Slides based on

Jurafsky and Martin

"Speech and Language Processing"

## Semantics 1/3 (Intro & representing meanings)

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#### **Semantics: 3 parts**

#### 1 - Representing meanings:

- The meaning of linguistic utterances, represented with formal structures
- Based on representation languages
- Often, a Knowledge Base is used to describe the "world" that represents the "context"

#### 2 - Semantic analysis:

- Semantic Analysis: from linguistic utterances to meaning representations
- Information Extraction
- Both SA and IE do not consider the ambiguity of words

#### 3 - Lexical semantics:

- How to represent meanings of words
- How to solve ambiguities (Word Sense Disambiguation)

## Representing meanings: Languages

#### Desiderata for representations

- Verifiable: It should be possible to verify that the sentence (i.e. the meaning of it) is true or false "Does Maharani serve vegetarian food?"
- Unambiguous: Ambiguities should be solved "I wanna eat someplace that's close to Politecnico"

(like Godzilla???)

"I want to eat Italian food" (which kind of italian food?)

• Canonical form: Different sentences with the same meaning should produce the same representation "Does Maharani serve vegetarian food?"

"Do they have vegetarian food at Maharani?"

### Desiderata for representations

• Inferences and variables: Inferences from the representation should be possible

"Can vegetarians eat at Maharani?"

→ true/false

"I'd like to find a restaurant where I can get vegetarian food"

- → list of vegeratian-food restaurants...
- Expressiveness: representation should be able to represent any natural language utterance
- Ok, let's start... First of all, what is the structure of human languages?

### Meaning structure of language

- Assumption: all human languages are based on a form of predicate-arguments arrangement (i.e., subj-verb-complements)
  - The grammar organizes this structure
  - Computational representations rely on this structure
- Verbs dictate specific constraints on args (frames)

"Robert wants Italian food"

"Robert wants to spend less than five dollars"

"Robert wants it to be close by here"

Verbal frames

NP want NP

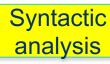
NP want Inf-VP

NP want NP Inf-VP

"NP[Robert] wants NP[Italian food]"

"NP[Robert] wants Inf-VP[to spend less than five dollars]"

"NP[Robert] wants NP[it] Inf-VP[to be close by here]"



## Meaning structure of language: Linking and roles

- In all those examples:
  - Someone wanting → want → something wanted
  - Thematic/case roles associated with the verb "to want"
- Verbal frames permit to link verbal arguments (the surface structure) with thematic/case roles (the underlying semantic representation)

<sup>&</sup>quot;wanting[Robert] wants wanted[Italian food]"
"wanting[Robert] wants wanted[to spend less than five dollars]"
"wanting[Robert] wants wanted[it to be close by here]"

## Meaning structure of language: Restrictions

- The verb "to want" restricts its arguments:
  - Only certain categories of concepts can play the role of "wanting" ("a stone wants..." ???)
  - Only certain categories of concepts can play the role of "wanted"
- Selectional restriction: constraints that verbs pose on their arguments
- More on roles and restrictions in 3. "Lexical semantics"...
- Summing up, verbs define:
  - A surface form → frame
  - An underlying semantic structure → thematic/case roles
  - Restrictions on their arguments

#### Representation languages

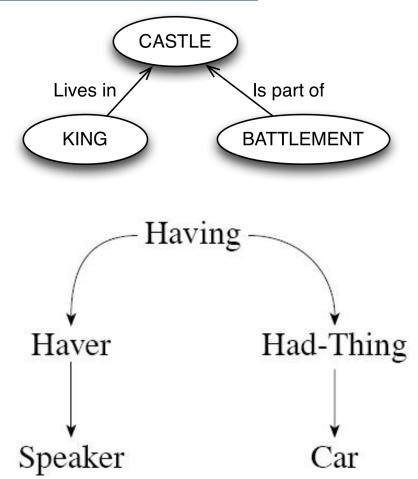
- Once understood the meaning structure of languages, we need a way to represent it
- We present some of them
- Languages representing meaning of utterances:
  - Semantic Networks
  - Logics
  - Frame-based representations
- Languages representing meaning of words:
  - Lexical databases

