

# MEDIATION WITHOUT A GLOBAL SCHEMA

## *Matching Queries and Local Schemas Through an Ontology*

Michel Schneider, Damien Thevenet  
*LIMOS, Complexe des Cézeaux, 63170 AUBIERE Cedex*

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Abstract: Approaches by mediation to make multiple sources interoperable were essentially investigated when one is able to resolve a priori the heterogeneity problems. This requires that a global schema must be elaborated or that mappings between local schemas must be established before any request can be posed. The object of this paper is to study to what extent a mediation approach can be envisaged when none of these features are a priori available. Our solution consists in matching a query with each of the local schema by using an ontology of the domain. Such a solution is particularly suitable when sources are liable to evolve all the time. We are investigating this solution by considering the mediation of heterogeneous XML sources. Local schemas are represented in the OWL language. Queries are formulated using an XQUERY-like language. Matching of names is solved by using the an ontology of the domain. We have developed a prototype and conducted a number of experiments to evaluate the capacity of the approach.

## 1 INTRODUCTION

The interoperability of multiple heterogeneous sources represents an important challenge considering the proliferation of numerous information sources both in private networks (intranet) and in public networks (internet). Heterogeneity is the consequence of the autonomy: sources are designed, implemented and used independently. Heterogeneity can appear for different reasons: different types of data, different representations of data, different management software packages. The interoperability consists in allowing the simultaneous manipulation of these sources so as to link the data which they contain. It is necessary to make different sources interoperable in numerous domains such as electronic business, the environment, the economy, medicine, genomics. Interoperability problems occur in very different ways depending on whether sources are structured (data bases), semi-structured (HTML or XML pages), non-structured (any file). The access interfaces also influences the possibilities of interoperability. For example two data bases can be difficult to make interoperable when they are only accessible through specific web interfaces.

One interoperability approach which has been studied for several years is based on mediation (Wiederhold, 1992), (Garcia-Molina, 1997). A

mediator analyzes the query of a user, breaks it down into sub-queries for the various sources and re-assembles the results of sub-queries to present them in a homogeneous way. The majority of mediation systems operate in a closed world where one knows a priori the sources to make interoperable. There are several advantages to this. First it is possible to build an integrated schema which constitutes a reference frame for the users to formulate their queries. Then it is possible to supply the mediator with various information which are necessary for the interoperability and particularly to resolve heterogeneity problems. The different kinds of heterogeneity to be resolved are now clearly identified: heterogeneity of concepts or intentional semantic heterogeneity; heterogeneity of data structures or structural semantic heterogeneity; heterogeneity of values or extensional semantic heterogeneity. Different solutions has been studied and experimented on to solve these problems. For example we can cite the work of (Hull, 1997) and (Kedad, 1999). From these initial investigations, very numerous works intervened to propose automatic approaches of integration of schemas. An approach was particularly investigated: the mapping of schemas. It led to the elaboration of several systems such as SEMINT, LSD, SKAT, DIKE, COMA, GLUE, CUPID. One will find analyses and comparisons of such systems in (Rahm, 2001) or

(Hai Do, 2002) or (Mohsenzadeh, 2005). The practical aspects of the application of such systems are discussed in (Berstein, 2004). The role of ontologies was also investigated. In (Cui, 2001) and (Missikoff, 2004), the interest of ontologies for the semantic interoperability is underlined. Several approaches of integration of information based on ontologies were suggested. One will find a synthesis of it in (Wache, 2001). It is necessary also to quote the work of (Lenzerini, 2005) suggesting a logical frame for the integration of data. In every case, the objective is to build a global schema which integrates all the local schemas.

When one operates in an evolutionary world where sources can evolve all the time, the elaboration of a global schema is a difficult task. It would be necessary to be able to reconstruct the integrated schema each time a new source is considered or each time an actual source makes a number of changes. In this paper we suggest an approach which does not require a preliminary integration of sources schemas but which is based on a matching between the user query and each source schema. The user query is formulated with regard to a domain specified through an ontology. Only the sources whose schemas match with the query are considered. The user query is rewritten for each of these sources according to its information capacity. These sources are then interrogated. Results are formatted and integrated.

This approach offers several advantages. Integration is processed only on the schemas of the results and not on the entire schemas of all potential sources. The rewriting process is simpler.

