Abstract

In this paper we describe a WSD experiment based on bilingual English-Spanish comparable corpora in which individual noun phrases have been identified and aligned with their respective counterparts in the other language. The evaluation of the experiment has been carried out against SemCor.

We show that, with the alignment algorithm employed, potential precision is high (74.3%), however the coverage of the method is low (2.7%), due to alignments being far less frequent than we expected. Contrary to our intuition, precision does not rise consistently with the number of alignments. The coverage is low due to several factors; there are important domain differences, and English and Spanish are too close languages for this approach to be able to discriminate efficiently between senses, rendering it unsuitable for WSD, although the method may prove more productive in machine translation.

1 Introduction

Word Sense Disambiguation (WSD) could be defined as the task of assigning the right sense to a word in context given a sense inventory. This is a problem in artificial intelligence reported at least since the nineteen fifties. There is general consensus in that although it is not a very interesting question in itself in many areas (lexicography being the obvious exception) deeper understanding of lexical ambiguity would greatly help to solve some applications of natural language processing and clarify new ones still to be uncovered.

Here we present a WSD experiment based on bilingual English-Spanish comparable corpora of news collections in which individual noun phrases have been identified and aligned to their counterparts in the other corpus. WordNet (Miller, 1995) is a lexical database for English which includes a sense inventory among many other things. This sense inventory relies in the synset concept. A synset is a synonym set of words with a particular meaning, for instance two synsets associated with different senses of church are {church, Christian church, Christianity} and {church, church building}. An extension of WordNet is EuroWordNet (Vossen, 1997). EuroWordNet has a very similar structure to WordNet, but comprises several European languages. In addition, there are links between the concepts in different languages. The evaluation of the experiment has been carried out against SemCor (Francis and Kucera, 1967). SemCor is a collection of English texts which has been manually annotated with WordNet senses and for this reason has often been used as a test collection for WSD algorithms.

In a first step, the noun phrases obtained from the English news articles corpus are searched for in SemCor. Next, we associate each of theses phrases with the corresponding aligned phrases in Spanish, together with the observed alignment frequency in the news collections. In this alignments, there is usu-
ally a cognate or at least, one word which is a direct translation of the other, but the rest of the words in the phrase can give us a clue about the correct sense.

The most relevant factors to consider about this experiment with respect to previous research are the following:

- Parallel vs. Comparable corpora. Many WSD algorithms use a supervised approach that relies on manually tagged examples to learn a classification algorithm. This manual tagging is very costly, leading to what has been called the knowledge acquisition bottleneck. A relatively popular approach has been to use parallel texts to extract knowledge automatically. The problem with parallel corpora is that it is also very scarce. Comparable corpora offers some of the advantages of parallel corpora with a much higher availability but at the cost of obtaining inferior quality knowledge.

- Phrase detection. It is not straightforward to detect noun phrases in different languages. We don’t know how big the impact of errors in detection is for the accuracy of this approach.

- Phrase aligning. Again, the precision of the alignments between phrases (about 73%), might affect the performance of the system.

- The domain problem. It is well-known that extracting knowledge in one domain and trying to apply it in another one is generally a bad idea. Ideally one should use the same domain for both tasks, however, it is unlikely for large unrestricted domain comparable corpora to be widely available in the near future.

- It is generally accepted that one important obstacle for WSD is that cross language linguistic effort has traditionally focused on bilingual dictionaries and the like, which work at word level or higher and that a reliable cross language man-made tool at the sense level would greatly contribute to the solution of the problem. Fortunately such resources now exist; the set of interlingual indices in EuroWordNet is an example.

In the second section we discuss previous work in the field together with a motivating example. In the following section we describe the experiment. In the third section we present the evaluation and results, with several successful and unsuccessful examples and in the fourth section we draw our conclusions and suggest future work.

2 Previous work

The basic idea, is similar to the approach in (Gale et al., 1993) which uses the English-French parallel corpus of the Canadian Hansards, although the fundamental unit from which information is extracted is not the word but the noun phrase, much less ambiguous in general. It allows discarding the senses of individual words when translating with a bilingual dictionary. Our approach is more related to the work in (Dagan et al., 1991; Dagan and Itai, 1994) which uses pairs of syntactically related words.

