



Universität Augsburg
Philologisch-Historische
Fakultät

Aspects of Knowledge Representation for Discourse Relation Annotation

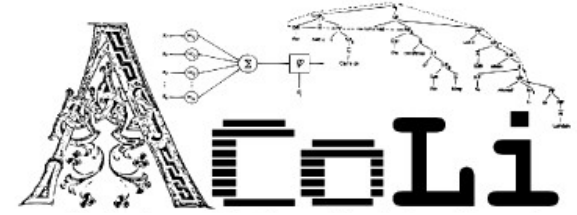
Christian Chiarcos

Applied Computational Linguistics (ACoLi)

University of Augsburg, Germany

Institute of Computer Science, Polish Academy of
Sciences, Warsaw, Nov 21, 2024

Agenda



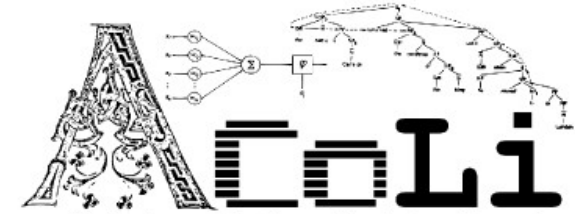
- 1** Semantic Technologies: Knowledge Graphs and Language Technology
- 2** Discourse and Discourse Relations
- 3** Formalizing Discourse Relations
- 4** Linking Discourse Marker Inventories
- 5** Inducing Discourse Marker Inventories
- 6** Annotation Engineering with Knowledge Graph Technologies
- 7** Towards a Multilingual Corpus of Discourse and Reference



Semantic Technologies

Knowledge Graphs and Language Technology

Semantic Technologies



Two main aspects

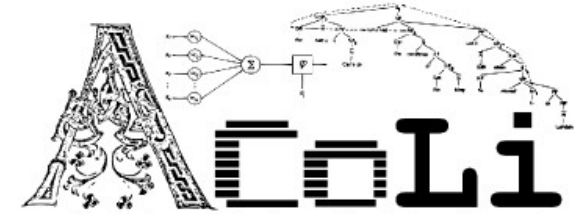
Provide and Process Structured Information

- Knowledge Representation
- technologies and protocols for sharing, accessing and inference over knowledge graphs
 - Ontologies, Property Graphs, graph technologies
- grounded on web (W3C) standards
 - HTTP, URI, RDF, SPARQL, OWL
 - **federation & interoperability**: integrate information that is provided by independent sources with heterogeneous technologies from different locations
- The field formerly known as Semantic Web
(What you expect at ISWC, ESWC, etc.)

Identify Information in Natural Language

- Natural Language Understanding (NLU)
- given natural language input, provide a structured representation of its information according to a specific representation formalism
- traditionally (mostly) supervised learning problems
- addressed in long-standing series of Shared Tasks devoted to individual sub-problems
 - Syntactic Parsing, Named Entity Recognition, Entity Linking, Co-Reference, Information Extraction, Semantic Role Labelling, Semantic Parsing, ...
- One of the primary concerns of the NLP community
(What you expect at ACL, EMNLP, etc.)

Semantic Technologies



Two main aspects

Provide and Process Structured Information

- Knowledge Representation

Identify Information in Natural Language

- Natural Language Understanding (NLU)

Linguistic Data Science

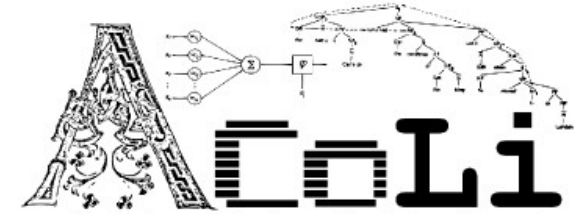
bringing together both
aspects/communities/worlds

use knowledge representation standards to
solve interoperability problems in NLP
(NLU), e.g., in the creation of training data

- The field formerly known as Semantic Web
(What you expect at ISWC, ESWC, etc.)