The paper is organised as follows. In section 2 we give an overall presentation of our approach. Section 3 is devoted to the query language and section 4 to the OWL representation of sources. In section 5 we explain the main features of our matching algorithm. Section 6 is relative to the rewriting of a query. Section 7 is devoted to some experiments with a prototype of the system. Section 8 presents a number of conclusions and perspectives.

## 2 PRESENTATION OF THE APPROACH

The approach which we propose does not use a global schema. The user thus formulates his query by using his implicit knowledge of the domain or by making an explicit reference to an ontology of the domain.

The matcher is the central element of the system. It receives the user query, and has the task to determine if this query can be applied to a data

source (figure 1). To achieve this processing, it possesses a representation of each data source in a common formalism (we propose OWL to support this formalism, cf section 3). It must search for a correspondence between the query and each source by taking into account the terms and the structure of the query. Intuitively, so that a source can answer a query, the terms of the query must correspond to those of the source and the structure of the query must correspond to that of the source.

We propose a query language based on a simplified version of XQUERY. The structure of a query is thus defined through the various paths which appear in the clauses FOR, LET, WHERE. A correspondence is established with a source if each of these paths has a correspondent in the OWL representation of the source. More exactly, let  $E_1, E_2, \dots, E_k$  be a path. There is correspondence if one can find in the OWL representation classes  $C_1, C_2, \dots, C_k$  such that  $C_j$  is a synonym or hyponym of  $E_j$  for  $j \in [1, k]$  and such that every couple of classes  $C_i, C_{i+1}$  for  $i \in [1, k-1]$  is connected by a composition of properties in the OWL representation. In other words, it is necessary to find a subset of the OWL representation which is subsumed by the path. The notions of synonym and hyponym are defined through the ontology of the domain.

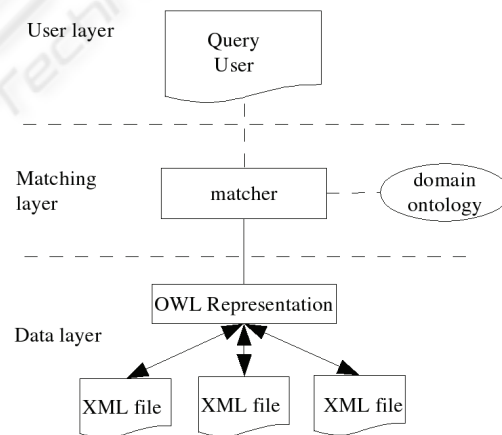


Figure 1: The architecture of our system.

For example consider the following query specified with our simplified XQUERY language :

```
Q : for $a in supplier, $b in customer
    where $b/name = "Ronald" and
          $a/region = $b/region
    return <b> {$b}</b>
```

It looks for customers with name "Ronald" and living in the same region as a supplier.

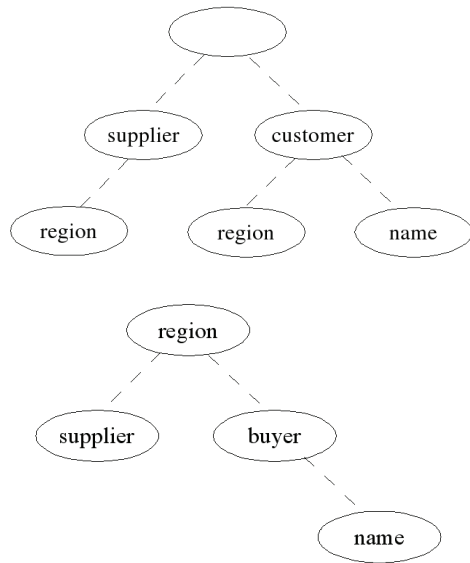


Figure 2: Two semi-structured sources.

Consider the two semi-structured sources of figure 2.

It is straightforward to infer that the query matches with the first source since the supplier element and the customer element both have a son element the name of which is region. The matcher must then check that this son element occurs only once. A matching for second source cannot be

inferred so easily. First the matcher must discover that buyer is an hyponym of customer. Then it must scan the hierarchy upward in order to establish that a supplier and a buyer are both connected to a unique region. So this second source is also a candidate for a rewriting of query Q.

We are now able to comment on the working of our system which is shown by the UML diagram of figure 3.

In the first phase the system initializes the connection with the ontology and gets back the names of the classes and the properties in the OWL representation. The system is then ready to handle queries.

