Ontology Learning

Lama Saeeda

saeeda.lama@fel.cvut.cz

December 17, 2020



Overview

- Ontology and Ontology Learning
- 2 Methods of Ontology Learning
- Ontology Evaluation
- 4 Ontology Learning Tools
- Conclusion



- Ontology and Ontology Learning
- Methods of Ontology Learning
- Ontology Evaluation
- 4 Ontology Learning Tools
- 5 Conclusion

Ontology and Ontology Learning



Basics

- Ontology and Ontology Learning
- Methods of Ontology Learning
- Ontology Evaluation
 - Ontology Learning Tools
- 5 Conclusion



- Applications with ontology
- Jim Hendler: "a little semantic goes a long way"
- Availability, suitability, completeness







Manual ontology creation is expensive







Ontology Definition

- Ontology and Ontology Learning
- Methods of Ontology Learning
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What is ontology?

- "Specification of a conceptualization" Tom Gruber
- "A description of things that exist and how they relate to each other"
 Chris Welty



Ontology components

- Ontology and Ontology Learning
- Methods of Ontology Learning
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What are the components of the ontology?

Ontology can be defined as a tuple:

$$\vartheta = (C, R, H^C, rel, A^{\vartheta})$$

- C is the set of ontology concepts. The concepts represent the entities of the domain being modeled. They are designated by one or more natural language terms and are normally referenced inside the ontology by a unique identifier.
- $H^C \subseteq C \times C$ is a set of taxonomic relationships between the concepts. Such relationships define the concept hierarchy.
- R is the set of non-taxonomic relationships. The function rel: $R \to C \times C$ maps the relation identifiers to the actual relationships.
- A^{ϑ} is a set of axioms, usually formalized into some logic language. These axioms specify additional constraints on the ontology and can be used in ontology consistency checking and for inferring new knowledge from the ontology through some inference mechanism.

Ontology Learning

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Methods of Ontology Learning



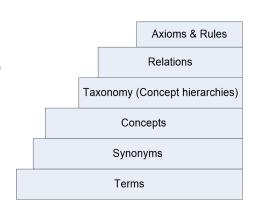
Layer-cake model

- Ontology and Ontology Learning
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Layer-cake model for learning ontology

 $\forall x, y (sufferFrom(x, y) \rightarrow ill(x))$ cure (domain:Doctor, range:Disease) is_a (Doctor, Person) Disease := $\langle I, E, L \rangle$ {disease, illness} disease, illness, hospital





input

- Ontology and Ontology Learning
- Methods of Ontology Learning
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Possible input sources

- Structured data database schemes
- Semi-structured data dictionaries like WordNet
- Unstructured data natural language text documents, like the majority of the HTML based web-pages



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Learning methods

- Linguistic
- Statistical
- Rule-Based
- Logical



Learning terms

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Terms extraction

Axioms & Rules
Relations
Taxonomy (Concept hierarchies)
Concepts
Synonyms
Terms

disease, illness, hospital



Terms extraction - Linguistic processing

Natural Language Processing (NLP), deep language analysis or information retrieval methods for term indexing.

- Identifies sentences, determined by periods or other punctuation marks
- Tokenization separates text into tokens which are the basic units
- Normalizes tokens to lower case to provide case-insensitive indexing
- Stemming: (fishing, fished, fisher) one stem: fish
- Stop-words removing: Meaningless tokens, (there, so, other, etc..)
- POS tagging: the book on the table (noun), to book a flight (verb)



Terms extraction - Statistical metrics

- TF: Term Frequency, how frequently a term occurs in one document.
 - TF = (Number of times term t appears in a document / Total number of terms in the document)
- IDF: Inverse Document Frequency, how important a term is in the corpus IDF = log (Total number of documents / Number of documents with term t in it)



Terms extraction - Statistical metrics

$$tfidf(w) = tf(w).log(\frac{N}{df(w)})$$

The word is more popular when it appears several times in a document

The word is more important if it appears in less documents

- tf(w) → term frequency (number of words occurrences in a document)
- $df(w) \rightarrow$ document frequency (number of documents containing the word
- $N \rightarrow$ number of all documents
- ullet tfidf o relative importance of the word in the document



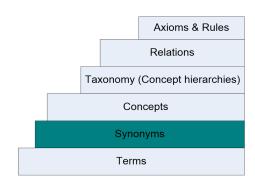
Learning synonyms

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Synonyms extraction

{disease, illness}





Synonyms extraction

Identification of terms that share semantics, i.e., potentially refer to the same concept

- Latent Semantic Indexing (LSI): Assumes that words that are close in meaning will occur in similar pieces of text (the distributional hypothesis).
- Wordnet



Synonyms extraction - Wordnet overview

What is wordnet?

