Semantic Measures and Ontology Matching: an Overview and some Examples



Jorge Gracia del Río



Ontology Engineering Group (OEG)

Departamento de Inteligencia Artificial Facultad de Informática

Universidad Politécnica de Madrid (UPM), Spain

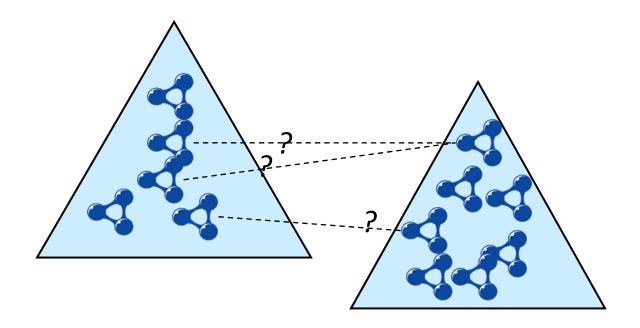
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Outline

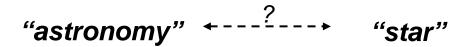
- Introduction
- Semantic Measures
- Ontology Matching
- References

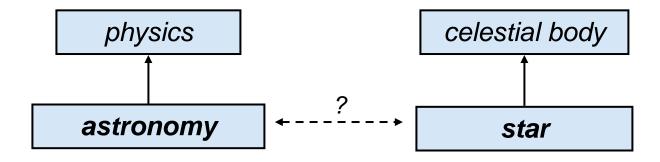
- The heterogeneity problem [Euzenat & Shvaiko, 2007]:
 - Resources being expressed in different ways must being reconciled before being used
 - Mismatch between different formalized knowledge may occur when:
 - Different languages are used
 - Different terminologies are used
 - Different modelling is used

Ontology Matching (OM) is the task of discovering correspondences between terms from different ontologies.



- Semantic Measures
 - Evaluate numerically how semantically related are two entities (words, ontology terms, etc.):





- Semantic Measures. Applications:
 - Ontology matching
 - Word sense disambiguation
 - Information retrieval
 - Annotation
 - Automatic indexing
 - Analysis of structure of texts
 - Spelling correction

- Semantic similarity
 - Hierarchy-based relationships
 - doctor is highly similar to nurse
 - doctor is not similar to hospital
- Semantic relatedness
 - Considers similarity + any other relationship
 - doctor is highly related to hospital
- Semantic distance
 - Inverse of relatedness

Semantic relationships

Exercise: Put in the right places...

Name	Example
Identity	
Hypernymy	
Hyponymy	
Synonymy	
Compatibility	
Taxonomical connection	
Disjointness	
Antonymy	
Holonymy	
Meronymy	
Positive association	

Semantic relationships

Semantic relatedness measure

Symbol	Name	Example
R _i	Identity	(person, person)
R_{H}	Hypernymy	(vehicle, car)
R_h	Hyponymy	(hospital, building)
R _{syn}	Synonymy	(doctor, physician)
R _{com}	Compatibility	(resort, private accommodation)
R _{conn}	Taxonomical connection	(hospital, monastery)
R _{dis}	Disjointness	(liquid, solid)
R _{ant}	Antonymy	(hot, cold)
\mathbf{R}_{hol}	Holonymy	(keyboard, key)
\mathbf{R}_{mer}	Meronymy	(finger, hand)
R _{assoc}	Positive association	(penguin, Antarctica)

Semantic similarity measure

Some definitions [Gracia 09]

- Considering the previous semantic relationships as mathematical binary relations, and given E a set of entities with an associated semantics, we define:
- Semantic relatedness relation: set of ordered pairs R ⊆ E x E such that

$$\mathbf{R} = \mathbf{R_i} \cup \mathbf{R_H} \cup \mathbf{R_h} \cup \mathbf{R_{syn}} \cup \mathbf{R_{conn}} \cup \mathbf{R_{com}} \cup \mathbf{R_{ant}} \cup \mathbf{R_{dis}} \cup \mathbf{R_{hol}} \cup \mathbf{R_{mer}} \cup \mathbf{R_{assoc}}$$