In the second phase, when the system receives a query, it first interrogates the ontology to retrieve the synonyms and the hyponyms of the terms of the request. It then initiates the operation of matching for each of the paths of the query. Several rewriting possibilities can be proposed. To avoid inconvenient rewritings, we consider only the hyponyms of levels 1 to 3.

The third phase corresponds to the execution of one of these rewritings on the source concerned. It may be necessary to transform the rewriting. This operation is performed with a wrapper associated to the source.

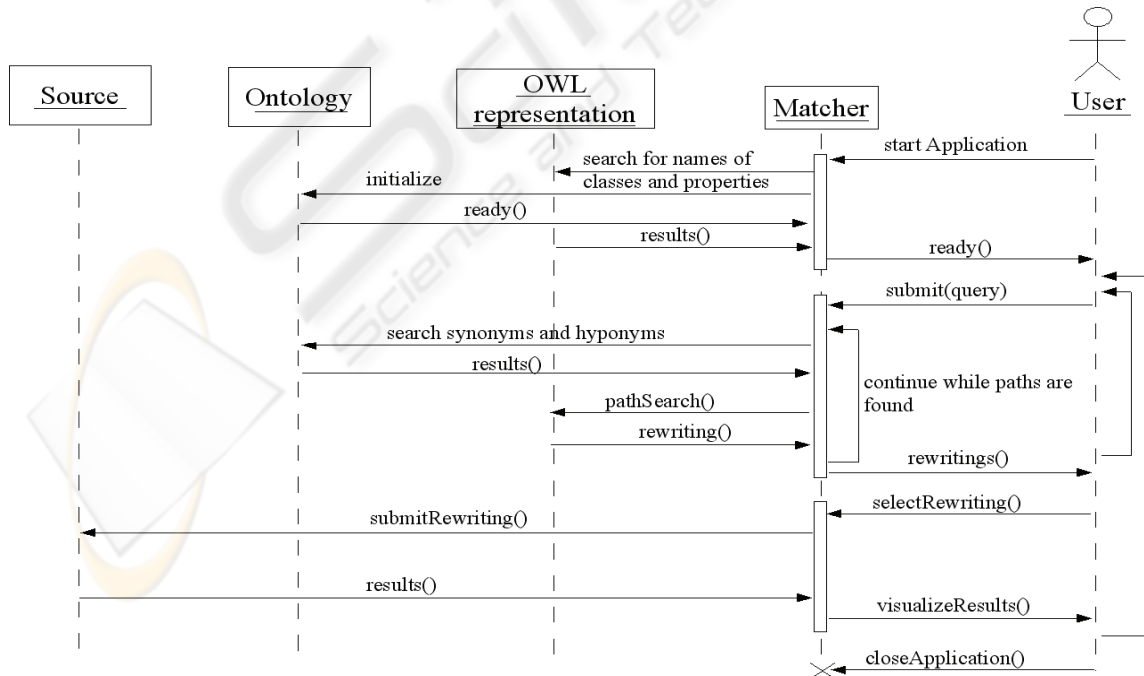


Figure 3: The working of our system.

### 3 QUERY LANGUAGE

Our query language is a simplified version of XQuery. The user will have the possibility of using the FLWR construct of XQUERY with limitations indicated below.

- Since the user does not know the documents which can provide an answer, names of documents are omitted. So the root of a path is the name of an element. The system will make searches in all the documents having the root names in their description.

- Since the user does not know the structure of the data sources, it is not possible for him to decide if a term corresponds to an element or to an attribute. However he has the possibility of using the symbol '@' to indicate that he wants to search for an attribute. The system will first look for a correspondence with an attribute, but if this is not possible, it will continue its search on elements. If the symbol '@' is not present, the search will be made at the same moment on elements and attributes.

- Also, it is impossible for the user to know whether two elements are directly connected or if there are one or several intermediate elements. So it is not possible to differentiate the descent of one level "/" and the descent of several levels "//". So the system will be responsible for testing the descent at several levels.

- The functions of XPATH are not implemented in the simplified version.

- No difference is made in the query user between lower case and upper case letters. The system will make sure the exact writing of a term is retrieved for the rewriting of the request.

