



A Semantic-based Ontology Matching Process for PDMS

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http://www.cin.ufpe.br/~speed/SemMatch/index.htm







Motivation

- Peer Data Management System (PDMS) [Adjiman et al., 2007]
 - Each peer is an autonomous data source that makes available a local schema
 - Schema mappings (correspondences between schema elements) are generated to allow information exchange between peers
- Ontologies
 - Make explicit the content of data sources (peer ontologies)
 - Enhance information integration

Motivation

- Peer ontologies
 - Designed and developed autonomously
 - Contain several forms of heterogeneity
- Ontology matching techniques [Euzenat and Shvaiko, 2007]
 - Deal with the diverse concept meanings existing in peer ontologies
 - Reconcile peer ontologies and find correspondences between their elements

Goal

- Propose a semantic-based ontology matching process which has been instantiated in a PDMS
- Contributions
 - Identification of semantic correspondences between two peer ontologies
 - Taking into account a domain ontology as background knowledge
 - Determination of the global similarity between two peer ontologies

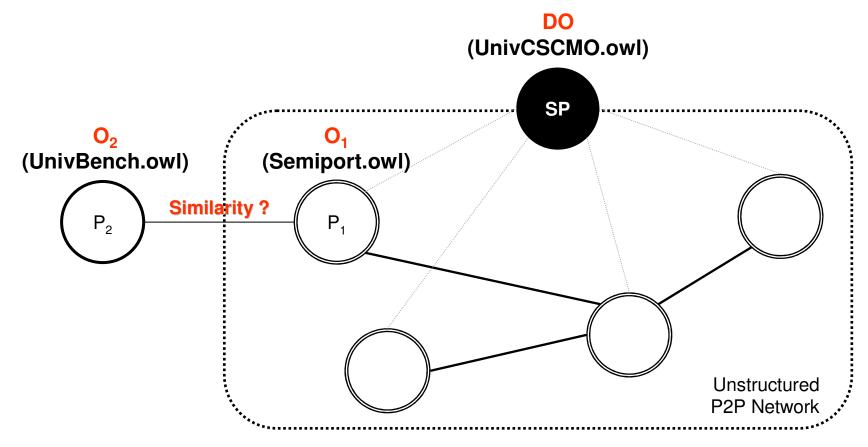
Outline

- Ontology Matching
- Using a Domain Ontology to Define Semantic Correspondences
- Semantic-based Ontology Matching Process
- Calculating the Global Similarity Measure
- Experiments and Results
- Related Work
- Conclusions and Further Work

Ontology Matching

- Process of finding correspondences between elements of different ontologies [Euzenat and Shvaiko, 2007]
 - Normally describing the same or similar domains
- An element is a concept, property or instance
- Ontology Alignment
 - Set of correspondences indicating which elements of two ontologies logically correspond to each other
 - Produced by one or more matchers which are executed sequentially or in parallel

Working Scenario



Semantic Community of Peers

* Education Domain *

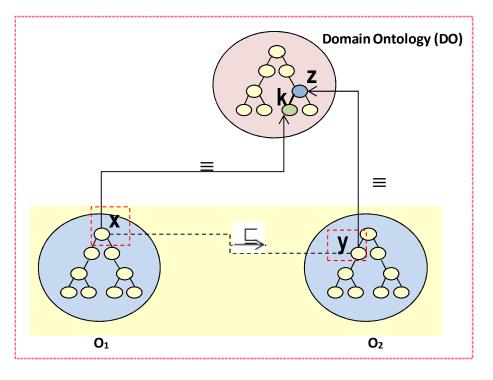
Working Scenario

Our focus

- Identify semantic correspondences between
 O₁ and O₂ elements
- Determine if P₁ and P₂ are semantic neighbors
 - Two peers are semantic neighbors if their global similarity is higher than a certain threshold

Using a Domain Ontology to Define Semantic Correspondences

- Domain Ontology (DO)
 - Reliable reference available on the Web
 - Used as Background Knowledge
 - Bridge the conceptual differences or similarities between two peer ontologies

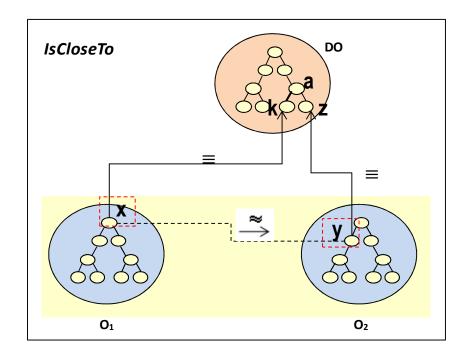


Using a Domain Ontology to Define Semantic Correspondences

- Definition. A semantic correspondence is represented by one of the following expressions:
 - \bigcirc $O_1:x \Rightarrow O_2:y$, an *isEquivalentTo* correspondence
 - \circ $O_1:x \to O_2:y$, an *isSubConceptOf* correspondence
 - \bigcirc $O_1:x \supseteq O_2:y$, an *isSuperConceptOf* correspondence
 - \bigcirc $O_1:x \triangleright O_2:y$, an *isPartOf* correspondence
 - \circ $O_1:x o O_2:y$, an *isWholeOf* correspondence
 - $O_1:x \approx O_2:y$, an *isCloseTo* correspondence
 - \bigcirc $O_1:x \to O_2:y$, an *isDisjointWith* correspondence
 - (*) x and y are elements belonging to the peer ontologies