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Semantic Technologies



Two main aspects

Provide and Process Structured Information

Identify Information in Natural Language



LDK 2023 – 4th Conference on Language, Data and Knowledge



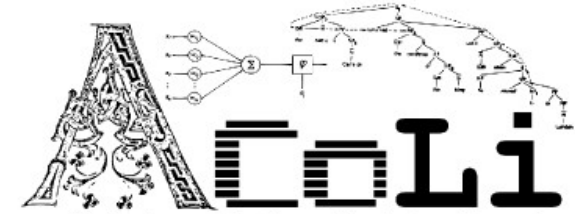
Linguistic Data Science

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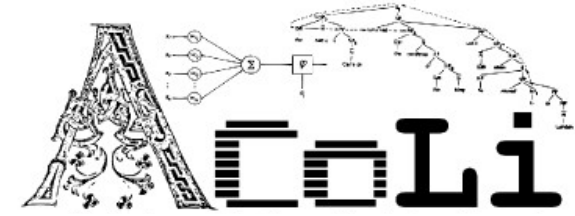
Semantic Technologies



Parts of our Technology Stack (also see <https://github.com/acoli-repo/>)

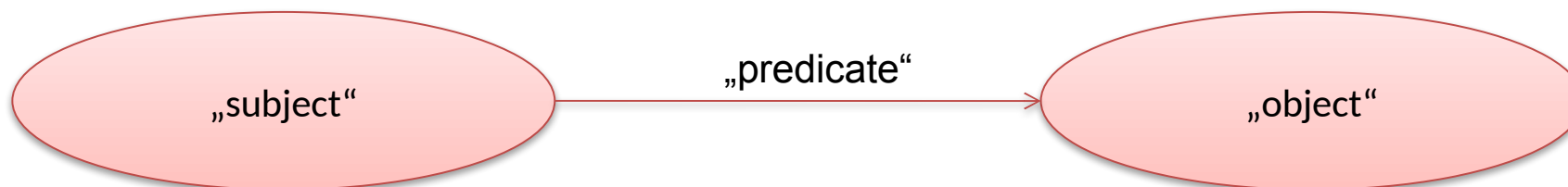
- preprocessing
 - CoNLL-Merge: merge TSV files, normalize tokenization, merge annotations (Chiarcos & Schenk@LDK-2019)
- RDF conversion and enrichment
 - FINTAN: Flexible Integrated Transformation and Annotation eEngineering platform
 - more in a moment ;)(Fäth et al.@ LREC-2020)
 - CoNLL-RDF: FINTAN customization for CoNLL/TSV files (Chiarcos & Fäth@LDK-2017)
- selected knowledge graphs
 - ACoLi Dictionary Graph: 3000+ bilingual dictionaries (Chiarcos et al.@LREC-2020)
 - Ontologies of Linguistic Annotation: 100+ annotation schemes (Chiarcos & Sukhareva, SWJ, 2015)
- foundational standards cf. Cimiano, Chiarcos, Gracia & McCrae (2020), *Linguistic Linked Data*. Springer, Cham
 - W3C standards: URI, HTTP, RDF
 - community standards: OntoLex, NLP Interchange Format, CoNLL-RDF data model

Resource Description Framework (RDF)



<https://www.w3.org/RDF/>

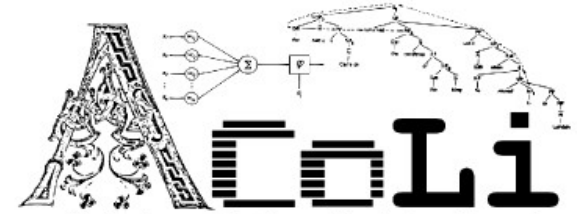
- a (labeled directed multi-) graph
 - nodes („RDF resources“)
 - anything we want to provide information about
 - edges („RDF properties“)
 - assigns a source node („subject“) a target node („object“) or a value („literal“)
 - nodes and edges are unambiguously identified
 - Uniform Resource Identifiers (URIs), e.g., URLs



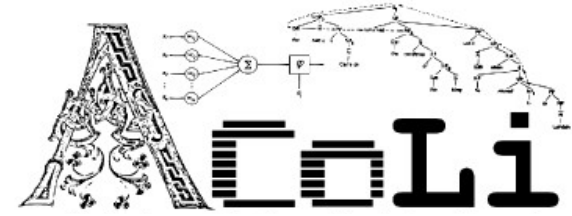
Resource Description Framework (RDF)

<https://www.w3.org/RDF/>

glottolog:bamb1269 rdf:type dcterms:LinguisticSystem.