### 4 OWL REPRESENTATION OF THE SOURCES

We chose to represent the sources schemas with OWL for various reasons. First it is possible to transform semi-structured schemas (XML documents) and structured schemas (relational databases, object databases) into OWL and OWL thus appears to be a good candidate for a pivot language. Then, with a view to our matching operation, it is easy to determine the connections between classes in an OWL file (as stated above, the matcher must discover paths in the source which are subsumed by a path in the query). Finally, with OWL it is possible to take advantage of the formal frame of the description logics.

We elaborated an algorithm with which a DTD can be mapped into an OWL representation. This mapping is bijective: from the OWL representation, it is possible to regenerate the DTD.

The main idea is to represent every element of the DTD by an OWL class. Every father-son link between two elements is then represented by an OWL property. An attribute is also represented by a property. When a father element has only a single son element, the cardinality of this son is represented by creating a restriction on the property connecting the two elements. When the father element is a complex element, we add an intermediate class to be able to express correctly all the cardinalities.

Agreements for the names of classes and properties are as following. The class representing an element will be named with the name of the element. For an intermediate class (associated to a

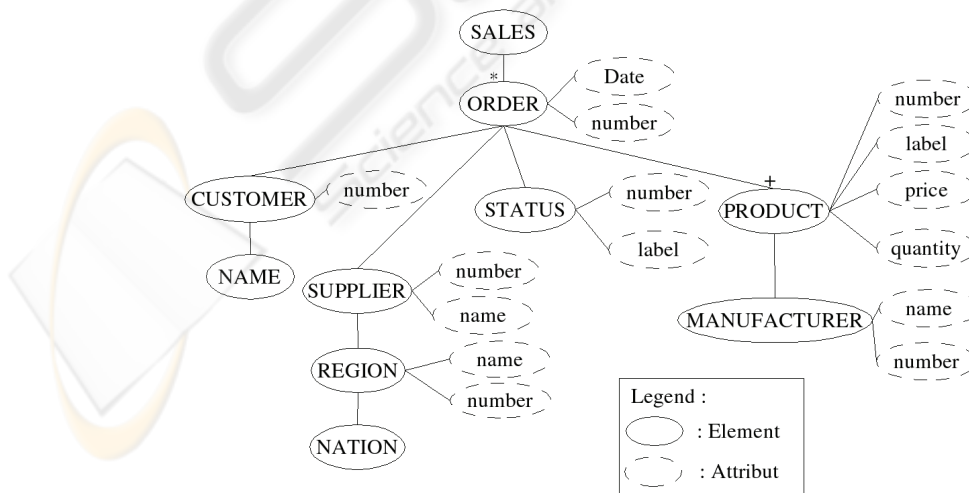


Figure 4: The semi-structured source A.

```

<owl:Class rdf:ID="ORDER"><rdfs:subClassOf> <owl:Restriction>
<owl:onProperty rdf:resource="#ORDER.&complex1;"/>
<owl:cardinality rdf:datatype="&xsd;nonNegativeInteger"> 1 </owl:cardinality>
</owl:Restriction> </rdfs:subClassOf> </owl:Class>
<rdf:Property rdf:ID="ORDER.&complex1;"/>
<rdfs:domain rdf:resource="#ORDER"/> <rdfs:range rdf:resource="#&complex1;"/>
</rdf:Property>
<owl:Class rdf:ID="&complex1;"/>
<rdfs:subClassOf> <owl:Restriction>
<owl:OnProperty rdf:resource="#&complex1;CUSTOMER"/> <owl:cardinality
rdf:datatype="&xsd;nonNegativeInteger"> 1 </owl:cardinality> </owl:Restriction> </rdfs:subClassOf>
.....
<rdfs:subClassOf> <owl:Restriction>
<owl:OnProperty rdf:resource="#&complex1;PRODUCT"/> <owl:minCardinality
rdf:datatype="&xsd;nonNegativeInteger"> 1 </owl:minCardinality> </owl:Restriction> </rdfs:subClassOf>
</owl:Class>
<rdf:Property rdf:ID="&complex1;CUSTOMER"/>
<rdfs:domain rdf:resource="#&complex1;"/>
<rdfs:range rdf:resource="#CUSTOMER"/>
</rdf:Property>
    
```

Figure 5: OWL representation for the ORDER element and its sons in source A.

complex element), the name of the class will contain the names of elements with their separator, quite in brackets. When this name is long, an entity can be used. A property between two classes will carry the two names separated by a point. For attributes, the symbol '@' is used to separate the name of the class and the name of the attribute.

