Harvesting Multi-Word Expressions from Parallel Corpora

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Abstract

The paper presents a set of approaches to extend the automatically created Slovene wordnet with nominal multi-word expressions. In the first approach multi-word expressions from Princeton WordNet are translated with a technique that is based on word-alignment and lexicosyntactic patterns. This is followed by extracting new terms from a monolingual corpus using keywordness ranking and contextual patterns. Finally, the multi-word expressions are assigned a hypernym and added to our wordnet. Manual evaluation and comparison of the results shows that the translation approach is the most straightforward and accurate. However, it is successfully complemented by the two monolingual approaches which are able to identify more term candidates in the corpus that would otherwise go unnoticed. Some weaknesses of the proposed wordnet extension techniques are also addressed.

1. Introduction

WordNet (Fellbaum 1998) is an extensive lexical database in which words are divided by part of speech and organized into a hierarchy of nodes. Each node represents a concept and words denoting the same concept are grouped into a synset with a unique id (e.g. ENG20-02853224-n: {car, auto, automobile, machine, motorcar}). Concepts are defined by a short gloss (e.g. 4-wheeled motor vehicle; usually propelled by an internal combustion engine) and are also linked to other relevant synsets in the database (e.g. hypernym: {motor vehicle, automotive vehicle}, hyponym: {cab, hack, taxi, taxicab}). Over time, WordNet has become one of the most valuable resources for a wide range of NLP applications, which initiated the development of wordnets for many other languages as well1.

One of such enterprises is the building of Slovene wordnet (Erjavec and Fišer 2006). The approach is based on extracting translation equivalents from parallel corpora with automatic word-alignment and disambiguating the polysemous words with wordnets that already exist for other languages. This paper is an extension of our previous experiment and tries to overcome the limitation of the alignment approach with a complementary method to add multi-word expressions harvested from the JRC-Acquis parallel corpus (Steinberger et al. 2006) to Slovene wordnet.

Multi-word expressions (MWE) are lexical units that include a range of linguistic phenomena, such as nominal compounds (e.g. blood vessel), phrasal verbs (e.g. put up), adverbal and prepositional locutions (e.g. on purpose, in front of) and other institutionalized phrases (e.g. de facto). MWEs constitute a substantial part of the lexicon, since they express ideas and concepts that cannot be compressed into a single word. Moreover, they are frequently used to designate complex or novel concepts. As a consequence, their inclusion into wordnet is of crucial importance, because any kind of semantic application without appropriate handling of MWEs is severely limited.

For the purpose of MWE identification, various syntactical (Bourigault 1993), statistical (Tomokiyo and Hurst 2003) and hybrid semantic-syntactic-statistical methodologies (Piao et al. 2003, Dias and Nunes 2004) have been proposed, to name but a few. Since the majority of MWEs included in the Princeton WordNet are nominal (see Table 1 below) and compositional, our approach is based on syntactic features of MWEs.

The rest of the paper is organized as follows: first, the Slovene WordNet Project is described. Section 3 describes the procedure used to extract multi-word expressions from the corpus and their mapping to the wordnet hierarchy. The results are presented and evaluated in Section 4, and the paper ends with concluding thoughts and plans for future work.

2. The Slovene WordNet Project

The first version of the Slovene wordnet was created on the basis of the Serbian wordnet (Krstev et al. 2004) was translated into Slovene with a Serbian-Slovene dictionary. The main advantages of this approach were the direct mapping of the obtained synsets to wordnets in other languages and the density of the created network. The main disadvantage was the inadequate disambiguation of polysemous words, therefore requiring extensive manual editing of the results. The core Slovene wordnet contains 4,688 synsets, all from Base Concept Sets 1 and 2.

In the process of extending the core Slovene wordnet we tried to leverage the resources we had available, which are mainly corpora. Based on the assumption that translations are a plausible source of semantics we used multilingual parallel corpora such as the Multext-East (Erjavec and Ide 1998) and the JRC-Acquis corpus (Steinberger et al. 2006) to extract semantically relevant information (Fišer 2007).

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1 See http://www.globalwordnet.org/gwa/wordnet_table.htm [15.03.2008]
We assumed that the multilingual alignment based approach can either convey sense distinctions of a polysemous source word or yield synonym sets based on the following criteria (cf. Dyvik 1998, Diab 2000 and Ide et al. 2000):

(a) senses of ambiguous words in one language are often translated into distinct words in another language (e.g. Slovene equivalent for the English word ‘school’ meaning educational institution is ‘šola’ and ‘jata’ for a large group of fish);
(b) if two or more words are translated into the same word in another language, then they often share some element of meaning (e.g. the English word ‘boy’ meaning a young male person can be translated into Slovene as either ‘fant’ or ‘deček’).