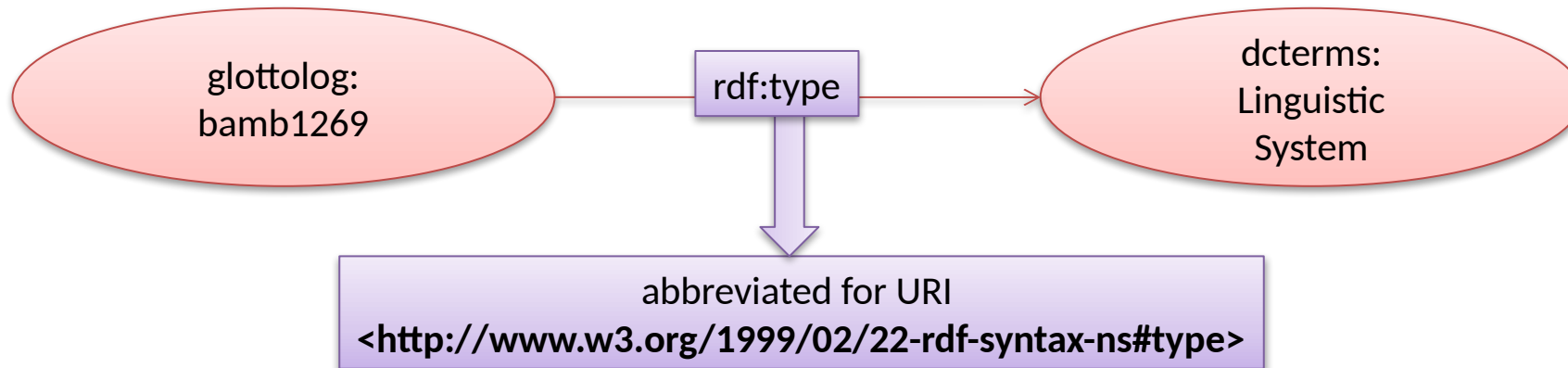


Resource Description Framework (RDF)



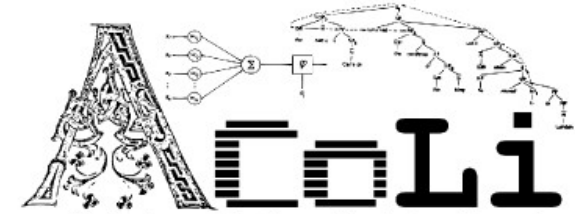
<https://www.w3.org/RDF/>

glottolog:bamb1269 rdf:type dcterms:LinguisticSystem.



