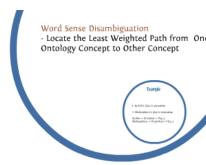


Semantic Web

Making the Web more readable for machines





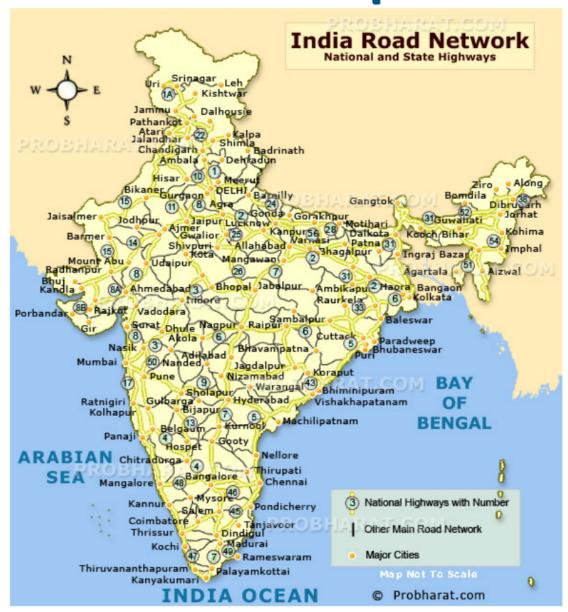


Semantic Web

Making the Web more readable for machines



Road Map





Our Roadmap

Web 2.0
Semantic Web

Representation and Languages Structures and Ontologies

NLP and Semantic Web Conclusion



Web 2.0



Limitation of HTML

Used to represent information NOT DATA

Visual Structured Information

Crux of the Problem

Machine didn't understand what the user actually wants

even if it did

It Didn't understand how to get it

Example Scenario

Online Shopping



As this point in the prolection of the point had the based for its study of the specific and passes, commercing print of points and man. As the second points of the second points and the specific of the second points and the study of the second points are all the man. States was, you have as all the With the langers, With, post to another spice. The sould make an potentions that a componented approach to a componented approach to with field in the option for you, and place you make.

Structuring Required



Type something on a search engine



It just matches the keywords in your query



Lot of work !!!!

~ 1 million hits on an avg query



Limitation of HTML

Used to represent information NOT DATA

Visual Structured Information



Example Scenario

Online Shopping



At this point in the evolution of the Web, your best bet would be to look at different retailers' web pages, comparing prices and shipping times and rates. You could also look for a site that will compare prices and shipping options from several retailers all at once. Either way, you have to do most of the virtual legwork, then make your buying decision and place your order yourself.

With the Semantic Web, you'd have another option. You could enter your preferences into a computerized agent, which would search the Web, find the best option for you, and place your order.





@2006 HowStuffWorks

At this point in the evolution of the Web, your best bet would be to look at different retailers' web pages, comparing prices and shipping times and rates. You could also look for a site that will compare prices and shipping options from several retailers all at once. Either way, you have to do most of the virtual legwork, then make your buying decision and place your order yourself.

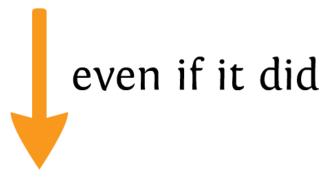


With the Semantic Web, you'd have another option. You could enter your preferences into a computerized agent, which would search the Web, find the best option for you, and place your order.



Crux of the Problem

Machine didn't understand what the user actually wants



It Didn't understand how to get it



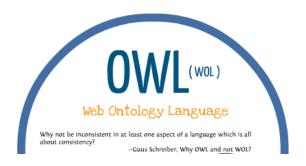
Structuring Required





Representation and Languages

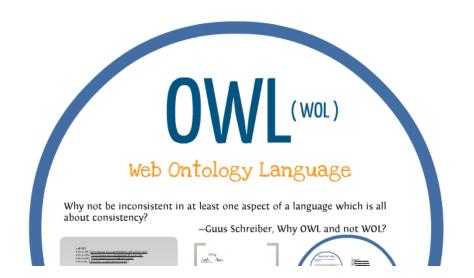








Representation and Languages



Motivation



RDF

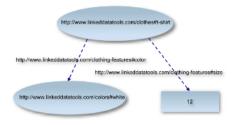
Resource Description Format





Example

-rdf.R0F xmin.sndf= http://www.ms.nm/1999/02/22-ddf-smitzs-ns=" xmin.feature= http://www.indeddatafous.com/dathing-features="2-ddf-features-feature-fe



Reification

Involves representation of factual assertions that are representated by some other assertions

To compare logical assertions from different witnesses in order to determine their credibility

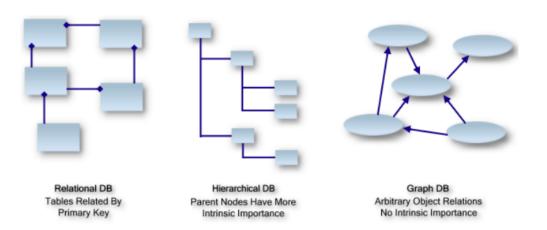
Drawback

RDF, whilst the foundation of defining data structures for the semantic web, does not in itself describe the semantics, or meaning, behind the data

> for that we need Schema or Ontology



Graph Database





Syntax

Defines data in the form of a <Subject,Predicate,Object> triple



Example

```
<rdf:RDF xmlns:rdf="http://www.w3.org/1999/02/22-rdf-syntax-ns#"
    xmlns:feature="http://www.linkeddatatools.com/clothing-features#">
    <rdf:Description rdf:about="http://www.linkeddatatools.com/clothes#t-shirt">
        <feature:size>12</feature:size>
        <feature:color rdf:resource="http://www.linkeddatatools.com/colors#white"/>
        </rdf:Description>
</rdf:RDF>
```



http://www.linkeddatatools.com/clothes#t-shirt http://www.linkeddatatools.com/clothing-features#color http://www.linkeddatatools.com/clothing-features#size http://www.linkeddatatools.com/colors#white 12



Reification

Involves representation of factual assertions that are representated by some other assertions

To compare logical assertions from different witnesses in order to determine their credibility



Drawback

RDF, whilst the foundation of defining data structures for the semantic web, does not in itself describe the semantics, or meaning, behind the data

for that we need Schema or Ontology



DESCRIPTION LOGIC

Motivation

Conceptual Models are required in

- · Artificial Intelligence
- Database Design
- Software Engineering
- · Information Integration

Fundamental Ontology

Conceptual model is populated by

- · Individuals
- related by binary relationships (called roles & features)
- · grouped into classes (concepts)

So we need ability to describe concepts, relationships and individuals

FOL cannot be used as it is not decidable

Description logic is characterized by a set of constructors that allow to build complex concepts and roles from atomic ones.

concept - class / set of objects

role - binary relations on objects











Motivation

Conceptual Models are required in

- Artificial Intelligence
- Database Design
- Software Engineering
- Information Integration



Fundamental Ontology

Conceptual model is populated by

- Individuals
- related by binary relationships (called roles & features)
- grouped into classes (concepts)

So we need ability to describe concepts, relationships and individuals

FOL cannot be used as it is not decidable



Description logic is characterized by a set of constructors that allow to build complex concepts and roles from atomic ones.

concept - class / set of objects

role - binary relations on objects



TBox and ABox

TBox (Terminological Box) : contains sentences describing concept hierarchies i.e; relations between concepts

ABox (Assertional Box) : contains ground sentences stating where in the hierarchy individual belongs (relation between concepts and individuals)

Why two separate boxes?

