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Semantic Relations for an Oral and Interactive Question-Answering System

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Abstract. RITEL is an oral and QA dialogue system, which aims to enable a human to refine his research interactively. Analysis is based on typing of chunks, according to several categories: named, linguistic or specific entities. Since the system currently obtains promising results by using a research based on a common presence of the typed chunks in the documents and in the query, we are expecting better score quality by detecting semantic relations between the chunks.

After a short description of the RITEL system, this paper describes the first step of our research to define and to detect generic semantic relations in the user's spoken utterances, by coping with difficulties of such an analysis: necessary quickness, recognition errors, asyntactical utterances. We also give the first results we have obtained, and some perspectives of our future works to extend the number of detected relations and to extend this detection to the documents.

1 Introduction

Semantic level processing is currently considered a promising way to significantly improve the performance of Question-Answering (QA) systems.

In this paper, we present the first steps of a work which aims to automatically detect semantic relations in user queries within the framework of RITEL, an oral and interactive QA system.

After a short description of RITEL and its current utterance analysis (section 2), we describe our objectives and the methodology we have chosen to represent and detect semantic relations in the specific context of that QA system (section 3). Then we describe what is currently implemented of this detection and give some prospects for our future works (section 4).

2 The RITEL system utterance analysis

The RITEL project aims at integrating a spoken language dialogue system and an open-domain information retrieval system in order to enable a human to ask a general question and to refine his research interactively. In such a system, there are two essential requirements: overall speed should be very fast, and vocabulary should be possibly unlimited.

The dialogue and QA modules of that system are in part based on *non-contextual analysis* the aim of which is to extract, from both user utterances and documents, what is considered to be *pertinent information*. The output of the analysis is twofold: a chunking which groups together series of words with coherent meanings and the typing of these chunks. The types can belong to several categories: named entities (person, location, time, organisation...), linguistic entities (e.g. verbs, prepositions), or specific entities (e.g. scores, colors...). Specific types of entities were added to improve communication management (e.g. `<Dopening>hello</Dopening>`; `<Qdial>I'm looking for</Qdial>`), and specific linguistic phenomena such as negation (e.g. `<Qnegdial>I'm not looking for</Qnegdial>` `an <Qneg-info>animal</Qneg-info>` `but for a type of car ; <Qneg-sys>I did'nt ask for that</Qneg-sys>`). There currently are 266 categories. These categories unify NE categories, semantic classes, dialogic markers and PoS tags. On these 266 categories, about 10 are PoS tags. The general objective of these analysis is to find and type the bits of information that can be of use for search and extraction. The system works in intertwining different level of analysis as it has been shown useful in previous works [BRU]. When the system is unable to find the type of a chunk, a backoff solution, PoS tagging, is used. Our QA system, adapted to English, has participated with success to the QAST track (Question Answering on Speech Transcriptions [TUR] of the CLEF 2007 evaluation [ROS]). The analysis is robust to spoken language, including automatic speech recognition output, and written language. The mean time per utterance or sentence is roughly 4ms. Within a dialogue, the history of the dialogue is used to complement that NCA result to obtain an in-context representation of the user utterance. Anaphora and ellipsis are handled at that level. This process is fully described in [SCH]. Figure 1 gives a classification of the used tags. Figure 2 shows an example of such an analysis.

3 Objectives and methodology

The RITEL QA system is entirely based on the presence in the documents of typed chunks common with the query. Candidate answer scoring is computed from the proximity with these chunks. Our hypothesis (shared with other authors [CUI,BUC]) is that adding relations between chunks will give us a better score quality than simple proximity. Two main classes of relations exist: syntactic and semantic. We chose the latter. Ultimately, the scoring should be improved by unification between the relations found in the query and the documents.

3.1 Related work

Many works related to the detection of semantic relations in QA systems are based on the robust syntactic analysis of whole sentences, for example [NAR,PRA], which aims at detecting of semantic roles. Such robust syntactic parsing systems

named entities	<org> <i>NIST</i> </> <eve> <i>festival Cannes 2007</i> </> <cit> <i>veni vidi vici</i> </>
indistinct entities	<Eve> <i>Cannes festival</i> </> <i>the</i> <Pers> <i>president</i> </>
extended entities multi-levels	functions, titles (bishop, president, professor, ...) colors, animals...
hierarchical super-classes	bishop → religious function → hierarchical function
thematic markers	<literature> <i>novels</i> </> <sport> <i>tennis</i> </>
inquiring markers	<Qqui> <i>who</i> </> <i>has ...</i> <Qmeasure> <i>How many</i> </> <i>days</i>
interaction markers	<DA_close> <i>goodbye</i> </> <DA_yes> <i>yes please</i> </>
compounds	<NN> <i>data base</i> </>
verbal chunks	<i>they</i> <action> <i>take part</i> </> <i>to...</i>
linguistic entities	<stat_objet_plus> <i>the biggest</i> </> <i>exporter</i> <i>it</i> <adv> <i>often</i> </> <i>occurs</i>