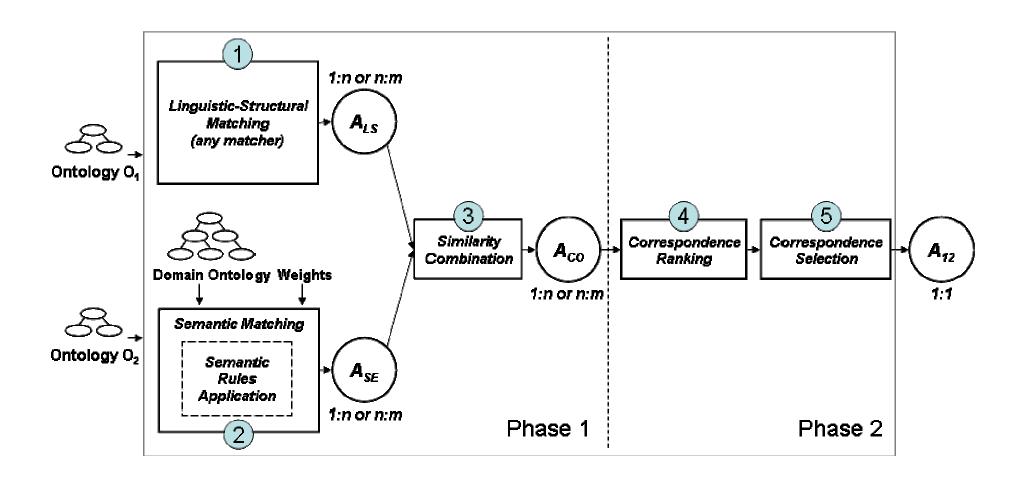
Using a Domain Ontology to Define Semantic Correspondences

- O₁:k and O₂:z are close if
 - They share a common ancestor in the DO
 - □ The common ancestor is not the root (⊤)
 - The concepts do not hold neither subsumption nor disjointness
 - The measured depths
 (thresholdRoot and thresholdCommonAncestor)
 are evaluated to true



Example: O_1 .Notebook \cong O_2 .MacintoshPC

Semantic-based Ontology Matching Process



Semantic-based Ontology Matching Process

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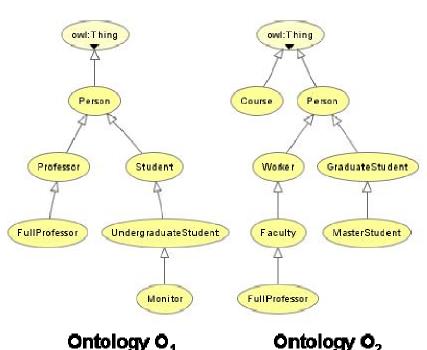
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Calculating the Global Similarity Measure

- Uses the alignment sets A₁₂ and A₂₁
- Existing similarity measures can be adapted
 - □ Dice [Aümuller et al., 2005], Weighted average [Castano et al., 1998] and overlap [Rijsbergen, 1979]
- All of them consider the size of the input ontologies
- The size of an ontology (|O|) is determined by the number of its elements

Calculating the Global Similarity Measure

Weighted Average
$$(O_1, O_2) = \frac{\sum_{i=1}^{|A|2|} n + \sum_{j=1}^{|A|2|} n}{|O_1| + |O_2|}$$



Alignment A₁₂

- (1, Person, Person, isEquivalentTo, 1.0)
- (2, FullProfessor, FullProfessor, isEquivalentTo, 1.0)
- (3, UndergraduateStudent, Course, isPartOf, 0.3)
- (4, Student, Person, isSubConceptOf, 0.8)
- (5, Professor, Faculty, isSubConceptOf, 0.8)

