Task: Information Extraction for the Semantic Web Solution: Integration of PDT Tools with GATE and Inductive Logic Programming

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- Information Extraction
- Semantic Annotation
- Example Tasks
- 2 Solution
 - Basic Idea
 - Linguistics we Are Using
 - Manually Created Rules
 - Semantic Interpretation
 - Learning of Rules
 - Evaluation
- Implementation Details
 - Integration of Linguistic Tools (GATE)
 - Integration with Semantic Tools
 - Conclusion

Information Extraction and the Semantic Web

- The Task of Information Extraction
 - Automatically find the information you're looking for.
 - Pick out the most useful bits.
 - Present it in preferred manner, at the right level of detail.
- Semantic Web
 - Web as universal medium for the exchange of information.
 - Not only for humans but also for software agents.
 - Main problem today: lack of semantic data on the Web.
- Extraction of information for the Semantic Web.
 - Let's use information extraction to produce semantic data.

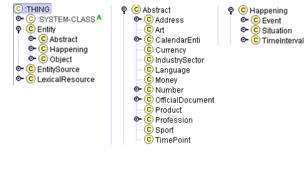
Semantic Web Introduction

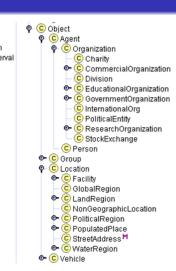
We use semantic web ontologies to express the semantics.

- RDF, OWL languages
- Motivated by description logics
- Concepts or Classes
- Predicates or Relations
- Individuals or Instances
- RDF triples: <Subject> <Predicate> <Object>
- RDF triples form a named oriented graph
 - Basic data structure of the Semantic Web.

Information Extraction

Ontology (example)





PROTON (PROTo ONtology)

http://proton.semanticweb.org/



- Information Extraction
- Semantic Annotation
- Example Tasks

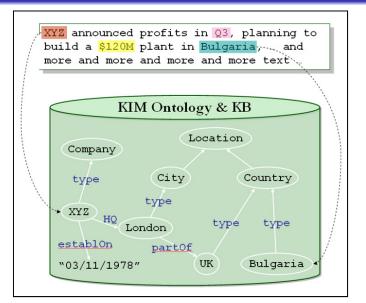
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Semantic Annotation (http://www.ontotext.com/kim/)





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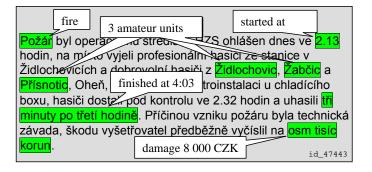
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Example of the web-page with a report of a fire department



Text of an Accident Report and Contained Information

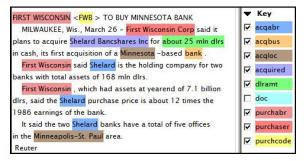


Information to be extracted is decorated.

Example Tasks

Acquisitions Corpus

- Corporate Acquisition Events
- Acquisitions v1.1 version¹

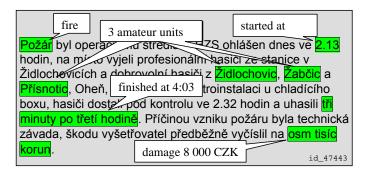


¹from the Dot.kom project's resources:

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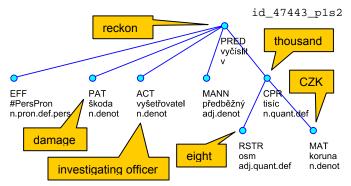
Problem 0000000 Basic Idea

How to extract the information about the damage of the accident?



- How to extract the information about the damage of the accident?
- See the last sentence on the next slide.

Corresponding linguistic tree



- ..., škodu vyšetřovatel předběžně vyčíslil na osm tisíc korun.
- ..., investigating officer preliminarily reckoned the damage to be 8 000 CZK.
- Basic Idea: use tree queries (tree patterns) to extract the information.