*could be opened in a browser
resolvable URIs may provide further information*

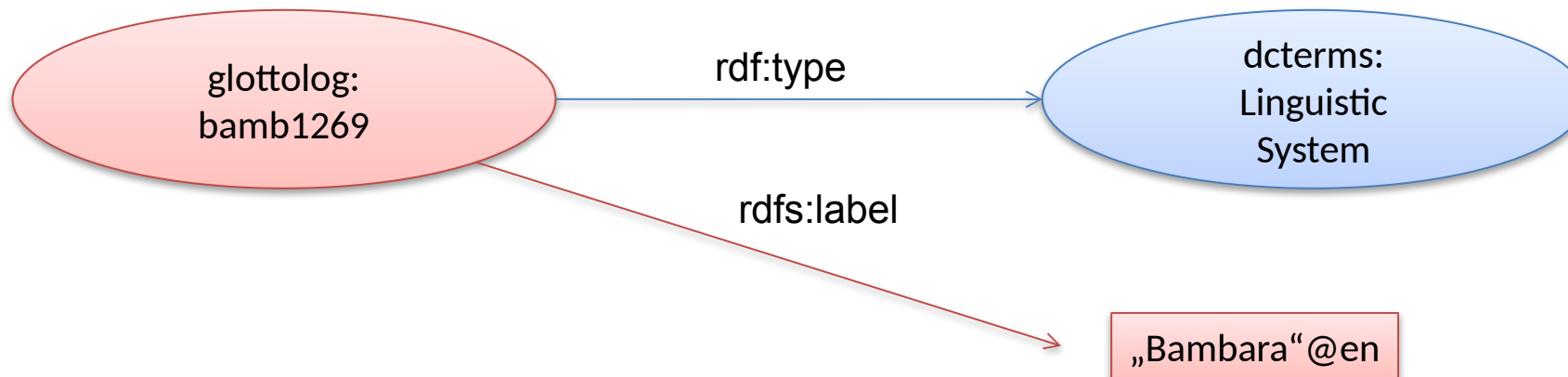
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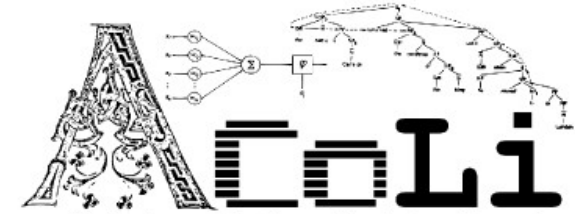
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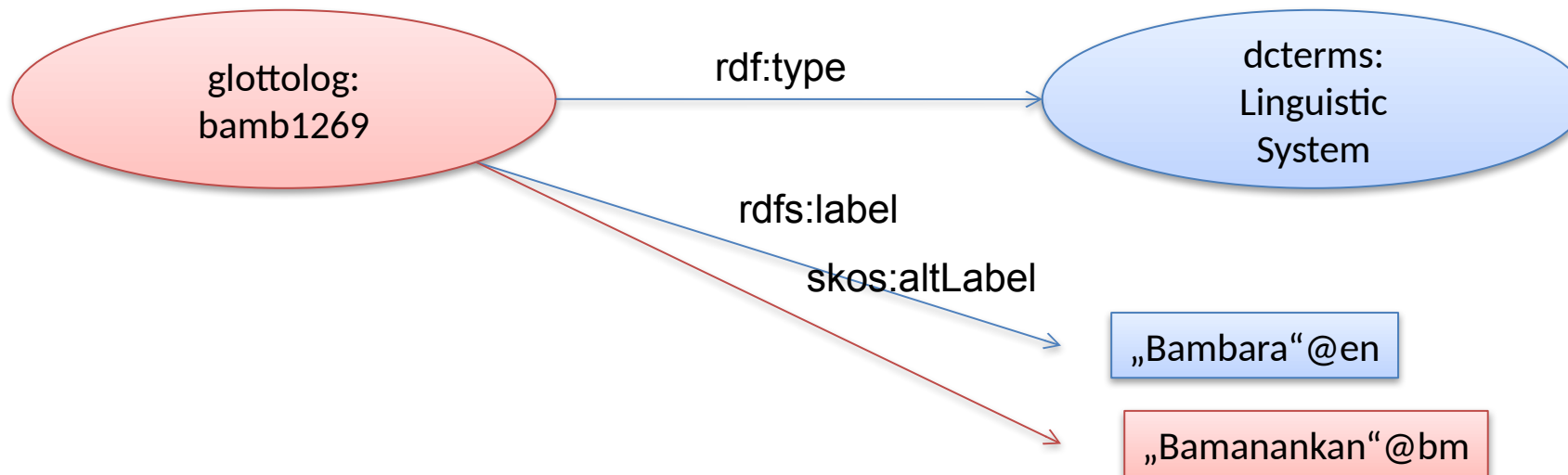


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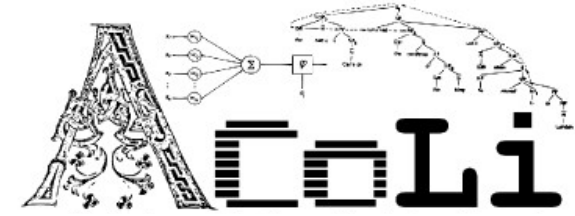
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glottolog:bamb1269 skos:altLabel „Bamanankan“@bm.