### **Semantic Networks (SNs)**

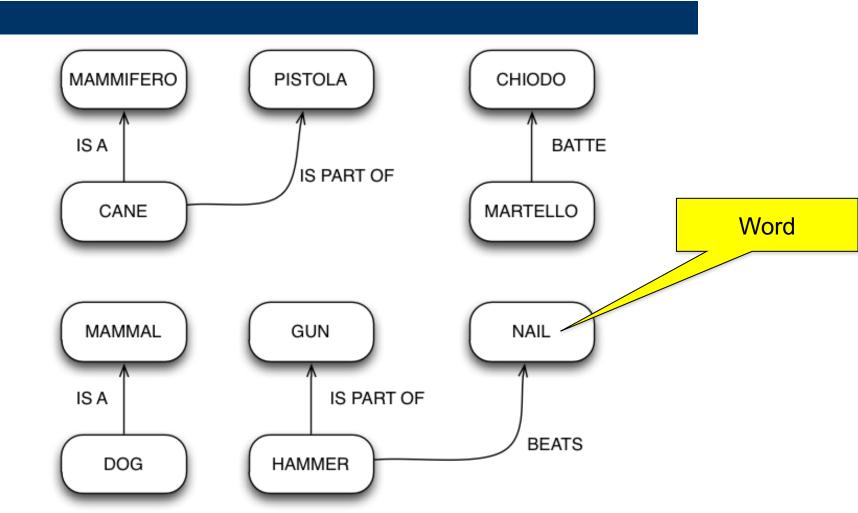
- Graphic notation introduced to model the organization of human semantic memory, or memory for word concepts
  - Node: word concept
  - Arc: relationship between word concepts
- In this approach, *concepts* coincide with *words* 
  - Knowledge modeling is language-dependent
- A word meaning is defined by the connections to other words

### Semantic Networks: An example

- A network representing a domain (a KB)
  - Defines meanings of concepts
  - Description is languagedependent
- A network representing a sentence: "I have a car"
  - Defines the meaning of a sentence



# Semantic Networks are language dependent



#### **Semantic Networks: characteristics**

- Can describe both utterances and KBs
  - But such KBs are language dependent
- Do not have any formal basis

http://conceptnet5.media.mit.edu

ConceptNet is an example of (somehow extended) SN

/s/contributor/globalmind/us/max

Info on arcs
 Relationship
 → surface text
 dataset: /d/globalmind weight: 1.0 surfaceText: [[a car]] can be used for [[driving]]. source: and
 /s/activity/globalmind/assert

#### Logics

- Languages:
  - First-Order Logics (FOL)
  - Description Logic (DL)
  - Temporal logics
- Reasoning
- Implementations:
  - Production-rule systems
  - PROLOG
  - Ontologies

# First Order Predicate Calculus (FOPC) aka First Order Logic (FOL)

```
Formula \rightarrow AtomicFormula
                           Formula Connective Formula
                           Quantifier Variable, . . . Formula
                           \neg Formula
                           (Formula)
AtomicFormula \rightarrow Predicate(Term,...)
             Term \rightarrow Function(Term,...)
                           Constant
                           Variable
     Connective \rightarrow \land |\lor| \Rightarrow
      Quantifier \rightarrow \forall \mid \exists
        Constant \rightarrow A \mid VegetarianFood \mid Maharani \cdots
         Variable \rightarrow x \mid y \mid \cdots
       Predicate \rightarrow Serves \mid Near \mid \cdots
        Function \rightarrow LocationOf \mid CuisineOf \mid \cdots
```

### Logical connectives & quantifiers

#### Logical connectives

P	Q	$\neg P$	$P \wedge Q$	$P \lor Q$	$P \Rightarrow Q$
False	False	True	False	False	True
False	True	True	False	True	True
True	False	False	False	True	False
True	True	False	True	True	True

#### Quantifiers

"A restaurant that serves Mexican food at Politecnico"

 $\exists x Restaurant(x) \land Serves(x, MexicanFood)$ 

 $\land$  Near(LocationOf(x),LocationOf(Politecnico))