This idea that different senses of the same word often translate to different words in a second language was also an argument to suggest a new method of evaluating WSD systems in (Resnik and Yarowsky, 1999). The paper presents a formula to calculate the relatedness of two word senses according to the translations to a second language. Another novelty is the generalization of the method to several pairs of languages instead of just one.

The noun phrases in English and Spanish have been taken from work described in (Penas, 2002) and the alignment between them is explained in (López-Ostenero et al., 2002; López-Ostenero, 2002). The alignment algorithm used has the advantage that corpora doesn’t have to be parallel, just comparable. The phrases were presented to Spanish human evaluators in the interactive track of the CLEF’02 competition. The evaluators had to find documents in a database relevant to a query in English with the aid of text fragments in Spanish. Using the phrases in Spanish aligned with the phrases in the English documents as aid fragments considerably outperformed SYS-TRAN automatic translations of the documents. These good results motivated the crossover attempt to WSD.

For the sake of clarity, we sketch the procedure followed to create the dictionary of aligned phrases.
The CLEF collections used to extract the phrases are a Spanish corpus made of 1994 news from Spanish news agency EFE and an English corpus containing articles published by Los Angeles Times also in 1994.

The first step is to identify the noun phrases. For Spanish, words are lemmatized and POS tagged. After that process, chunks of words fitting the following pattern are automatically considered noun phrases.

\[(\text{noun} \mid \text{adjective}) (\text{noun} \mid \text{adjective} \mid \text{preposition} \mid \text{determiner} \mid \text{conjunction})* (\text{noun} \mid \text{adjective})\]

Only phrases with two or three open-class words are considered because the amount of longer phrases that can be aligned rapidly decreases. This process identified more than twenty-seven million alleged noun phrases in the corpus.

As far as English is concerned, each word was assigned the prior most likely POS tag. The pattern for identifying patterns was the same as in Spanish. More than nine million noun phrases for English were identified this way.

The alignment has been carried out with a bilingual resource. The phrases with two open-class words are aligned with two open-class word phrases in the other language and so on for three open-class word phrases. The other constraint required to align is that each open-class word in a phrase translates to an open-class word in the candidate phrase in the other language. The real alignment algorithm is somewhat more complicated but the details can be found in the referenced articles.

It has been shown in these articles that the precision in recognizing noun phrases is high. The precision of the alignments has been estimated in excess of 73%, and the correction in the alignment correlates with the absolute frequency of the phrases, that is, an alignment between commoner phrases is more likely to be correct.

We illustrate the idea with the following example.

We want to disambiguate \textit{issue}, which can be translated in Spanish as: \textit{asunto}, \textit{tema}, \textit{número}, \textit{emisión}, \textit{expedición}, \textit{descendencia}, \textit{publicar}, \textit{emitir}, \textit{expedir}, \textit{dar y promulgar}. At this point we detect that the context of the word indicates that it is part of the phrase \textit{abortion issue}. This phrase has been aligned with the phrase in Spanish \textit{tema del aborto}.

If we were doing machine translation, we would be satisfied with this translation, however, in the framework of WSD we would like to discard the senses of \textit{issue} not corresponding to the \textit{tema} translation. Unfortunately, WordNet structure does not permit obtaining that information easily. The key then, is to associate individual word senses to translations in the other language, going one step further from the word to word translation.

3 The experiment

In this section we describe the experiment. We start with the resource of the bilingual phrases and the interlingual indices (ILI) in EuroWordNet and we perform WSD in the SemCor collection.