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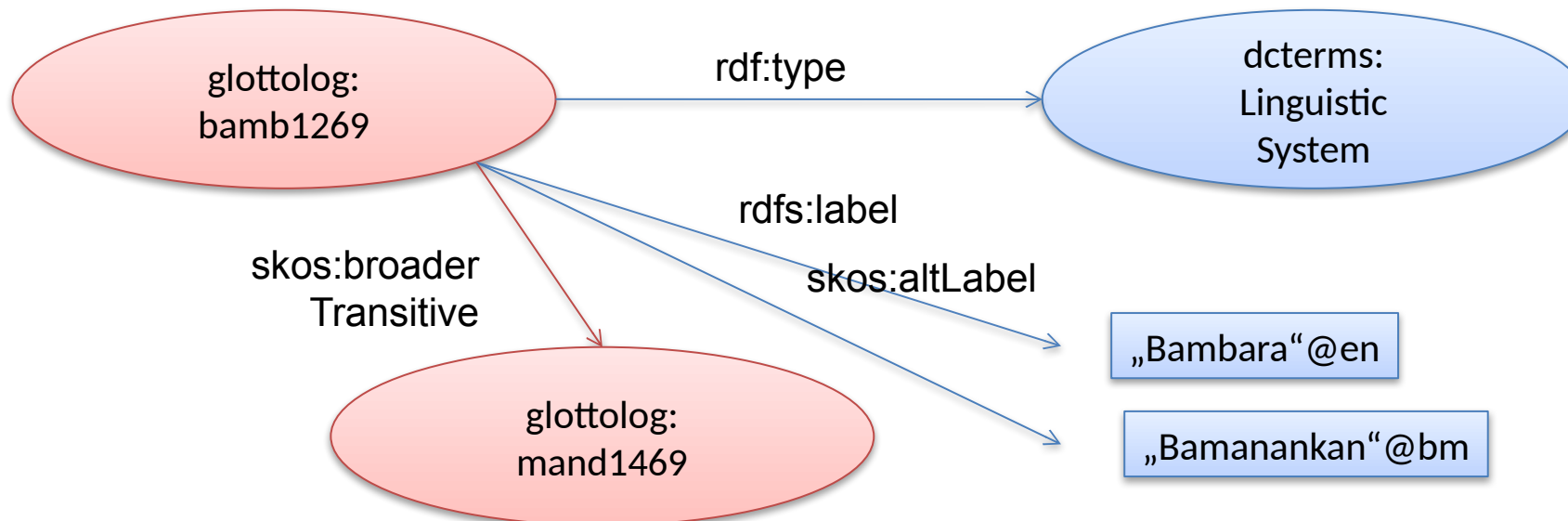
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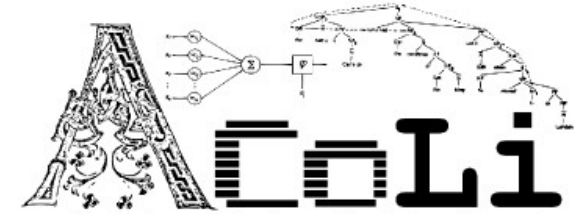
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glottolog:bamb1269 skos:broaderTransitive glottolog:mand1469.



Resource Description Framework (RDF)



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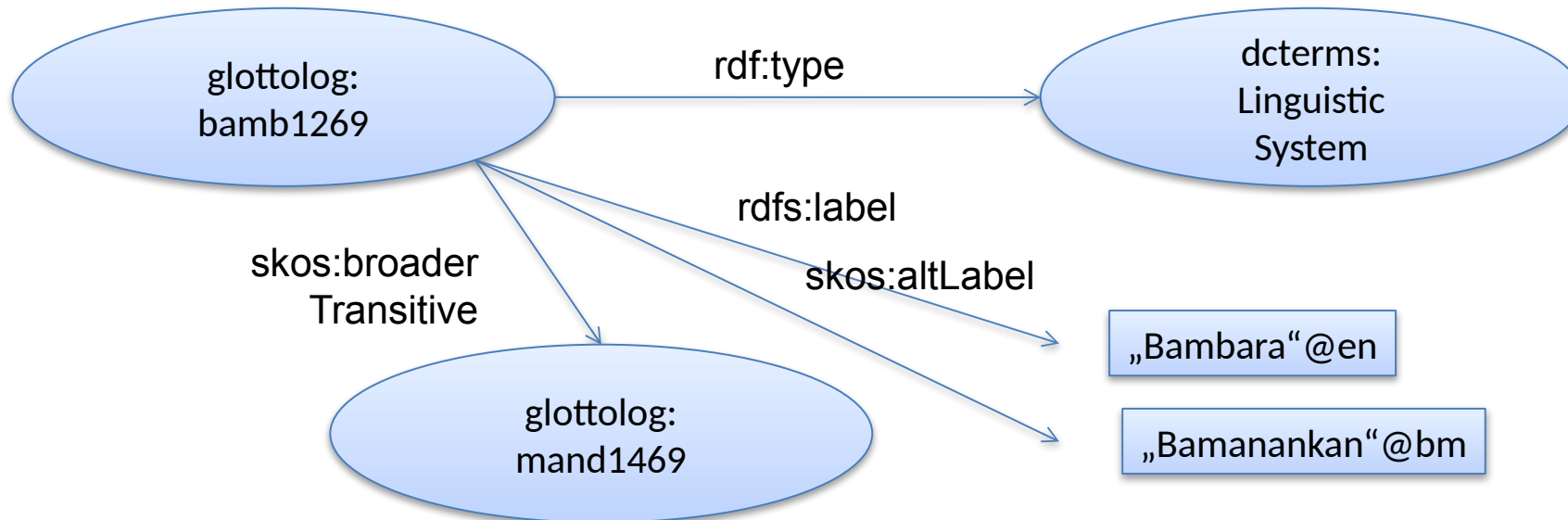
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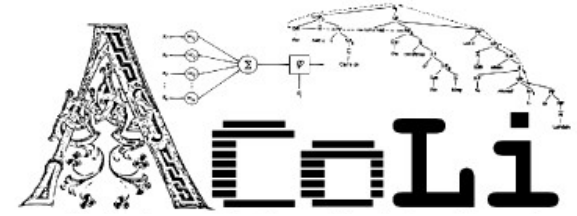
triple
notation
(Turtle)



graphical
notation

Querying it with SPARQL

<https://www.w3.org/TR/sparql11-query/>



`glottolog:bamb1269 rdfs:label „Bambara"@en.`

triple
notation
(Turtle)

