SIABO

Semantic Information Access through Biomedical Ontologies

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Abstract: The scientific aim of the project presented in this paper is to provide an approach to representing, organizing, and accessing conceptual content of biomedical texts using a formal ontology. The ontology is based on UMLS resources supplemented with domain ontologies developed in the project. The approach introduces the notion of 'generative ontologies', i.e., ontologies providing increasingly specialized concepts reflecting the phrase structure of natural language. Furthermore, we propose a novel so-called 'ontological semantics' which maps noun phrases from texts and queries into nodes in the generative ontology. This enables an advanced form of data mining of texts identifying paraphrases and concept relations and measuring distances between key concepts in texts. Thus, the project gains its identity in its attempt to provide a formal underpinning of conceptual similarity or relatedness of meaning.

1 INTRODUCTION

Search in texts is progressing beyond conventional keyword search in order to make it less syntactic and more semantically oriented. This paper presents endeavours in the SIABO project aiming at achieving content-based text search within the application area of biomedicine.

Our main thesis is that a content-based search functionality can be achieved by computerised text analysis using ontologies enhanced with domain models and language processing.

The remainder of this section describes the aims of the SIABO project in general, section 2 introduces the notion of 'generative ontology', section 3 presents the kind of domain modelling carried out in the project, section 4 sets out two approaches to concept extraction which we are currently testing, one synthetic, and the other pattern-based. Section 5 addresses the problems related to querying information and knowledge, and finally, in section 6, we present our conclusions.

1.1 The SIABO Project

The aim of the SIABO project is to provide an approach to representing, organizing, and accessing the conceptual content of biomedical texts using a formal ontology.

In order to be competitive, companies need to have access to the contents of the increasing amount of documentation about their products, processes and projects. Retrieval of information and knowledge from huge, diverse resources is vital, and only a semantics-based approach to information management is adequate to that task.

This project presents an approach in which the meaning content of each document is described as a set of arbitrarily complex conceptual feature structures facilitating detailed comparison of the

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content of documents. The properties of an ontology-based system lead to easier access to data sources, locally as well as globally.

Ontologies are formal tools for structuring the concepts of a scientific domain by means of relationships between concepts, e.g., along the specialization/generalization dimension. The SIABO approach introduces the notion of generative ontologies, i.e., infinite ontologies providing increasingly specialized concepts. The project sets up a novel, so-called 'ontological semantics', which maps the conceptual content of phrases into points in the generative ontology. Text chunks with identical meaning but different linguistic forms are to be mapped to the same node in the generative ontology. Thus, the approach facilitates identification of conceptual relationships paraphrases. and measurement of distances between key concepts in texts. The project focuses on ontological engineering of biomedical ontologies applying lattices and relation-algebras, and has clear affinities to contemporary research in the Semantic Web area, description logic as well as XML approaches. However, it gains its distinct innovative scientific profile by means of the above-mentioned notions.

2 GENERATIVE ONTOLOGY

A generative ontology is based on a finite ontology with the *isa* concept inclusion relation (called the 'skeleton ontology'), enriched with a set of semantic relations providing generativity. For instance, the skeleton ontology may specify the inclusion path:

insulin secretion isa secretion isa process isa event

A generative ontology is to be understood as a non-finite set of concepts. We move from finite ontologies to infinite systems of concepts, thereby reflecting the recursive productivity of the phrase structures in natural language. This makes it possible to map complex linguistic structures into correspondingly complex concepts associated with nodes in the ontology.

Semantic relations provide feature structures such as *disease*[*CausedBy*: *lack*[*WithRespectTo*: *insulin*]], which corresponds to linguistic forms found in a text or a query, such as *diseases caused by insulin lack, diseases induced by insulin deficiency, insulin deficiency disease*, etc.

2.1 Concept Feature Structures

Concept feature structures are recursive structures, taking the following form:

$$c[r_1:c_1, r_2:c_2, ..., r_n:c_n]$$

where *c* is a concept from the skeleton ontology, and $r_1, r_2, ...$ are semantic relations, and $c_1, c_2, ... c_n$ are concept feature structures. Note that an atomic concept is also a concept feature structure.