"All vegetarian restaurants serve vegetarian food"

 $\forall x VegetarianRestaurant(x) \Rightarrow Serves(x, VegetarianFood)$ 

### First Order Logic (FOL)

Can represent utterances:

```
"A restaurant that serves Mexican food at Politecnico" \exists x Restaurant(x) \land Serves(x, MexicanFood)
```

 $\land$  Near(LocationOf(x),LocationOf(Politecnico))

"All vegetarian restaurants serve vegetarian food"

```
\forall x VegetarianRestaurant(x) \Rightarrow Serves(x, VegetarianFood)
```

"I have a car" ("I" is the speaker)

```
\exists e, y \ Having(e) \land Haver(e, Speaker) \land HadThing(e, y) \land Car(y)
```

- And, as SNs, can represent concepts (a KB)
  - Such description is language-independent
- Formally defined
- Reasoning

### Reasoning

- FOL is semidecidable:
  - If a formula is true, there is a mechanical way to determine this result;
  - if a formula is false, however, such procedure may give a negative result or no result at all
- Restrictions of FOL exist that are decidable

#### Inference: Modus ponens

Modus ponens: The most important inference method

VegetarianRestaurant(Rudys)

 $\forall x VegetarianRestaurant(x) \Rightarrow Serves(x, VegetarianFood)$ 

Serves(Rudys,VegetarianFood)

It is used as forward chaining or backward chaining

### Forward Chaining reasoning

- A new proposition (a fact) added to the KB:
   VegetarianRestaurant(Rudys)
- A *rule* in the KB:

```
\forall x VegetarianRestaurant(x) \Rightarrow Serves(x, VegetarianFood)
```

- As the new fact is added, the rule fires:
  - because the antecedent matches the new fact
  - variable x is instantiated: x=Rudys
  - then, the consequent asserts a new fact into the KB:
    - Serves(Rudys, Vegetarian Food)

### **Backward Chaining reasoning (1/2)**

- Fact in the KB: VegetarianRestaurant(Rudys)
- Rule in the KB:
  - $\forall x VegetarianRestaurant(x) \Rightarrow Serves(x, VegetarianFood)$
- Query: Serves(Rudys, VegetarianFood) ?
- Consequent matches the query: x=Rudys ...
- ... then: if antecedent is true, the query is true
  - Searching for: VegetarianRestaurant(Rudys)
- Antecedent is true (it is in the KB), then consequent is true, then query is true

## **Backward Chaining reasoning (2/2)**

- Fact in the KB: VegetarianRestaurant(Rudys)
- Rule in the KB; consequent matches the query...  $\forall x VegetarianRestaurant(x) \Rightarrow Serves(x, VegetarianFood)$
- Query: Serves(x, Vegetarian Food) ?
- ... then: query has answer if antecedent matches
  - Searching for: VegetarianRestaurant(x)
- Antecedent matches, the answer is: x = Rudys

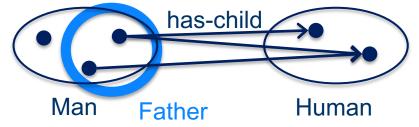
#### Inferences

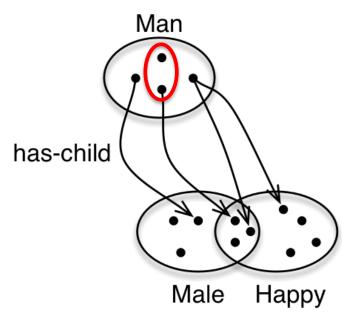
- Neither forward nor backward reasoning are complete
  - I.e., there are valid inferences that cannot be found
- Complete reasoning techniques exist, but they are much more time consuming
  - E.g., the Tableau Calculus

#### FatherOfHappyMaleChildren

### **Description Logics**

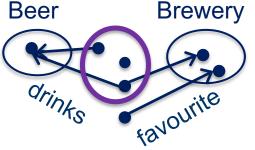
- Subset of FOL
- Decidable
- $Father = Man \sqcap \exists has\text{-}child.Human$