Introduction of Our Solution

- Extraction of semantic information from texts.
- Exploiting of linguistic tools.
 - Mainly "from" the Prague Dependency Treebank project.
 - Related tools language analyzers (TectoMT), Netgraph, etc.
 - Experiments with the Czech WordNet.
- Rule based extraction method.
 - Extraction rules ≈ tree queries
 - ILP learning of extraction rules

Schema of the extraction process



1) Extraction of text



2) Linguistic annotation



3) Data extraction

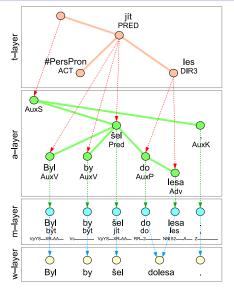


4) Semantic representation

Ontology

- Extraction of text
 - Using RSS feed to download pages.
 - Regular expression to extract text.
- 2 Linguistic annotation
 - Using chain of 6 linguistic tools (see on next slides).
- O Data extraction
 - Exploitation of linguistic trees.
 - Using extraction rules.
- Semantic representation of data
 - Ontology needed.
 - Semantic interpretation of rules.
 - Far from finished in current state.

Layers of linguistic annotation in PDT



- Tectogrammatical layer
- Analytical layer
- Morphological layer
- PDT 2.0 on-line:

http://ufal.mff.cuni.cz/pdt2.0/

Sentence:

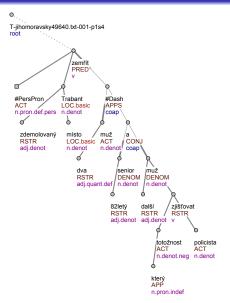
Byl by šel dolesa. He-was would went toforest.

Tools for machine linguistic annotation

- Segmentation and tokenization
- Morphological analysis
- Morphological tagging
- McDonnald's Maximum Spanning Tree parser
 - Czech adaptation
- Analytical function assignment
- Tectogrammatical analysis
 - Developed by Václav Klimeš
 - Available within the TectoMT² project

²http://ufal.mff.cuni.cz/tectomt/

Example of an output tectogrammatical tree



- Lemmas
- Functors
- Semantic parts of speech

Sentence:

Ve zdemolovaném trabantu na místě zemřeli dva muži – 82letý senior a další muž, jehož totožnost zjišťují policisté.

Two men died on the spot in demolished trabant – . . .



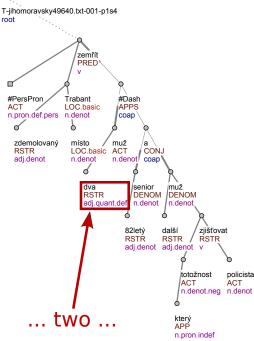
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- Example Tasks

Solution

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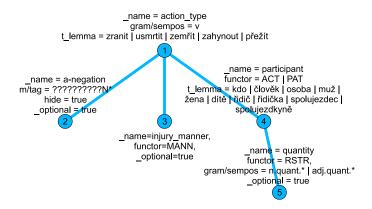
Implementation Details

- Integration with Semantic Tools



 How to extract the information about two dead people?

Extraction rules – Netgraph queries



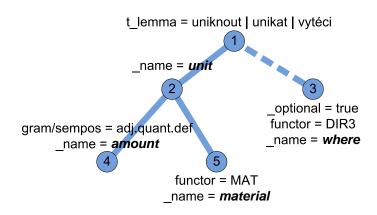
- Tree patterns on shape and nodes (on node attributes).
- Evaluation gives actual matches of particular nodes.
- Names of nodes allow use of references.

