A High Quality Partial Parser for Annotating German Text Corpora

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Abstract

In this paper, a two-stage partial parser for untagged German sentences is presented. In the first stage, the sentence is segmented into better parsable units according to the Topological Field Model. In the second stage, minimal phrases of NPs, DPs and PPs as well as nominal multiword units are identified in each of the recognized fields. In this paper, we discuss the results of the second stage. We evaluated 500 parsed sentences of a newspaper corpus. The achieved recall and precision rates are better than the ones of comparable systems as reported in literature so far.

1. Introduction

Because ambiguity is still one of the central problems for full parsing, partial parsing is used for many NLP tasks. For the task of information extraction, it is easier to extract domain and scenario specific information with a partial parser instead of determining the correct syntactic analysis from a huge set of ambiguous analyses.

Furthermore, partial parses are of great value for corpus-based computational lexicography. For the extraction of subcategorization frames of content words (e.g., verbs) and the identification of light verb constructions, corpora annotated with partial parses are very fruitful. Partial parses are also helpful for syntactic grammar refinement, for disambiguation tasks like PP-attachment (Hindle and Rooth, 1993), semantic clustering (Riloff and Shepherd, 1997) as well as an input source for building fully parsed treebanks (Skut et al., 1998).

In (Klatt, 1997), we suggested a strategy for parsing German sentences consisting of three stages. The partial parser described here is a slight modification of the first two stages. In the first stage, a sentence is segmented according to the Topological Field Model for German (cf. (Drach, 1937), (Höhle, 1986)), shortly TFM. In the second stage, so-called minimal phrases are recognized in each recognized topological field. In the third stage, a fully parsed structure is assigned – first on the field level, then on the sentence level. But here, we’re confronted with the problem of ambiguity – one of the major problems in parsing – where disambiguation strategies using treebank information seems to be the best solution. Unfortunately, the size of the existing treebanks for German (e.g., the TIGER project (Brants et al., 2002)) is too small to be applicable for such a task.

This problem is one motivation (beneath the other applications mentioned before) for the construction of our partial parser, that is based on the analysis technique Pattern-Matching Easy-First Planning, shortly PEP (Klatt, 1997). In opposite to mainstream techniques a sentence is not strictly processed from left to right. Instead we prefer an easy-first strategy, doing the easier decisions before the harder ones, as described in (Abney, 1996).

In the next two chapters, we introduce the structures we want to recognize and illustrate how this could be done with PEP. In the fourth chapter, we describe the recognition process for finding minimal phrases in the so-far received segmented fields of the first parsing stage.1 In the fifth chapter, we present the evaluation of the identified structures, before we show in the sixth chapter some worthwhile applications and extensions.

2. Structures to be recognized

2.1. Topological Fields

For the segmentation of a sentence into its topological fields, we make use of an extension of the TFM by Rehbein (cf. (Rehbein, 1992)). The extended TFM splits up a sentence into seven fields. A so-called sentence bracket (SK) consisting of a left (LK) and a right part (RK) segments a sentence into a top, middle and bottom field (in German ’Vorfeld’ (VF), ’Mittelfeld’ (MF) and ’Nachfeld’ (NF)). Giving coordinations and punctuations a home, Rehbein extends this model by the fields ’Satzanfangsrahmen’ (SAR) and ’Satzenderahmen’ (SER).

Because we have to process real-life sentences, we extend the TFM by some more fields. E.g., SKEL marks a sentence bracket, where all verbs were elided. Furthermore, we use different SK-annotations with respect to their clausal subtype: SKI marks a verb-first- or a part of a verb-second-clause (cf. (4))2.

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1 A detailed description and evaluation of this parsing stage is beyond the scope of this paper and will be given in a own publication.

2 Note that empty fields are omitted in some of the examples.
2.2. Intra-clausal minimal phrases

In the second parsing stage, we determine so-called minimal phrases of selected categorial heads (nouns, determiners, prepositions) of each recognized VF, MF and NP as well as phrases with an adjectival head modified by special adverbials. A minimal phrase could be considered as a structured chunk comparable to the chunk definition in (Kermes, 2003). The traditional notion of a chunk is that of a flat, non-recursive structure (Abney, 1991) (cf. (6)). Kermes extends this definition by two aspects (i) recursive embedding and (ii) post-head embedding (cf. (7)).

(6) [NP das erste Tor] [NP des Turniers] the first goal of the tournament

(7) [NP [NP das [AP erste] Tor] [NP des Turniers]] the first goal of the tournament

In (8), we see the minimal phrases, we assign to this text. First, we consider a determiner as a governor of a NP according to the DP hypothesis (Abney, 1987)\(^3\). Second, we make no post-head embedding, since such a decision is dependent from the lexical verb (cf. (9) and (10)).

