

On word disambiguation features

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November 29, 2015

1 Integrating Multiple Knowledge Sources to Disambiguate Word Sense: An Exemplar-Based Approach

<http://arxiv.org/pdf/cmp-lg/9606032.pdf>

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Problem: Word Sense Disambiguation (WSD).

Input: set of unrestricted, real-world English sentences. WARNING: the disambiguation is therefore internal to each sentence. Those sentences are pre-tagged with parts-of-speech (POS) tags (N , V , Adj , Adv), using existing software. Also: a list of ambiguous words mapped to their different meanings, those meanings being referenced to with numbers.

Output: each ambiguous word w tagged with its correct sense number i (context-dependent).

1.1 Feature extraction: $L_3, L_2, L_1, R_1, R_2, R_3, M, K_1, \dots, K_m, C_1, \dots, C_9, V$

L_i : POS of the word i -th position to the left of w in the sentence s .

R_i : POS of the word i -th position to the right of w in s .

M : morphological form of w in s (if w is a noun (N), then sing./plur.; if w is a verb (V), then prst. 3^d pers. sing./past participle/etc).

K_1, \dots, K_m : features corresponding to a set of keywords that frequently co-occur with w in the same sentence. This set is determined by conditional probabilities (see below). For a sentence s , the value of feature K_i is 1 if K_i appears in s , 0 otherwise.

V : verb-object syntactic relation (useful in disambiguation of nouns).

2 Some observations

Other features we might take into consideration: types of prepositions in the sentence (place, time, ...).