Fig. 1. Entity types

<_Qneg_dial> je ne veux pas d' informations </> <_neg_info> <_prep> sur </> <_pers> Benedetti </> </> <_Qdial> je voudrais <1> une </> information sur </> <_det> le </> <_range_objet> dernier </> <_prix> <_Prix> prix Nobel </> <_type_prix> de la paix </> </>
--

Fig. 2. Annotation of a user utterance: *je ne veux pas d' informations sur Benedetti je voudrais une information sur le dernier prix Nobel de la paix* (I am not looking for information about Benedetti, I want information on the last recipient of the Nobel Peace Prize)

are efficient, and have already proven their worth in the domaine of information extraction from text documents [HAG,AIT]. Other works have shown that pattern matching is also a good way to extract some semantic relations [RAV].

Spoken data is different from textual data in various ways: it contains disfluencies, false starts, speaker corrections, truncated words. The grammatical structure of spontaneous speech is quite different than for written discourse. Moreover, automatic speech recognition introduces errors. These factors make parsers designed for written text unsuitable for that kind of input.

Aït-Motkhar and his colleagues argue that an incremental parsing approach allows to design deep language parsing while preserving robustness [AIT]. Our approach can be viewed as an adaptation of such an approach to spontaneous spoken language parsing. A similar approach has been already used in a previous work [VIL].

3.2 Representation and detection of the semantic relations

We have chosen to represent semantic relations as logical formula over the chunks on the annotated utterance. This formula is built from predicates with arguments linked with logical connectives. The name of the predicate gives the nature of the relation and the arguments refer to a list (which can potentially be empty) of chunks of the utterance. To be useful, the predicates should represent a reasonably simple and generic semantic relation. This will help unifying queries and documents which a larger number of more specific predicates would impede. The detection of the relations is currently split into three steps. They are based to both the types of the chunks and a direct detection of some words or of their syntactic categories. All of these steps are so heavily dependent on the previously detected entities.

1. **Grouping of some of the chunks into syntactic groups**, in particular nominal and verbal phrases. Reducing the number of chunks that must be dealt with makes the following steps easier; moreover, linking grammatical words to the content word to which they are linked is a way towards syntactic or semantic disambiguation. Examples of syntactic groups are given in Figure 3 (prepositional nominal phrase, <PNG_on>) and in Figure 5 (verbal phrase, <VG_SA>, where P means *past* (verb tense) and A means *active* (verb mood)).
2. **Detecting local predicates from syntactic and semantic rules.** These predicates specify semantic links between two or three consecutive chunks. Their detection is almost always triggered by an annotated type. The other cases start from syntactic clues.

For example, the predicate *rank_of(Arg1, Arg2)* specifies that *Arg2* is the rank of *Arg1*. That predicate triggers the <range_objet> tag, which marks chunks such as “*last*”, “*the first two*” and so on. The predicate links such tags to the associated nominal phrase.

There are about twenty such relations. Some specify the object of the query, as in the example of Figure 5, where the relation *number_of* specifies which chunk is related to the “*how many*” question. Others expose a semantic link between two tags: for example, the relation *type_of* links the two tags of the following phrase:

```
<GN> <det> the </det>
      <topic_cinema><pers_act><Acinema>
        director </Acinema></pers_act></topic_cinema></topic>
</GN>
<pers> Fritz Lang </pers>
```

These local relations include an initial detection of coordinations (*and*, *or*, etc.). These coordinations link chunks with identical tags. In the example of Figure 3, the scope of the coordination *and* has been detected from the presence of the same tag (<*pers*>), which marks the main chunks of the previous and following nominal phrases. For very simple queries, for instance

requesting a country's capital or a date of birth, these local relations are sufficient to represent the whole semantic relation.

0	<Dneg> no it's not </Dneg> <PNG_on>	1
1	<prep> about </prep>	2
2	<neg_info> <pers> Fritz Lang </pers> </neg_info> </PNG_on>	3
3	<conj> but </conj> <PNG_on>	4
4	<prep> about </prep>	5
5	<pers> Freud </pers> </PNG_on>	6

Coordinations = and(not(1-2),4-5)

Fig. 3. Example of coordination of chunks

3. Detecting global predicates.

Global relations aim at modeling patterns of generic queries. Their detections are based on both semantic and syntactic clues. Four predicates are currently defined and tested:

- **geo_record** involves queries such as “*what is the highest mountain of Soudan?*”. The predicate has three arguments: the type of the record (*highest*), the object of the record (*mountain*) and its domain (*Soudan*). The question mark in an argument specifies an association to the user query (*mountain*). The search for such a relation is initiated by the tag <stat_objet_plus> (typing chunks which express a superlative) joined to a geographical tagged object as a mountain, a river, etc. Figure 4 shows the analysis and the formula related to this query.

