Dynamic Ontology Evolution in Open Environments

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Introduction

Interoperation between knowledge-based systems or agents requires common ontologies to facilitate successful information exchange. However, the openness of the Semantic Web means that the notion of there being common domain ontologies sufficient to cater for the requirements of a diverse range of consumers and producers of services has become untenable. In these types of environments it is necessary to consider that no ontology can be expected to remain unchanged throughout its lifetime. However, the dynamism and the large scale of the environment prevent the use of traditional ontology evolution techniques, where changes are mediated by a knowledge engineer [3]. We argue that the ability to estimate the impact of change a priori, i.e. before performing the change itself, is crucial, since this estimate can be used to assess the usefulness of the change. We assume that agents are capable of rational behaviour, and that they decide whether to change the ontology they commit to if the cost of the change (in terms of reclassification of knowledge) is offset by the benefits derived from the ability of a system to acquire new capabilities and therefore to achieve new tasks (or answer new queries, in the case of knowledge based systems). However, the agent's decision making process follows the principle of bounded rationality [5]: agents operate with limited computational resources, and with partial knowledge of the environment [4]. We present an approach that evaluates the impact of change on an ontology a priori, without using reasoning, by estimating which set of axioms in an ontology is impacted by the change.

Change Evaluation

OWL DL ontologies may undergo modifications for many different reasons; the simplest cases are those in which new assertions are added, or new axioms further detailing the domain knowledge. Another important source of modifications for an ontology comes from alignment with another ontology; the problem of ontology alignment has received wide interest in recent years, and many alignment methodologies and systems have been developed [2], together with methods to evaluate the results of different alignment techniques¹. Not much attention has been devoted to the way a knowledge base should cope with the additions introduced by alignments to another ontology; in an open environment this may cause the knowledge base to grow without control. The way a knowledge base can be affected by alignments to another ontology is illustrated by the following example: Two agents, A_O and A_K , commit to two different ontologies

O and K, which differ on the definition of a specific concept,

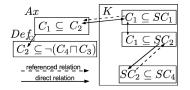


Figure 1: The group rooted at C_1 contains the whole knowledge base K

named C_2 in O, and which is not present in K. By means of an alignment technique, a concept C_1 in K is discovered as the best match for C_2 , with the relation between them being an axiom $Ax = C_1 \subseteq C_2$. This enables agent A_O to issue queries against the knowledge base of agent A_K . By asking for instances of C_2 , the result will contain all the instances of C_1 ; this has the clear advantage of returning valid results to A_O , while issuing the same query without previous alignment would return no results. K may be affected by the alignment: let us consider the case in which the definition of C_2 (Def_{C_2}) is added to K, with Ax acting as a connection between Def_{C_2} and the original K, as depicted in Figure 1; the impact this has on the future performance of K cannot be easily foreseen. The simplest heuristic is based on the assumption that $K \sqcup Ax \sqcup Def_{C_2}$ will behave in the same way as its expressivity class. K has expressivity \mathcal{AL} ; adding Axand Def_{C_2} changes the expressivity of the knowledge base to ALC. In a larger knowledge base, such an increase in expressivity may be restricted to a portion of the knowledge base. This is true for real world ontologies. The impact of a change is measured using the very simple heuristic of the number of axioms and assertions contained in the section of the knowledge base affected by the change, weighed with the expressivity of the knowledge base. An axiom A in the framework represents a DL axiom; let us give an example referring to Ax:

A represents the abstraction over an OWL axiom or assertion, which has a *signature*, and is represented as a node in a directed multigraph whose edges represent *relations* between *axioms*; it has a *signature s* which is the set of named and unnamed concepts and roles mentioned in an *axiom*; A has $s = \{C_1, C_2\}$, and a *main concept* or *main role m* (also called *main node*) which is the concept or role being defined by the *axiom*²; m_A is C_1 in the example.

Relations between axioms correspond to intersections be-

¹http://www.ontologymatching.org/evaluation.html

²An OWL axiom is represented internally as a set of RDF statements; the standard RDF mapping for OWL is defined in http://www.w3.org/TR/owl-semantics/mapping.html

Ontology	Expressivity	Groups	Size
			range
Galen full	\mathcal{AL}	72	1/45
# axioms:	\mathcal{ALC}	14303	61723/63625
82030	\mathcal{ALHIF}	28	13/27
# groups:	$\mathcal{ALR} + \mathcal{HI}$	1	1
37408	ALR + HIF	313	53/1739
avg size:	SHIF	24711	93/81815
9649			
Galen fragment	\mathcal{AL}	552	1/115
# axioms:	$\mathcal{AL}(\mathcal{D})$	441	3/255
9915	\mathcal{ALC}	101	3/151
# groups:	$\mathcal{ALC}(\mathcal{D})$	1257	4/7576
4500	$\mathcal{ALCHF}(\mathcal{D})$	95	16/7904
avg size:	$\mathcal{ALCH}(\mathcal{D})$	29	20/7370
8529	$\mathcal{SHF}(\mathcal{D})$	427	47/9889