The approach we have taken uses EuroWordNet interlingual indices. These indices map the \textit{synsets} from one EuroWordNet language to another so that, in the previous example, we could use them to look up the synsets associated with \textit{issue} in Spanish and find out which of them hold the word \textit{tema}. One drawback with this approach is that EuroWordNet taxonomy is linked in English with the concepts in WordNet-1.5, which is a little outdated. We want to apply our system to the senses in WordNet-1.7. This will allow the system, to be tested in the short term with the latest SENSEVAL (Kilgarriff, 1998) collections and thus compared with state-of-the-art participating systems. To overcome this version conflict we use the mappings from versions 1.5 to 1.6 and from 1.6 to 1.7 of WordNet developed in (Daudé et al., 2000; Daudé et al., 2001).

As we want to disambiguate a target word, we first look at the context to determine if the word belongs to one of the phrases with alignments in our knowledge base. We construct a simple automaton to implement a detection algorithm which takes into account inflectional variants. We create a forest in which trees have words as labels. Each tree contains all the phrases beginning with a certain word. For each possible continuation of a phrase beginning with that word we have a child node with the corresponding label. Some nodes as also labeled as acceptance nodes, marking the end of a legal phrase (although there might be longer phrases with the

\[1\] Adapted from López-Ostenero et al., 2002.
same prefix). To detect the phrases in the text, a
word is read and looked up in the heads of the trees
to check for a match (two words match if their lemu-
mas are the same). If there is a the next word is read
on and tried to match with the label of one of the
children nodes, and so forth until no more matches
are possible. If an acceptance node has been tra-
versed then a noun phrase has been found and the
unused portion of the input is restored to the input
buffer in order for the search to continue. Longer
phrases are preferred over shorter ones in case sev-
eral acceptance nodes have been traversed. This al-
gorithm takes linear time to detect the phrases.

If the word belongs to a noun phrase, we traverse
the list of the phrases aligned with it in Spanish.
For every word in each Spanish phrase, we look for
the synsets associated with their senses and, using
the ILI, their associated English synsets. If any of
these synsets contains the target word, then the cor-
responding sense is kept, otherwise it is discarded.

The first step to prepare the experiment has con-
sisted of automatically re-annotating SemCor. Sem-
Cor has manual annotations in SGML of lemma,
part of speech, wordNet sense and even compound
words, among others, but of course the information
about our dictionary of phrases is not included so
we translated it into XML form and added, for the
words belonging to aligned phrases, one attribute,
phrase, which indicates the detected phrase, and an-
other, alignments, which shows a list of admissible
senses with respect to the algorithm just described,
along with the frequency with which the corpora al-
lowed a particular alignment supporting that sense.
This information is important, because reliability of
alignment is supposed to directly depend on its fre-
quency.

So, for instance, this SemCor fragment:

\begin{verbatim}
<wf cmd="done" pos="NN" lemma="number"
wnsn="2" lexsn="1:23:00::":">number</wf>

<wf cmd="ignore" pos="IN">of</wf>

<wf cmd="done" pos="NN" lemma="voter"
wnsn="1" lexsn="1:18:00::":">voters</wf>

Would now look like this:

\begin{verbatim}
<wf alignments="number%1:07:00:::51
number%1:10:00:::51 number%1:10:01:::51
number%1:10:02:::51 number%1:10:03:::51
number%1:10:04:::51 number%1:10:05:::51
number%1:23:00:::51" cmd="done"
lemma="number" lexsn="1:23:00::":" phrase="number of voters" pos="NN"
wnsn="2">number</wf>

<wf cmd="ignore" pos="IN">of</wf>

<wf cmd="done" lemma="voter"
lexsn="1:18:00:::" pos="NN" wnsn="1">
voters</wf>

It is interesting to note that number has eleven
senses in WordNet-1.7, of which now only eight are
equally amenable to be chosen.

4 Evaluation and results

We have evaluated this approach against SemCor.
This decision is supported by the fact that it is a
test collection whose size allows drawing more rep-
resentative conclusions than from other, smaller-
sized collections, such as those in SENSEVAL
(Kilgarriff, 1998).

In the process of re-tagging the collection, out of
the 192840 words amenable for disambiguation in
brown-1 and brown-2 segments, we detected 10787
English phrases, which make up for 5.6% of the
words. This phrases have alignments in Spanish in
5290 cases, so we filtered senses for this number of
words, 2.74% of the total. Among them, the right
sense has remained unfiltered in 3922 cases. That
is, the filtering process has a potential precision of
74.33%.