• Semantic similarity relation: set of ordered pairs $R_S \subseteq E \times E$ such that

$$R_S = R_i \cup R_H \cup R_h \cup R_{syn} \cup R_{conn} \cup R_{com}$$

Semantic relatedness measure. Given R, E

rel: $E \times E \to \mathbb{R}$ such that:

```
\forall x, y \in E, rel(x,y) \ge 0 (positiveness)<br/>
\forall x, y, z \in E, rel(x,x) \ge rel(y,z) (maximality)<br/>
\forall x, y \in E, rel(x,y) = rel(y,x) (symmetry)<br/>
\forall x, y \in E, (x, y) \notin R, rel(x,y) = 0
```

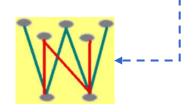
Semantic similarity measure. Given R_S, E

sim: $E \times E \rightarrow \mathbb{R}$ such that:

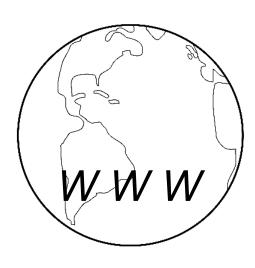
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\forall x, y \in E, sim(x,y) \ge 0 (positiveness)<br/>
\forall x, y, z \in E, sim(x,x) \ge sim(y,z) (maximality)<br/>
\forall x, y \in E, sim(x,y) = sim(y,x) (symmetry)<br/>
\forall x, y \in E, (x, y) \notin R_S, sim(x,y) = 0
```

Use of external resources

Few named entities: "The Rolling Stones" | Few very specialized terms: "exocitosys" | Many relations remain hidden: "aspirin" + "stomach disease"







Lexical resources (WordNet, corpus, ...)

Wikipedia

The Web

increasing coverage

- Two examples of semantic measures, designed for its use on the Semantic Web [Gracia 09, Gracia & Mena 08, Trillo et al. 07]:
 - Context and inference-based semantic similarity measure
 - Web-based semantic relatedness measure

Semantic Measures: Similarity

- Semantic similarity between two classes
 - 1st) Extract Ontological Contexts
 - 2nd) Enrich them by applying inference
 - 3rd) Compute the following:

```
sim(t_1, t_2) = Linear Combination of
```

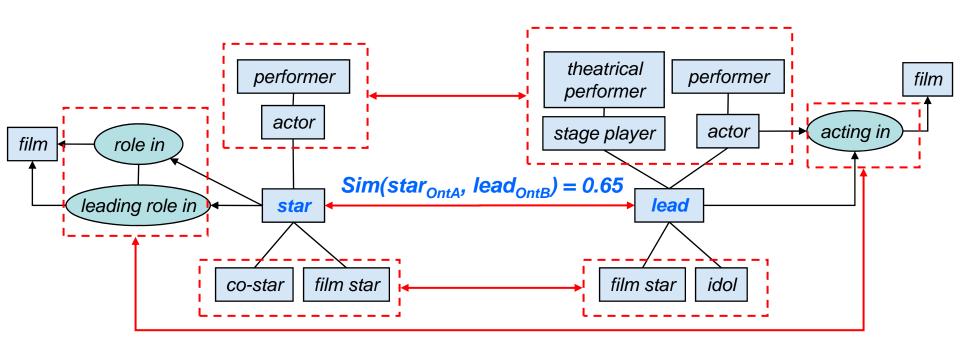
- lexical_similarity (synonyms₁, synonyms₂)
- VSM (description₁, description₂)
- VSM (properties₁, properties₂)
- Graph_similarity (graph₁, graph₂)