In the experiment, corpora for up to five languages (English, Slovene, Czech, Bulgarian and Romanian) were word-aligned with Uplug (Tiedemann 2003) used to generate a multilingual lexicon that contained all translation variants found in the corpus. The lexicon was then compared to the existing wordnets in other languages. For English, the Princeton WordNet (Fellbaum 1998) was used while for Czech, Romanian and Bulgarian, wordnets developed in the BalkaNet project (Tufiš 2000) were used. If a match between the lexicon and wordnets across all the languages was found, the Slovene translation was assigned the appropriate synset id. In the end, all the Slovene words sharing the same synset ids were grouped into a synset.

The results obtained in the experiment were evaluated automatically against a manually created gold standard. A sample of the generated synsets was also checked by hand. The results were encouraging, especially for nouns with f-measure ranging between 69 and 81%, depending on the datasets and settings used in the experiment. However, the approach had two serious limitations: first, the automatically generated network contains gaps in the hierarchy where no match was found between the lexicon and the existing wordnets, and second, the alignment was limited to single-word literals, thus leaving out all the multi-word expressions.

3. Adding Multi-Word Expressions to Slovene Wordnet

This section presents the three approaches we used to add multi-word expressions to Slovene wordnet. The first approach is bilingual and uses a lexicon we extracted from the automatically word-aligned JRC-Acquis corpus and lexico-syntactic patterns to find Slovene translations of English multi-word expressions that are in Princeton WordNet (PWN).

The second two approaches are monolingual and aim to complement the translation approach by identifying additional multi-word expressions that are not yet in PWN. Domain-specific MWEs were extracted from a fishing subcorpus of the JRC-Acquis and filtered using fishing-related seedwords found in Eurovoc.

In the final approach, we use lexico-syntactic patterns to extract semantically related expressions (hypernyms and hyponyms) from the corpus and then map them to Slovene wordnet.

3.1 Translation-Based Approach: Adding MWEs Found in The Princeton WordNet

First, a list of multi-word expressions PWN was extracted. As Table 1 shows, most of them were nouns. Since our alignment-based approach proved to work best on nouns, we decided to limit our work to those (about 61,000). A large majority of the expressions do not belong to Base Concept Sets (about 64,000) that are the main constituents of the current version of Slovene wordnet. This means that this approach will generate a lot of new synsets, extending our wordnet even further (see Table 2).

| POS     | Freq. |
|---------|-------|
| nouns   | 60,931|
| verbs   | 4,315 |
| adverbs | 955   |
| adjectives | 739 |
| total   | 66,940|

Table 1. The distribution of MWEs in PWN across POS

Table 3 shows that the polysemy of MWEs is not nearly as high as polysemy of single-word literals: a large majority of expressions were found in only one synset and very few were highly polysemous. This is why we were able to restrict the number of languages used in the experiment to English and Slovene only.

| no. of synsets | no. of MWE’s |
|----------------|--------------|
| 1              | 56,989       |
| 2              | 3,335        |
| 3              | 453          |
| 4              | 108          |
| 5              | 40           |
| 6              | 6            |
| total          | 60931        |

Table 3. Polysemy of MWEs in PWN

Our first task was to match English MWEs found in PWN with their Slovene counterparts in the JRC-Acquis corpus. The chosen approach combines methods of bilingual lexicon extraction (Tiedeman 2003) with those of terminology extraction (e.g. Bourigault et al. 1996, Martin and Heid 2001).

2 http://europa.eu/eurovoc/ [15.03.2008]
For each source MWE from PWN, we extracted all sentence pairs from the parallel corpus that contain the source term. Also, for each single word from the source MWE we extract all possible translation equivalents from the bilingual lexicon. We then use lexico-grammatical patterns to identify potential multi-word terms in the target language and employ the bag-of-equivalents approach for the selection of the best equivalent, which is the candidate with the most matches for each constituent word in the bilingual lexicon (Vintar 2004).

For each source MWE this approach yields a list of target MWE’s, ranked according to the quality of single-word translations inherited from the word-alignment system. Adding the translated MWE’s to the Slovene wordnet is trivial in this case because each English MWE already has a unique synset id from PWN that is retained in Slovene wordnet (e.g MWE ‘art exhibition’ with PWN id ‘ENG20-07896855-n’ is translated into Slovene as ‘umetnostna razstava’ and keeps the same id in Slovene wordnet).