- $FatherOfHappyMaleChildren = Man \sqcap \forall has-child.(Male \sqcap Happy)$
- $\exists favourite.Brewery \sqsubseteq \exists drinks.Beer$

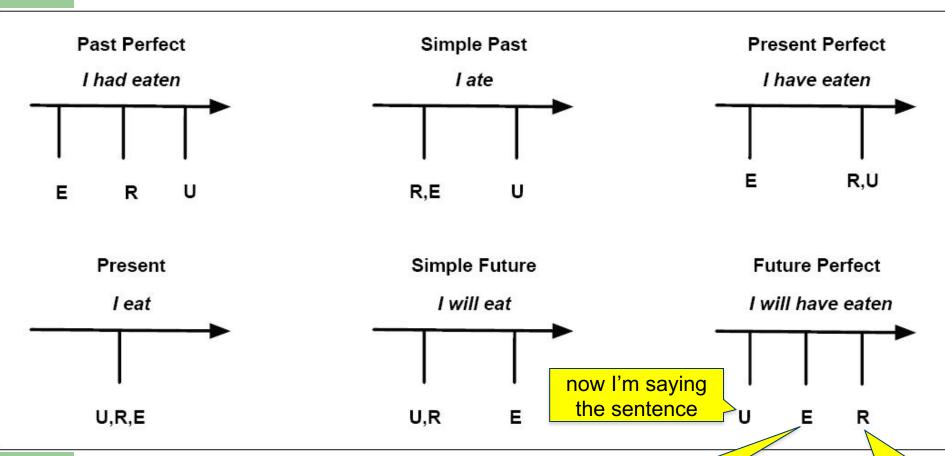




## Temporal logics (tense logic)

- Events are related to the concept of time
  - Time points or time intervals
  - Temporal logics try to represent just that
- Time of the event (E)
- Time of the utterance (U)
- Time of the *reference point* (R)
  - "When Mary's flight departed, I had eaten lunch"
  - Lunch occurred before flight departure

## English tenses (Reichenbach, 1947)



# Temporal logics (Linear temporal Logic)

- "If you have passport and ticket, you can board the flight"
- □ ((¬passport ∧ ¬ticket) ⇒ ¬board\_flight)
- = is true in all future moments
- = is true in the next future moment

#### Frame-based representation

- Represents objects as feature-structures
  - Features (slots)
  - Values (fillers)
  - Values can, in turn, be a frame
- "I believe Mary ate British food"

BELIEVING
BELIEVER SPEAKER

EATING
EATER MARY
EATEN BRITISHFOOD

(NB "I" is the Speaker)

# Representing meanings: Tools

#### **Implementations**

- Production-rule systems
- PROLOG
- Ontologies

- Both production-rule systems and PROLOG implement a subset of FOL (a decidable subset)
- Most common ontologies are based on DL

#### **Production rules systems**

Demo

- Implement forward chaining inference
  - CLIPS, Jess, Drools, ...

```
(deffacts startup "Refrigerator Status"
  (refrigerator 1 light on)
  (refrigerator 1 door open)
  (refrigerator 2 door open))
```

#### Fact list

(refrigerator 1 light on)(refrigerator 1 door open)(refrigerator 2 door open)

 $\forall x \ Refrigerator(x, open) \land Refrigerator(x, light \ on) \Rightarrow$ 

 $Refrigerator(x, food\ spoiled)$ 

```
(defrule example-rule "example"
(refrigerator ?x light on)
(refrigerator ?x door open)
=>
(assert (refrigerator ?x food spoiled)))
```

#### Fact list

(refrigerator 1 light on)(refrigerator 1 door open)(refrigerator 2 door open)(refrigerator 1 food spoiled)

### The PROLOG language

Demo

- Implements backward chaining inference
- A lot of different implementations exist...

```
KB
```

```
mortal(X) :- human(X).
human(socrates).
```

```
\frac{\forall x \; Human(socrates)}{Mortal(socrates)}
\frac{\forall x \; Human(x) \Rightarrow Mortal(x)}{Mortal(socrates)}
```