Various inference methods require different boxes

Classification will require TBox and Instance checking requires ABox.

Example

Every TA is a student (TBox)

Nikhil is a TA (ABox)

Other Significant Properties

No UNA (Unique Name Assumption)

Two concepts/roles having different names may be shown to be equivalent by inference e.g; married_to(X,Y) is equivalent to spouse_of(X,Y)

OWA (Open World Assumption)

Lack of knowledge of a fact does not imply negation of the fact



TBox and ABox

TBox (Terminological Box): contains sentences describing concept hierarchies i.e; relations between concepts

ABox (Assertional Box): contains ground sentences stating where in the hierarchy individual belongs (relation between concepts and individuals)



Example

Every TA is a student (TBox)

Nikhil is a TA (ABox)



Why two separate boxes?

Various inference methods require different boxes

Classification will require TBox and Instance checking requires ABox.



Other Significant Properties

No UNA (Unique Name Assumption)

Two concepts/roles having different names may be shown to be equivalent by inference e.g; married_to(X,Y) is equivalent to spouse_of(X,Y)

OWA (Open World Assumption)

Lack of knowledge of a fact does not imply negation of the fact



(WOL)

Web Ontology Language

Why not be inconsistent in at least one aspect of a language which is all about consistency?

-Guus Schreiber, Why OWL and not WOL?

ordf:RDF
wmlnsrdf-"http://www.wi.org/1999/02/22-rdf-vvntax-nst"
wmlnsrdf-"http://www.wj.org/2000/01/rdf-schemast"
wmlnscwd-"http://www.wj.org/2000/01/rdf-schemast"
wmlnscwd-"http://www.wj.org/2002/07/wwte"
wmlnscwd-"http://www.gd.dcielements/u"
d-DWL Header Example ->
dwbfontology rdfabous-"http://www.linkeddatatools.com/plants">
dcitite:The LinkedbaraTools.com Dample Plant Ontology/dc-titledcddescription.wh.example ontology written for the
LinkedbaraTools.com BDF's & OWL introduction
tutorial/dc-descriptions
d-wid Ontology
d-- Remainder Of Document Omitted For Brevity. ->
d/rdf:RDF>



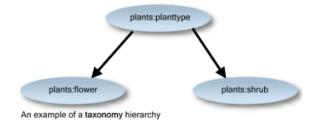






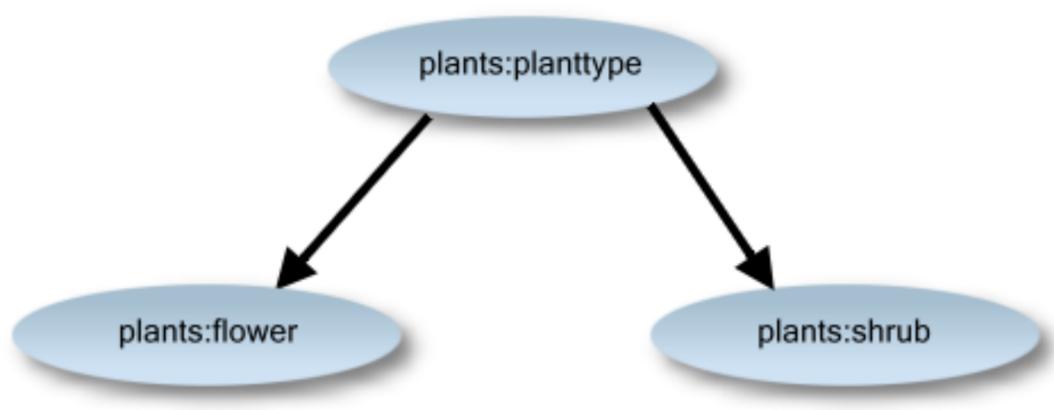
```
<rdf:RDF
xmlns:rdf="http://www.w3.org/1999/02/22-rdf-syntax-ns#"
xmlns:rdfs="http://www.w3.org/2000/01/rdf-schema#"
xmlns:owl="http://www.w3.org/2002/07/owl#"
xmlns:dc="http://purl.org/dc/elements/1.1/">
<!-- OWL Header Example -->
<owl:Ontology rdf:about="http://www.linkeddatatools.com/plants">
<dc:title>The LinkedDataTools.com Example Plant Ontology</dc:title>
<dc:description>An example ontology written for the
LinkedDataTools.com RDFS & OWL introduction
tutorial</dc:description>
</owl:Ontology>
<!-- Remainder Of Document Omitted For Brevity... -->
</rdf:RDF>
```





<rdf:RDF <!-- OWL Subclass Definition - Flower --> <owl:Class rdf:about="http://www.linkeddatatools.com/plants#flowers"> <!-- Flowers is a subclassification of planttype --> <rdfs:subClassOf rdf:resource="http://www.linkeddatatools.com/plants#planttype"/> <rdfs:label>Flowering plants</rdfs:label> <rdfs:comment>Flowering plants, also known as angiosperms.</rdfs:comment> </owl:Class> <!-- OWL Subclass Definition - Shrub --> <owl:Class rdf:about="http://www.linkeddatatools.com/plants#shrubs"> <!-- Shrubs is a subclassification of planttype --> <rdfs:subClassOf rdf:resource="http://www.linkeddatatools.com/plants#planttype"/> <rdfs:label>Shrubbery</rdfs:label> <rdfs:comment>Shrubs, a type of plant which branches from the base.</rdfs:comment> </owl:Class> <!-- Individual (Instance) Example RDF Statement --> <rdf:Description rdf:about="http://www.linkeddatatools.com/plants#magnolia"> <!-- Magnolia is a type (instance) of the flowers classification --> <rdf:type rdf:resource="http://www.linkeddatatools.com/plants#flowers"/> </rdf:Description> </rdf:RDF>





An example of a taxonomy hierarchy



```
<rdf:RDF
<!-- OWL Subclass Definition - Flower -->
<owl:Class rdf:about="http://www.linkeddatatools.com/plants#flowers">
<!-- Flowers is a subclassification of planttype -->
<rdfs:subClassOf rdf:resource="http://www.linkeddatatools.com/plants#planttype"/>
<rdfs:label>Flowering plants</rdfs:label>
<rdfs:comment>Flowering plants, also known as angiosperms.</rdfs:comment>
</owl:Class>
<!-- OWL Subclass Definition - Shrub -->
<owl:Class rdf:about="http://www.linkeddatatools.com/plants#shrubs">
<!-- Shrubs is a subclassification of planttype -->
<rdfs:subClassOf rdf:resource="http://www.linkeddatatools.com/plants#planttype"/>
<rdfs:label>Shrubbery</rdfs:label>
<rdfs:comment>Shrubs, a type of plant which branches from the
base.</rdfs:comment>
</owl:Class>
<!-- Individual (Instance) Example RDF Statement -->
<rdf:Description rdf:about="http://www.linkeddatatools.com/plants#magnolia">
<!-- Magnolia is a type (instance) of the flowers classification -->
<rdf:type rdf:resource="http://www.linkeddatatools.com/plants#flowers"/>
</rdf:Description>
</rdf:RDF>
```



Properties / roles

Individuals in OWL are related by properties

Object properties (owl:ObjectProperty) relates individuals (instances) of two OWL classes.

