Ontology-Based Applications in the Age of the Semantic Web

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The Rise of Semantics

Semantic Technologies Center

Semantic Technologies are designed to extend the capabilities of infrastructures and to be networked in meaningful ways. The adoption of standards like XML, RDF (Resource Description Framework), and Ontology Language serve as foundation technologies to advancing applications.

Oracle Spatial 10g introduces the industry's first open, scalable, and graph data model. RDF triples are persisted, indexed and quickly retrieved. The 10g RDF database ensures that application developers benefit from secure semantic applications. Application areas include:

Gartner

Finding and Exploiting Value in Semantic Technologies on the Web

9 May 2007

Revolutionary online protection

DataPatrol is a new monthly monitoring service that finds, tracks, and prevents the latest threats.

In this article:
- Technologies that make up the Semantic Web
- Ontologies form the backbone of a whole new way to understand online data

Level: Introductory
  <mediapro:People>
    <rdf:Bag>
      <rdf:li>Jim Hendler</rdf:li>
      <rdf:li>Enrico</rdf:li>
    </rdf:Bag>
  </mediapro:People>
</rdf:Description>
Key Propositions

• The SW is less and less an aspiration and more and more a reality
• This emerging large scale semantics opens up new scenarios and introduces a number of implications for:
  – the practice of ontology engineering
  – the kind of functionalities that ontology engineering tools ought to support
  – the kind of ontology-based applications we can now develop
• In addition, it may also provide a solution to one of the holy grails of AI research: the availability of large-scale background knowledge to enable intelligent behaviour
Large Scale Semantics and Ontology Engineering
Ontology Engineering in the Age of the Semantic Web

- The availability of large scale semantics (millions of docs and tens of thousands of ontologies) opens up the following opportunities
  - to make cost-effective the development of large scale semantic applications out of reusable resources
  - to move away from monolithic ontologies and characterise ontology engineering as the process of constructing and managing networked ontologies

- The goal of the NeOn project is precisely to provide a methodology and a novel infrastructure for ontology engineering in line with this vision of the next generation of ontology-based applications
Networked Ontologies

- \( O_1 \) is related with \( O_2 \)
- \( M_1 \) extends \( O_1 \)
- \( M_2 \) depends on \( O_1 \)
- \( O_1 \) is prior version of \( O_1' \)
- \( O_1' \) extends \( O_1'' \)
- \( O_4 \) extends \( O_3 \)
- \( O_3 \) is incompatible with \( O_1'' \)
- \( O_2 \) is source of \( M_1 \)
- \( M_2 \) is source of \( O_1' \)
First Year Outputs: Some Highlights

- Meta model and initial methods for reasoning with networked ontologies
- A formal, ontological framework for characterizing collaborative ontology design workflows
- Formalization of context and initial methods and software for generating mappings which contextualise ontologies
- New methods for ontology alignment, selection and modularization

**A task-centric user study highlighting limitations of current tools in tackling typical NeOn development scenarios**

- Initial modelling components for NeOn methodology
- NeOn Architecture design and initial infrastructure components
- Initial Version of the NeOn Toolkit
- Analysis and design of NeOn testbeds