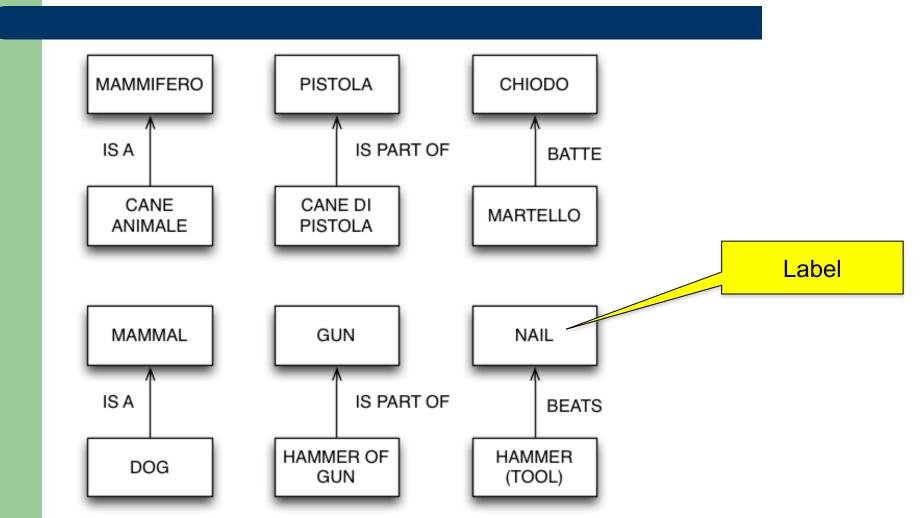
```
?- mortal(socrates).
yes
```

```
goal: search KB for mortal(socrates) → not found
matches consequent: mortal(X=socrates) :- human(X=socrates)
new goal: search KB for: human(socrates) → found
answer: yes
```

### **Ontologies**

- Formal definition of concepts belonging to a domain
- Abstract definition of concepts: does not depend on the language
- Most ontologies are composed of:
  - Classes (e.g. Wine, Winery)
  - Individuals (e.g. champagne)
  - Attributes (e.g. *price*)
  - Relationships (e.g. Winery produces Wine)
- Ontologies described here rely on DL

## Ontologies are language independent



### **Ontologies: characteristics (1/2)**

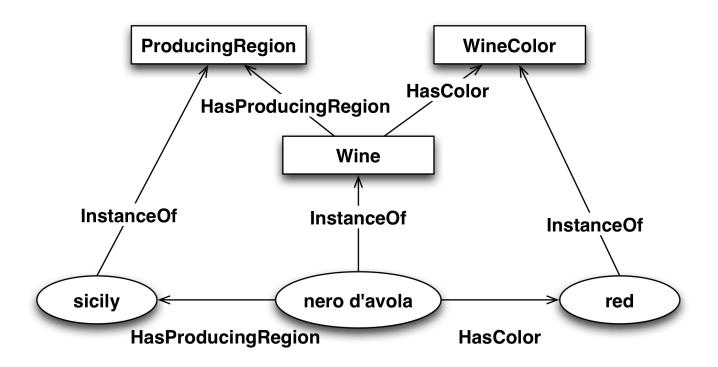
- Class (e.g. Wine):
  - A set of elements; a concept; a type (i.e., a class)
- Individual (e.g. Nero\_Avola):
  - An element; a concept; an instance (i.e., an object)
- Attribute (e.g. price):
  - A simple property; a class field
  - Has a primitive data type (e.g. string)
- Data type (e.g. string)
  - A primitive data type

## Ontologies: characteristics (2/2)

- Relationship (e.g. Winery Produces Wine)
  - Semantic characterization of a concept; relationships among classes or among individuals
- Restriction
  - For example, on attribute values (e.g., ">0")
- Logical rules further specifying the domain hasParent(?x1,?x2) ^ hasBrother(?x2,?x3) → hasUncle(?x1,?x3)