Raw data extraction output

```
<QueryMatches>
  <Match root id="T-vvsocina63466.txt-001-pls4" match string="2:0.7:3.8:4.11:2">
    <Sentence>
      Při požáru byla jedna osoba lehce zraněna - jednalo se
      o majitele domu, který si vykloubil rameno.
    </Sentence>
    <Data>
      <Value variable name="action type" attribute name="t lemma">zranit</Value>
      <Value variable name="injury manner" attribute name="t lemma">lehký</Value>
      <Value variable name="participant" attribute name="t lemma">osoba</Value>
      <Value variable name="quantity" attribute name="t lemma">jeden</Value>
    </Data>
  </Match>
  <Match root_id="T-jihomoravsky49640.txt-001-p1s4" match_string="1:0,13:3,14:4">
    <Sentence>
      Ve zdemolovaném trabantu na místě zemřeli dva muži - 82letý senior
      a další muž, jehož totožnost zjišťují policisté.
    </Sentence>
    <Data>
     <Value variable name="action type" attribute name="t lemma">zemřít</Value>
      <Value variable name="participant" attribute name="t lemma">muž</Value>
      <Value variable name="guantity" attribute name="t lemma">dva</Value>
    </Data>
  </Match>
  <Match root id="T-jihomoravsky49736.txt-001-p4s3" match string="1:0.3:3.7:1">
    <Sentence>Čtyřiatřicetiletý řidič nebyl zraněn.
    <Data>
      <Value variable name="action type" attribute name="t lemma">zranit</Value>
      <Value variable name="a-negation" attribute name="m/tag">VpYS---XR-(N)A---
      </Value>
      <Value variable name="participant" attribute name="t lemma">řidič</Value>
    </Data>
  </Match>
</OuervMatches>
```

SELECT action type.t lemma, a-negation.mtag, injury manner.t lemma, participant.t lemma, quantity.t lemma FROM ***extraction rule***

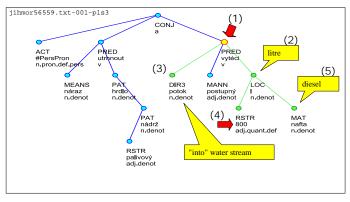
Extraction rules – Environment Protection Use Case



Matching Tree

"Due to the clash the throat of fuel tank tore off and 800 litres of oil (diesel) has run out to a stream."

"Nárazem se utrhl hrdlo palivové nádrže a do potoka postupně vyteklo na 800 litrů nafty."



Manually Created Rules

Raw data extraction output

```
<OuervMatches>
 <Match root id="jihmor56559.txt-001-p1s3" match string="15:0,16:4,22:1,23:2,27:3">
   <Sentence>Nárazem se utrhl hrdlo palivové nádrže a do potoka postupně vyteklo na
800 litrů nafty.</Sentence>
                                                   litre
   <Data>
     <Value variable name="amount" attribute name="t lenga">800</Value>
     <Value variable name="unit" attribute name="t lemma">1/Value>
     <Value variable name="material" attribute name="t lemma">nafta</Value>
      <Value variable name="where" attribute name="t_lemma">potok</Value
   </Data>
                                       water stream
                                                                             diesel
 </Match>
 <Match root id="jihmor68220.txt-001-p1s3" match string="3:0,12:4,21:1,22:2,27:3">
   <Sentence>Z palivové nádrže vozidla uniklo do půdy v příkopu vedle silnice zhruba
350 litrů nafty, a proto byli o události informováni také pracovníci odboru životního
prostředí Městského úřadu ve Vyškově a České inspekce životního prostředí.</sentence>
   <Data>
     <Value variable name="amount" attribute name="t lemma">350</Value>
     <Value variable name="unit" attribute name="t lemma">1</Value>
      <Value variable name="material" attribute name="t lemma">nafta</Value>
     <Value variable name="where" attribute name="t lemma">puda</Value>
   </Data>
                                                                         soil
 </Match>
```

SELECT amount.t_lemma, unit.t_lemma, material.t_lemma, where.t_lemma



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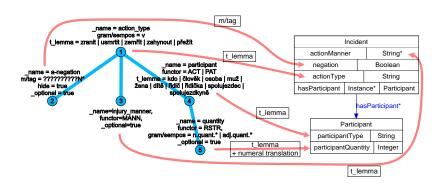
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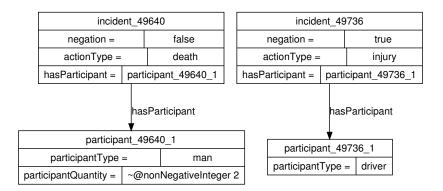
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Semantic interpretation of extraction rules



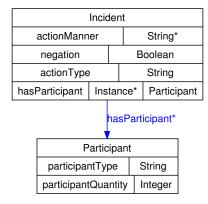
- Determines how particular values of attributes are used.
- Gives semantics to extraction rule.
- Gives semantics to extracted data.

Semantic data output



Two instances of two ontology classes.