`glottolog:bamb1269 skos:broaderTransitive glottolog:mand1469.`

```
SELECT ?language_name
```

```
WHERE {
```

```
  ?language rdfs:label ?language_name.
```

```
  ?language skos:broaderTransitive glottolog:mand1469.
```

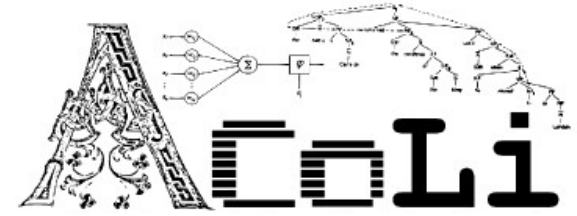
```
}
```

query
(SPARQL)

“give me the names of all
Mande (glottolog:mand1469)
languages”

Querying it with SPARQL*

<https://www.w3.org/TR/sparql11-query/>



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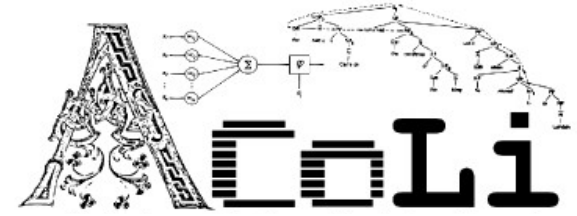
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  ?language skos:broaderTransitive glottolog:mand1469.
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query
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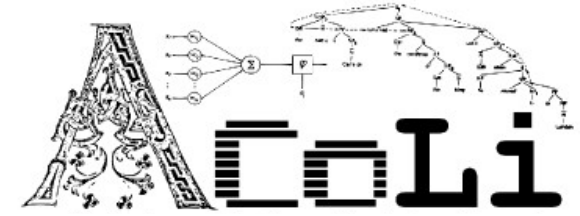
* with SELECT, we perform queries
with DELETE and INSERT, we perform updates



Rules of best practice for publishing data on the web

- use URIs as names for things (1)
links to external URIs retrieve more information
 - **if** they can be resolved via HTTP (2)
 - **and** provide information as RDF, SPARQL, etc. (3)
 - **and** they include links to other URIs (4)
- ⇒ **then**, this is Linked Data (informally)

<http://www.w3.org/DesignIssues/LinkedData.html>



Rules of best practice

=> Information integration

- ❑ Interoperability
 - => **the same query** to query different datasets

- ❑ Federation
 - data published on the web
 - ❑ with a query interface (SPARQL end point)
 - => **a single query** to query different datasets simultaneously