(8) das erste Tor des Turniers

(9) Ich habe [[das Auto] [meiner Frau]] gefahren. I have driven the car of my wife

(10) Ich habe [das Auto] [meiner Frau] geschenkt. I have donated the car to my wife

Third, for simple constructions like the two DPs in (8), we assign a full parse to each of them. For complex constructions like the one in (11), we assign a so-called closure, a not fully and also – strictly speaking – ill-formed structure. But this can be easily corrected later (e.g. if this structure is recognized as the only part of a VF) by two tree operations (cf. (12)).

(11) das in Portugal stattfindende Turnier

The tournament taking place in Portugal

(12) das in Portugal stattfindende Turnier

We define a minimal phrase as a parse tree reaching from the head of a phrase to its governed element that (i) leaves out post-head embedding and (ii) models recursive embedding of the same categorial head by a closure. These are the structures, we want to annotate in the second parser stage – as well as nominal multiword units (N-MWUs) as shown in (13).  

(13) [DP die Düssidten] [N-MWU Aung San Suu Kyi] the dissident Aung San Suu Kyi

\(^3\)Syntactic heads are printed by thick lines.

3. The Analysis Technique PEP

Pattern-Matching Easy-First Planning (Klatt, 1997), shortly PEP, is an analysis technique that can be used for several analysis tasks. We built a tokenizer (Klatt and Bohnet, 2004), a POS-tagger (Klatt, 2002) and this partial parser for German which all outperform existing systems. The analysis process of PEP is driven by traversing a transition network deterministically. A transition network is defined by several states that are linked by directed arcs. Every arc is associated with a so-called network function (NWF). PEP has a lot of NWFs, the most important ones are one corresponding pattern for finding adjacent elements (the NWF seq, cf. (14)) and one corresponding pattern for finding non-adjacent patterns with a left and right border element (the NWF seq, cf. (15)).

(14) Lisbon is a
d_e_s_t_a_i_n_D
_n_i_c_e
_t_w_o_n_i_n

PEP allows a bi-directional search for patterns from every token position inside the input text. There exist three different ways for searching a pattern: a categorial-driven search, a positional-driven search and a token-driven search. (17) is an example of a categorial-driven search pattern. Here, all adjacent determiners and nouns in (16) are detected and merged to a NP constituent (see (18)). (19) is an example of a positional-driven search pattern. Assuming that the word position pointer *s-top* is positioned at the beginning of the first word in (16), only the first determiner-noun-pair is detected (see (20)). In (19) the left context, which is associated with *s-top* (:LC *s-top*), is chosen as the anchor point of the pattern (:AP LC). For the token-driven search as well as additional features of PEP see (Klatt, 1997).

(16) The ball hit the bar.

(17) (nb ((m-cat DET)) ((m-cat N)) :match-as ((m-cat NP)))

(18) det The n ball hit det the n bar

(19) (nb ((m-cat DET)) ((m-cat N)) :AP LC :LC *s-top* :match-as ((m-cat NP)))

(20) det The n ball hit det the n bar

4. Minimal phrase recognition

In this chapter, we describe the recognition of minimal phrases in the second parsing stage in more detail. Usually every NP, DP and most PPs consist of a right-peripheral noun in German. We make use of this property by choosing each noun inside a VF, MF or NF as the starting point for the recognition of the minimal phrases. We process a field from left to right, go to the next noun (with a seq-pattern) and apply the following stages to it.

4.1. Recognizing fully parsed subtrees

Determining the left adjacent element of the noun (with a nb-pattern), we build a binary branching tree\(^4\), if the left

\(^4\)The merged structures are printed in boldface.
neighbour is a proper determiner, preposition or attributive adjective (cf. (21)). If the left neighbour is an uppercase written word, but no known first name, we apply a corpus-based test to recognize N-MWUs in the surrounding context. This test identifies Aung San Suu Kyi in (22) as a N-MWU. After that, we iterate the process for the newly built structure and its left neighbour (cf. (23) and (24)).

(21) mit dem [NP ersten Friedensnobelpreis] with the first Nobel Peace Prize
(22) die Dissidenten [N-MWU Aung San Suu Kyi] the dissident Aung San Suu Kyi
(23) mit [DP dem [NP ersten Friedensnobelpreis]]
(24) [PP mit [DP dem [NP ersten Friedensnobelpreis]]]

This substage also leads to a partial recognition of closures (cf. (25)) that will be continued in the next substage.

(25) mit der [PP in Birma] [NP lebenden Frau] with the woman living in Birma

4.2. Recognizing closures of subtrees

In this substage, we process all topological fields in a second run by identifying the first motherless NP with a left adjacent attributive adjective (ADJA-N-pair) from left to right. Next, we identify the nearest proper non-adjacent possible governor (determiner or preposition) to its left. If the governor is motherless, we mark these two elements as the border elements and continue with the strategy of the substage described before to merge left adjacent elements with the closure (cf. (26)).