Table 1: Expressivity Metrics

tween their signatures and main nodes; they can be: **direct**: a direct relation between an axiom A and an axiom B exists if m_B belongs to the s_A ; two axioms with the same m have a bidirectional direct relation; **indirect**: an indirect relation between two axioms A and B holds when s_A overlaps with s_B (e.g. $D_1 \subseteq \exists R.C$ and $D_2 \subseteq \exists R.D$ share a reference to the role R); an indirect relation is bidirectional; **referenced**: a referenced relation is the inverse of a direct relation; such a relation is implicitly defined also for indirect relation. The relations define three graphs: called O the set of axioms in O, O_d is the graph < O, DR > where DR is the set of direct relations; O_i is the graph < O, IR > where IR is the set of indirect relations; O_r is the graph < O, RR > where RR is the set of referenced relations.

We define a *Group G rooted at an axiom A* as the set of axioms resulting from the union of the sets of axioms S_d , S_i and S_r explored during the exhaustive visit of O_d , O_i and O_r respectively, starting from A and following the relations. **Evaluation**

The grouping framework is implemented in Java; it uses Jena [1] and the SPARQL ³ implementation ARQ⁴ to perform the axiom extraction. The OWL DL reasoning engine Pellet [6] has been used to check the expressivity of a group. Here the results for the Galen ontology translated to OWL⁵ and a fragment of Galen⁶, smaller than the original and often used to test reasoners, are reported. The measures taken into account are: the number of axioms in the ontology, the average number of axioms in a group and the expressivity of each group; for each expressivity level, the number of groups (duplicate groups are counted as one) and the size range are reported in Table 1. Table 2 presents in more detail the results obtained on the Galen fragment ontology. Out of 9915 axioms, more than 4500 groups were computed, by selecting only axioms with named concepts as m. The groups overlap very often, due to the high detail of the ontology, which is has a deep role hierarchy. In the following, Int will be the intersection of the 5 largest groups, labelled G_2 to G_6 , and will denote $G_i \setminus Int$ the set difference between one of these groups and Int. Only two groups are significantly larger than Int, G_5 and G_6 , by 1923 and 2420 axioms respectively. Table 2 reports the size of Int; for each set, then, the size of the difference $G_i \setminus Int$ and its expressivity is reported.

The expressivity of the groups is \mathcal{ALC} for the four smaller

	$\# \ axioms$		Expressivity	
Int	7469		ALC	
$G_2 \setminus Int$	32		\mathcal{AL}	
$G_3 \setminus Int$	19		\mathcal{AL}	
$G_4 \setminus Int$	34		\mathcal{AL}	
$G_5 \setminus Int$	1923		\mathcal{SHF}	
$G_6 \setminus Int$	2420		\mathcal{SHF}	
		I(A) impact value	
$A \in Int$		47089.25		
$A \notin Int: A \in G_2$		6086.75		
$A \in G_3$		6081.75		
$A \in G_4$		6088.75		
$A \in G_5$		1	14252.00	

Table 2: Expressivity of group overlaps and results for impact computation

14580.00

groups, and SHF for the two larger ones; the difference in expressivity is therefore related to roles.

If a change to an ontology is proposed (by the agent's decision making process), we can now assess which part of the ontology the change will impact. For example, if the removal of an $axiom\ A$ from the Galen fragment is the proposed change, the agent can assess if this axiom is located at the intersection of the five largest groups, or if it is located within one specific group. In order to estimate the impact of the change, we propose an impact function that is computed considering the size of the distinct groups containing A:

$$I(A) = \sum_{A \in G} size(G) * expr(G)$$

size(): Given \mathcal{G} the set of all groups G over \mathcal{O} , size(G): $\mathcal{G} \to \mathbb{N}$ is the number of axioms contained in G.

expr(): Given (G) the set of all groups G over \mathcal{O} , expr(G): $\mathcal{G} \to \mathbb{R}$ is the function that computes ec, the DL expressivity of G, and maps ec into a real numeric value.

The values for I(A) are reported in Table 2; on this basis, the decision making process may choose to accept or reject the change involving A.

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³http://www.w3.org/TR/rdf-sparql-query/

⁴http://jena.sourceforge.net/ARQ/

⁵http://www.co-ode.org/galen/

⁶http://www.daml.org/ontologies/400