```
Graph_similarity (graph<sub>1</sub>, graph<sub>2</sub>) = Linear Combination of
```

- VSM (hyper₁ \forall directHyper₁, hyper₂ \forall directHyper₂)
- VSM (hypo₁ \biguplus directHypo₁, hypo₂ \biguplus directHypo₂)

Semantic Measures: Similarity

- Analogously to compute similarity between properties and between individuals.
- Example of Semantic Similarity between Classes:



• Hypothesis [Cilibrasi and Vitànyi 07] : For each $x, y \in S$

x, y are semantically related \Leftrightarrow x, y appear on the same web page

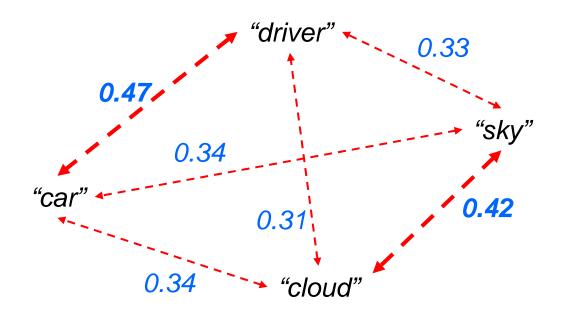
Semantic relatedness between two words

$$relWeb(x, y) = e^{-2NWD(x,y)}$$

where NWD is Cilibrasi and Vitànyi's Google distance

An example of web-based relatedness computation

http://horus.cps.unizar.es:28080/Relatedness/Relatedness.html



- Semantic relatedness between two ontology terms
 - Level 0: Labels and synonyms