- General lexical knowledge base
- Contains 150,000 words (noun, verb, adj, adv)
- A word can have multiple senses: "plant" as a noun has 4 senses
- Each concept (under each sense and PoS) is represented by a set of synonyms (a syn-set).
- Semantic relations such as hypernym/antonym/meronym of a syn-set are represented



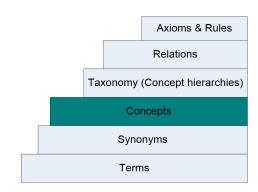
Learning concepts

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Concepts extraction

Disease := <I, E, L>





Concepts

Controversial as it is not clear what exactly constitutes a concept

A term may indicate a concept, if we define its:

- Intension (In)formal definition of the objects this concept describes ex: a disease is an impairment of health or a condition of abnormal functioning
- Extension Set of objects described by this concept (ontology population)
 - ex: influenza, cancer, heart disease
- **Lexical Realizations** The term itself and its multilingual synonyms *ex:* disease, illness, maladie



Concepts forming approaches

The detection of synonyms can help to cluster terms to groups of terms sharing (almost) the same meaning, thus representing ontological classes.

- Learning the extension of concepts
 - Unsupervised hierarchical clustering techniques known from machine learning research
 Clusters of related terms (overlaps almost completely with term and
 - synonym extraction)
 - for example "all movie actors appearing on the Web"
- Learning the intension of concepts
 - Acquisition of informal definition
 Textual description,i.e.a gloss of the concept (ex. from Wordnet), normative document, etc.
 - Acquisition of formal definition
 Disovering formal constraints that define a class membership



Concepts labeling

Hearst's patterns

- Find hypernym candidates for each class members
- Then select the top candidate related to the largest number of class members

Web search

- Proper query of concatenating the child concepts
- Return top 10 results + NLP
- Select the most frequent phrase
 - Challenging problem
 - Assign meaningful name to the newly-created parent node
- General enough to cover the scope of all the child concepts
- Specific enough to just cover that of them

Ex: president becomes the parent of concepts Bush and Reagan.



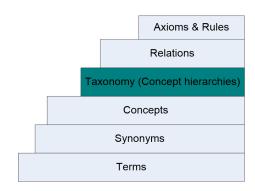
Learning concept hierarchies

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Taxonomy

is_a (Doctor, Person)





Learning concept hierarchies

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Taxonomy

"is-a" hierarchy on concepts

Existing approaches

- Hearst Patterns (Lexico-syntactic patterns)
- 4 Hierarchical Clustering
- Ocument-based subsumption



Taxonomy - Hearst Patterns

- The acquisition of hyponym lexical relations from text
- Uses a set of predefined lexico-syntactic patterns which:
- Occur frequently and in many text genres
- Indicate the relation of interest
- Can be recognized with little or no pre-encoded knowledge
 - Principle idea: match these patterns in texts to retrieve is-a relations
 - Reasonable precision, very low recall



Taxonomy - Hearst Patterns

- Vehicles such as cars, trucks and bikes
- Such fruits as oranges or apples
- Swimming, running and other activities
- Swimming, running or other activities
- Publications, especially papers and books
- NP_{hyper} such as $\{NP_{hypo},\}^*\{(and|or)\}NP_{hypo}$
- such NP_{hyper} as $\{NP_{hypo},\}^*\{(and|or)\}NP_{hypo}$
- $NP_{hypo}\{, NP\}^*\{,\}$ or other NP_{hyper}
- $NP_{hypo}\{, NP\}^*\{,\}$ and other NP_{hyper}
- NP_{hyper} especially $\{NP_{hypo},\}^*\{(and|or)\}NP_{hypo}$



Taxonomy - Hierarchical Clustering

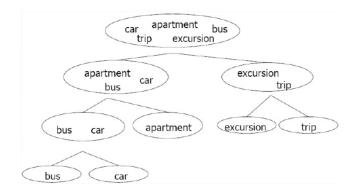
	$\operatorname{Book}_{obj}$	$Rent_{obj}$	Drive_{obj}	$Ride_{obj}$	$Join_{obj}$
Hotel	X				
Apartment	x	X			
Car	x	X	x		
Bike	x	X	x	x	
Excursion	X				X
Trip	X				X

Jaccard coefficient distance $= |\frac{A \wedge B}{A \vee B}|$

	Hotel	Apartment	Car	Bike	Excursion	Trip
Hotel	1.0	0.5	0.33	0.25	0.5	0.5
Apartment		1.0	0.66	0.5	0.33	0.33
Car			1.0	0.75	0.25	0.25
Bike				1.0	0.2	0.2
Excursion					1.0	1.0
Trip						1.0