Alignment A₂₁

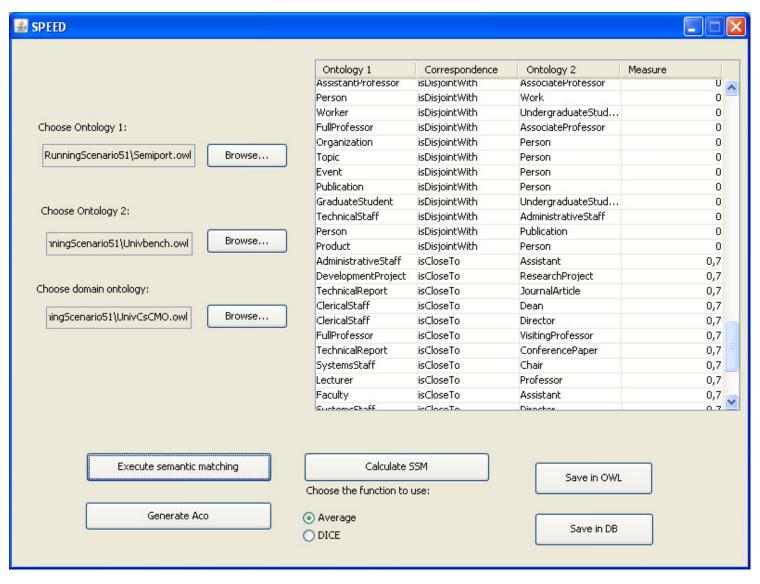
- (1, Person, Person, isEquivalentTo, 1.0)
- (2, FullProfessor, FullProfessor, isEquivalentTo, 1.0)
- (3, Course, UndergraduateStudent, isWholeOf, 0.3)
- (4, Worker, Person, isSubConceptOf, 0.8)
- (5, GraduateStudent, UndergraduateStudent, isDisjointWith, 0.0)
- (6, Faculty, Professor, isSuperConceptOf, 0.8)
- (7, MasterStudent, Student, isSubConceptOf, 0,8)

Weighted Average
$$(O_1, O_2) = \frac{(1.0 + 1.0 + 0.3 + 0.8 + 0.8) + (1.0 + 1.0 + 0.3 + 0.8 + 0.0 + 0.8 + 0.8)}{|6| + |7|} = 0.66$$

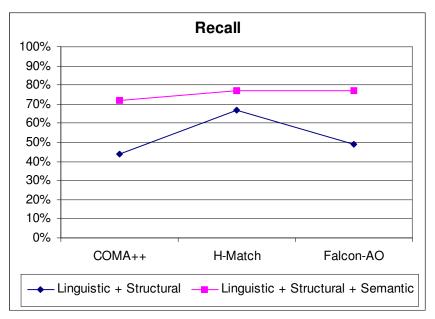
Experiments and Results

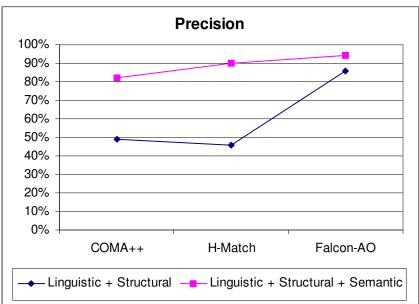
- The semantic-based ontology matching tool
 - Implemented in Java
 - Jena has been used to provide ontology manipulation and reasoning
 - H-Match has been used as the hybrid matcher
- Correspondence identification has been restricted to concepts
 - Properties are not included
- http://www.cin.ufpe.br/~speed/SemMatch/index.htm

Experiments and Results



Experiments and Results





Correspondences for O ₁ :Faculty							
O_1 :Faculty $\equiv O_2$:Faculty	O_1 :Faculty $\supseteq O_2$:PostDoc						
O_1 :Faculty $\sqsubseteq O_2$:Worker	O_1 :Faculty $\approx O_2$:Assistant						
O_1 :Faculty $\supseteq O_2$:Professor	O_1 :Faculty $\approx O_2$:AdministrativeStaff						

Related Work

- Only a few semantic-based approaches consider the use of background knowledge to improve ontology matching
 - S-Match, TaxoMap, CTXMatch
- Correspondences are usually restricted to equivalence
 - CTXMatch considers other ones (specialization and generalization)
- We also identify other types of semantic correspondences
 - E.g., disjointness and closeness

Related Work

- Global Similarity Measure
 - Not produced by the previous works
 - [Castano et al., 1998] propose a kind of such measure
 - Concerned with ER schemas
 - □ COMA++ [Aümuller et al., 2005] argues that calculates a global measure
 - Considering the version we performed our tests, we were not able to find out such feature explicitly

Conclusions and Further Work

- Our matching process tries to overcome limitations of linguistic and structural approaches by using domain ontologies as background knowledge
 - A semantic matcher identifies, besides traditional types of correspondences, other ones (e.g., closeness and disjointness)
 - Determination of a global similarity measure between two ontologies (not only between their elements)

Conclusions and Further Work

- The combination of different matchers can improve the alignments produced by ontology matchings tools
 - Taking out incorrect or meaningless correspondences and including relevant ones
- Further work
 - Extend our tool to consider properties
 - Include an alignment-reuse matcher





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