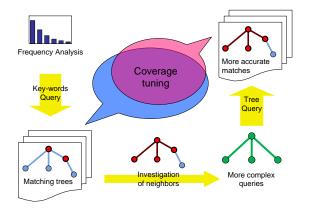
The experimental ontology



- Two classes
 - Incident and Participant
- One object property relation
 - hasParticipant
- Five datatype property relations
 - actionManner (light or heavy injury)
 - negation
 - actionType (injury or death)
 - participantType (man, woman, driver, etc.)
 - participantQuantity

Semantic Interpretation

Design of extraction rules – iterative process



- Frequency analysis → representative key-words.
- Investigating of matching trees \rightarrow tuning of tree query.
- Complexity of the query \cong complexity of extracted data.

Corpus of Fire-department articles

- Fire-department articles
- Published by The Ministry of Interior of the Czech Republic³
- Processed more than 800 articles from different regions of Czech Republic
- 1.2 MB of textual data
- Linguistic tools produced 10 MB of annotations, run time 3.5 hours
- Extracting information about injured and killed people
- 470 matches of the extraction rule,
 200 numeric values of quantity (described later)

http://www.mvcr.cz/rss/regionhzs.html



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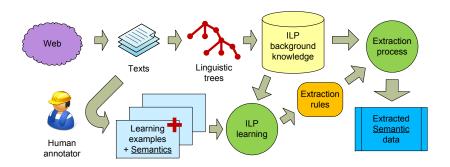
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Inductive Logic Programming

- Inductive Logic Programming (ILP)
 - is a Machine Learning procedure for multirelational learning
 - Heuristic and iterative method, learning is usually slow
 - It is capable to deal with graph or tree structures naturally
 - Learns form positive and negative examples
 - Positive and negative tree nodes
 - It is necessary to label tree nodes from corresponding labeled text (not trivial problem)
 - Learned rules are strict (no weights, probabilities, etc.)
 - Easier human understanding, modification
 - Possibility of sharing of rules amongst different tools
 - Lower performance (precision, recall)

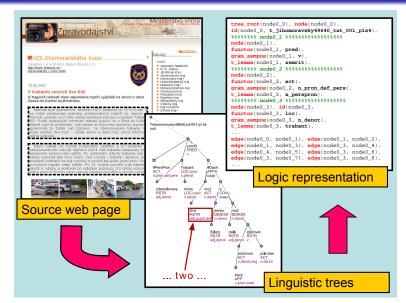
Integration of ILP in our extraction process



- Main point: transformation of trees to logic representation.
- Human annotator does not need to be a linguistic expert.

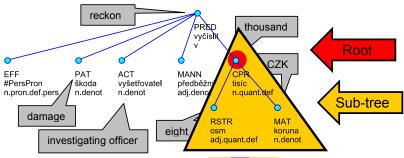
Learning of Rules

Logic representation of linguistic trees



Problem

Root/Subtree Preprocessing/Postprocessing (Chunk learning)



škodu vyšetřovatel předběžně vyčíslil na osm tisíc korun.

investigating officer preliminarily reckoned the damage to be eight thousand Crowns (CZK).

Learning of Rules

Examples of learned rules, Czech words are translated.

Example

```
[Rule 1] [Pos cover = 14 Neg cover = 0]
damage_root(A) :- lex_rf(B,A), has_sempos(B,'n.quant.def'),
   tDependency (C, B), tDependency (C, D),
   has_t_lemma(D,'investigator').
[Rule 2] [Pos cover = 13 Neg cover = 0]
damage_root(A) :- lex_rf(B,A), has_functor(B,'TOWH'),
   tDependency(C,B), tDependency(C,D), has t lemma(D,'damage').
[Rule 1] [Pos cover = 7 Neg cover = 0]
injuries(A) :- lex_rf(B,A), has_functor(B,'PAT'),
   has_gender(B, anim), tDependency(B,C), has_t_lemma(C,'injured')
[Rule 8] [Pos cover = 6 Neg cover = 0]
injuries (A) :- lex_rf(B,A), has_gender(B,anim), tDependency(C,B),
   has t lemma(C,'injure'), has negation(C,neg0).
```