One example in which the algorithm doesn’t work
as expected is in disambiguating friend in the phrase
friend of mine. The alignment in Spanish was cono-
cido de las minas (which could be translated as ac-
quaintance of the mines). There are two relevant ob-
servations. First of all, the Spanish phrase proba-
bly refers to a well-known flamenco festival which
is a proper noun and should therefore not be aligned
with a common one. An entity recognition module,
even one as simple as considering initial capital let-
ters, should have ruled this alignment out. Second,
one has to wonder how high can the degree of overlapping between the news in both collections be.

It is obvious that alignment techniques need to be improved. However, since these two phrases only were aligned once we felt the need to test the correlation between frequency of alignments and potential precision susceptible of being achieved.

In order to shed some light on the subject we repeated the experiment adding a threshold. This time we only disambiguate words in phrases having alignments with Spanish phrases when the alignment frequency is over the threshold. Results can be seen in figure 1.

![Figure 1: Relation between threshold, coverage and potential precision](image)

The results are surprising: Potential precision does not really increase with increasing threshold values. Up to 3000 occurrences\(^2\), there is hardly any difference in potential precision. From there, the number of phrases with alignments is so low (with a threshold of 3000 occurrences the number of phrases with alignments in SemCor is just 38) that the information is useless. Potential precision of 100\% from a threshold value of 8713 until the end (8836) corresponds solely to alignments of the phrase year old, which equally support all four senses of year, so, at that point the information coming from the aligned phrases is totally irrelevant.

As we can see, the coverage of the approach is rather low, but the method really works, even when alignments are only of modest quality, to say the least. As a remarkable example, the noun phrase head of the family was aligned eight times with the phrase responsable de la cámara. Head has 132 different senses as a noun in WordNet-1.7. Cámara doesn’t help but responsible has two senses. One of them is \{autor, culpable, perpetrator, responsable\} which the ILI links to the English synset \{culprit, perpetrator\}, which doesn’t support any of the many senses of head. However, the other sense of responsible corresponds to the synset \{responsible\}, linked by ILI to the English synset \{head, chief, top dog\} supporting the correct sense of head in the original phrase.

The mappings between WordNet versions and the Interlingual indices get sometimes in the way of the success. For the sake of clarity we will use the sense notation in WordNet instead of the associated synsets. The sense notation refers to a set of lexicographer tematic files. For an illustration of both problems in one example, consider the phrase art studies. It is aligned with estudios de arte (again, a reasonably related phrase, but highly unsuitable as a translation). Arte has four senses: arte%1:04:00:: which goes through ILI plus the mappings to art%1:04:00::, a second one arte%1:06:00:: which points to art%1:06:00::, a third one, arte%1:09:00:: which ILI points to art%1:09:00::: but the mappings just don’t map to anything at all, and the last sense which starts out as arte%1:10:00:: and ends up as art%1:10:00:: as expected.

The mappings are not complete and therefore occasionally fail to upgrade a sense to the newer version, but it is more disturbing to verify that art and arte are given the exact same semantic structure in EuroWordNet as far as the algorithm is concerned.

5 Conclusions and future work

We had hoped that use of comparable corpora would help alleviate the knowledge acquisition bottleneck. The corpora, according to the millions of noun phrases detected, seemed indeed bigger than many parallel corpora available, and the scaling possibilities are obvious, just add more years to the news collections. Nevertheless, the scarceness of alignments has produced an extremely low coverage for the WSD algorithm. It is thus, very unclear that

\(^2\)Not shown in the graphic due to the almost negligible percentage of phrases aligned
bigger comparable corpora would help WSD in this specific approach.