As an example let us consider the element ORDER of the source A, the schema of which is shown in figure 4.

In the DTD, the definition of this element is:  
 <!ELEMENT ORDER(CUSTOMER, STATUS, SUPPLIER, PRODUCT+)>

In order to obtain its OWL representation, a class ORDER is created and also an intermediate class the name of which is (CUSTOMER, STATUS, SUPPLIER, PRODUCT+). For clearer understanding, the entity &complex1 is introduced to replace this name in the OWL file. Then a property connecting ORDER with the complex class is created, and the cardinality in the class ORDER is restricted. In the definition of the complex class the limitations of cardinalities are introduced for each of the elements. For CUSTOMER, STATUS and SUPPLIER, the cardinality is forced to be 1. Then properties are created to connect the complex class with each of the classes CUSTOMER, STATUS, SUPPLIER and PRODUCT (figure 5).

Using the same principles, it is possible to design an algorithm which maps a relational schema into a similar OWL representation. So our approach can be extended to deal also with relational sources.

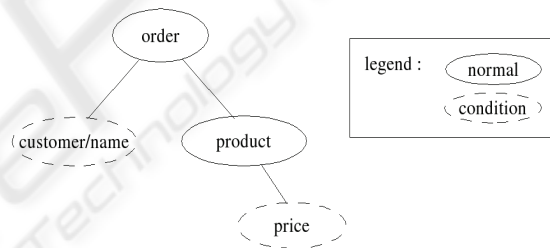


Figure 6: The tree corresponding to a path.

## 5 MATCHING ALGORITHM

We have to find a matching for each of the paths of the query.

To make the matching we represent a path as a tree having normal nodes and condition nodes. For example the path

order[customer/name="Pierre"] /product[price>15]

is represented by the tree in figure 6. A simple path composed only of a succession of terms separated by the symbol / is associated to every node.

The matching of a path is then made through two main functions matchSimplePath(SP) and matchPath(P).

The function matchSimplePath(SP<sub>i</sub>) looks in the OWL representation for the simple paths SP<sub>ij</sub> which have a matching with SP<sub>i</sub>. For example, let

$SP_i = E_1/E_2/E_3$ . A path  $SP_{i1}$  matches with  $SP_i$  if  $E_1, E_2, E_3$  have correspondents  $C_1, C_2, C_3$  in the source and if  $C_1$  is connected to  $C_2$  and  $C_2$  is connected to  $C_3$ .  $E_j$  corresponds to  $C_j$  if  $E_j$  or an  $E_j$ 's synonym or an  $E_j$ 's hyponym is identical to  $C_j$ .  $C_1$  is connected to  $C_2$  if  $C_1$  and  $C_2$  are connected either by a direct or inverse property or either by a composition of direct or inverse properties.

In the tree of the path  $P_i$ , the set of paths  $SP_{ij}$  which have a matching with  $P_i$  is associated to every node  $N_i(SP_i)$ . A node will thus be represented by  $N_i(SP_i, SP_{ij}, j \in [1, k])$ .

The function  $matchPath(P)$  tests whether if a correct assembly of simple paths can be found which corresponds to the tree of  $P$ . Let  $N_i$  be a node of the tree and  $N_{i+1}$  one of its sons. One says that the assembly between  $N_i$  and  $N_{i+1}$  is correct if the last element of one of the simple paths  $SP_{im}$  is connected with the first element of one of the simple paths  $SP_{i+1,n}$ . The function  $matchPath(P)$  supplies all the possible correct assemblies. Each of these assemblies represents a path in the source which has a matching with  $P$ .

## 6 REWRITINGS OF THE QUERY

To rewrite a query with regard to a source one looks for a rewriting of each of its paths. The rewriting of a path  $P_i$  then consists in replacing it in the query by one of the paths  $P_{ik}$  which matches with  $P_i$  and in inserting the navigation operators between the elements. When in  $P_{ik}$  one moves from a class  $C_1$  to a class  $C_2$  by a direct property, we only insert the descent operator  $//$  between the corresponding elements into  $P_i$ . If one moves from  $C_1$  to  $C_2$  by an inverse property, then the situation is more complicated. In most circumstances the query must be rewritten in depth.

At the end of this stage one can obtain several rewritings for a query.

