

Semantic enrichment of user-generated educational scenarios with spatial concepts and entities

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Abstract

The paper presents ongoing work on how to semantically enrich user-generated content of an educational authoring environment with spatial concepts and entities. The main purpose of such an enrichment is to improve search results for content. Analysis of existing user-generated educational scenarios has shown that although they contain a wealth of spatial concepts and entities, the search mechanisms supporting keyword-, map-, and graph-based search return positive searches only for a limited number of concepts and entities based on those attached to scenarios which are very limited as opposed to what is mentioned in the content. The methodology makes use of NLP techniques and open tools and exploits the environment’s semantic network of spatial thinking concepts to further annotate user-generated educational scenarios. Even though, the work is in its initial phases, results indicate that the approach has potential in improving semantic search and discovery.

Keywords: semantic enrichment, spatial concepts, user-generated content, semantic search, semantic metadata.

1 Introduction

XXX is an online, collaborative repository and authoring environment for the collection, creation, and sharing of educational resources for enhancing spatial thinking developed in the framework of a European project through an innovative ICT-based approach.

Resources are either simple educational objects (such as documents, images, videos, webpages, etc.) or more elaborate educational scenarios that describe a whole lesson plan or educational activity. Currently the platform has 714 registered members who have contributed in different ways to the development of 437 resources.

A semantic network of 327 concepts and 802 taxonomic relations supports the development of educational scenarios in 6 languages. The semantic network supports semantic organization of spatial knowledge but also enables the graph-based search of educational resources. Furthermore, it supports the visualization of scenarios by highlighting concepts (nodes) of the overall network present in the scenario.

Another functionality of the platform is the visualization of educational scenarios in geographic space. This functionality is supported by dynamically drawing 2,158,751 geographic entities from GeoNames geographical database which may be attached by users to scenarios as instances of certain concepts (e.g. country, lake, city etc.).

The platform provides various possibilities for searching the content, such as keyword-search, metadata search (based on metadata provided by the authors such as target audience,

language, discipline, etc.), map-based search (based on geographic entities attached to a scenario), and graph-based search (based on the spatial concepts attached to a scenario).

The ability to attach spatial concepts and instances to educational scenarios is an important feature of the platform because it supports the semantic and map-based search and visualization of content. However, this feature was not widely used by users limiting thus the potential for advanced search and visualization of the scenarios in semantic and map space. Although various spatial concepts and entities are mentioned in the majority of educational scenarios, users have not actually attached these in their scenarios for allowing the subsequent semantic and map visualization. For example, a semantic search on “geological phenomena in the Mediterranean” would ideally return results about “deforestation in Cyprus” or “degradation in Greece”.

The semantic enrichment of user-generated scenarios would greatly improve the search and discovery capabilities provided by the system, for example by supporting the following:

1. Search for scenarios based on multiple concepts to provide more specific answers to the users’ search
2. Identification of similar scenarios based on the sets of concepts treated therein.
3. Linking of related/similar content
4. Enrichment of the search results by also taking into account the relations among concepts. For example, equivalence relations may provide for identifying relevant content using synonymous terms that refer to the same concepts. Hierarchical relations may contribute in expanding or limiting the exploration of relevant content.

The remainder of the paper is organized as follows. Section 2 reviews relevant literature regarding semantic enrichment and Section 3 outlines the elaboration of the existing semantic infrastructure into an ontology. Section 4 presents the method for the semantic enrichment of user-generated educational scenarios and Section 5 draws conclusions and discusses future directions.

2 Related work

Semantic enrichment aims at enhancing content interlinkage, search, and discovery by adding well-defined semantic metadata that help machines make sense of the content and reveal latent relations. Semantic enrichment is used for information organization and retrieval, semantic search and knowledge discovery, ontology development and population.