```
rel_0(a, b) = average of relWeb between synonyms of a and b
```

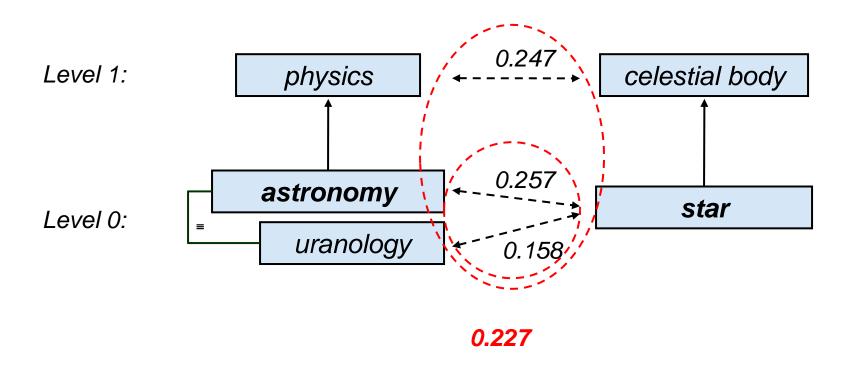
■ Level 1: OC^m Minimal Ontological Context

E.g.: Java
$$\rightarrow$$
 Island \rightarrow Land \rightarrow Thing

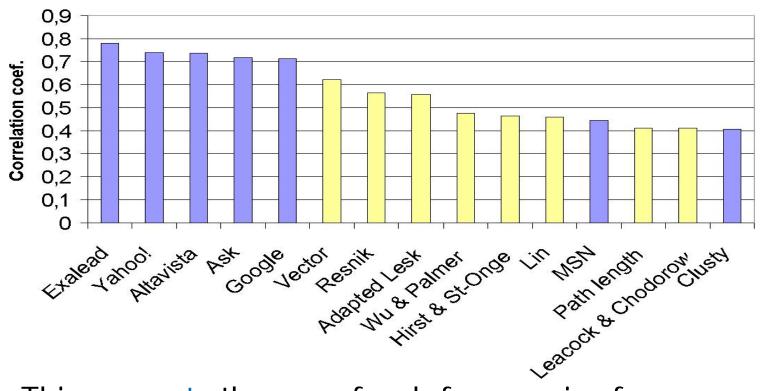
 $rel_1(a, b)$ = average of rel_0 between the elements of the OC_a^m and OC_b^m

 $rel(a, b) = linear combination of <math>rel_0(a, b)$ and $rel_1(a, b)$

Example: relatedness between ontology terms



- Experiment: relatedness between words
 - Web-based (blue), WordNet-based (yellow) measures



This supports the use of web frequencies for relatedness computation

- Semantic Measures: other approaches
 - Based on Thesauri and Lexical Resources [Jiang and Conrath, Banerjee and Pedersen, Resnik, etc.]
 - Based on Wikipedia [Gabrilovich and Markovitch, Strube and Ponzetto, etc.]
 - Based on the Web [Bollegala et al., Chen et al., Sahami et al., OntoNL, etc.]
 - Based on DL constructs [D'Amato, et. al., Borgida et al., Janowicz et al., Hu et. al., etc.]

A couple of definitions [borrowed from Euzenat & Shvaiko's OM tutorial]

Definition (Correspondence)

Given two ontologies o and o', a **correspondence** between o and o' is a 5-uple: $\langle id, e, e', r, n \rangle$ such that:

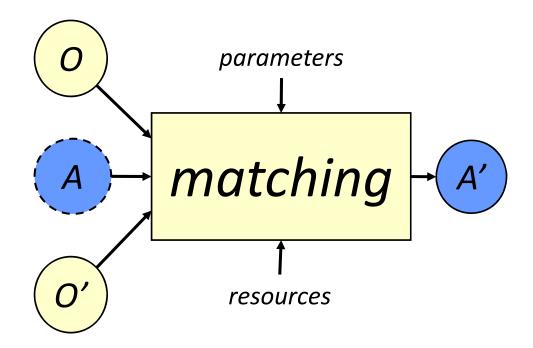
- ▶ id is a unique identifier of the correspondence
- ightharpoonup e and e' are entities of o and o' (e.g., XML elements, classes)
- ▶ r is a relation (e.g., equivalence (=), more general (\supseteq), disjointness (\perp))
- n is a confidence measure in some mathematical structure (typically in the [0 1] range)

Definition (Alignment)

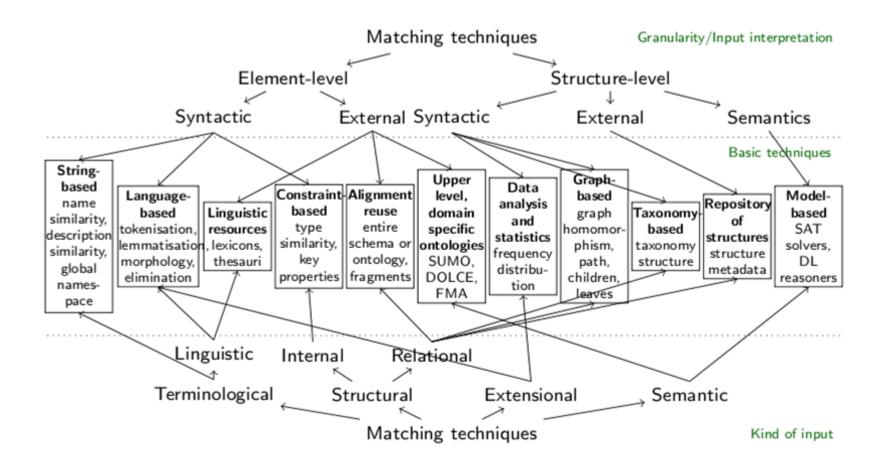
Given two ontologies o and o', an **alignment** (A) between o and o':

- ightharpoonup is a set of correspondences on o and o'
- with some additional metadata (multiplicity: 1-1, 1-*, method, date, properties, etc.)

General scheme of an Ontology Matching process, being O and O' are the input ontologies, A an initial set of known correspondences (optional), and A' the resultant alignment (the found correspondences):



Matching techniques [Euzenat & Shvaiko, 2007]



- Matching techniques [borrowed from Euzenat & Shvaiko's OM tutorial]:
 - Name of the entities
 - comments, alternate names, names of related entities
 - \implies NLP, IR, etc.
 - Structure
 - internal structure: constraints on relations, typing
 - external structure: relations between entities
 - ⇒ Data mining, Discrete mathematics
 - Extension
 - Instances themselves
 - Related resources: annotated documents, exchanges messages or queries
 - ⇒ Statistics, data analysis, data mining, machine learning
 - ► Semantics (models)
 - ⇒ Reasoning techniques
 - ► Background knowledge
 - the web
 - ontologies
 - wordnet, etc.

How does an alignment looks like? E.g. (in RDF):

