labelling method to the supertagging task, e.g. HMMs, CRFs, maximum entropy models,...
- But the task is much harder than PTB pos tagging, since the label set is typically an order of magnitude larger
- We will use a maximum entropy tagger (Ratnaparkhi, 1996), because of its ability to incorporate a large amount of contextual information for disambiguation

He	goes	on	the	road	with	his	piano
NP	$(\overline{S[dcl] \backslash NP)/PP}$	$\overline{PP/NP}$	NP/N	N	$(\overline{(S \backslash NP) \backslash (S \backslash NP))/NP}$	NP/N	N

A	bitter	conflict	with	global	implications
$\overline{NP/N}$	N/N	N	$(NP \setminus NP)/NP$	N/N	N

- Categories in blue are all for prepositions
- Need to distinguish between complements and adjuncts, as well as make attachment decisions

- PP attachment is known to be one of the hardest parsing sub-problems

- Over 500 labels in the grammar used by the C&C parser, compared with around 50 PTB pos tags
- Useful baseline: for each word in the test set, assign the label most frequently seen with that word in the training data (and for unknown words assign N, say)
- For PTB pos tagging, this baseline is around 90%
- For CCG supertagging, this baseline is around 72%

- Assigning a single category to each word (using Viterbi) results in around 92% per-word accuracy (using a standard MaxEnt tagger with a 5-word window, with words and pos tags as features)
 - around 2-3 mistakes per sentence!
- 92% is not accurate enough for reliable parsing
- We need to allow the supertagger to assign more than one category when the context cannot reliably disambiguate

 $P(\texttt{category}|\texttt{sentence}) = \sum\limits_{s} P(\texttt{category}, S|\texttt{sentence})$

where S ranges over all lexical category sequences for the sentence

- The Forward-Backward Algorithm is a DP algorithm for efficiently performing this sum
- Assign all categories to a word whose probability is greater than some dynamic threshold:

assign category C if $P(C|\text{sentence}) > \beta$. $P(C_{\text{max}}|\text{sentence})$

where Cmax is the category with the highest probability for that word

- The number of categories assigned to each word on average is a crucial factor in the speed of the parser
- The following "adaptive" strategy has been found to work very well:
 - Start with a high β value/low ambiguity
 - If the parser fails to find an analysis, decrease β
 - Repeat until spanning analysis is found (success) or parsing is taking too long (failure)
- See Section 10.3 of Clark and Curran (2007) for experimental details of the adaptive strategy

- Bangalore and Joshi. Supertagging: an approach to almost parsing. Computational Linguistics, 22:1-29, 1998
- Adwait Ratnaparkhi. A Maximum Entropy Model for Part-Of-Speech Tagging. In Proceedings of the Empirical Methods in Natural Language Processing Conference (EMNLP), 1996
- Stephen Clark and James R. Curran. Wide-Coverage Efficient Statistical Parsing with CCG and Log-Linear Models. Computational Linguistics, 33(4), 2007
- Stephen Clark and James R. Curran. The Importance of Supertagging for Wide-Coverage CCG Parsing. Proceedings of the 20th International Conference on Computational Linguistics (COLING-04), Geneva, Switzerland, 2004