http://www.neon-project.org/
Outline of the study

• 2 ontology engineering tools
  – TopBraid, Protégé
• 3 ontologies
  – Copyright (85 C; 49 P; 128 Re)
  – AKT Support (14 C; 15 P; 0 Re)
  – AKT Portal (162 C; 120 P; 130 Re)
• 28 participants
  – Mixed w.r.t. expertise with ontologies and tools
  – Actually most users had designed ontologies in the past, but usually not in OWL
• 3 tasks
  – Task 1: Simple class/subclass relationship across ontologies
  – Task 2: Import two ontologies and change axioms
  – Task 3: Import concepts and redefine them
<table>
<thead>
<tr>
<th>Question (existing feature or ‘proposed fix’)</th>
<th>Avg. marks</th>
<th>-1</th>
<th>0</th>
<th>+1</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Existing support for ontology re-use</td>
<td>-0.097 (not very good / reasonable)</td>
<td>26%</td>
<td>58%</td>
<td>16%</td>
<td>31</td>
</tr>
<tr>
<td>Support for partial re-use of ontologies</td>
<td>-0.739 (not very good)</td>
<td>62%</td>
<td>14%</td>
<td>4%</td>
<td>29</td>
</tr>
<tr>
<td>flag chunks of ontologies or concept worked with</td>
<td>+0.674 (would be useful)</td>
<td>20%</td>
<td>24%</td>
<td>56%</td>
<td>25</td>
</tr>
<tr>
<td>hide selected (irrelevant?) parts of ontologies</td>
<td>+0.465 (would be reasonable / useful)</td>
<td>25%</td>
<td>38%</td>
<td>38%</td>
<td>24</td>
</tr>
<tr>
<td>Existing support for mappings and contextual boundaries</td>
<td>-0.065 (not very good / reasonable)</td>
<td>19%</td>
<td>68%</td>
<td>13%</td>
<td>31</td>
</tr>
<tr>
<td>Management and assistance with any mappings</td>
<td>-0.480 (not very good / reasonable)</td>
<td>48%</td>
<td>52%</td>
<td>0%</td>
<td>26</td>
</tr>
<tr>
<td>propose mappings &amp; ensure their consistency</td>
<td>+0.433 (would be reasonable/useful)</td>
<td>3%</td>
<td>50%</td>
<td>47%</td>
<td>30</td>
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<tr>
<td>using trial queries to see consequences of mappings</td>
<td>+0.045 (would be reasonable)</td>
<td>9%</td>
<td>77%</td>
<td>14%</td>
<td>23</td>
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<td>Existing support for versioning, alternatives</td>
<td>-0.200 (not very good)</td>
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<td>20%</td>
<td>30%</td>
<td>11</td>
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<tr>
<td>Existing visualizing capabilities &amp; their adaptation</td>
<td>-0.536 (not very good)</td>
<td>57%</td>
<td>39%</td>
<td>4%</td>
<td>28</td>
</tr>
<tr>
<td>propagate changes between alternative versions</td>
<td>+0.519 (would be reasonable / useful)</td>
<td>7%</td>
<td>33%</td>
<td>60%</td>
<td>28</td>
</tr>
<tr>
<td>compare/visualize different interpretations/versions</td>
<td>+0.700 (would be useful)</td>
<td>6%</td>
<td>17%</td>
<td>77%</td>
<td>30</td>
</tr>
<tr>
<td>performing operations in graphical/textual mode</td>
<td>+0.414 (would be reasonable / useful)</td>
<td>7%</td>
<td>45%</td>
<td>48%</td>
<td>29</td>
</tr>
<tr>
<td>visualize on the level of ontologies (not just concepts)</td>
<td>+0.357 (would be reasonable / useful)</td>
<td>7%</td>
<td>50%</td>
<td>43%</td>
<td>28</td>
</tr>
</tbody>
</table>
Implications for ontology engineering infrastructure

• Empirical findings confirm intuition that existing tools need new functionalities to support the NeOn vision
• This is potentially a critical issue as the tension between what is feasible in principle and what is supported by the current infrastructure may generate a “software crisis”.
• Problems are clearly harder for less expert users, which actually provide the key industrial target audience
Implications for Ontology Engineering Practice

- Reuse rather than ad hoc design of ontology elements
- Potential for making the Ont. Dvpt. process more robust
  - Cfr. similar paradigm shift for KBS thanks to work on Problem Solving Methods (1985 - onwards)
- The NeOn vision nicely complements ongoing work on design patterns for ontology engineering
  - Meta-level nature of design patterns vs. object-level nature of direct reuse of definitions
  - NeOn methodology is indeed based on work on design patterns
- Ontology engineering generates new kinds of outputs
  - Networked ontologies
    - when process creates connections between distributed pre-existing ontologies
  - Faceted ontologies
    - when process consists of creating a new ontology out of massively distributed ‘ontology snippets’
Faceted Ontologies