Semantic enrichment has been used to add semantic metadata to different types of content, such as unstructured documents (Pernelle, 2016), maps (Hu et al., 2015), images (Ennis et al., 2015; Tardy et al., 2016), and video (Nixon et al., 2013). Semantic enrichment mainly focuses on extracting entities and concepts; however, there are approaches targeting at extracting semantic metadata for more specific types of features such as events (Romero and Becker, 2016), movement data (Fileto et al., 2015), and places (Alves et al., 2009; Tardy et al., 2016).

The paper focuses on the semantic enrichment of user-generated educational scenarios describing lesson plans for enhancing students’ spatial thinking abilities. These include both unstructured and structured parts. Unstructured parts are documents describing the lesson plans whereas structured parts include tags, metadata, concepts, and geographic entities attached to the scenarios. The work aims at extracting semantic information from the unstructured parts scenarios in order to connect it with the structured parts, primarily with concepts and geographic entities. Text enrichment with geographic entities is more straightforward supported by semantic annotation tools and gazetteers. On the other hand, semantic enrichment of unstructured text with concepts is not a straightforward process, since existing semantically-enabled tools are based on their own underlying ontologies and do not support the extraction of specialized concepts.

For example, the sentence:

“Across your country (in east-west direction) measure the time of sun rise and/or sun set. Investigate Earth rotation and the origin of day and night on Earth”

includes spatial concepts such as: country, east, west direction, rotation and Earth. Figures 1 and 2 show the semantic enrichment of this sentence as performed by DBpedia Spotlight (Daiber et al., 2013) and Open Calais (Butuc, 2009) respectively. DBpedia Spotlight annotates the sentence with the concept Earth, whereas Open Calais does not identify any semantic annotations in this sentence.

3 From semantic network to ontology

A set of 327 spatial concepts with definitions derived from WordNet (2010) were organised in a three-level hierarchy that reflects clusters of basic notions and subject areas such as

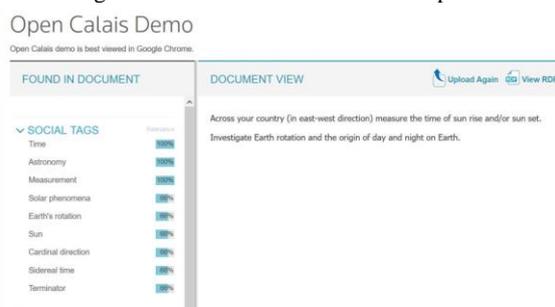
geometric primitives, spatial relations, space-time primitives, geography etc. Four major clusters form the upper level hierarchy, further dived into 15 clusters that represent the second level and finally the 327 concepts are organised within them. WordNet synsets were scrutinized to identify lemmas and definitions that better served the spatial thinking domain.

Figure 1. Semantic annotation with DBpedia Spotlight.



Source: DBpedia Spotlight (<http://demo.dbpedia-spotlight.org/>).

Figure 2. Semantic annotation with Open Calais.



Source: Open Calais (<http://www.opencalais.com/>).

In an attempt to keep the structure as simple as possible, concepts were semantically interlinked in a rather “loose” way through 802 unlabelled relations forming a semantic network.

However, a deeper WordNet analysis provided further information that was not originally used for the development of the semantic network but would greatly enhance the platform’s capabilities in terms of knowledge search and discovery. More specifically, the analysis showed excessive interlinkage of concepts in terms of semantic relations (subtype / supertype, part-of / has part, sister terms, etc.) and indicated 26 domains that the 327 concepts belong to. A concept belongs to more than one domains as shown by the analysis. In average, concepts belong to three or four domains, while very few of them belong only to one domain. This finding reveals additional 461 relations among concepts and provides ground for developing an ontology in a more rigid and formal way.

An example shown in Figures 3 and 4 indicates how the semantic network is elaborated once the relations between concepts are explicitly denoted and domains are added to the

structure. Figure 3 shows how the concept Cartesian coordinate is interrelated with other concepts in the semantic network. Straight lines denote the hierarchy of the concepts, while curves denote relations, indicating that certain concepts are somehow interrelated. On the contrary, Figure 4 constitutes an excerpt of the ontology visualization where all relations are explicitly stated and enriched with ontological domains not previously included in the semantic network.