## 7 PROTOTYPE AND EXPERIMENTS

The prototype which we built implements the architecture presented in figure 1. We incorporated the tool SAXON-B (Saxon) to access the OWL representations. We used the ontology WORDNET as the domain ontology. Since WORDNET is in fact a general ontology, we shall use sources for our experiments which do not contain highly specialized terms. Access to WORDNET is made through the JAVA API Java WordNet Library (JWNL). The

body of the matcher is written in JAVA. We have implemented the two matching functions described in section 5. However we did not generate the rewritings that require ascents in the XML files.

Our experiments were conducted on source A already presented in figure 4 and on source B presented in figure 7.

We have submitted different queries to the prototype. We show the results obtained with the two sources A and B.

### Query 1 :

```
{order[customer/name="Pierre"]/product[price>15]}
```

Rewritings for source A:

```
1: {//ORDER[./CUSTOMER/NAME = "Pierre"]
//PRODUCT[./@price>15]}
```

Rewritings for source B:

1:

For this query the matcher proposes a correct rewriting for source A. It does not use any synonyms or hyponyms. No rewriting is proposed for source B.

**Query 2 :** {for \$a in supplier where \$a/nation="FRANCE" return \$a}

Rewritings for source A:

```
1: {for $a in //SUPPLIER where $a/NATION =
"FRANCE" return $a}
```

Rewritings for source B:

```
1: {for $a in //PROVIDER where $a/@Nation =
"FRANCE" return $a}
```

In source A, NATION is an element and in source B, it is an attribute. In both cases, the matcher provides the correct rewriting.

**Query 3 :** {for \$a in supplier, \$b in manufacturer where \$a/name=\$b/name and \$a/nation = "FRANCE" return \$a}

Rewritings for source A:

```
1: {for $a in //SUPPLIER, $b in
//MANUFACTURER where $a/@name =
$b/@name and $a/NATION = "FRANCE" return
$a}
```

Rewritings for source B:

1:

For source B, the matcher does not provide any rewriting. For source A, it proposes a unique pertinent rewriting.

A rewriting such as:

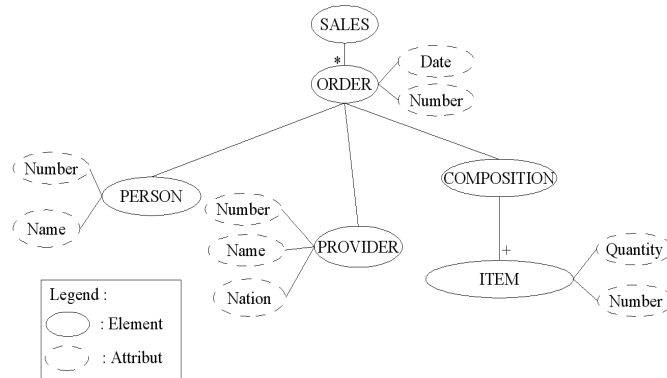


Figure 7: The semi-structured source B.

$\{ \text{for } \$a \text{ in } //\text{SUPPLIER}, \$b \text{ in } //\text{MANUFACTURER}$   
 where  $\$a//\text{REGION}/@name = \$b/@name$  and  
 $\$a//\text{NATION} = \text{"FRANCE"}$  return  $\$a\}$   
 is provided by our matching algorithm. This  
 rewriting comes from the fact that there exists  
 another attribute “name” of element REGION  
 which can be reached from SUPPLIER. This  
 rewriting contains strictly rewriting 1 and is  
 not pertinent for the user. It is filtered in  
 an additional step by using the following  
 rule: “if a rewriting path P1 is a sub-path  
 of another rewriting path P2 with the same  
 starting node, delete P2 from the set of  
 solutions”.

**Query 4 :**  $\{ \text{for } \$a \text{ in person}, \$b \text{ in supplier}$   
 where  $\$a/name = \$b/name$  return  $\$a\}$

Rewritings for source A:

1:  $\{ \text{for } \$a \text{ in } //\text{MANUFACTURER}, \$b \text{ in } //\text{SUPPLIER}$   
 where  $\$a/@name = \$b/@name$  return  $\$a\}$

2:  $\{ \text{for } \$a \text{ in } //\text{CUSTOMER}, \$b \text{ in } //\text{SUPPLIER}$   
 where  $\$a//\text{NAME} = \$b/@name$  return  $\$a\}$