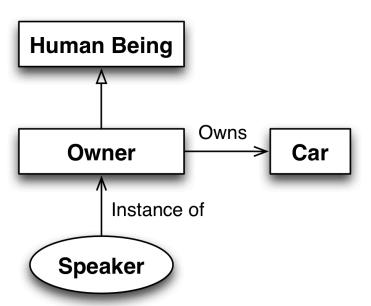
### Ontologies as a graph

Ontologies can be used to represent KB



### Ontologies as a graph

- Ontologies can be used to represent sentences
  - "I have a car"("I" is the Speaker)



# **OWL (Web Ontology Language)**

- W3C specification
- Based on RDF (triple: subject predicate object)
- Three languages:
  - OWL Lite: taxomonies and simple constraints
  - OWL DL: permits to represent a Description Logic (decidable subset of First-Order Logic)
  - OWL Full: Higher order logics (not decidable)
- Inference rules: SWRL
- Query: SPARQL

### Open and closed world

- The "closed world" assumption (e.g. SQL):
  - Any statement that is not known to be true (i.e. a tuple/fact not found in DB) is false (negation as failure)
  - The system is assumed to have complete knowledge
  - PROLOG, production rules engines use "closed world"
- The "open world" assumption (FOL, OWL):
  - Any statement that is not known to be true is... unknown
  - The system does not have enough info to decide
  - Pros and cons:
    - limits the kinds of inferences an agent can make
    - + represents the notion that no single agent has complete knowledge

#### Open and closed world

- E.g. propositions in KB:
  - "Giovanni is an architect"
  - "Giovanni is not a physicist"
- Query: "Is Giovanni an engineer?"
  - Closed world answer: no (proposition not found in KB)
  - Open world answer: unknown
- Query: "Is Giovanni a physicist?"
  - Closed world answer: no (proposition not found in KB)
  - Open world answer: no (negated proposition found)

## The OWL language

- A set of axioms
- The language:
  - Class
  - Property
    - Object property: relationship
    - Datatype property: attribute
  - subclass, disjoint, equivalent relationships
  - Restrictions on properties (type, cardinality)
  - Characteristics of object properties (transitive, ...)
  - Individuals
- Domain
   Object/Datatype Property

# Example (1/3)

```
<owl:Class rdf:ID="ConsumableThing"/>
<owl:Class rdf:ID="PotableLiquid">
  <rdfs:subClassOf rdf:resource="#ConsumableThing" />
</owl:Class>
<owl:Class rdf:ID="Wine">
  <rdfs:subClassOf rdf:resource="#PotableLiquid"/>
  <rdfs:label xml:lang="en">wine</rdfs:label>
  <rdfs:label xml:lang="fr">vin</rdfs:label>
</owl:Class>
```

# Example (2/3)

```
<owl:Class rdf:ID="Grape">
</owl:Class>
<owl:Class rdf:ID="WineGrape">
  <rdfs:subClassOf rdf:resource="#Grape" />
</owl:Class>
<WineGrape rdf:ID="ChardonnayGrape">
</WineGrape >
```

# Example (3/3)

```
<owl:ObjectProperty rdf:ID="madeFromGrape">
  <rdfs:domain rdf:resource="#Wine"/>
  <rdfs:range rdf:resource="#WineGrape"/>
</owl:ObjectProperty>
<owl:DatatypeProperty rdf:about="#year">
  <rdfs:domain rdf:resource="#Wine"/>
  <rdf:range rdf:resource="&xsd:nonNegativeInteger"/>
</owl:DatatypeProperty>
<Wine rdf:ID="LindemansBin05Chardonnay">
  <madeFromGrape rdf:resource="#ChardonnayGrape" />
  <year rdf:datatype="&xsd:nonNegativeInteger">2005</year>
</Wine>
```

### **Query: SPARQL**

- Query language on RDF (W3C standard)
  - Matching on triples

```
PREFIX food: <http://somewhere/example#>
SELECT ?wine
WHERE
{
    ?wine food:madeFromGrape food:ChardonnayGrape .
    ?wine food:year ?year .
    FILTER (?year > 2004)
}
```

LindemansBin05Chardonnay

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# Reasoning: SWRL

• Inference rules: SWRL (W3C standard)