Datatype properties (owl:DatatypeProperty) relates individuals (instances) of OWL classes to literal values.

```
crdf;RDF

{!- Define the family property ->
cowl-DatastypeProperty rdf;about="http://www.linkeddatatools.com/plants#family"/;

{!- Define the similarlyPopularTo property ->
cowl-ObjectProperty rdf;about="http://www.linkeddatatools.com/plants#similarlyPopularTo"/;

{!- Define the Orchid class instance ->
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#orchid">
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
cl- Define the Magnolia class instance ->
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
crdf;Description rdf;about="http://www.linkeddatatools.com/plants#magnolia">
crdf;Description rdf;Description rdf;Descript
```

ord RDF

In Jeffin by Inning property—

towed BrazappeProperty

rife abouts—Inning beam Linkeddatatools compliances family / /

refil Description

for abouts—Inning beam Linkeddatatools compliances magnolia*

de-Magnolia is a type (material of the flowers class—

de-Magnolia is a type (material of the flowers class—

for resource—Furth University linkeddatatools compliances flowers/

de-The magnolia is port of the Magnoliacose (amily —

opinite Samily Magnoliacose dylants family wird! Description

of the DDF.



Properties / roles

Individuals in OWL are related by properties

Object properties (owl:ObjectProperty) relates individuals (instances) of two OWL classes.

Datatype properties (owl:DatatypeProperty) relates individuals (instances) of OWL classes to literal values.



```
<rdf:RDF
<!-- Define the family property -->
<owl:DatatypeProperty</pre>
rdf:about="http://www.linkeddatatools.com/plants#family"/>
<rdf:Description
rdf:about="http://www.linkeddatatools.com/plants#magnolia">
<!-- Magnolia is a type (instance) of the flowers class -->
<rdf:type
rdf:resource="http://www.linkeddatatools.com/plants#flowers"/>
<!-- The magnolia is part of the 'Magnoliaceae' family -->
<plants:family>Magnoliaceae</plants:family>
</rdf:Description>
</rdf:RDF>
```



```
<rdf:RDF
<!-- Define the family property -->
<owl:DatatypeProperty rdf:about="http://www.linkeddatatools.com/plants#family"/>
<!-- Define the similarlyPopularTo property -->
<owl:ObjectProperty rdf:about="http://www.linkeddatatools.com/plants#similarlyPopularTo"/>
<!-- Define the Orchid class instance -->
<rdf:Description rdf:about="http://www.linkeddatatools.com/plants#orchid">
</rdf:Description>
<!-- Define the Magnolia class instance -->
<rdf:Description rdf:about="http://www.linkeddatatools.com/plants#magnolia">
<!-- Magnolia is an individual (instance) of the flowers class -->
<rdf:type rdf:resource="http://www.linkeddatatools.com/plants#flowers"/>
<!-- The magnolia is part of the 'Magnoliaceae' family -->
<plants:family>Magnoliaceae</plants:family>
<!-- The magnolia is similarly popular to the orchid -->
<plants:similarlyPopularTo rdf:resource="http://www.linkeddatatools.com/plants#orchid"/>
</rdf:Description>
</rdf:RDF>
```



LEMON

Lexicon model for ontologies

- Separate Lexicon models and Ontologies
- Linking them can aid in lot of NLP applications like Q/A, Machine Translation



Lexicon is the vocabulary of the language

Uses RDF for defining links

Meaning of a word given by reference

Reference (Ontology) capable of representing more complex semantic information Lemon model provides a principled chain between the semantic representation and its linguistic realization



Lexicon is the vocabulary of the language

Uses RDF for defining links

Meaning of a word given by reference

Reference (Ontology) capable of representing more complex semantic information



Lemon model provides a principled chain between the semantic representation and its linguistic realization



POWLA

Interoperability of Linguistic corpora

Motivation

Same text, Different annotations POS tags, Parse trees, etc

A processing stage may require different annotations to the same text

Common representation which provides access to all the linguistic information conveyed in the annotations

Structural Interoperatibility

Use RDF to represent all the annotations of the corpus in an interoperable way, integrate their information without restrictions and query the information

Use OWL/DL to specify and verify formal constraints on the correct representation of linguistic corpora in RDF

Conceptual Interoperatibility

Heterogenous annotation schemes

Terminological Reference repository

Provide interlingua that allow mapping form scheme A to scheme B

GOLD



Motivation

Same text, Different annotations POS tags, Parse trees, etc

A processing stage may require different annotations to the same text

Common representation which provides access to all the linguistic information conveyed in the annotations



Structural Interoperatibility

Use RDF to represent all the annotations of the corpus in an interoperable way, integrate their information without restrictions and query the information

Use OWL/DL to specify and verify formal constraints on the correct representation of linguistic corpora in RDF



Conceptual Interoperatibility

Heterogenous annotation schemes

Terminological Reference repository

Provide interlingua that allow mapping form scheme A to scheme B

GOLD



us

Semantic Web

Structures and Ontologies

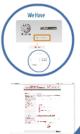
BABLENET

refulinged but alled a warsh return? attended in a said by bridge the largest refulinged if the recombinational between the termination of the computational between it the









INDO WORDNET

riini bei ultimendolge haar of varalines et si deulekel languages et teolia mannen, langka belon languasi, Pitelij Karensi kateniat, Frankai Vollagiane, manjangi Varant igalij Dejes, Perjaks Garaksi, Land, Uduge an





NOTAL SELLTY

CHARLES SELLY

CHARLES

Cyc (pronounced like syke)

en i un meritad tarbiga ann des de dinggio e des del proposicio de principalità fon el merita como













BABLENET

multilingual lexicalized semantic network. automatically created by linking the largest multilingual Web encyclopedia - i.e., Wikipedia - to the most popular computational lexicon of the English language - i.e., Wordnet

MOTIVATION

- Manually maintaining and updating lexical knowledge resources is expensive and timeconsuming.
- consuming.
 Second, such resources are typically lexicographic, and thus contain mainly concepts and only a few named entities
- concepts and only a few named entities.

 Third, resources for non-English languages often have a much poorer coverage since the construction effort must be repeated for every language of interest.

How to tackle these?

POTENTIAL

Experiments show that this fully-internated approach produces a large-scale lexical reserver

The researce includes willions of semantic relations, mainly from Wikipedia flowever, Wordhiet relations are labeled, and contains almost 5 million concepts (6.7 labels percently on average).

Such coverage is much wider than that of coloning worders in non-Inglish languages. While failefiller correctly includes 6 languages, in in freely-worldbille worders can introducely be catalohibed by athleting the English WordVict as an

Stabilities can be extended to virtually any language of interest. The translation method allogs it to cowith any translation poor language.

APPLICATIONS

Multilingual Natural Language Processing
-Semantic relatedness
-Multilingual word sense disambiguation.