Therefore, in order to semantically enriching user-generated content and improve semantic search and knowledge discovery, an ontology was developed that relates the 327 original concepts with explicit relations among them and domains that they belong to.

Figure 3. Relations currently in the semantic network.

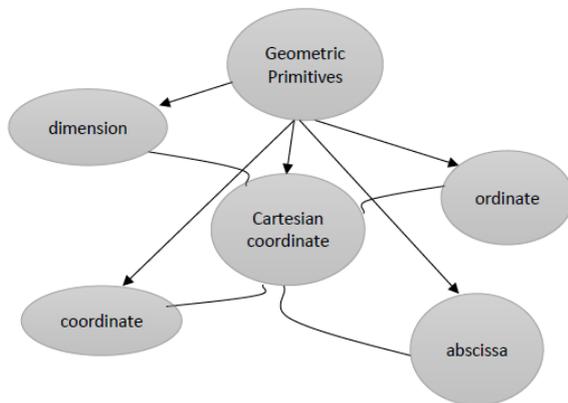
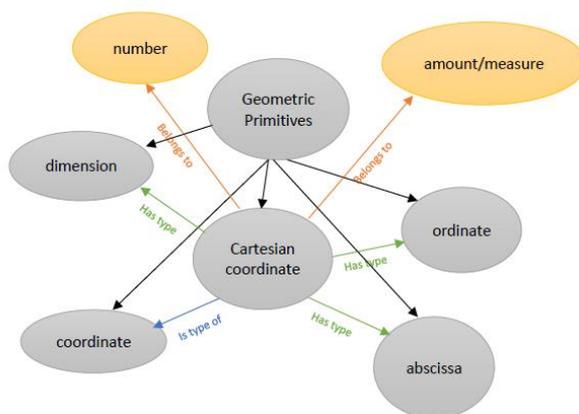


Figure 4. How the ontology changes the semantic network.



4 Semantic enrichment of user-generated scenarios with spatial concepts and entities

An examination of the implemented scenarios reveals that although users have attached concepts and instances to the scenarios (2947 and 1475 attachments of concepts and

instances respectively) they have not fully exploited this potential.

There are various scenarios with less than five attached concepts that mention a multitude of concepts in their descriptions. These free text descriptions could be semantically annotated and enriched with these concepts to enforce search and discovery results. Moreover, an even larger number of scenarios do not have any geographic entity attached to them, although various geographic entities are included in the corresponding free-text descriptions. These geographic entities can be discovered through the platform’s keyword-based search mechanisms, but cannot be retrieved in the map-search.

In what follows, the procedure of such an enrichment is detailed. To perform the task, we used the GATE Information Extraction System (Cunningham et al., 2017) that integrates several NLP systems apart from having its own, ANNIE, as default.

4.1 Enrichment process

The enrichment endeavour comprises of four phases.

4.1.1 Text selection

Text collection constitutes the pre-processing phase which considers user-generated scenarios to spot those with a limited number of concepts that usually have no instances of concepts attached to them as well.

4.1.2 Preprocessing a sample

Once the texts to be enriched have been identified, a sample was used as input to GATE to be tested and analyzed. This formed a corpus of the scenarios involved.

The authoring environment allows downloading of the content in the form of pdf files which when uploaded on GATE present some problems in structure (e.g. connection or repetition of words etc.). Before running the NLP process, the files were processed manually to tackle these issues.

4.1.3 Processing the corpus

For the NLP processing, we ran two tests. First, we used ANNIE, GATE’s default system with its set of NLP tools (tokenizer, gazetteer, sentence splitter, and POS tagger). In a separate session, we used the OpenNLP (2010) package, an open source library for processing texts using machine learning tools developed by the Apache Software Foundation. OpenNLP is fully integrated in GATE and offers an alternative to ANNIE. Both toolkits may be used together in a single processing task with modules taken from each in any combination.