Example: Integrating SW and Web2.0
• Tagging as opposed to rigid classification
• Dynamic vocabulary does not require much annotation effort and evolves easily
• Shared vocabulary emerge over time
  – certain tags become particularly popular
Limitations of tagging

- Different granularity of tagging
  - rome vs colosseum vs roman monument
  - Flower vs tulip
  - Etc..
- Multilinguality
- Spelling errors, different terminology, plural vs singular, etc...

This has a number of negative implications for the effective use of tagged resources
  - e.g., Search exhibits very poor recall
Giving meaning to tags

All time most popular tags

06 ateca amsterdam animals architecture art august australia autumn baby barberona beach berlin birthday black blackandwhite blue boston bw california cameraphone camping canada canon car cat cats chicago china christmas church city clouds color concert day dc dog england europe fall family festival film florida flower flowers food france friends fun garden geotagged germany girl graffiti green halloween hawaii hiking holiday home honeymoon hongkong house india ireland island italy japan july june kids lake landscape light live london losangeles macro may me mexico mountain mountains museum music nature new newyork newyorkcity newzealand night nikon nyc ocean october paris park party people portrait red river roadtrip rock rome san sanfrancisco school scotland sea seattle september show sky snow spain spring street summer sun sunset sydney taiwan texas thailand tokyo toronto travel tree trees trip uk urban usa vacation vancouver washington water wedding white winter yellow york zoo
What does it mean to add semantics to tags?

1. Mapping a tag to a SW element
   "japan"
   <akt:Country Japan>

2. Linking two "SW tags" using semantic relations
   {japan, asia} → <japan subRegionOf asia>
Applications of the approach

• To improve recall in keyword search

• To support annotation by dynamically suggesting relevant tags or visualizing the structure of relevant tags

• To enable formal queries over a space of tags
  – Hence, going beyond keyword search

• To support new forms of intelligent navigation
  – i.e., using the 'semantic layer' to support navigation
Pre-processing

Tags

Clean tags

Group similar tags

Filter infrequent tags

Concise tags

Folksonomy

Clustering

Analyze co-occurrence of tags

Co-occurrence matrix

Cluster tags

Cluster1

Cluster2

Cluster_n

Remaining tags?

Yes

2 “related” tags

Find mappings & relation for pair of tags

Wikipedia

Google

SW search engine

No

<concept, relation, concept>

Concept and relation identification

END
Cluster_1: {admin application archive collection component control developer dom example form innovation interface layout planning program repository resource sourcecode}
**Examples**

**Cluster_2**: \{college commerce corporate course education high instructing learn learning lms school student\}

Faceted Ontology

- Ontology creation and maintenance is automated
- Ontology evolution is driven by task features and by user changes
- Large scale integration of ontology elements from massively distributed online ontologies
- Very different from traditional top-down-designed ontologies
The example given provides an example of a new generation of SW applications, with the following features:
- Dynamic use of online knowledge
- SW is used as a large scale repository providing background knowledge to an intelligent problem solver
- No single ontology driving data integration

The new class of systems enabled by the SW is fundamentally different in many respects both from traditional KBS and even from early SW applications.