0	<Qquel> what is the </Qquel>	1
1	<stat_objet_plus> highest </stat_objet_plus>	2
2	<loc> <montagne> mountain </montagne> </loc> <PNG_of>	3
3	<prep> of </prep>	4
4	<loc><pays> Soudan </pays></loc>	5
5	</PNG_of>	6

Formula = geo_record(1-2, 2-3?, 3-5)

Fig. 4. Example of global relation

- **record** is a generalization of the previous. Its detection is tied to the tag `<stat_object_plus>`, when search for *geo_record* has failed.
- **to_create** involves the creation of an artistic work or a concept, or the discovery of an object. The three arguments are: the creator, the created object and the circumstances.
In the example in Figure 5, only the first two arguments are instantiated: the first one (*Fritz Lang*) is the creator. The second one (*how many movies*) is associated to the created object and to the user’s query. The third argument is uninstantiated. In this example, the detected predicates are linked with the logical connective *and*.
- **to_receive** has four arguments: the first is the person who receives (if any), the second is the object which receives (if any), the third is the reward received and the last the circumstances. For example, for the query: *what is the movie which won the palme d’or in Cannes in 2001*, the first argument is uninstantiated, the second is *movie*, the third is *palme d’or* and the last is the list {*Cannes, 2001*}.

0	<code><Qdial> I want to know </Qdial></code>	1
1	<code><Qnombre> How many </Qnombre></code>	2
2	<code><topic><topic_cinema><Tcinema></code> <i>movies</i> <code></Tcinema></topic_cinema></topic></code>	3
3	<code><pers> Fritz Lang </pers></code> <code><VG_PA></code>	4
4	<code><aux> has </aux></code>	5
5	<code><action> written </action></code> <code></VG_PA></code>	6
6	<code><conjc> and</conjc></code>	7
7	<code><action> directed </action></code>	8

Detected relations:

`number_of?(2 – 3) and and(4 – 5, 7 – 8) and to_create(3 – 4, 1 – 2?, _)`

Fig. 5. An example of detected semantic relations from a user utterance: “*I want to know how many movies Fritz Lang has written and directed*”.

4 Present and Future works

For now, we only have implemented the detection of the described relations in user queries. For performance reasons, all these searches are based on exploration of n-tree structures, and are written in C. Figure 6 shows the precision and recall, related to the detection of global relations; they are calculated from about 2500

user utterances extracted from the RITEL corpus [ROC]. These results are only indicative, since an independent annotation of the corpus related to the semantic relations is not available.

Relation	Precision P	Recall R	F_measure $2RP/(R+P)$
geopol	99%	90%	94%
to_create	93%	62%	74%
geo_record and record	98%	87%	92%
to_receive	Insufficient data		

Fig. 6. Number of detections for global relations. The local relation *geopol* encodes information requests about a country’s capital.

The described relations are very frequent in our corpus. The relations of figure 6 are detected about 650 times in 4948 user utterances of the corpus, including dialogue interaction utterances, where there are no semantic relations to be detected. The dialogue interaction utterances are those only containing opening, closing and rejection markers. Still, the coverage is obviously insufficient.

First we will have to expand the kind and number of detected relations in the user utterances. We will also have to apply this detection to the documents in order to test the contribution of the approach to the QA system. While this detection may be based on the same approach than for spoken documents, it can also use others methods such as the ones used for extracting informations from texts (patterns matching for instance, see 3.1). We plan to test them by detecting semantic relations in biographies. Of course, we will also have to integrate semantic relations into the history of the dialogue.

Otherwise, while keeping simple generic predicates helps to unify the semantic relation between queries and documents, it is in no way sufficient. For instance, *Henry IV was murdered by Ravaillac in 1610* can be represented by the logical formula

$$kill(Ravaillac, Henry IV, date = 1610)$$

while the query *When did Henry IV die?* can be represented as

$$dead(Henry IV, date =?).$$

Without knowledge of the world, the information does not unify with the question and no answer is found. So we plan to add such knowledge under the form of deduction rules as has been done for the very simple example given here:

$$\forall X \forall Y (kill(X, Y, Z) \Rightarrow dead(Y, Z))$$

Our project is quite ambitious, perhaps even a little too ambitious, but given that the QA system already works as-is, any additional semantic information

we manage to add can only help the performance. As such, we will be able to directly evaluate the contributions of our approach and ideas.

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