3.2 Seed-Word-Based Approach: Extracting Domain-Specific MWEs

With the second approach we tried to identify completely new MWEs that were not found in PWN and add them to Slovene wordnet. Synsets that were added in this way were assigned PWN-like synset ids that retain the source information (e.g. MWE ‘fish, fishing, fishery, net, ship, fleet’). This corpus contains over 12,000 ‘kot so (such as)’ contexts, however not all yield structures suitable for inclusion into Slovene wordnet. The challenge in extracting relevant candidates for the taxonomic is-a relation lies especially in accurate identification of constituent terms. From the RJC-Acquis corpus was created a subcorpus of nearly 2,000 documents was created from the Slovene part of the JRC-Acquis.

From this corpus we extracted multi-word terms using a hybrid method based on part-of-speech patterns and keyworsdness ranking. The patterns included noun phrases such as A N N (adjective + noun + noun) or A N S N (adjective + noun + preposition + noun). The extracted noun phrases were then ranked according to their keyworsdness, where each constituent word of a MWE is assigned its keyworsdness by comparing its relative frequency in the fishing corpus to its relative frequency in the FidaPlus corpus, a 600-million-word reference corpus of Slovene (Arhar et al. 2007).

This yields quite a large number of terms related to fishing, but since we needed to establish a link to existing synsets in the Slovene Wordnet, we selected a small number of fishing-related seed words from the multilingual EUROVOC descriptors (riba, ribolov, ribištvo, mreža, ladja, ladjevje, plovilo [fish, fishing, fishery, net, ship, fleet, vessel]). It was now possible to select all extracted MWEs that contained one of the seed words. If the seed word was the headword of the extracted MWE (e.g. vlečna mreža [trawl net] – hyponym of – mreža [net]), we presuppose the MWE to be a hyponym of the seed word, whereas if the seed word appears in a pre- or postmodifying position, the hyponym of the MWE is its headword (e.g. posadka plovila [vessel crew] – hyponym of – posadka [crew]).

3.3 Pattern-Based Approach: Extracting New Semantically Related MWEs

In our final approach, we extracted knowledge-rich contexts from the entire JRC-Acquis corpus, requiring manual inspection.

[NP , kot so ([NP] , )+ [NP] inalji [NP] ([NP] , such as ([NP] , )+ [NP] and/or [NP])
([NP] , kot na primer ([NP] , )+ [NP] inalji [NP] ([NP] , for example ([NP] , )+ [NP] and/or [NP])
([NP] , )+ [NP] inalji drug [NP]
([NP] , )+ [NP] and/or other [NP])

Such contexts typically contain a set of semantically related terms, usually with one term representing a superordinate (parent) concept (e.g. regional fishing organisation) and one or several other terms representing subordinate (child) concepts (e.g. NAFO, ICCAT, CCAMLR, NEAFC). Since no parser for Slovene is publicly available at the time, we used a self-made shallow parser to identify noun phrases. The JRC-Acquis corpus contains over 12,000 ‘kot so (such as)’ contexts, however not all yield structures suitable for inclusion into Slovene wordnet. The challenge in extracting relevant candidates for the taxonomic is-a relation lies especially in accurate identification of constituent terms.

Far from trivial is also the task of finding the corresponding wordnet synset into which the terms extracted from the corpus should be included. This was done in several steps:

- First, we checked whether the hypernym was already in Slovene Wordnet. In this case, its hyponyms extracted from the corpus were just added to it.
- If the hypernym was not found in wordnet, it was decomposed step by step until a match was found. For example, the term ‘mednarodna pomorska organizacija’ [international naval organization] was first decomposed to ‘pomorska organizacija’ [naval organization]. Because there was no match, the term was further decomposed to ‘organizacija’ [organization] and this expression was then considered as the hypernym of our term. If there were more possible hypernyms for a given term, the term was added to all hypernym candidates with a mark requiring manual inspection.
- If the hypernym could not be found in Slovene Wordnet at all, we checked whether any hyponyms were already in wordnet. The rest of the extracted hyponyms were treated as its siblings and attached to the same parent node.