The difference between 1st and 2nd generation SW applications can be seen as that between “corporate semantic webs” and “open semantic web”
Overview: NR Shadbolt

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CoP
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DH Sleeman
DR
Robertson
Stephen
Harris
Hugh Glaser
M Eisenstadt
mkv
E Motta
Kieron
O'Hara
W Hall
A Tate
Ian Millard
Les Carr
Y Wilks
SW as Enabler of Intelligent Behaviour

Intelligent Behaviour
Intelligence as a function of possessing domain knowledge

Large Body of Knowledge

Intelligent Behaviour
The Knowledge Acquisition Bottleneck

**Large Body of Knowledge**

**KA Bottleneck**

**Knowledge**

**Intelligent Behaviour**
The SW may well provide a solution to one of the classic AI challenges: how to acquire and manage large volumes of knowledge to develop truly intelligent problem solvers and address the brittleness of traditional KBS.
Infrastructure for 2G SW Applications
Architecture of NGSW Apps

Semantic Web Gateway
Current Gateway to the Semantic Web

Swoogle

semantic web search 2006
Limitations of Swoogle

- **Limited quality control mechanisms**
  - Many ontologies are duplicated

- **Limited Query/Search mechanisms**
  - Only keyword search; no distinction between types of elements
  - No support for formal query languages (such as SPARQL)

- **Limited range of ontology ranking mechanisms**
  - Swoogle only uses a 'popularity-based' one

- **Limited API**

- **No support for ontology modularization**
A New Gateway to the Semantic Web

http://watson.kmi.open.ac.uk
• Sophisticated quality control mechanism
  – Detects duplications
  – Fixes obvious syntax problems
    • E.g., duplicated ontology IDs, namespaces, etc..
• Structures ontologies in a network
  – Using relations such as: extends, inconsistentWith, duplicates
• Provides sophisticated API
• Supports formal queries (SPARQL)
• Supports a variety of ontology ranking mechanisms
• Modularization support
• Plug-ins for Protégé and NeOn Toolkit (both under devpt.)
• Very cool logo!
Existing descriptions of the class duck


  ```
  subclassOf  animal
  label       "duck"
  comment     
  ```

- http://cohe.semanticweb.org/ontologies/people#duck

  ```
  subclassOf  animal
  Huey       type
  Dewey      type
  Louie      type
  creationDate "2003-12-03T15:38:19Z"
  label      "duck"
  creator    "seanbo"
  comment    
  ```

- http://www.atl.mco.com/projects/ontology/ontologies/people+pets/people+petsA.owl#duck

  ```
  subclassOf  animal
  Huey       type
  Dewey      type
  Louie      type
  label      "duck"
  comment    
  ```
Distribution of SW documents according to the number of entities, classes and individuals

– SW is characterized by a large number of small documents and a small number of large ones
– This is true for both ontological knowledge (classes) and factual data (individuals)
Great variety: Some topics are almost not covered (e.g. Adult), while some are over represented (e.g. Society, Computers)

As we can expect, a large number of narrow coverage documents and a small number of large coverage ones.
### Density of the online knowledge

<table>
<thead>
<tr>
<th>Measures</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of classes</td>
<td>161,264</td>
</tr>
<tr>
<td>Number of properties</td>
<td>76,350</td>
</tr>
<tr>
<td>Number of individuals</td>
<td>984,526</td>
</tr>
<tr>
<td>Number of sub-class relations</td>
<td>106,729</td>
</tr>
<tr>
<td>P-density (average number of properties per class)</td>
<td>0.47</td>
</tr>
<tr>
<td>H-density (average number of super-classes per class)</td>
<td>0.66</td>
</tr>
<tr>
<td>I-density (average number of instances per class)</td>
<td>6.1</td>
</tr>
</tbody>
</table>
• **Usage of URIs for ontologies: lack of clear recommendation!**
  - Most of the ontologies do not declare their URI
  - URI duplication and reuse:
    • Different versions of an ontology having the same URI (e.g. http://lsdis.cs.uga.edu/proj/semdis/testbed/ used 4 times for 4 different versions, all available)
    • Mistaken use of a well known namespace (e.g. http://www.w3.org/2002/07/owl used as the URI of ontologies)
    • Default URI given by the ontology editor (e.g. http://a.com/ontology, the default URI in the OWL plugin of Protégé, used more than 20 times for ontologies having nothing to do together).
Example #2: Ontology Matching
New paradigm: use of background knowledge

Background Knowledge
(external source)
Proposal:

- rely on online ontologies (Semantic Web) to derive mappings
- ontologies are *dynamically* discovered and combined

Does not rely on any pre-selected knowledge sources.