We Have







BABLENET

multilingual lexicalized semantic network. automatically created by linking the largest multilingual Web encyclopedia - i.e., Wikipedia - to the most popular computational lexicon of the English language - i.e., Wordnet



multilingual lexicalized semantic network. automatically created by linking the largest multilingual Web encyclopedia - i.e., Wikipedia - to the most popular computational lexicon of the English language - i.e., Wordnet



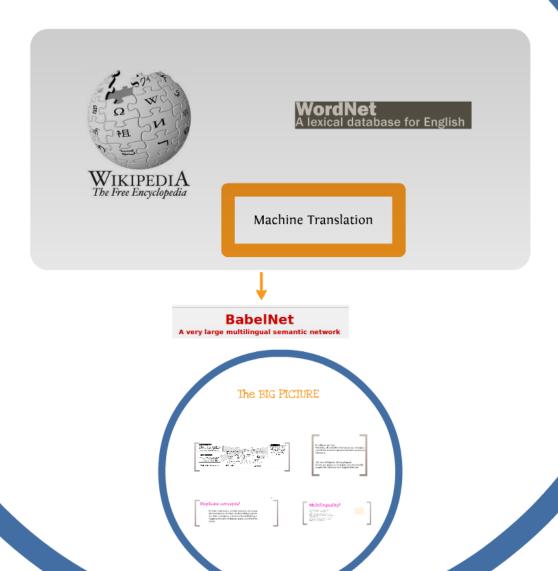
MOTIVATION

- Manually maintaining and updating lexical knowledge resources is expensive and timeconsuming.
- Second, such resources are typically lexicographic, and thus contain mainly concepts and only a few named entities.
- Third, resources for non-English languages often have a much poorer coverage since the construction effort must be repeated for every language of interest.

How to tackle these?

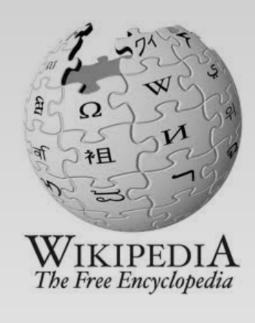


We Have





e Have



WordNet A lexical database for English

Machine Translation



BabelNet

A very large multilingual semantic network



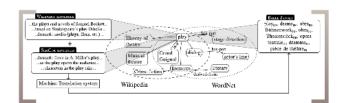


BabelNet

A very large multilingual semantic network



The BIG PICTURE



It collectss (a) from WordNet, all available word senses (as concepts) and all the semantic pointers between synsets (as relations):

(b) from Wikipedia, all encyclopedic entries (i.e. pages, as concepts) and semantically unspecified relations from hyperlinked text.

Duplicate concepts?

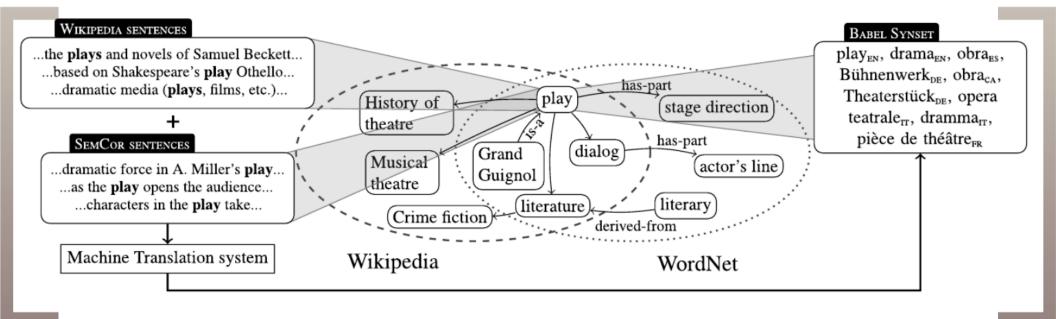
In order to provide a unified resource, we merge the intersection of these two knowledge sources (i.e. their concepts in common) by establishing a mapping between Wikipedia pages and WordNet senses

Multilinguality?

The less call resilvations of the available concepts in different languages by using (a) the homeon generated translations provided in whitepoint phenoconfluinter-language lends, as well as (b) a meeting translation system as translation concerns of the concepts within sense-consenses of the concepts within sense-consense of concepts and the sense.







It collectss (a) from WordNet, all available word senses (as concepts) and all the semantic pointers between synsets (as relations);

(b) from Wikipedia, all encyclopedic entries (i.e. pages, as concepts) and semantically unspecified relations from hyperlinked text.



Duplicate concepts?

In order to provide a unified resource, we merge the intersection of these two knowledge sources (i.e. their concepts in common) by establishing a mapping between Wikipedia pages and WordNet senses



Multilinguality?

The lexical realizations of the available concepts in different languages by using

- (a) the human-generated translations provided in Wikipedia (the so-called inter-language links), as well as
- (b) a machine translation system to translate occurrences of the concepts within sensetagged corpora, namely SemCor







Multiinguaii

The lexical realizations of the available concepts in different languages by using

- (a) the human-generated translations provided in Wikipedia (the so-called inter-language links), as well as
- (b) a machine translation system to translate occurrences of the concepts within sensetagged corpora, namely SemCor



Print/export

Languages

العربية

Azərbaycanca

Беларуская

Беларуская (тарашкевіца)

Български

Brezhoneg

Català

Česky

Cymraeg

Dansk

Deutsch

Eesti

Ελληνικά

Español

Esperanto

Euskara



Terminology



Some marine diatoms - a key phytoplankton group

The name property capable of currents. The currents of the cur

Within the plant larval stage

Plankton al

The study of

The adjection more strictly the root of the strictly the root of the strictly the s

Trophic groups

Plankton are primarily divided into broad functional (or

 Phytoplankton (from Greek phyton, or plant), auto important groups are the diatoms, cyanobacteria, d

- Zoonlankton (from Crook zoon, or animal), small

Multiinguaii

The lexical realizations of the available concepts in different languages by using

- (a) the human-generated translations provided in Wikipedia (the so-called inter-language links), as well as
- (b) a machine translation system to translate occurrences of the concepts within sensetagged corpora, namely SemCor