The difference between this approach and the approach described in Section 3.2 is that we could not filter the contexts with any domain-sensitive filters in order to reduce noise. As it turns out, a lot of the extracted contexts were useless for our purpose because the semantic relations between the words were highly contextual and therefore not suitable for inclusion to the taxonomy (e.g. postavke, kot so status, privilegij in imuniteta [items, such as status, privilege and immunity]). Also, many
In this approach, we first extracted a list of multi-word expressions from the fishing subcorpus, which was then filtered using a list of seed terms taken from Eurovoc (see Section 3.2). Here, all the expressions were extracted correctly and are indeed fishing-related terms, because the restriction was that they must contain at least one seed word. We obtained however only 251 terms, 75 of which contain the seed word as the headword and thus directly imply their hyponymic relation to it, while 176 candidates contain the seed word either as a premodifier or postmodifier.

Hypernym assignment for terms that contain a seedword was almost perfect. All but 2 headwords were correctly identified in Slovene wordnet and the terms were added to them as their hyponyms. The two headwords that were missing in the wordnet (‘plovilo’ [vessel] and ‘mreža’ [net]) were compensated by finding one of their hyponyms in the network (‘ribaško plovilo’ [fishing vessel] and ‘ribaška mreža’ [fishing net]) and treating the rest of the terms as their cohyponyms.

The approach was slightly less successful in cases where seedwords were pre- or postmodifiers of MWEs. Wordnet contained hypernyms for 126 out of such 165 MWEs and 5 of the 175 hyponyms. However, in about a third of the cases there were more possible hypernym candidates. Since no automatic WSD was available, the correct hypernym was therefore selected by hand.

While our approach presupposes that MWE’s are compositional, i.e. that a MWE’s headword is at the same time its hypernym, this is not always the case, as the examples below show:

- morski meč – *hyponym of – meč [sea sword (silver scabbardfish)] – [sword]
- morski pes – *hyponym of – pes [sea dog (shark)] – [dog]

It turned out that only 6 out of 175 MWEs were non-compositional. For these the appropriate hypernym could not be identified from their constituent words and could therefore not be correctly added to wordnet.

### 4.3 Pattern-Based Approach

In the last approach we used lexico-syntactic patterns to extract words from the corpus that are in a hypernym-hyponym relation. The extracted candidates were then mapped to wordnet (see Section 3.3). Table 5 gives the results of candidate extraction and mapping. It must be noted here that not all NP’s we harvested from the corpus were MWE’s and are therefore not included in evaluation and discussion in this paper.

| Pattern | such as | for example | and other |
|---------|---------|-------------|-----------|
| Extracted | 941 | 38 | 3344 |
| Mapped | 503 | 29 | 1817 |
| MWE | 217 | 14 | 1385 |

Table 5: Yield of lexico-grammatical patterns
The biggest drawback of this approach was the nature of semantic relatedness of words matching the pattern, which is highly contextual in the JRC-Acquis corpus and too vague to be included in wordnet. Also, many hypernyms that were successfully identified in both the extracted terms and wordnet are too general to be of any real taxonomic significance. This is why many (almost half) of the extracted patterns were not mapped to Slovene wordnet at all.

4.4 Comparison of the Approaches

As table 6 shows, by far the most MWEs were obtained from the pattern approach. But very few of them were (correctly) added to wordnet. On the other hand, the seedword approach generated relatively few MWE candidates but a large majority of those were successfully mapped to wordnet. The best approach is clearly the first one in which all of the correctly translated MWEs were also successfully added to wordnet. This is hardly surprising because the English MWEs were extracted from PWN and therefore already had appropriate synset ids which were simply inherited by their Slovene translation equivalents. This approach was also attractive because it simply reused the existing resources that were created in our previous experiments.

The value of the second two approaches lies in the fact that there are many more useful MWEs in corpora than in PWN. These cannot be handled by the first approach that relies on PWN, which is why complementary monolingual techniques can be handy in identification of additional MWEs, thus generating more synset candidates for the extension of the wordnet.

Table 6: Comparison of the approaches according to the number of correctly extracted MWEs and the number of correctly added MWEs to Slovene wordnet

|                        | translation approach | seed-word approach | pattern approach | total     |
|------------------------|----------------------|--------------------|-----------------|-----------|
| correctly extracted    | 1,059                | 251                | 4,323           | 5,597     |
| correctly added to wordnet | 1,059 (100%) | 206 (82%) | 1616 (37%) | 2,881 (52%) |

Another necessary improvement of the procedure is the automation of the selection of the best hypernym candidate that is currently being done by hand, which is tedious and slow.

The problem that many semantic relations extracted from the corpus are contextual rather than taxonomic remains unresolved here. Also, the density of the hierarchy remains an issue and will have to be dealt with in our future work. A possible solution would be to use more encyclopaedic-like resources such as Wikipedia in which the semantic relations between words are more regular.

6. References

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