Can Predicate Lexicalizations Help in Named Entity Disambiguation?

Position Paper

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Abstract. Most named entity disambiguation approaches use various resources, such as surface form catalogues and relations of entities in the target knowledge base. In contrast, predicates that describe relations between the entity mentions in text are only scarcely exploited. In this position paper, we argue how predicates, i.e., surface forms for relations in the target knowledge base, can potentially help to improve named entity disambiguation results.

Keywords: Named Entity Disambiguation, Ontology Lexicon, Knowledge Base Lexicalization, DBpedia

1 Motivation

The identification of entities in text usually comprises two steps. First, mentions of entities are recognized, which often involves a big amount of ambiguity. For example, the expression Heidi could refer to the model Heidi Klum, the Swiss children book, and so on. Therefore, the recognized mentions need to be disambiguated. This second step is often called named entity disambiguation (NED) or entity linking, as it involves linking mentions to unique identifiers in a knowledge base. For example, the entity mention Heidi in a sentence such as Heidi and her husband Seal live in Vegas would be linked to the DBpedia [2] resource dbr:Heidi_Klum, while the same mention in a sentence such as Heidi was written by Swiss author Johanna Spyri would be linked to the DBpedia resource dbr:Heidi, representing the children book.

Named entity disambiguation often uses dictionaries which collect textual surface forms of entities, e.g. mapping the forms New York, NCY, and Big Apple to the DBpedia entity dbr:New_York_City. In many cases, also co-occurrences and relations between entities are taken into account for disambiguation. For example, in the sentence Cairo was the code name for a project at Microsoft from 1991 to 1996, the co-occurrence of Cairo with Microsoft allows to link it to the operating system instead of the Egypt city. However, co-occurring entities are not always sufficient for disambiguation. For example, in the sentence While Apple is an electronics company, Mango is a clothing one,
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the co-occurrence of Apple and Mango does not provide enough context to distinguish between companies and fruits.

To the best of our knowledge, NED approaches usually do not exploit predicates occurring in texts along with entities, such as husband or written by in the Heidi example, or company in the Apple and Mango example. In this paper, we argue that such predicates are actually a helpful source of knowledge to improve NED, especially when little other context is available for disambiguation. We demonstrate this using examples from the KORE50 benchmark [1], exploiting property lexicalizations. For example, for the property spouse, typical lexicalizations are married to, husband of, and wife of.

Such lexicalizations can help named entity disambiguation in two respects. First, properties in knowledge bases such as DBpedia often specify domain and range information in their ontologies, i.e., valid classes of entities that can appear in the subject and object position of a statement using that property. This domain and range information can be used to discard NED candidates that are inconsistent with the ontology. For example, consider the following KORE50 sentence:

David and Victoria named their children Brooklyn, Romeo, Cruz, and Harper Seven.

Here, Brooklyn is easily confused with the New York City borough Brooklyn by NED tools. However, taking the predicate children into account, which is a lexicalization of the property child, we can discard this misleading option because the domain and range of child are persons, while Brooklyn is a place.

The second possible use of property lexicalizations is that we can explicitly search for relations between entities in the knowledge base. For example, in the above case, we would already have learned that the mentioned entities (Brooklyn, Romeo, etc.) are persons. Given that we already correctly disambiguated one of the entities, we can use this information to search for entities that stand in the child relation to it. For example, if we already linked David to David Beckham, we can use the DBpedia triple

dbr:David_Beckham dbo:child dbr:Brooklyn_Beckham .

to link Brooklyn to dbr:Brooklyn_Beckham, instead of any other person.

In order to exploit such information, lexicalizations of properties are required. One such collection is DBlexipedia.

2 Predicate Lexicalizations in DBlexipedia

DBlexipedia [7] is an ontology lexicon that connects properties in the DBpedia ontology to common surface forms that express them in a particular natural language, together with linguistic information about their morpho-syntactic properties.

The lexicon published on http://dblexipedia.org is the result of applying the automatic ontology lexicon induction method M-ATOLL [5, 6], which creates ontology lexica in lemon [3] format as follows. It takes as input an ontology and dataset (here, DBpedia) and a dependency parsed text corpus in the target language (here, English Wikipedia). As first step, M-ATOLL retrieves all triples for a given property from the dataset. For example, the results for the property spouse include the triple
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In 1969 the singer Lulu married Maurice Gibb. It searches for predefined patterns in those sentences, in order to extract candidate lexicalizations of the property, such as to marry.