M. Sabou, M. d’Aquin, E. Motta, “Using the Semantic Web as Background Knowledge in Ontology Mapping”, Ontology Mapping Workshop, ISWC’06. **Best Paper Award**
Ex1: *Chicken* **Vs.** *Food*

\[
\begin{align*}
\text{Chicken} & \subseteq \text{Poultry} \quad \text{(midlevel-onto)} \\
\text{Poultry} & \subseteq \text{Food} \quad \text{(Tap)}
\end{align*}
\]

\[
\frac{\text{Same results for Duck, Goose, Turkey}}{(r1)}
\]

Ex2: *Ham* **Vs.** *Food*

\[
\begin{align*}
\text{Ham} & \subseteq \text{Meat} \quad \text{(pizza-to-go)} \\
\text{Meat} & \subseteq \text{Food} \quad \text{(SUMO)}
\end{align*}
\]

\[
\frac{\text{(r1)}}{(r1)}
\]

Ex3: *Ham* **Vs.** *Seafood*

\[
\begin{align*}
\text{Ham} & \subseteq \text{Meat} \quad \text{(pizza-to-go)} \\
\text{Meat} & \perp \text{Seafood} \quad \text{(wine.owl)}
\end{align*}
\]

\[
\frac{\text{(r3)}}{(r3)}
\]
Large Scale Evaluation

Matching AGROVOC (16k terms) and NALT(41k terms)

<table>
<thead>
<tr>
<th></th>
<th>Nr.</th>
<th>Examples</th>
</tr>
</thead>
<tbody>
<tr>
<td>Subclass ($\sqsubseteq$)</td>
<td>1477</td>
<td>Lamb $\sqsubseteq$ Sheep, Soap $\sqsubseteq$ Detergent, Asbestos $\sqsubseteq$ Pollutant</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Oasis $\sqsubseteq$ Ecosystem, RAM $\sqsubseteq$ ComputerEquipment</td>
</tr>
<tr>
<td>SuperClass ($\supseteq$)</td>
<td>1857</td>
<td>Shop $\supseteq$ Supermarket, Spice $\supseteq$ BlackPepper, Valley $\supseteq$ Canyon</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Infrastructure $\supseteq$ Highway, Storm $\supseteq$ Tornado, Rock $\supseteq$ Crystal</td>
</tr>
<tr>
<td>Disjoint ($\sqsubset$)</td>
<td>229</td>
<td>Fluid $\sqsubset$ Solid, Fluid $\sqsubset$ Gas, Pond $\sqsubset$ River, Plant $\sqsubset$ Animal</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Newspaper $\sqsubset$ Journal, Fruit $\sqsubset$ Vegetable, Female $\sqsubset$ Male</td>
</tr>
<tr>
<td>Total</td>
<td></td>
<td>3563</td>
</tr>
</tbody>
</table>

(derived from 180 different ontologies)

Evaluation: 1600 mappings, two teams

Overall performance comparable to best in class (over 70%)

Ontologies (180) used to derive mappings.