The attributions (feature-value pairs) $[r_1:c_1, r_2:c_2, ...]$ consist of relations and concept arguments, and function as conceptual restrictions on the head concept c. This means that $c[r_1:c_1]$ is always situated below the node c in the ontology. In this way, new paths stretch towards more specialised concepts in the ontology. However, the generative ontology does not admit arbitrary combinations of relations and concepts: The relations function as case roles (cf. Fillmore, 1968) expressing ontologically admissible ways of combining concepts, according to so-called 'ontological affinities'. Currently, logical affinities are specified as triples $\langle c', r, c' \rangle$. The affinities are specified so as to rule out category mistakes. In our context of ontologies for scientific texts within biomedicine, we concentrate on physical-chemicalbiological categories and disregard metaphors.

3 DOMAIN ONTOLOGIES

As a validated fragment of the generative ontology, we construct domain ontologies supplementing and refining already existing ontologies for the domain, such as UMLS (Unified Medical Language System). Validated domain ontologies are needed because UMLS is not specific enough as regards concepts and concept relations, and in many cases, the existing resources are imprecise. By adopting the principles of terminological ontologies and by consulting domain experts, we arrive at validated concept structures. As the basis of the domain ontologies, a small text corpus has been used to produce a list of term candidates, and in cooperation with a domain expert, central concepts have been identified. Furthermore, UMLS resources have been consulted.

3.1 Terminological Ontologies

In this project, the domain ontologies are terminological ontologies, i.e., their structure is based on characteristics and subdivision criteria Madsen et al., 2005), and we use an extended set of



Figure 2: Extract of the domain ontology of insulin production.

concept relations (Madsen et al., 2002).

Terminological ontologies are not strictly speaking formal ontologies but may be transformed into such. The graph in Figure 1 is thus a transformation into concept feature structures of the concepts in Figure 2.

In the generative ontology, the concepts are represented by concept feature structures, such as for example: *secretion*[*RES: insulin*]. In terminological ontologies the concepts are represented by terms (linguistic expressions), e.g. *insulin secretion*, and feature specifications expressing characteristics of the concept are introduced on each concept, e.g.:

RESULT: insulin on the concept *insulin secretion*. On the basis of concept relations and characteristics, the concept feature structures of the generative ontology may be generated, e.g. the concept *insulin secretion* has the superordinate concept *secretion* and the characteristic *RESULT: insulin* resulting in the concept feature structure: *secretion*[*RES: insulin*]. The two representations are closely interlinked in that any terminological representation can be translated into a generative representation expressing the semantic content for each concept in the terminological ontology.

3.2 Ontology of Insulin Production

In Figure 2, we present an extract of one of the resulting ontologies. Boxes with text in capital letters represent subdivision criteria, the other boxes represent concepts. The lines without arrows represent *isa* relations, the arrow lines represent other relations. Characteristics are given in the form of feature specifications below the concept boxes.

Based on an analysis of the characteristics of the concepts *stimulation* and *inhibition*, these concepts were grouped under the subdivision criterion *INFLUENCE*, where the distinct characteristics clearly show the difference between them. Where appropriate, concepts have been mapped to UMLS in order to obtain one coherent ontology.

4 EXTRACTION OF CONCEPTS FROM TEXT

The SIABO project investigates two approaches to concept extraction in parallel: a synthesis-approach relying on the generative ontology, and a patternbased approach which relies on knowledge extracted from a variety of lexical resources. The two approaches are described below.

4.1 The Synthesis Approach

The computerised text analysis employed in this approach is conducted chiefly by the generative ontology assisted by conventional grammars. Ideally, a sentence is turned into one concept feature structure in the generative ontology, which is supposed to represent the ontological meaning content of the sentence. This is in contrast to other approaches to the characterization of propositional content which take into account determiners, negations, and logical conjunctions (e.g., Nirenburg and Raskin, 2004).

The ontology-driven rather than syntax-driven text processing is performed by a so-called 'ontograbber', which, in principle in a top-down manner, generates feature structures to be matched against the target sentence in the text. However, since, in general, parts of sentences have to be skipped as unrecognisable, the current ontograbber prototype conducts a bottom-up analysis for synthesising onto-terms according to the generative ontology. In addition, the ontograbber is guided by conventional grammar rules. However, many potential syntactical analyses brought about by structural ambiguities, are never actualised since they are dismissed as category mistakes by the ontological affinities specified on the set of ontological relations.

In this synthesis approach, adjectives and prepositional phrases give rise to feature structure contributions to be attached to the concept coming from the head noun in noun phrases. Verbs are dealt with by nominalisation. Crucially, the ontograbber admits partial, incomplete analysis, which in the worst case falls back on keywords found in a sentence and being present in the generative ontology.