The domain problem has undeniably and heavily affected the experiment. The Brown Corpus of which SemCor is a portion, was compiled from texts printed in 1961. The news thirty-four years later surely cover different topics, many of which didn’t even exist back in 1961. One of the alignment examples cited in (López-Ostenero, 2002) is free trade agreement which aligns with tratado de libre comercio. Even worse is that the noun phrases present in SemCor hardly occur in the LAT '94 collection. This is not a problem of the noun-phrase approach, but a serious domain problem. The question is that SemCor is big enough to allow interesting conclusions to be extracted from experiments as far as statistics are concerned, but very old with respect to modern texts. There are more recent hand annotated collections, however they are much smaller-sized and thus unfit for statistically relevant purposes.

Also, regarding domain, it is reasonable to suspect that the differences between the news domain used to gather the phrases and the SemCor collection, a part of the Brown Corpus, which was collected with the aim of being domain-free, might have influenced the results.

Another interesting question regarding coverage is how comparable comparable corpora really are. Ostenero reports 38% of the English two-open-class-word noun-phrases to have been aligned to Spanish ones so the corpora seem moderately comparable.

The mappings used to convert WordNet-1.5 synsets to version 1.7 have been assessed by their authors to have a precision around 90%. Since we have applied two mappings (1.5 → 1.6 and 1.6 → 1.7) we can estimate the probability of correctly mapping a sense as being .9*.9=.81. That would account for 19% of the senses disappearing in the process, so the real reduction of ambiguity due to the bilingual noun-phrase approach has to be lower than the overall figures apparently indicate.

The ILI has proven to be somewhat disappointing. In spite of being heavily advertised as one the of the most outstanding achievements in EuroWordNet, it turns out that the language neutral representation of nominal entries to which the ILI point is precisely the nominal structure of the original English WordNet-1.5. Moreover, in the Spanish nominal structure the contents of the synonyms sets differ from the English ones but the network structure of hypernyms and other relations is the same except when there is no equivalent concept in Spanish or the obvious linking is inapplicable. That, plus the fact the the ILI is semi-automatically constructed and only manually revised, amounts to the English and Spanish nominal structures being so close that the ILI coincides more often than desirable with the identity function. This fact is quite clear in the sense file notation, although the synset-offset number notation provides a rather awkward encoding for this approximation of the function f(x)=x. This may not be an issue with pairs of languages other than English but in this case is a factor that requires further research.

The phrase detection algorithm is in its first stages of development and there is much room for improvement, although it is unknown if such improvements will effectively help WSD.

The alignment technique employed is also not exempt of problems. The algorithm seems very sound with respect to finding correct alignments, although we suspect that there is a considerable amount of false positives. If this problem was solved, potential precision could raise a bit, however it would definitively lower a coverage that is already rather tiny. On the other hand there are cases in which less-than-spectacular quality alignments have proven useful for the task.

This series of facts lead us to conclude, in the first place, that although this method constitutes an a priori interesting filter in terms of precision, the rather low coverage of the method produces nearly negligible results for WSD.

Apart from that conclusion, the most interesting result is that, contrary to our intuition, potential precision does not rise consistently with the number of alignments. Since the precision of the alignments has been shown to correlate with the frequency of such alignments, the only explanation is that these high-frequency alignments are not productive in terms of filtering senses due to exactly equal mapping of senses to words in the two languages. We observed this behaviour in the case of the most aligned phrase, year old.

So, in the case of English and Spanish, it was
easy to predict that there would be many pathological cases. For instance, the phrases containing the word *art* in English align with phrases containing *arte* in Spanish, something which is not productive at all, since all the senses of art can be translated as *arte* and so the method does not discard any senses. Alignments between more heterogeneous pairs of languages may improve performance as well as adding together the results for comparable corpora in multiple pairs of languages. That would take advantage of the reasonable hypothesis that ambiguities will be different across different pairs of languages. The method may prove more productive in machine translation, where many different word senses may translate to the same word in the target language. Of course, the alignments are not directly acceptable translations.

The potential precision concept used for the evaluation is certainly somewhat fuzzy, in that reduction of ambiguity is not specified. Potential precision is not to be confronted with actual precision, since this approach only aspires to efficiently discard some senses of the words, not to perform full disambiguation. Anyway, the low coverage of the method allows to discard it for WSD purposes whatever the actual ambiguity reduction obtained.