a formalism to
„build bridges“
=> more (re-)usable resources and technologies

coupled with the dynamics of the
open source / open data
movement

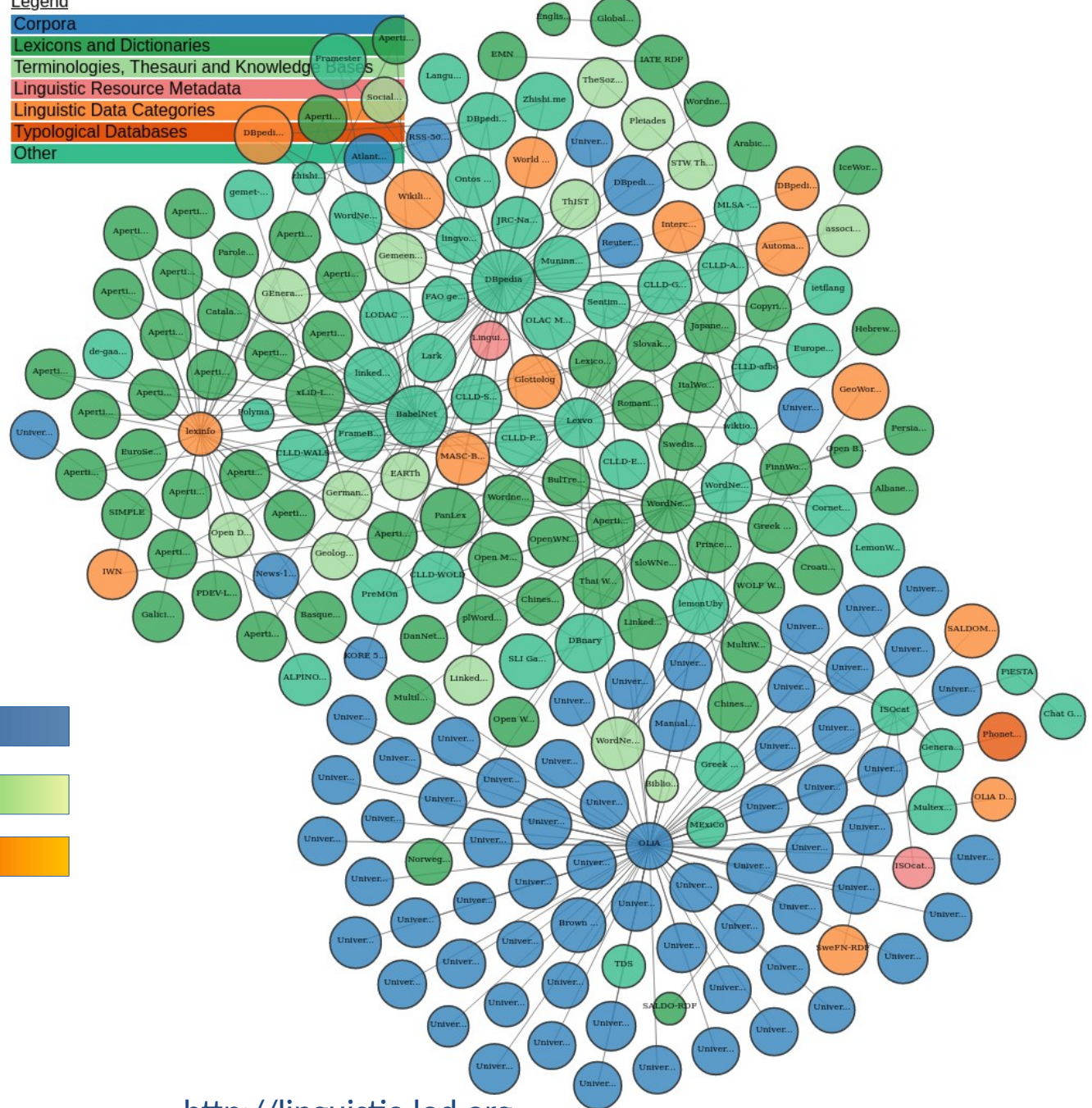
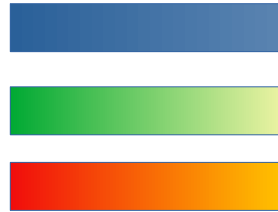
Linked Open Data

Linguistic Linked Open Data (LLOD)

LLOD cloud diagram

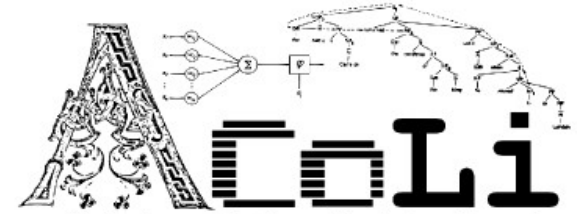
sub-diagram of the Linked Open Data (LOD) cloud diagram

- open** resources for
- linguistic annotation
 - lexical/conceptual knowledge
 - linguistically relevant metadata



<http://linguistic-lod.org>

Limitations and Potential



- Linked Data and RDF tech are not (and probably won't ever be) sufficiently user-friendly for end users (say, a linguist)

- **BUT**

- Most users won't have to work with it directly, but only mediated through software tools.

- **IF**

- The data can be prepared by/for them

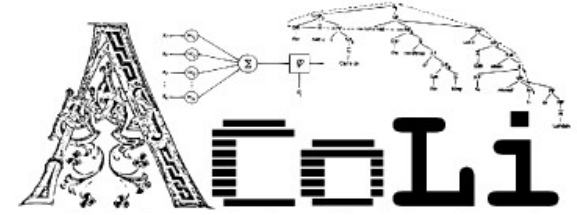
Illustrated here for aspects of discourse annotation

A background network diagram consisting of numerous small grey circular nodes connected by thin, light grey lines. The nodes are distributed across the entire frame, with a higher density in the lower right quadrant. The overall appearance is that of a complex, interconnected web or graph.