In our case, the NLP processing gave better results using OpenNLP toolkit (Figure 5). OpenNLP identified nearly all named entities in the file (e.g. London, Mediterranean, Thera, etc.) that could serve as instances of certain concepts during

the enrichment phase that follows. OpenNLP has its own models for the following named entities: location, time, organization, person, money, date, percentage. In our case, location is the one that we need the most to identify instances for concepts such as city, country, continent, island, etc.

4.1.4 Enriching the scenario

Upon completion of the NLP processing phase with noun phrases being identified, we are able to annotate concepts and geographic entities in the text by uploading the developed ontology in GATE using the ontology tools plugin.

The ontology was uploaded as an xml file in order to identify concepts in the processed corpus. Furthermore, instances of certain concepts, identified in step 3 as named entities, were annotated as well, populating thus the ontology itself. Figure 6 shows an excerpt of such an enrichment in terms of annotated concepts and instances, while Figure 7 shows an excerpt of the resulting populated ontology.

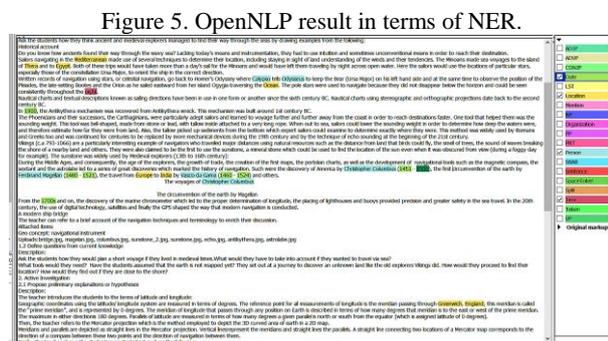


Figure 5. OpenNLP result in terms of NER.

Source: OpenNLP (2010)



Figure 6. Annotation result.

4.2 Use of enriched scenarios for improving semantic search.

The enrichment process ends with an xml file of the annotated scenario. This is then uploaded on the platform and forms an edited (new version) of the processed scenario.

The difference is quite obvious once the semantic networks in each version are examined. Figures 8 and 9 show the semantic networks of the two scenario versions (before and after the semantic enrichment process).

A great advantage is the creation on the platform of an “image” of the existing scenario that has not altered the original content (no content was added or deleted from the original lesson plan), but was only semantically enriched in terms of concepts and instances (Figure 10 shows the instances added in the new version for map-based search).

The enriched version results in a scenario that can contribute to keyword-, graph-, and map-based search capabilities of the platform and allow users to browse more efficiently its content.

Figure 7. Populating the ontology.

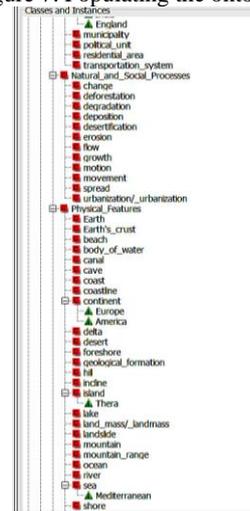
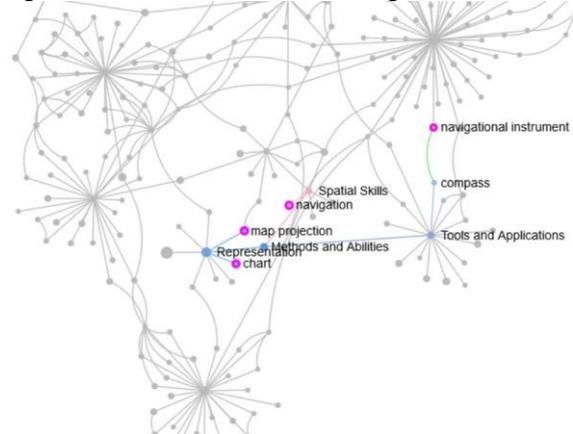


Figure 8. Semantic network of the user-generated scenario.



5 Conclusions

Semantic enrichment is currently an important feature in content retrieval, big data management, semantic web technologies etc.

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