So far, M-ATOLL covers entries that describe transitive verbs (e.g. to cross), intransitive verbs with a prepositional object (e.g. to live in), relational nouns with prepositional object (e.g. capital of), and relational adjectives (e.g. similar to).

3 Preliminary Experiment

To analyze the potential value of property lexicalizations for the NED task, we analyzed the 50 sentences of the KORE50 corpus. We processed each of those sentences using DBpedia Spotlight [4] in the standard settings. Out of the 50 cases, DBpedia Spotlight performed a wrong disambiguation for at least one entity in 37 cases.

Next, we looked at the errors made, and analyzed whether the error could potentially be solved by using information on predicates occurring in the sentence. To that end, we looked up the predicate in DBlexipedia. If we found it as lexicalization of a DBpedia property, we marked the error as potentially solvable if

- a wrongly disambiguated entity had a type which was inconsistent with the respective property’s domain or range, or
- a wrongly disambiguated entity had a direct connection through the found property to the correct entity.

For example, in the following KORE50 sentence, DBpedia Spotlight correctly links Angelina to Angelina_Jolie, but fails to link Jon and Brad to the correct entities.

Angelina, her father Jon, and her partner Brad never played together in the same movie.

The predicate father, however, can be found in DBlexipedia as lexicalization of the property child, which links Angelina_Jolie to Jon_Voight, the correct linking for Jon.

Similarly, in the following sentence, both Hurricane and Desire are incorrectly linked by DBpedia Spotlight.

Dylan performed Hurricane about the black fighter Carter, from his album Desire.

Here, the predicate perform is found as lexicalization of the property musicalArtist with domain Single and range MusicalArtist, which helps disambiguating Hurricane to the Bob Dylan single. Furthermore, album is found as lexicalization of the property artist, which relates Bob_Dylan with his album Desire.

Also the mention John in the following sentence is incorrectly linked.

Pixar produced Cars, and John directed it.
But it could be correctly linked using the lexical knowledge from DBlexipedia that \textit{direct} expresses the property \texttt{director}, and the factual knowledge from DBpedia that the correctly identified movie \texttt{Cars} stands in the \texttt{director} relation to the entity \texttt{John Lasseter}.

In total, in 17 out of the 37 cases where DBpedia Spotlight performed a wrong disambiguation, the error could have been identified with either one of the two strategies.\footnote{However, there may be more than one error in the sentences, and in some cases, we would not be able to address all of those. Hence, this should not be misread as “half of the errors can be identified.”} These 17 cases are distributed quite equally across domains:

- CEL (Celebrities): 4/8
- MUS (Music): 4/8
- BUS (Business): 1/8
- SPO (Sports): 4/7
- POL (Politics): 4/6

In addition, we can identify cases where the proposed approach cannot help. First, it can happen that a predicate is not contained in DBlexipedia. For example, neither \texttt{drop out} nor \texttt{join} are listed as lexicalizations of any property, so they cannot be used for disambiguating \texttt{Steve} in \texttt{Steve dropped out of Stanford to join Microsoft}.

Second, it can happen that a lexicalization is found but either does not point to the correct property, or the corresponding triple in DBpedia is missing. For example, in the phrase \texttt{Steve, the former CEO of Apple}, DBlexipedia does list \texttt{CEO of} as lexicalization of the property \texttt{keyPerson}, but in DBpedia \texttt{Steve Jobs} is related to \texttt{Apple Inc.} by means of \texttt{board} and \texttt{occupation}.

Third, there are sentences without an explicit predicate between entity mentions, as the following one:

\begin{quote}
Steve, Bill, Sergey, and Larry have drawn a great deal of admiration these days for their pioneering successes that changed the world we live in.
\end{quote}

Analogously, there are sentences that contain predicates but the expressed relation is not modelled in DBpedia. For example, the sentence \texttt{M"uller scored a hattrick against England} contains the predicate \texttt{score against}, which does not correspond to any property in DBpedia. Similar cases affect predicates that are modeled through more complex constructs, such as property paths or reifications.

\section{Conclusion}

This preliminary experiment shows that predicates, i.e. natural language lexicalizations of properties in the knowledge base, are a valuable source of knowledge when trying to improve the results of NED in cases where only little context is available for disambiguation. Although a formal evaluation on an actual implementation is still missing, the findings from the experiments are quite promising.
References