 (s snum=3) <wf cmd=ignore pos=DT>The</wf> <wf cmd=done pos=NN lemma=september wnsn=1</p> lexsn=1:28:00::>September</wf> <wf cmd=done pos=NN lemma=october wnsn=1 lexsn=1:28:00::>October</wf> <wf cmd=done pos=NN lemma=term wnsn=2</p> lexsn=1:28:00::>term</wf> <wf cmd=done pos=NN lemma=jury wnsn=1 lexsn=1:14:00::>jury</wf> <wf cmd=done pos=VBD ot=notag>had</wf> <wf cmd=done pos=VBN ot=notag>been</wf> <wf cmd=done pos=VB lemma=charge wnsn=5</p> lexsn=2:41:00::>charged</wf> <wf cmd=ignore pos=IN>by</wf> <wf cmd=done rdf=location pos=NNP</p> lemma=location wnsn=1 lexsn=1:03:00:: pn=location>Fulton</wf> <wf cmd=done rdf=person pos=NNP lemma=person</p> wnsn=1 lexsn=1:03:00:: pn=person>Superior_Court_Judge_Durwood_Pye</w <wf cmd=ignore pos=TO>to</wf> <wf cmd=done pos=VB lemma=investigate wnsn=2 lexsn=2:32:01::>investigate</wf> <wf cmd=done pos=NN lemma=report wnsn=2</p> lexsn=1:10:00::>reports</wf> <wf cmd=ignore pos=IN>of</wf> <wf cmd=done pos=JJ lemma=possible wnsn=2</pre> lexsn=3:00:04::>possible</wf> <punc>``</punc> <wf cmd=done pos=NN lemma=irregularity wnsn=1 lexsn=1:04:00::>irregularities(/wf> <punc>"</punc> <wf cmd=ignore pos=IN>in</wf> <wf cmd=ignore pos=DT>the</wf> <wf cmd=done pos=IJ lemma=hard-fought wnsn=1</pre> lexsn=5:00:00:difficult:00>hard-fought</wf> <wf cmd=done pos=NN lemma=primary wnsn=1 lexsn=1:04:00::>primary</wf> <wf cmd=ignore pos=WDT>which</wf> <wf cmd=done pos=VBD ot=notag>was</wf> <wf cmd=done pos=VB lemma=win wnsn=1 lexsn=2:33:00::>won</wf> <wf cmd=ignore pos=IN>by</wf> <wf cmd=done rdf=person pos=NNP lemma=person</p> wnsn=1 lexsn=1:03:00:: pn=person>Mayornominate_Ivan_Allen_Ir.</wf> <punc>.</punc> </s>



POTENTIAL

Experiments show that this fully-automated approach produces a large-scale lexical resource with high accuracy.

The resource includes millions of semantic relations, mainly from Wikipedia (however, WordNet relations are labeled), and contains almost 3 million concepts (6.7 labels per concept on average).

Such coverage is much wider than that of existing wordnets in non-English languages. While BabelNet currently includes 6 languages, links to freely-available wordnets can immediately be established by utilizing the English WordNet as an interlanguage index.

BabelNet can be extended to virtually any language of interest. The translation method allows it to cope with any resource-poor language.

APPLICATIONS

Multilingual Natural Language Processing

- -Semantic relatedness
- -Multilingual word sense disambiguation.









BabelNet

A very large multilingual semantic network

BabelNetXplorer, an online interface, is available here.



Papers

Roberto Navigli & Simone Paolo Ponzetto BabelNet: Building a very large multilingual semantic network

In: Proceedings of the 48th Annual Meeting of the Association for Computational Linguistics, Uppsala, Sweden, 11-16 July 2010, pp. 216-225.





Roberto Navigli & Simone Paolo Ponzetto **BabelNet: The Automatic** Construction, Evaluation and Application of a Wide-Coverage **Multilingual Semantic Network**

Artificial Intelligence, 193, Elsevier, 2012, pp. 217-250.





Roberto Navigli & Simone Paolo Ponzetto BabelNetXplorer: A Platform for Multilingual Lexical Knowledge Base Access and Exploration

In: Companion Volume to the Proceedings of the 21st World Wide Web Conference, Lyon, France, 16-20 April 2012, pp. 393-396.





PDF BibTeX

Roberto Navigli & Simone Paolo Ponzetto Multilingual WSD with Just a Few Lines of Code: The BabelNet API

In: Proceedings of the 50th Annual Meeting of the Association for Computational Linguistics, Jeju Island, South Korea, 8-14 July 2012.







Overview

From this page you can download: BabelNet, a very large multilingual semantic network with millions of concepts obtained from:

- · an integration of WordNet and Wikipedia based on an automatic mapping
- · translations of the concepts (i.e. English Wikipedia pages and WordNet synsets) based on Wikipedia cross-language links and the output of a machine translation

The resource was automatically generated using the methods presented in our ACL-10 and AlJ papers (see "Papers" in the left box).

Download



BabelNet API version 1.0.1 (August 2012)



BabelNet precompiled index version 1.0 (July 2012)



BabelNet dump version 1.0 (July 2012)



abelNet glosses version 1.0 (July 2012)

How to get started with BabelNet

The best way to start is to download the BabelNet API and follow the instructions found in the README file. You can also download the 'raw' data (i.e., the BabelNet dump and glosses files) and use them with other software (e.g., your favourite API). Details on the APIs and the data format can be found in our demo papers (again, please refer to the papers found in the left box).

Acknowledgments

BabelNet is an output of the ERC Starting Grant MultiJEDI No. 259234.







INDO WORDNET

linked lexical knowledge base of wordnets of 18 scheduled languages of India

Assamese, Bangla, Bodo, Gujarati, Hindi, Kannada, Kashmiri, Konkani, Malayalam, Manipuri, Marathi, Nepali, Oriya, Punjabi, Sanskrit, Tamil, Telugu and Urdu.

MOTIVATION

Indian languages form a very significant component of the languages landscape of the world.

Many languages rank within top to in the world in terms of the population speaking them, e.g., Hindi-Urdu 5th, Bangla 7th, Marathi 12th and so on as per the List of languages by number of native speakers.



Indowordnet project started with creations of Hindi WordNet by the Natural Language Processing group at the Center for Indian Language Technology (CFILT) in the Computer Science and Engineering Department at IIT Rombay

AVAILABILITY

Publicaly browsable at http://www.cfilt.iitb.ac.in/indowordnet Available under GNU licence



INDO WORDNET

linked lexical knowledge base of wordnets of 18 scheduled languages of India

Assamese, Bangla, Bodo, Gujarati, Hindi, Kannada, Kashmiri, Konkani, Malayalam, Manipuri, Marathi, Nepali, Oriya, Punjabi, Sanskrit, Tamil, Telugu and Urdu.





linked lexical knowledge base of wordnets of 18 scheduled languages of India

Assamese, Bangla, Bodo, Gujarati, Hindi, Kannada, Kashmiri, Konkani, Malayalam, Manipuri, Marathi, Nepali, Oriya, Punjabi, Sanskrit, Tamil, Telugu and Urdu.



MOTIVATION

Indian languages form a very significant component of the languages landscape of the world.

Many languages rank within top 10 in the world in terms of the population speaking them, e.g., Hindi-Urdu 5th, Bangla 7th, Marathi 12th and so on as per the List of languages by number of native speakers.





Indowordnet project started with creations of Hindi WordNet by the Natural Language Processing group at the Center for Indian Language Technology (CFILT) in the Computer Science and Engineering Department at IIT Bombay.