Taxonomy - Hierarchical Clustering



see [staab2010handbook]



Taxonomy - Document-based subsumption

Term t_1 subsumes term t_2 [is-a(t2,t1)] if t_1 appears in all the documents in which t_2 appears

$$P(x|y) = \frac{n(x,y)}{n(y)}$$

Term x subsumes term y iff P(x|y) = 1, where $n(x,y) \rightarrow$ the number of documents in which x and y co-occur $n(y) \rightarrow$ the number of documents that contain y



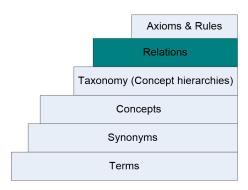
Learning relations

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Relation extraction

cure (domain:Doctor, range:Disease)





Relation extraction - Specific Relations

Discover anonymous associations between words

- X consists of Y (part-of)
 The framework for OL consists of information extraction, ontology discovery and ontology organization
- X is used for Y (purpose)OL is used for OE
- X leads to Y (causation)
 Good OL methods lead to good OE



Relation extraction

OntoLT

Syntactic analysis: Maps a *subject* to the **domain**, the *predicate* or *verb* to **relation** and the *object* to its **range**.

The player kicked the ball to the net relation: kick (domain: player, range: ball)

TextToOnto

 $love(man; woman) \land love(kid; mother) \land love(kid; grandfather)$

 \Rightarrow

love(person; person)

However, different verbs can represent the same (or a similar To) relation Clustering \rightarrow advise, teach, instruct

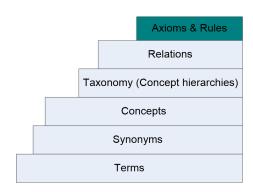
Learning rules and axioms

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Rules and axioms extraction

$$\forall x, y (sufferFrom(x, y) \rightarrow ill(x))$$





Rule Extraction

DIRT - Discovery of Inference Rules from Text (Lin and Pantel, 2001)

- Let X be an algorithm which solves a problem Y
- Using similar constructions like X solves Y, Y is solved by X, X resolves Y
- $\forall x, y(solves(x, y) \Rightarrow isSolvedBy(y, x))$ (Inverse object property)
- $\forall x, y(solves(x, y) \Rightarrow resolves(x, y))$ (Equivalent object property)



Axiom Extraction

Automated Evaluation of Ontologies - AEON (Völker et al., 2008)

Axioms are extracted (using lexico-syntatic patterns) from a Web Corpus

• Dealing with uncertainty and inconsistency (Haase and Völker, 2005)

 ${\sf Disjointness\ axioms} \to {\sf disjoint(man,woman)}$



	1	Terms	Concepts	Taxonomic relations	Non-taxonomic relations
statistic methods	Text pre-processing	X			
	POS tagging	X			
	Sentence parsing	X			
	Latent semantic		X		
	Cooccurrence	X	X		
	Clustering	1	X	X	
	Term subsumption	1		X	
	Association rules	1			
Linguistic methods	Seed words	X			
	Semantic lexicon	1	X	×	X
	Sub-categorization frames	X	X		
	Syntactic structure	X			X
	Dependency analysis	X			×
	Semantic templates	1		×	X
	Lexico-syntactic paterns			X	X
	Axiom templates	1			
Logical methods	Logical inference	1		X	X
	Inductive Logic				

Table: Ontology learning tasks and subtasks and the state-of-art techniques applied for each



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Ontology Evaluation



Quality criteria

- Accuracy Does the ontology accurately model the domain?
- Adaptability Can the ontology easily be adapted to various uses?
- Clarity Is the meaning implied by the ontology clear?
- Completeness Does the ontology richly or thoroughly cover the domain?
- Computational efficiency How easily can automatic reasoners perform typical tasks?
- Conciseness Does the ontology include unnecessary axioms or assumptions?
- Consistency Does the ontology lead to logical errors or contradictions?
- Organisational fitness Is the ontology easily deployed in the application context? Is it easy to access? Does it align to other ontologiesalready in use?