 $hasParent(?x1,?x2) \land hasBrother(?x2,?x3) \rightarrow hasUncle(?x1,?x3)$ 

```
<ruleml:imp>
  <rulem1: rlab rulem1:href="#example1"/>
  <ruleml: body>
    <swrlx:individualPropertyAtom swrlx:property="hasParent">
      <ruleml:var>x1</ruleml:var>
      <ruleml:var>x2</ruleml:var>
    </swrlx:individualPropertyAtom>
    <swrlx:individualPropertyAtom swrlx:property="hasBrother">
      <ruleml:var>x2</ruleml:var>
      <ruleml:var>x3</ruleml:var>
    </swrlx:individualPropertyAtom>
  </ruleml: body>
  <ruleml: head>
                                   swrlx:property="hasUncle">
    <swrlx:individualPropertyAtom</pre>
      <ruleml:var>x1</ruleml:var>
      <ruleml:var>x3</ruleml:var>
    </swrlx:individualPropertyAtom>
  </ruleml: head>
</ruleml:imp>
```

# **Top & domain ontology**

- Usually, ontologies describe a specific domain
- Top ontologies: describe basic concepts (e.g. time, space, ...)
  - LKIF
- Some ontologies try to represent a vast quantity of fundamental human knowledge
  - CYC

## **Ontologies and NLP**

- Enable searching for concepts
  - Abstract and language independent
  - Search through ontology navigation
  - Search through key-words (how can we map words and concepts?)
  - Reasoning (search for sub-classes, part-of classes, individuals, ..., of a given concept)
  - Question answering
- Calculating document similarity (distance)
- Mapping ontologies

#### **Tools for OWL**

- JENA: a free and open source Java framework for processing RDF data (OWL is based on RDF)
  - Ontology navigation
  - Query (SPARQL)
  - Reasoning (embedded; third-party engines; e.g., Pellet)
- Protégé: An ontology editor
  - OWL, SPARQL
  - Extensible via plug-ins

# REFERENCES

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- C. D. Manning, H. Schütze, Foundations of Statistical Natural Language Processing, MIT Press.
- S. Bird, E. Klein, and Edward Loper, Natural Language Processing with Python, O'Reilly
- S. Russell, P. Norvig, Artificial Intelligence: A Modern Approach, Prentice Hall.

#### On semantics

- OWL
  - http://www.w3.org/2004/OWL/
- SPARQL
  - http://www.w3.org/TR/rdf-sparql-query/
- LKIF
  - http://www.estrellaproject.org/lkif-core/
- Cyc
  - http://www.cyc.com/
- Protégé
  - http://protege.stanford.edu/
  - http://protegewiki.stanford.edu/index.php/Protege\_Ontology\_Library

#### On semantics

- Inference engine Pellet
  - http://clarkparsia.com/pellet
- Inference rule language SWRL
  - http://www.w3.org/Submission/SWRL/
- Quary language SQWRL:
  - http://protege.cim3.net/cgi-bin/wiki.pl?SQWRL
- OWL Jena
  - http://jena.sourceforge.net/
- Production rule engines:
  - http://www.jessrules.com/
  - http://clipsrules.sourceforge.net/
  - http://www.jboss.org/drools

#### On semantics

- WordNet
  - http://clarkparsia.com/pellet
- MultiWordNet
  - http://multiwordnet.fbk.eu
- EuroWordNet
  - http://www.illc.uva.nl/EuroWordNet/
- Global WordNet
  - http://globalwordnet.org

#### **General references**

- Tools:
  - http://www-nlp.stanford.edu/links/statnlp.html
  - http://www.comp.nus.edu.sg/~rpnlpir/
- General resources:
  - http://www-nlp.stanford.edu/links/linguistics.html
  - http://www.math.tau.ac.il/~shimsh/text\_domain.html

### **NLP** in Italy

- ILC-CNR (Pisa, Genova)
  - http://www.ilc.cnr.it/indexflash.html
  - http://www.ge.ilc.cnr.it/
- ISTC-CNR (Roma, Padova, Trento)
  - http://www.istc.cnr.it/
  - http://www.loa-cnr.it/
  - http://www.pd.istc.cnr.it/home.htm
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