AVAILABILITY

Publicaly browsable at http://www.cfilt.iitb.ac.in/indowordnet Available under GNU licence

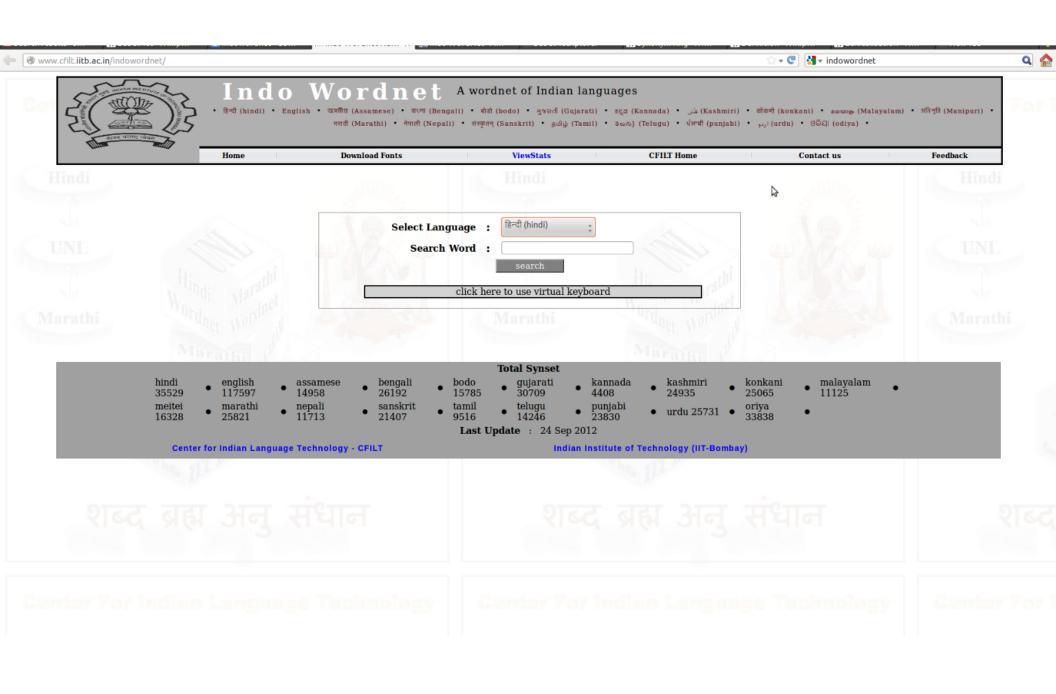




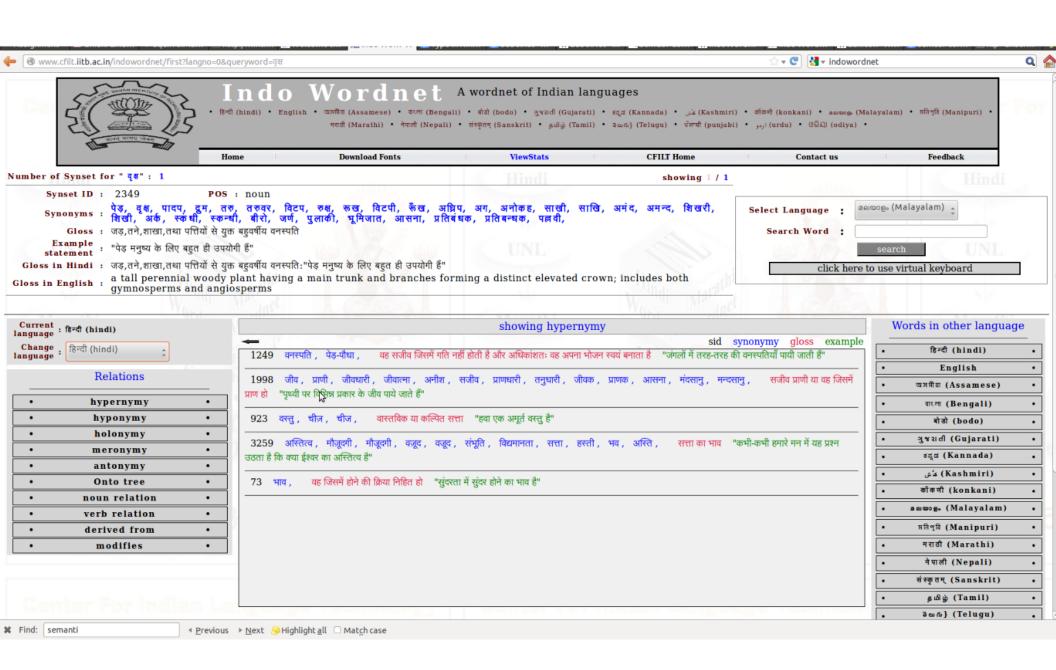




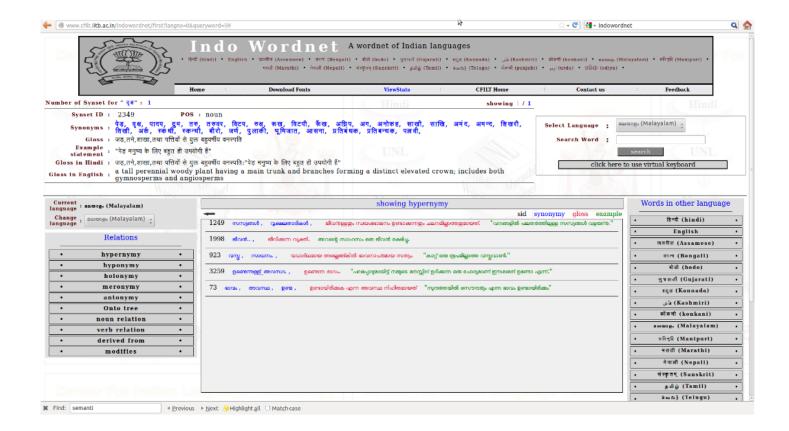














Cyc (pronounced like syke)

Cyc is an artificial intelligence project that attempts to assemble a comprehensive ontology and knowledge base of everyday common sense knowledge.

MOTIVATION

To codify, in machine-usable form, millions of pieces of knowledge that compose human common sense.

To enable AI applications to perform humanlike reasoning.



INFERENCE ENGINE

An inference engine is a computer program that tries to derive answers from a knowledge base. The Cyc inference engine performs general logical deduction (including modus porens, modus tollens, universal quantification and existential quantification).

Typical pieces of knowledge represented in the fatabase are "Every tree is a plant" and "Plants dis eventually". When asked whether trees die, the inference engine can draw the obvious conclusion and assume the execution convention.

AVAILABILITY

The original knowledge base is proprietary, but a smaller version of the knowledge base, intended to establish a common vocabulary for automatic reasoning, was released as OpenCyc under an open source (Apache) license. More recently, Cyc has been made available to Al researchers under a researchpurposes license as ResearchCyc.

APPLICATIONS





ronouncedlike

Cyc is an artificial intelligence project that attempts to assemble a comprehensive ontology and knowledge base of everyday common sense knowledge



MOTIVATION

To codify, in machine-usable form, millions of pieces of knowledge that compose human common sense.

To enable AI applications to perform human-like reasoning.



KNOWLEDGE BASE

The Knowledge Base (KB) contains over one million human-defined assertions, rules or common sense ideas.

REPRESENTATION

The concept names in Cyc are known as constant Constants start with an optional "#\$" and are case-sensitive. There are constants for:

Individual items known as individuals, such as #Stillclintum or #Strance.

Collections, each as #STree-ThePlant (containing all trees) or #SEquivalenceRelation (containing all equivalence relations). A member of a collection is called an instance of that collection.

Truth Functions which can be applied to one or more other concepts and return either true or false. For example #5stblings is the stbling relationship, true if the two arguments are stblings.

Functions, which produce new serms from given enes, for example, estimatin, when provided with an argument describing a type (or collection) of plants, will return the collection of its fruits, by convention, function constants start with an upper-case letter and end with the string 'Fin'.

