Analyzing German Noun Compounds using a Web-Scale Dataset – Final presentation UIMA Software Project WS 2010/2011



Outline



- Motivation
- Problem definition
- Splitting algorithm
- ▶ Ranking algorithm
- ► Lessons learned / Conclusion

Motivation



Let's start with an experiment

- ► Take a search engine
- Search for: Blumensträuße (flower bouquet)
- As result you will receive documents with the word Blumensträuße
- ► You will also see document with the words Blumen (flower) and Sträuße

Result

► The search engine is intelligent and knows that *Blumensträuße* is a noun-compound

Problem definition (1) - Noun Compounding



Definition

A noun compound word is the combination of one or more individual words to a new word

Example

Blumensträuße -> Blumen + Sträuße

- Compounds are formed with nouns, verbs and adjectives.
- Compound words can be compound with other
- ▶ Linking morphemes are added between words: *Tag(es)+ration*
- ▶ Different context for different splits: *Tag(es)+ration* vs. *Tag(es)+rat+ion*

Problem definition (2) - Noun Decompounding



How to split a word? [ea08, Alfonseca et al.]

- 1. Calculate every possible way of splitting a word in one or more parts
- 2. Score those parts according to some weighting function
- Take the highest-scoring decomposition. If it contains one part, it means that the word in not a compound.
- ► The algorithm for task 1 and task 2 two are nearly independent
- But results of task 2 can never be better the result of task 1
- Task 3 is the combination of task 1 and task 2

Splitting Algorithm (1) - Requirements



Problem 1

How do we know when a word starts or ends?

Solution

- Use a dictionary
- The following algorithm use the IGerman98 dictionary. This is part of most spell-checkers today.

Problem 2

How to evaluate?

Solution

- ▶ Use a Marek' corpus [Mar06, Marek] (around 160,000 examples)
- ▶ If the correct split in the list of all possible splits we have a correct result

Splitting Algorithm (2) - Left to right algorithm



- ▶ Walk from left to right through the word an check if left part is a correct word
- Right part can be an unword
- Result of real implementation returns a tree -> can be visualized

```
function split(word)
result = List()
for (i = 0..word.len)
  left = word[0..i+1]
  right = word[i+1..word.len]

  if (Dictionary.contains(left)
      and (right.len > 2 or right.len == 0))
    result += (left, right)
```

Splitting Algorithm (3) - Data driven algorithm



- Uses "statistics" of the dictionary
- With statistics I mean to put the dictionary in a trie.
- Example extracted from [ea00, Martha Larson et al.]

f	r	i	е	d	е	n	S	р	0	ı	i	t	i	k
-	-	39	29	29	25	24	23	3	1	1	1	1	1	1
1	1	1	1	1	2	7	37	88	89	89	92	99	-	-

ſ	f	r	i	е	d	е	n	s	р	0	ı	i	t	i	k
[·	-	-	-	10	0	4	1	1	20	2	0	0	0	0	0
Г	0	0	0	0	0	5	30	51	1	0	3	7	-	-	-

f	r	i	е	d	е	n	S	р	0	I	i	t	i	k
-	-	-	-		*			*						
							*				-	-	-	-

Splitting Algorithm (4) - Evaluation



Algorithm	Correct with morphemes	Correct without morphemes
Left to right	0.813	0.888
Data driven	0.16	0.41

Legend

- Correct without morphemes means that only the splits are at the correct position, but the morphemes are not set correctly
- Correct with morphemes means that the splits are at the correct position and also the morphemes are set correctly

Result

- Left to right algorithm works a lot better then the other.
- ► Following always use the left to right algorithm

Ranking algorithm (1) - Requirements



Problem 1

To rank we need knowledge about the words

Solution

The Google Web1T corpus has frequency information about a lot of n-grams.

Problem 2

How to search the n-grams?

Solution

A lucene index was created. It has a size of 12.8 GB and is very slow on a normal hard disk

Ranking algorithm (2) - Frequency based [ea08, Alfonseca et al.]



$$F_s = \prod_{s_i \in S} freq(s_i))^{\frac{1}{|S|}} \tag{1}$$

- S: The split with split elements in it
- freq: Searches for n-grams with the given words and returns the frequency value from the corpus.
- For each split F_s is calculated.
- ▶ The split with the highest F_s wins

Ranking algorithm (3) - Probability based [ea08, Alfonseca et al.]



$$P_{s} = \sum_{s_{i} \in S} -log(\frac{freq(s_{i})}{F})$$
 (2)

- S: The split with split elements in it
- freq: Searches for n-grams with the given words and returns the frequency value from the corpus.
- F: The total amount of frequency values in the corpus (add all values)
- For each split P_s is calculated.
- ▶ The split with the lowest P_s wins

Ranking algorithm (4) - Mutual information based [ea08, Alfonseca et al.]



$$M(w_1, w_2) = log_2(\frac{F \times freq(w_1, w_2)}{freq(w_1) \times freq(w_2)})$$
(3)

- S: The split with split elements in it
- freq: Searches for n-grams with the given words and returns the frequency value from the corpus.
- F: The total amount of frequency values in the corpus (add all values)
- M will be calculated for neighbor pairs in the split
- The value of a split is the average of all M
- ► The split with the highest averaged *M* wins

Ranking algorithm (5) - Evaluation



Algorithm	Correct tree	Correct@1	Correct@2	Correct@3
Frequency	0.523 (0.720)	0.508 (0.703)	0.665 (0.785)	0.726 (0.829)
Probability	0.182 (0.252)	0.184 (0.256)	0.510 (0.658)	0.628 (0.743)
MI	0.295 (0.442)	0.369 (0.542)	0.516 (0.666)	0.587 (0.737)

- Correct tree: Correct result when ranking on a tree (subset of a list)
- ► Correct@1: The first of the ranked list is correct
- Correct@1: The first or second of the ranked list is correct
- Correct@1: The first, second or third of the ranked list is correct
- Values in brackets are correct without morphemes

Lessons Learned / Conclusion



- Lucene indexes on normal disks are very SLOW
- I need a better machine ;)
- Weekly documentation help to write final report, and to think about next steps
- Simplest algorithm returned best results

End



Questions

Ask now, or later.

More information

Code, documentation and slides are available on github:

https://github.com/jenshaase/noun-decompounds

References





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