Evaluation results

task/method	matching	missing	excess	overlap	prec.%	recall%	F1.0%
damage/ILP	14	0	7	6	51.85	70.00	59.57
damage/ILP - ler	74.07	100.00	85.11				
dam./ILP-roots	16	4	2	0	88.89	80.00	84.21
damage/Paum	20	0	6	0	76.92	100.00	86.96
injuries/ILP	15	18	11	0	57.69	45.45	50.85
injuries/Paum	25	8	54	0	31.65	75.76	44.64
inj./Paum-afun	24	9	38	0	38.71	72.73	50.53

- 10-fold cross validation
- Two tasks: 'damage' and 'injuries'
- Root/subtree preprocessing/postprocessing used for 'damage' task

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Integration of Linguistic Tools (GATE)

GATE

- GATE: General Architecture for Text Engineering
- The University of Sheffield
- http://gate.ac.uk/
- Implemented Batch TectoMT Language Analyzer
 - Transformation of PDT annotations to GATE
- Netgraph used as a tree viewer
 - Works also for Standford Depndencies

Integration of Linguistic Tools (GATE)

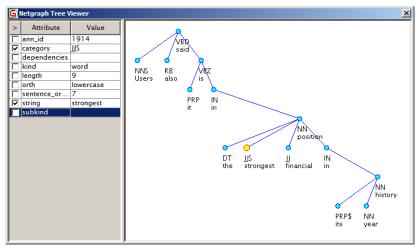
PDT in GATE

Požár byl operačnímu středisku HZS ohlášen dnes ve 2.13 hodin, na místo výjeli profesionální hasiči ze stanice v Židlochovicích a dobrovolní hasiči z Zidlochovic, Žabčic a Přísnotic, Oheň, který zasáhl elektroinstalaci u chladícího boxu, hasiči dostali pod kontrolu ve 2.32 hodin a uhasili tři minuty po třetí hodině. Příčinou vzniku požáru byla technická závada, škodu vyšetřovatel předběžně vyčísilí na osm tisíc korun.

A -							
Type	Set	Start	End	ld			
Token	TectoMT	2	7	2	(afun=Sb, ann_id=2, form=Požár, hidden=true, lemma=požár,≜		
tDependency	TectoMT	2	44	278	{args=[125, 108]}		
tToken	TectoMT	2	7	108	{ann_id=108, deepord=1, formeme=n:1, functor=PAT, gender		
aDependency	TectoMT	2	44	279	{args=[7, 2]}		
Sentence	TectoMT	2	319	- 1	8		
Token	TectoMT	8	-11	3	{afun=AuxV, ann_id=3, form=byl, hidden=true, lemma=být, or		
auxRfDependency	TectoMT	8	44	205	{args=[125, 3]}		
aDependency	TectoMT	8	44	280	{args=[7, 3]}		
Token	TectoMT	12	22	4	{afun=Atr, ann_id=4, form=operačnímu, hidden=true, lemma=		
tDependency	TectoMT	12	32	281	{args=[121, 119]}		
tToken	TectoMT	12	22	119	{ann_id=119, deepord=2, degcmp=pos, formeme=adj:attr, fu		
aDependency	TectoMT	12	32	282	{args=[5, 4]}		
Token	TectoMT	23	32	5	{afun=Obj, ann_id=5, form=středisku, hidden=true, lemma=sti		
tDependency	TectoMT	23	36	283	{args=[121, 123]}		
tDependency	TectoMT	23	44	284	{args=[125, 121]}		
tToken	TectoMT	23	32	121	{ann_id=121, deepord=3, functor=ADDR, gender=neut, lex.rf=		
aDependency	TectoMT	23	44	286	{args=[7, 5]}		
aDependency	TectoMT	23	36	285	{args=[5, 6]}		

	•	TectoMT				
- 1	哮	Sentence				
ci	哮	Token				
- 1	哮	aDependency				
- 1	哮	auxRfDependency				
_	哮	tDependency				
	哮	tToken				
_	•	· 🤾 🕠		•		>
	Т	oken				•
	С	afun	•	Sp	~	×
	С	ann_id	•	2	-	×
	С	form	•	Požár	\blacksquare	×
	С	hidden	_	true	~	×
	С	lemma	T	požár	~	×
	С	ord	T	1	~	×
	С	sentence_order	T	0	~	×
	С	tag	\blacksquare	NNIS1A	_	×
	С		\blacksquare		▼	×
	Þ	Open Search & An	notat	e tool		

Netgraph Tree Viewer in GATE (for Stanford Dependencies)



Sentence: Users also said it is in the strongest financial position in its 24-year history.