CycL

These are formulated in the language Cycl, which is based on predicate calculus and has a syntax similar to that of the Lisp programming language.

"All trees are plants".

og

(#8in #98iliCinton #8t/vitedStatesFresident)

Will Clinton belange to the collection of U.S.
presidency

(#8gonis #8free-TheFlant #8Flant)



KNOWLEDGE BASE

The Knowledge Base (KB) contains over one million human-defined assertions, rules or common sense ideas.



The concept names in Cyc are known as constants. Constants start with an optional "#\$" and are case-sensitive. There are constants for:

Individual items known as individuals, such as #\$BillClinton or #\$France.

Collections, such as #\$Tree-ThePlant (containing all trees) or #\$EquivalenceRelation (containing all equivalence relations). A member of a collection is called an instance of that collection.

Truth Functions which can be applied to one or more other concepts and return either true or false. For example #\$siblings is the sibling relationship, true if the two arguments are siblings.

Functions, which produce new terms from given ones. For example, #\$FruitFn, when provided with an argument describing a type (or collection) of plants, will return the collection of its fruits. By convention, function constants start with an upper-case letter and end with the string "Fn".



CycL

These are formulated in the language CycL, which is based on predicate calculus and has a syntax similar to that of the Lisp programming language.

```
eg:

(#$isa #$BillClinton #$UnitedStatesPresident)

"Bill Clinton belongs to the collection of U.S. presidents"

(#$genls #$Tree-ThePlant #$Plant)

"All trees are plants".
```



INFERENCE ENGINE

An inference engine is a computer program that tries to derive answers from a knowledge base. The Cyc inference engine performs general logical deduction (including modus ponens, modus tollens, universal quantification and existential quantification).

Typical pieces of knowledge represented in the database are "Every tree is a plant" and "Plants die eventually". When asked whether trees die, the inference engine can draw the obvious conclusion and answer the question correctly.



AVAILABILITY

The original knowledge base is proprietary, but a smaller version of the knowledge base, intended to establish a common vocabulary for automatic reasoning, was released as OpenCyc under an open source (Apache) license. More recently, Cyc has been made available to Al researchers under a researchpurposes license as ResearchCyc.



APPLICATIONS

Terrorism Knowledge Base

The comprehensive Terrorism Knowledge Base is an application of Cyc in development that will try to ultimately contain all relevant knowledge about "terrorist" groups, their members, leaders, ideology, founders, sponsors, affiliations, facilities, locations, finances, capabilities, intentions, behaviors, tactics, and full descriptions of specific terrorist events. The knowledge is stored as statements in mathematical logic, suitable for computer understanding and reasoning.

Cyclopedia

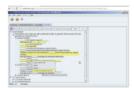
Cyclopedia is being developed; it superimposes Cyc keywords on pages taken from Wikipedia pages

Cleveland Clinic Foundation

The Cleveland Clinic has used Cyc to develop a natural language query interface of biomedical information.





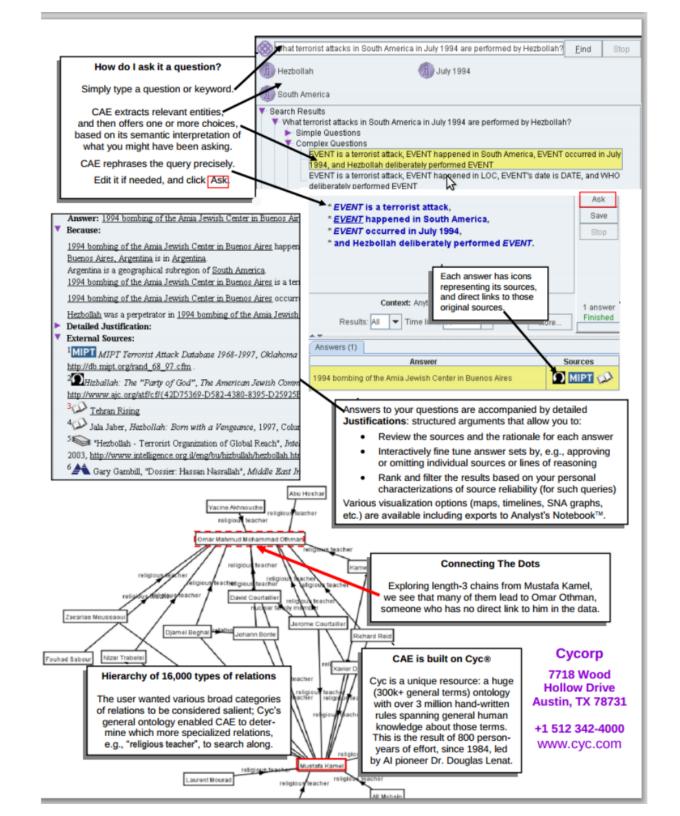




Terrorism Knowledge Base

The comprehensive Terrorism Knowledge Base is an application of Cyc in development that will try to ultimately contain all relevant knowledge about "terrorist" groups, their members, leaders, ideology, founders, sponsors, affiliations, facilities, locations, finances, capabilities, intentions, behaviors, tactics, and full descriptions of specific terrorist events. The knowledge is stored as statements in mathematical logic, suitable for computer understanding and reasoning.





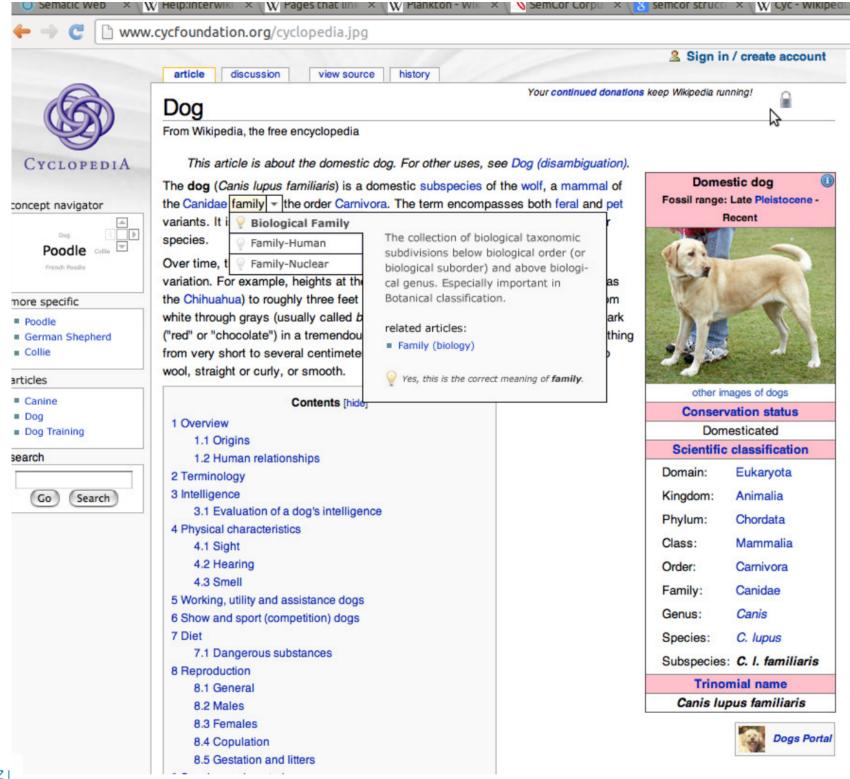


i casoiiiig.