- TAP
- CPE
- Mid-level-ontology.daml
- SUMO.daml
- Economy.daml
Using the SW to provide dynamically background knowledge to tackle the Agrovoc/NALT mapping problem provides the first ever test case in which the SW, viewed as a large scale heterogeneous resource, has been successfully used to address a real-world problem.
<table>
<thead>
<tr>
<th>Error Type</th>
<th>Nr./%</th>
<th>AGROVOC Concept</th>
<th>Labels</th>
<th>Rel.</th>
<th>NALT Concept</th>
<th>Labels</th>
</tr>
</thead>
<tbody>
<tr>
<td>Anchor</td>
<td>114,</td>
<td>c_{6443}</td>
<td>Rams, Tups</td>
<td>⊆</td>
<td>memory</td>
<td>memory</td>
</tr>
<tr>
<td></td>
<td>53%</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>O_1:ram ⊆ O_1:memory</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td></td>
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<td>O_1 = <a href="http://www.arches.uga.edu/~gonen/qos_bilal.owl">http://www.arches.uga.edu/~gonen/qos_bilal.owl</a></td>
<td></td>
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<tr>
<td>Subsumption as generic</td>
<td>40,</td>
<td>c_{3954}</td>
<td>Irrigation</td>
<td>⊆</td>
<td>agriculture</td>
<td>agriculture</td>
</tr>
<tr>
<td>relation</td>
<td>18%</td>
<td></td>
<td></td>
<td></td>
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</tr>
<tr>
<td></td>
<td></td>
<td>O_1:Irrigation ⊆ O_1:SoilCultivation ⊆ O_1:Agriculture</td>
<td></td>
<td></td>
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<td></td>
<td></td>
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<tr>
<td>Subsumption as part-whole</td>
<td>16,</td>
<td>c_{666}</td>
<td>Asia</td>
<td>⊆</td>
<td>Iran</td>
<td>Iran</td>
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<tr>
<td></td>
<td>7%</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td></td>
<td></td>
<td>O_1:Asia ⊆ O_1:WestAsia ⊆ O_1:Iran</td>
<td></td>
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<td>O_1 = <a href="http://islab.hanyang.ac.kr/damlis/Country.daml">http://islab.hanyang.ac.kr/damlis/Country.daml</a></td>
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<tr>
<td>Subsumption as role</td>
<td>11,</td>
<td>c_{11091}</td>
<td>Garlic</td>
<td>⊆</td>
<td>ingredients</td>
<td>ingredients</td>
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<tr>
<td></td>
<td>5%</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>O_1:garlic ⊆ O_1:vegetable ⊆ O_1:ingredient</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>Inaccurate labeling</td>
<td>12,</td>
<td>c_{1693}</td>
<td>Coal</td>
<td>⊆</td>
<td>industry</td>
<td>industry</td>
</tr>
<tr>
<td></td>
<td>5%</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>O_1:coal ⊆ O_1:industry</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
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<tr>
<td>Different View</td>
<td>12,</td>
<td>c_{2943}</td>
<td>Fishes</td>
<td>⊆</td>
<td>lobsters</td>
<td>lobsters</td>
</tr>
<tr>
<td></td>
<td>5%</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>O_1:Fish ⊆ O_1:MarineInvertebrate ⊆ O_1:Crustacean ⊆ O_1:Lobster</td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
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<td>O_1 = <a href="http://139.91.183.30:9090/RDF/VRP/Examples/tap.rdf">http://139.91.183.30:9090/RDF/VRP/Examples/tap.rdf</a></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
The claim that the information on the SW is of poor quality and therefore not useful to support intelligent problem solving is a myth not supported by concrete experience:

Our experience in the NALT/Agrovoc ontology matching benchmark problem shows that without any particularly intelligent filter, the info available on the SW already allows a 85% theoretical precision for our algorithm, well beyond the performance of any other ontology matching algorithm
Conclusions

- SW provides an unprecedented opportunity to build a new generation of intelligent systems, able to exploit large scale background knowledge.
- The large scale background knowledge provided by the SW may address one of the fundamental premises (and holy grails) of AI.
- The SW is not an aspiration: it is a concrete technology that is already in place today and is steadily becoming larger and more robust.
- This new scenario opens up new opportunities, however we also need new methods and tools to support the life-cycle of the envisaged applications, which is the goal of the NeOn project.
- The applications shown in this talk provide an initial taster of the kind of opportunities the SW will provide for intelligent problem solving.