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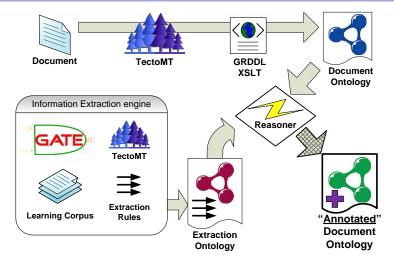
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Transformation of PML to RDF

- Quite simple XSLT transformation
- Allows working with PDT annotations inside Semantic Web tools
 - Ontology Editors
 - Reasoners
 - Query tools (graph queries)
 - ?Visualization and navigation tools?
- In our case interpretation of extraction rules by a OWL reasoner

Problem

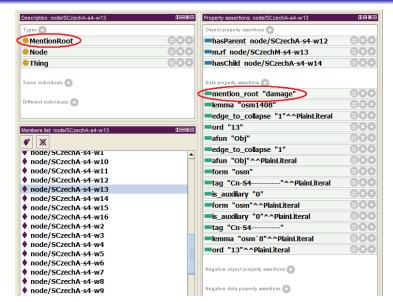
Extraction Rules Interpreted by OWL Reasoner



Tool independent extraction ontologies

Integration with Semantic Tools

PDT in The Protégé Ontology Editor



Examples of extraction rules in the native Prolog format.

```
[Rule 1] [Pos cover = 23 Neg cover = 6]
mention root (acquired, A) :-
   'lex.rf'(B,A), t_lemma(B,'Inc'), tDependency(C,B),
    tDependency (C,D), formeme (D,'n:in+X'), tDependency (E,C).
[Rule 11] [Pos cover = 25 Neg cover = 6]
mention root (acquired, A) :-
   'lex.rf'(B,A), t_lemma(B,'Inc'), tDependency(C,B),
    formeme(C,'n:obj'), tDependency(C,D), functor(D,'APP').
[Rule 75] [Pos cover = 14 Neg cover = 1]
mention root (acquired, A) :-
   'lex.rf'(B,A), t_lemma(B,'Inc'), functor(B,'APP'),
    tDependency (C, B), number (C, pl).
```

Examples of extraction rules in Protégé 4 – Rules View's format

```
[Rule 1]
lex.rf(?b, ?a), t_lemma(?b, "Inc"), tDependency(?c, ?b),
tDependency (?c, ?d), formeme (?d, "n:in+X"),
tDependency (?c, ?e)
      -> mention_root(?a, "acquired")
[Rule 11]
lex.rf(?b, ?a), t_lemma(?b, "Inc"), tDependency(?c, ?b),
formeme(?c, "n:obj"), tDependency(?c, ?d), functor(?d, "APP")
      -> mention root(?a, "acquired")
[Rule 75]
lex.rf(?b, ?a), t_lemma(?b, "Inc"), functor(?b, "APP"),
tDependency(?c, ?b), number(?c, "pl")
      -> mention root(?a, "acquired")
```



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Summary

- Implemented a system for extraction of semantic information
- Based on third party linguistic tools (TectoMT⁴)
- Extraction rules adopted from Netgraph⁵ application.
- ILP used for learning rules.
- All methods integrated inside GATE⁶.
- Main advantages:
 - Automated selection of learning features
 - "Language independent"
 - Rule based

⁴http://ufal.mff.cuni.cz/tectomt/

⁵http://quest.ms.mff.cuni.cz/netgraph/

⁶http://gate.ac.uk/

Conclusion

Future work

- Use some Knowledge Base (e.g. WordNet).
- Adaptation of this method on other languages.
- Evaluation of the method on other datasets.
- Be able to provide more semantics.
 - e.g. sophisticated semantic interpretation of extracted data