Cyclopedia

Cyclopedia is being developed; it superimposes Cyc keywords on pages taken from Wikipedia pages



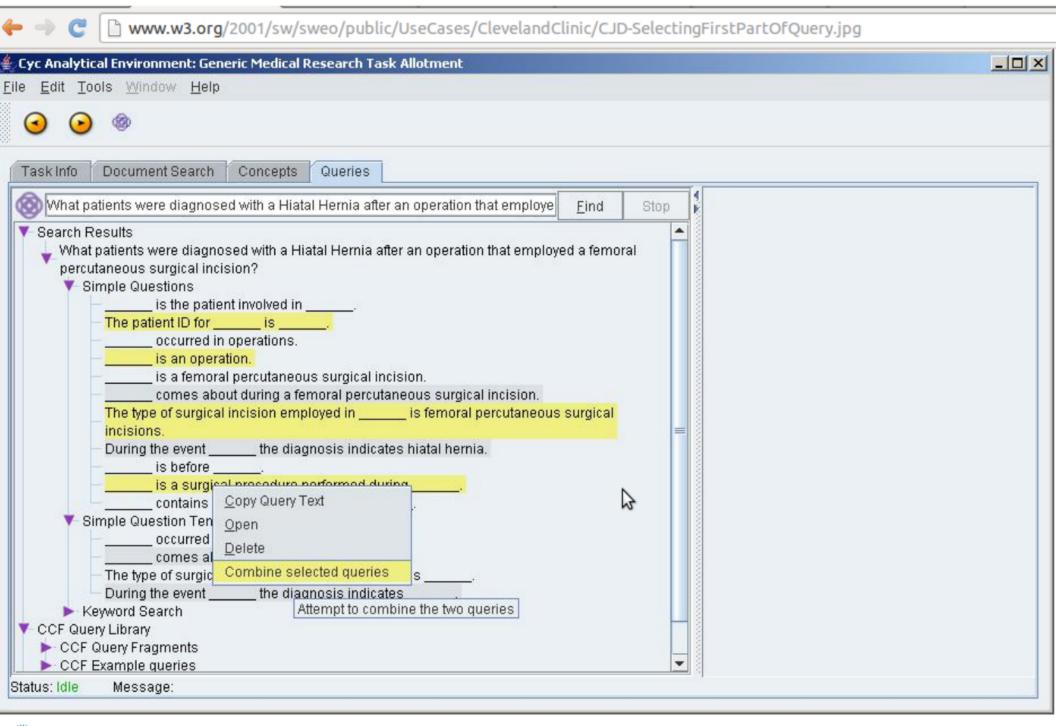




Cleveland Clinic Foundation

The Cleveland Clinic has used Cyc to develop a natural language query interface of biomedical information.







NLP & Semantic web

- Relationship between NLP & Semantic Web
- Mutual Benefit



Why NLP needed for Semantic Web?

Ontology Query



1.Ontology Query2.Ontology Learning3.MultilingualOntology Mapping



Ontology Query

- -Semantic Web has the quite complex structure and strict logical form
- It is not expected that Users will make a query logic form

NL Query: Who is wife of Rama?

Logical Query:

MarriedTo(?,Rama) and typeOf(?,Men)



Who is wife of Rama?

NL to Formal Query

MarriedTo(X,Rama) and typeOf(X,Women)

Semantic Web Agent

 $X = \langle ABCD \rangle$

Presentation

Wife of Rama is <ABCD>.



Ontology Learning



Unstructured sources Involves NLP techniques, morphological and syntactic analysis, etc.

Semi-structured source elicit an ontology from sources that have some predefined structure, such as XML Schema

Structured data

Extracting concepts and relations from knowledge contained in structured data, such as databases



Layer - Learning

Vx,y Suffering $(x,y) \rightarrow ill(x)$ Axioms & Rules Relations cure(doctor, disease) is_a(doctor, person) Taxonomy disease Concepts disease, illness Synonyms **Terms** Disease, illness, Hospital



Terms

- Linguistic realizations of domainspecific concepts
- Are the basis of the ontology learning process

Term Extraction

- -Run a Part-Of-Speech (POS) tagger over the domain corpus
- -Identify possible terms by constructing patterns, such as: Adj-Noun, Noun-noun, Adj-Noun-Noun,...
- -Ignore Names
- -Identify only the relevant to the text terms by applying statistical metrics



Taxonomy Extraction

1.With the use of WordNet

2.Co-occurrence Analysis

3.

Lexico-syntactic patterns



Taxonomy Extraction with wordnet

-Given two terms t1 and t2, check if they stand in a hypernym relation with regard to WordNet

-Normalize the number of hypernym paths by dividing by the number of senses of t1

Example: 4 different hypernym paths between synsets 'country' and 'region' - 'country' has 5 senses

value of isa (country, region) = 0.8



Lexico-syntactic patterns

- NPo such as {NP1, NP2,..., (and | or)} NPn Vehicles such as cars, trucks and bikes....
- such NP as {NP,} * { (or | and) } NP Such fruits as oranges, nectarines or apples...

S

- NP {, NP} * { , } { or | and } other NP Swimming, running, or/and other activities...
- NP { , } including {NP, } * { or | and } NP Injuries, including broken bones, wounds and bruises...



Relations

-General Relations

Exploiting linguistic structure

Ex: Author wrote a book.

Relation: write(Author, Book)



Why Semantic Web is needed for NLP?



mny semantic web is needed for NLP?

Word Sense Disambiguation

 Locate the Least Weighted Path from One Ontology Concept to Other Concept

Example

- 1. Sachin's play is awesome.
- 2. Shakesphere's play is awesome.



Example

- 1. Sachin's play is awesome.
- 2. Shakesphere's play is awesome.

Sachin -> Cricketer-> Play_1 Shakesphere -> Playwriter-> Play_2



Conclusion



Conclusion

- -Without any doubt, Semantic Web needs natural language technology
 - -To acquire knowledge from massive unstructured/ semi-structured Web 2.0
 - -To understand user's Natural Language Query
 - -Finally, To present answers to users
- -Semantic Web should improve the NLP's performance in IE and WSD
- -Almost all problems are in open to research, a long journey to go.



References

- 1. Philipp Cimiano and Bernardo Magnini, Ontology Learning from Text: An Overview, IOS Press, 2003
- 2. Ruiqiang Guo, Fuji Ren, Towards the Relationship Between Semantic Web and NLP, IEEE, 2009
- 3. C'ssia Trojahn , Paulo Quaresma , Renata Vieira, A Framework for Multilingual Ontology Mapping, The International Conference on Language Resources and Evaluation, 2008
- 4. Christian Chiarcos, Interoperability of Corpora and Annotations, DFG, 2011
- 5. John McCrae, Dennis Spohr, and Philipp Cimiano Linking Lexical Resources and Ontologies on the Semantic Web with lemon, AG Semantic Computing, CITEC, University of Bielefeld, 2012
- 6. Roberto Navigli, Simone Paolo Ponzetto, BabelNet: Building a Very Large Multilingual Semantic Network, ACL, 2012