How to evaluate OL

- Benchmark corpora and ontologies
- Evaluation of methods using different information sources



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Ontology Learning Tools



The festival attracts culture vultures to see live drama, dance and music

OntoLT

- festival and culture are class candidates using statistical analysis (TF-IDF)
- attracts is a relation between festival and culture using NLP



ASIUM - Acquisition of Semantic knowledge Using ML Methods

- Taxonomic relations among terms in technical texts
- Conceptual Clustering

OntoLearn

- Enrich a domain ontology with concepts and relations
- NLP and ML

Text-To-Onto

- Find taxonomic and non-taxonomic relations
- Statistics, Pruning Techniques and Association Rules
- Sucessor: Text2Onto tool



Text2Onto

- Ontology learning from textual documents framework
- System calculates a confidence for each learned object for better user interaction
- Updates the learned knowledge each time the corpus is changed and avoid processing it by scratch
- Interaction with end-users which is the central part of the architecture.
- Allows for easy
 - combination of algorithms,
 - execution of algorithms,
 - writing new algorithms



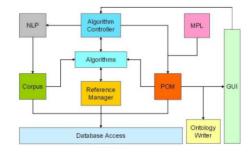
Text2Onto requirements

- Java 6 +
- WordNet
- GATE (General Architecture for Text Engineering)



Text2Onto components

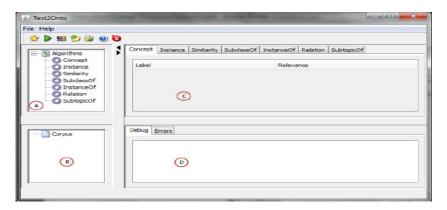
- NLP engine
- Algorithms
- Algorithm Controller
- (Probabilistic Ontology Model)
 POM



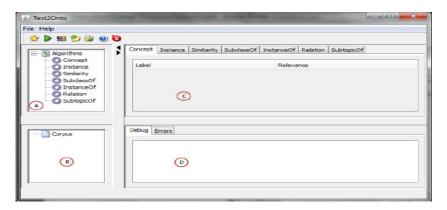
POM

container for learned objects.

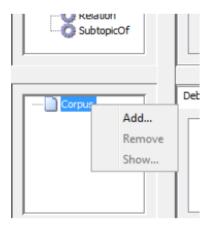
All objects are enhanced by calculated probabilities in such manner that a user can decide whether to include this object into the ontology or not.



- **(A) Controller view** where we specify which Algorithms to use and how to combine the results of these algorithms.
- (B) Corpus view from where adding / removing a corpus is done.



- **(C) POM view panel**. Displays the results of the current ontology learning process.
- (D) Displays debugging messages and error messages.



Step 1 - Add a Corpus

Right-click on the label Corpus on corpus view panel and add a corpus.

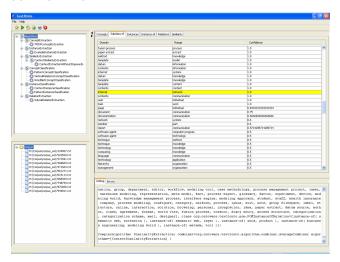


Step 2 - Specify algorithms to be applied

Right-click on the required entity on the controller view panel and click **add**. A list of available algorithms will appear. You can add one or more algorithms from here.

Step 3 - Run

Once all required algorithms have been specified, click the Run icon



The results will appear on the POM view panel (C).

contents	information	1.0
internet	system	1.0
datum	knowledge	1.0
template	knowledge	1.0
template	content	1.0
contents	content	1.0
internet	network	1.0
contents	communication	1.0
user	individual	1.0
task	work	1.0
page	individual	0.8333
document	communication	0.75
documentation	communication	0.6666
network	system	0.6
member	part	0.6
report	communication	0.5714
software agent	computer program	0.5
software agent	technology	0.5

Step 4 - Review the results

The results of Text2Onto may need to be filtered. We can do this by giving feedback to it. To give feedback, right-click on the required entity, go to feedback and set the appropriate feedback (True, False or Don't know).

Export the results

Results can be exported in KAON, RDFS or OWL format. To do this, go to File and click Export.

Text2Onto

Can Text2Onto **automatically** build an ontology by learning on a corpus of texts?

Can Text2Onto help a user to build an ontology?



FRED

- Transforms natural language text into RDF/OWL with linked data support
- Based on ontology design patterns
- Graph visualization
- Export the RDF result: RDF/XML, Turtle, N3, etc.
- Tasks:
 - Entity linking to Semantic Web data
 - Relation extraction between frames, events, concepts and entities
 - Negation representation
 - Temporal relation extraction
 - ... more



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Conclusion



- We need to build Ontology quickly, easily and they have to be reliable!
- Fully automated OL system that works perfectly, doesn't exist YET
- User revision and interaction is essential
- Methods are based mainly on NLP techniques complemented with statistical measures

Ontology Learning is the old new era of developing ontologies. It is linked with many CS fields and it is all about understanding the reality through the structure of things