```
<Alignment>
<xml>yes</xml>
<level>0</level>
<type>11</type>
<onto1>./models/Nitrogen cycling.owl</onto1>
<onto2>./ models/Phosphorus cycling.owl</onto2>
<map>
<Cell>
   <entity1 rdf:resource='http://www.dynalearn.eu/models/Nitrogen#plant'/>
   <entity2 rdf:resource=' http://www.dynalearn.eu/models/Phosphorus#plantas'/>
   <relation>=</relation>
   <measure rdf:datatype='http://www.w3.org/2001/XMLSchema#float'>0.8</measure>
</Cell>
</map>
</Alignment>
```

Two examples of ontology matching systems, designed for its use on the Semantic Web:

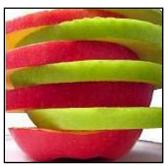
CIDER [Gracia & Mena 08b]: Context and Inference baseD alignER

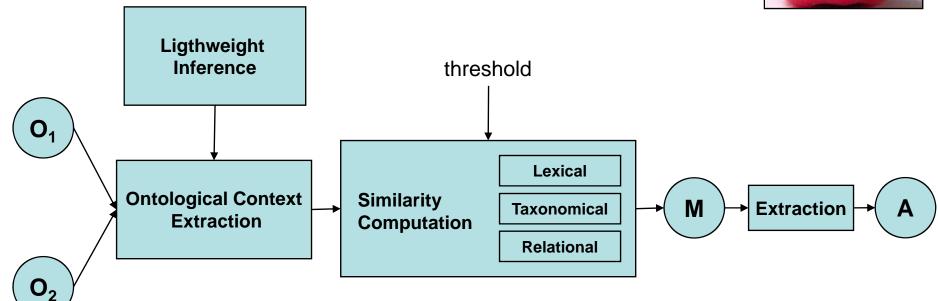
- Based on our semantic similarity measure
- Identifies semantic equivalences
- Scarlet [Sabou et al. 08]
 - Background knowledge-based ontology alignment
 - Identifies various types of semantic relationships
 - Hampered by semantic ambiguity problems



Ontology Matching: CIDER

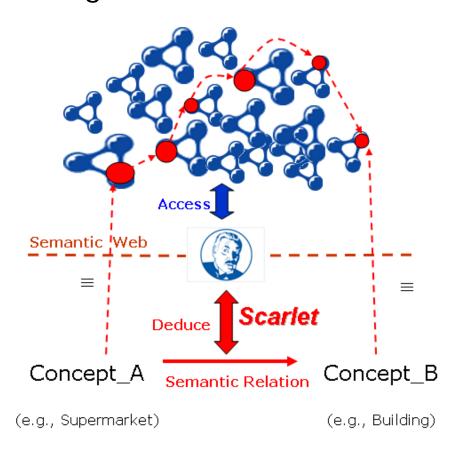
CIDER: Context and Inference baseD alignER





Ontology Matching: Spider

Scarlet system (KMi, Open University, United Kingdom): uses online ontologies as background knowldege for ontology matching



- Ontology Matching. Other approaches [Euzenat & Shvaiko 07]
 - Falcon-AO
 - ASMOV
 - Lily
 - Glue
 - S-Match
 - etc.

References

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Further Readings

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