These aligning techniques were successfully applied for the CLEF competition (López-Ostenero et al., 2002), in human-computer interaction scenario, however this success does not carry over to the automated WSD problem. It is interesting to note, however, that since in that work the phrases were detected, aligned and used on the same collections, there were no domain problems, thus obtaining much higher coverage.

Summing up, the logical future work, in order for this approach to WSD to reach viable status, would ideally comprise, among other things, finding large quantities of moderately parallel corpora in different pairs of languages (with one of them fixed as target language), of genre and age as close to those of the test collection as possible, preferably without any intervening mappings. The predictable much higher coverage of the method would then foster the need to measure the actual degree of reduction in ambiguity. The question on the usefulness of the ILI remains open.

Acknowledgements

We are indebted to Julio Gonzalo for coming up with the idea of applying the noun phrases plus the ILI to WSD and for his advice, and to Fernando Lopez Ostenero for his willing assistance with the bilingual noun phrase resource.

References

[Dagan and Itai1994] Ido Dagan and Alon Itai. 1994. Word Sense Disambiguation Using a Second Language Monolingual Corpus. *Computational Linguistics*, 20(4):563–596.

[Dagan et al.1991] Ido Dagan, Alon Itai, and Ulrike Schwall. 1991. Two Languages are More Informative than One. In *Proceedings of the 29th meeting of the Association for Computational Linguistics*, pages 130–137.

[Daudé et al.2000] J. Daudé, L. Padró, and G. Rigau. 2000. Mapping WordNets using structural Information. In *Proceedings of the 38th Annual Meeting of the Association for Computational Linguistics (ACL), Hong Kong*.

[Daudé et al.2001] J. Daudé, L. Padró, and G. Rigau. 2001. A Complete WN1.5 to WN1.6 Mapping. In *Proceedings of the NAACL Workshop WordNet and Other Lexical Resources : Applications Extensions and Customization, Pittsburg*.

[Francis and Kucera1967] S. Francis and H. Kucera. 1967. Computational Analysis of present-day American English. *Providence, Rhode Island: Brown University Press*.

[Gale et al.1993] William A. Gale, Kenneth W. Church, and David Yarowsky. 1993. A method for disambiguating word senses in a large corpus. *Computers and the Humanities*, 26(5):415–439.

[Kilgarriff1998] Adam Kilgarriff. 1998. SENSEVAL: An exercise in evaluating word sense disambiguation programs. In *Proceedings of the International Conference on Language Resources and Evaluation (LREC)*, pages 581–588, Granada, Spain.

[López-Ostenero et al.2002] F. López-Ostenero, J. Gonzalo, Pe nas Anselmo, and F. Verdejo. 2002. Noun phrase translations for Cross-Language Document Selection. In *Cross Language Evaluation Forum (CLEF)*, number 2406 in Lecture Notes in Computer Science. Springer-Verlag.

[López-Ostenero2002] F. López-Ostenero. 2002. *Un Sistema Interactivo para la Búsqueda de Información en*
Idiomas Desconocidos por el Usuario. Ph.D. thesis, Universidad Nacional de Educación a Distancia (UNED).

[Miller1995] George A. Miller. 1995. WordNet: A Lexical Database for English. Communications of the ACM, 38(11):39–41.

[Penas2002] Anselmo Penas. 2002. Website Term Browser. Un sistema interactivo y multilingüe de búsqueda textual basado en técnicas lingüísticas. Ph.D. thesis, Universidad Nacional de Educación a Distancia (UNED).

[Resnik and Yarowsky1999] Philip Resnik and David Yarowsky. 1999. Distinguishing Systems and Distinguishing Senses: New Evaluation Methods for Word Sense Disambiguation. In Journal of Natural Language Engineering, volume 5(2), pages 113–134.

[Vossen1997] Piek Vossen. 1997. Eurowordnet: A multilingual database for information retrieval. In Proceedings of the DELOS workshop on Cross-language Information Retrieval.