4.2 The Pattern-based Approach

In parallel with the synthesis approach, we explore a pattern-based approach to concept extraction. Like the synthesis approach, this approach allows us to match phrases in text with a view to mapping the conceptual content of these fragments into the generative ontology. The patterns are generated from information available from existing lexical ressources, currently the nominalisation lexicon NOMLEX-plus, the verb-lexicon VerbNet and WordNet. These resources provide syntactic argument realization rules for verbs and their arguments and nominalised forms of verbal expressions with semantic information in the form of semantic roles.

The patterns form part of a process which generates concept feature structures serving as an index for a given text. We thereby move from string or wordbased indexing to semantics-based indexing.

As mentioned, this approach includes nominalisation of relations expressed by verbs. We use this nominalisation strategy for an important reason: The set of well-formed feature structures in the generative ontology is determined by the set of atomic concepts in the skeleton ontology and the chosen set of relations. In order to construct a conceptual feature structure, we need at least one semantic relation. However, there is no natural upper limit to the number of possible semantic relations. Our claim is that conceptual indexing will lead to better results when using a small number of semantic relations. The primary purpose of conceptual indexing is to permit retrieval by way of matching descriptions of queries against descriptions of text. An increased number of semantic relations will lead to an increased number of possible concept descriptions. Due to the unavoidable imprecision in the concept extraction, this increased number of possible content descriptions will in turn lead to a

reduced probability of match between descriptions for conceptually similar but lexically or syntactically dissimilar linguistic expressions, and thereby lead to a decreased recall. Thus, we need a set of relations to express semantics, but we aim at keeping this set at a manageable size in order to obtain the best possible match.

5 QUERYING INFORMATION AND KNOWLEDGE

Given a domain ontology as shown above and a set of documents in which concepts have been identified, the task is to provide means for query interpretation and evaluation that draws on conceptual content and exploits the conceptualisation in the ontology.

In the present approach, query evaluation relies on comparison of a conceptual description of the query with conceptual descriptions of texts from the database. A conceptual description is a set of conceptual feature structures providing a mapping from the text or the query to the ontology. Search in a text collection indexed by concepts can employ concept similarity-measures so that conceptual reasoning can be replaced by simple similarity computation, thereby allowing for a scaling to very large information bases. Thus, a major challenge is to define conceptual description similarity in terms of the structure and relations in the ontology.

One obvious way to measure similarity in ontologies is to evaluate the distance in the graphical representation between the concepts being compared, where shorter distance implies higher similarity. A number of different ontological similarity measures have been proposed along these lines, for instance, Shortest Path Length (Rada, 1989), Information Content (Resnik, 1999), see also (Budanitsky and Hirst, 2006).

An essential part of document querying is to establish a mapping that, given a description for the query, indicates matching – or similar – descriptions for texts. One option is to let similarity reflect the skeleton ontology by deriving it from the syntactic derivation relation for conceptual feature structures, where longer derivation paths correspond to smaller degree of similarity. However, the comparison of conceptual descriptions should not be merely syntactic. Rather, description resemblance can be measured in terms of similarity derived from all concept relations in the ontology. Initially, in the processing of a query, a description is generated. Then this query description is compared, in principle, to every conceptual description of every document appearing in the database. Finally, documents are ranked by the degree to which their respective descriptions resemble the conceptual description of the query. The query answer is a ranking of the documents that are most similar to the query.

In a framework where the domain of texts is reflected in a knowledge base, as comprised by the ontology, obviously not only the texts, but also the domain ontology may in some cases be the target of interest for queries. Knowledge about existence of concepts, how concepts are related and about similarities between concepts is also relevant. In addition knowledge about the actual content of texts can be viewed through the ontology simply by means of revealing only concepts that exist in the texts. In other words, the ontology plays a specific role here, since it constitutes the means by which we can obtain a conceptual view of the texts content.

Thus as an additional functionality, the user may browse the generative ontology directly and then follow the links to the relevant text parts by descending to an ontological level of specialisation with a manageable number of links to the target text.

6 CONCLUSIONS

We have presented an approach to representing, organizing, and accessing conceptual content of biomedical texts using a formal ontology. In particular, we have presented the key ideas addressing exploitation of ontologies for carrying out content-based text search within a scientific domain recognising not only synonyms but also more general paraphrasations. Presently, we have working prototypes. However, the viability of the approach remains to be validated on a large scale, in particular whether the devised ontological text processing afford prototypes а significant improvement compared with conventional keyword search.

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