3:  $\{ \text{for } \$a \text{ in } //\text{NAME}, \$b \text{ in } //\text{SUPPLIER}$   
 where  $\$a = \$b/@name$  return  $\$a\}$

Rewritings for source B:

1:  $\{ \text{for } \$a \text{ in } //\text{PERSON}, \$b \text{ in } //\text{PROVIDER}$   
 where  $\$a/@Name = \$b/@Name$  return  $\$a\}$

The matcher provides three rewritings for  
 source A since SUPPLIER, MANUFACTURER,  
 NAME are hyponyms of PERSON (at level 3).  
 The rewritings 1 and 2 are both pertinent  
 and have immediate interpretation for a  
 user. It comes from the ambiguity of using  
 in the schema of source A an element such  
 NAME which is a hyponym of PERSON. In  
 fact this rewriting is redundant with  
 rewriting 2: its path is a sub-path of  
 rewriting 2 and both terminate at the same  
 element.

So, rewritings 2 and 3 give the same result  
 when executed on the source. We can use  
 another filtering rule based on sub-paths.  
 In this case we eliminate the rewriting 2  
 and keep only the rewriting 3.

Rewritings such as:

$\{ \$a \text{ in } //\text{MANUFACTURER}, \$b \text{ in } //\text{SUPPLIER}$   
 where  $\$a/@name = \$b//\text{REGION}/@name$   
 return  $\$a\}$  are provided by our matching  
 algorithm. Like for query 3, there are  
 filtered in the additional step.

Our matcher filters also rewritings such as:  
 $\{ \text{for } \$b \text{ in } //\text{SUPPLIER}, \$a \text{ in } //\text{SUPPLIER}$   
 where  $\$a/@name = \$b/@name$  return  $\$a\}$   
 which are correct but which correspond to  
 a truth assertion and do not give pertinent  
 results.

If we use only the hyponyms of level 1 for  
 this query, the matcher give no answer for  
 source A. This example shows clearly the  
 interest of hyponyms, but also the problems  
 which they can pose when confronting it to  
 ambiguous schemas.

For source B the matcher provides a unique  
 rewriting which is pertinent.

## 8 CONCLUSION AND PERSPECTIVES

Through the results obtained, it appears that  
 our mediation system is able to find data  
 from an intuition of the user, intuition  
 expressed through an implicit vision of the  
 domain compatible with the ontology.

The main difficulty results from the fact  
 that the system generally proposes several  
 rewritings for a query. Not all these  
 rewritings are relevant. We have suggested  
 a filtering based on sub-paths to treat this  
 problem. But this rule cannot solve all the  
 situations. One can also act on the  
 exploration depth in the ontology. We have  
 also noticed that some terms

(name, number) contribute to increase the number of irrelevant solutions. It would so be necessary to minimize their use in the database schemas and to resort to more precise terms. The quality of the ontology is also highly important to obtain relevant rewritings. Ontology WORDNET used for our experiments is too general and contributes to sending back too many solutions.

More elaborated solutions exist to deal with this problem. A solution which we are investigating at present consists in placing annotations in the OWL representation at the level of classes or properties. These annotations will be exploited by the matcher to take into account semantic features (sense of a term, meaning of a property). These annotations could be installed manually by the administrator of the source or automatically by the system by seeking the opinion of the users when several rewritings are possible. To help the matching one can ask the user to clarify his query if the system detects some ambiguities.

We think that these improvements could result in an efficient system.

The system can be extended to deal with other types of sources (relational, object).

The main advantage of our approach is its robustness with regard to the evolution of sources. When a new source is inserted, it is sufficient to elaborate its OWL representation so that it can be exploited by the system. When a source evolves, it is sufficient to reshape its OWL representation.

We are also engaged in another improvement of our prototype in order to allow the join of results coming from different sources. In that case a query is rewritten in several sub-queries, each sub-queries being relative to a different source. Our matching algorithm can be easily adapted for this more general situation. It is necessary to look for sub-paths in different sources and to impose a join condition between sub-paths (the terminal node of a sub-path must be compatible with the start node of another sub-path).

Such a system can be very useful for different applications. Incorporated into an intranet system, it would allow a user to reach the data sources without knowing their schemas, by being based only on the domain ontology. In a P2P system, it could be installed on some peers or on the super-peers to facilitate access to data by their semantics. The only obligation for a peer would be to publish its data by using the OWL representation.

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