(26) [PP mit [DP der [PP in Birma] [NP lebenden Frau]]]

After that, we start an iteration of this process with the next right ADJA-N-pair. If the governor stands in a non-motherless relation (cf. (27)-(29)), we build a closure, too. In this case, we assign a lower confidence factor to it preferring the alternative analysis. Sometimes, the higher ranked analysis is correctly eliminated by the ongoing analysis process, e.g. by identifying the closure in a VF-position (cf. (29)) or in respect to the subcat frame constraints of the lexical verb.

(27) dass er [DP die Blumen] [NP singenden Frauen] gab that he gave the flowers to the singing women
(28) dass er [DP die Lieder] [NP singenden Frauen] traf that he met the women singing songs
(29) [DP Die Lieder [NP singenden Frauen]] tanzten The women singing songs were dancing

4.3. Picking up the rest

At last, we’re picking up the rest, e.g. pronouns with left-peripheral prepositions (cf. (30)), NPs with a prenominal genitiv (cf. (31)), first names that followed by an uppercase written word. In the latter case, we don’t prefer one of the two assigned analyses, since both could be correct (cf. (32) and (33)).

(30) Er wartet [PP auf sie]. He’s waiting for her.
(31) [DP Birmas [NP erste Friedensnobelpreisträgerin]]

Birmas first Nobel Peace Prize laureate
(32) dass sich [NP Gottfried Dienst] irrite that Gottfried Dienst made a mistake
(33) dass [N Gottfried] [N Dienst] hatte

For the sake of evaluation, we applied a longest-match strategy to the recognized structures whose confidence factor matches a parameterizable threshold.

5. Evaluation

For the evaluation of the minimal phrase recognition, we’ve chosen the first 500 sentences of the REF-D-corpus. Table 1 shows a frequency distribution respective to the token length and the construction type of the identified NPs, DPs and PPs. Note that we didn’t count XPs of the word length 1.

| XP      | constr | freq | corr | miss | spur | Prec. | Rec. |
|---------|--------|------|------|------|------|-------|------|
| NP      | mwl    | 39   | 37   | 2    | 1    | 98.09 | 96.74 |
|         | nb     | 763  | 739  | 24   | 14   |       |      |
| DP      | nb     | 1260 | 1249 | 11   | 21   | 98.39 | 99.00 |
|         | seq    | 35   | 32   | 3    |      | 98.22 | 98.00 |
| PP      | nb     | 890  | 874  | 16   | 15   |       |      |
|         | seq    | 12   | 10   | 2    | 1    |       |      |

Table 1: Frequency Distribution of selected XPs

The evaluated results in terms of precision and recall are shown in Table 2. Recall and precision were computed as follows: \( \text{Rec} = \frac{\text{corr} \times 100}{\text{corr} + \text{miss}} \), \( \text{Prec} = \frac{\text{corr} \times 100}{\text{corr} + \text{spur}} \).

1. Discussion of NP results

We identified 38 N-MWUs in a rule-based fashion by special suffixes (e.g. Rudolf Hell GmbH) as well as by a corpus-based strategy (e.g. Assurances Generales de France). Only one of the N-MWUs wasn’t detected properly. Instead of assigning Heidelberger Zement AG a MWU-reading, we did this only for Zement AG assuming that Heidelberger has the reading of an adjective of origin. In (34) and (35), we see the most complex structured NPs. In (36), we see two spurious and one missing NP annotations, since we wrongly recognized Jahren wie Pilze as a noun coordination.

(34) <pp> in <dp> der <np> Auflinen- , <np> Sicherheits- , <np> Sozial- , <np> Wirtschafts- und Finanzpolitik </np> </np> </np> </np> </dp>
(35) <dp> die <np> adj } klaaprinzen und stinkenden </adj> <np> alten </np> Laster und Busse </np> </np> </np> </dp>
(36) <pp> in <dp> den <np> letzten </np> Jahren wie Pilze </np> </np> </np> / </dp>

5In our full parsing strategy, we regret of such a disambiguation, hoping that the ongoing analysis process will dismiss some ambiguous ill-formed structures.

6Thanks to the Institute for Natural Language Processing (IMS) of Stuttgart for making the corpus available to us.
5.2. Discussion of DP and PP results

(36) is also an example of a propagating error for the DP and PP recognition, which was the most frequent error source (in total: 10 missing and 5 spurious DPs, 13 missing and 9 spurious PPs). 23 of the 35 DP closures, recognized with a seq-pattern, possess one XP between the border elements (cf. (37)). In this case, we can easily generate the correct structure by labelling the adjective as governor of the left adjacent XP. Such a correction isn’t possible by two or more XPs in between the border elements (cf. (38)). In (39), we see two correctly annotated PPs.

5.3. Discussion of Poss and PP results

In a first step, it segments and annotates sentences in corpora into better parseable units according to the Topological Field Model. In a second step, it recognizes so-called minimal phrases in the previously segmented fields. An evaluation of the second stage demonstrated the high quality of the parser, that can be used for many analysis tasks as well as for the task of corpus annotation.

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