Discourse and Discourse Relations

Some Theoretical Background

Discourse

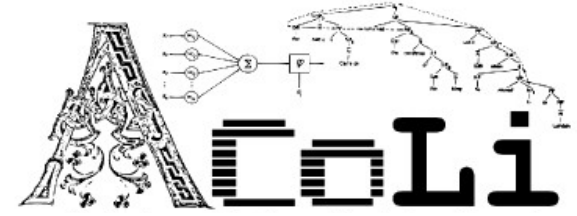


Some non-trivial aspects of Natural Language Understanding

- In Natural Language Understanding, the semantic analysis of individual sentences is an established field of research (and to a large extent, solved).
- But there is meaning between the lines (resp., sentences) ...
 - Peter pushed John.
 - He was hurt badly.

Who was hurt?

Discourse



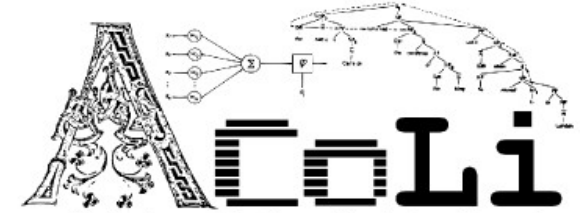
Some non-trivial aspects of Natural Language Understanding

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Probably

John was hurt
and this is the result of Peter pushing him

Discourse



Some non-trivial aspects of Natural Language Understanding

- In Natural Language Understanding, the semantic analysis of individual sentences is an established field of research (and to a large extent, solved).
- But there is meaning between the lines (resp., sentences) ...
 - Peter pushed John.
 - He was hurt badly.
 - Peter pushed John.
 - He was hurt badly.
 - And so, the retaliation felt good, for a while.

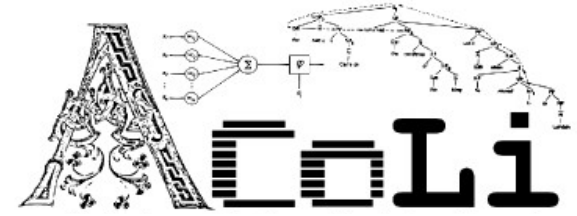
Probably

John was hurt
and this is the result of Peter pushing him

Could also be

Peter was hurt
and he pushed John for retaliation

Discourse Markers



How to make the meaning between the lines visible

- Make explicit how an utterance (clause, sentence, etc.) is linked to its discourse context
 - lexical expressions, mostly conjunctions, adverbs and PPs
 - John can't go. ...

And Mary can't go either.

additive

Therefore, Mary can't go either.

causal

However, Mary can't go either.

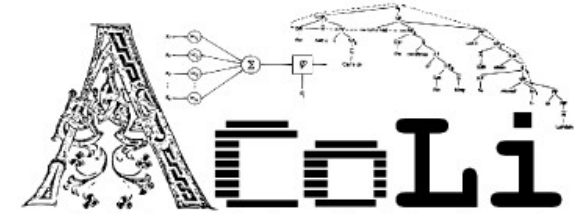
contrastive

Mary can't go either.

*implicit
(unmarked)*

relation

Discourse Relations



The meaning of discourse markers

■ Semantic, pragmatic or conversational relation holding between an utterance and its discourse context.

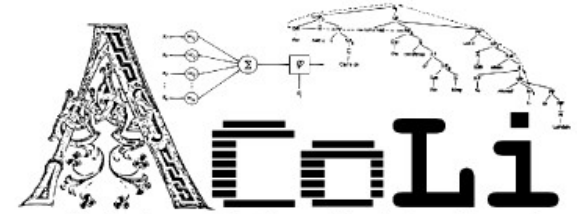
□ different theories and annotation frameworks

- Coherence relations (Hobbs, 1979)
- RST (Mann and Thompson, 1987)
- SDRT (Asher & Lascarides, 2003)
- PDTB (Prasad et al., 2008)

overlapping in intent and content, but not compatible with each other

We focus on RST and PDTB here, these provide the major corpora.

Discourse Parsing



- Identify relations to assess how utterances are connected for information extraction, text summarization, machine translation, etc.

off-the-shelf LLMs are still rather poor at such context-dependencies ;)

- Two primary (and incompatible) frameworks

Rhetorical Structure Theory (RST)

discourse relations constitute a tree structure that encompasses all utterances of a coherent discourse

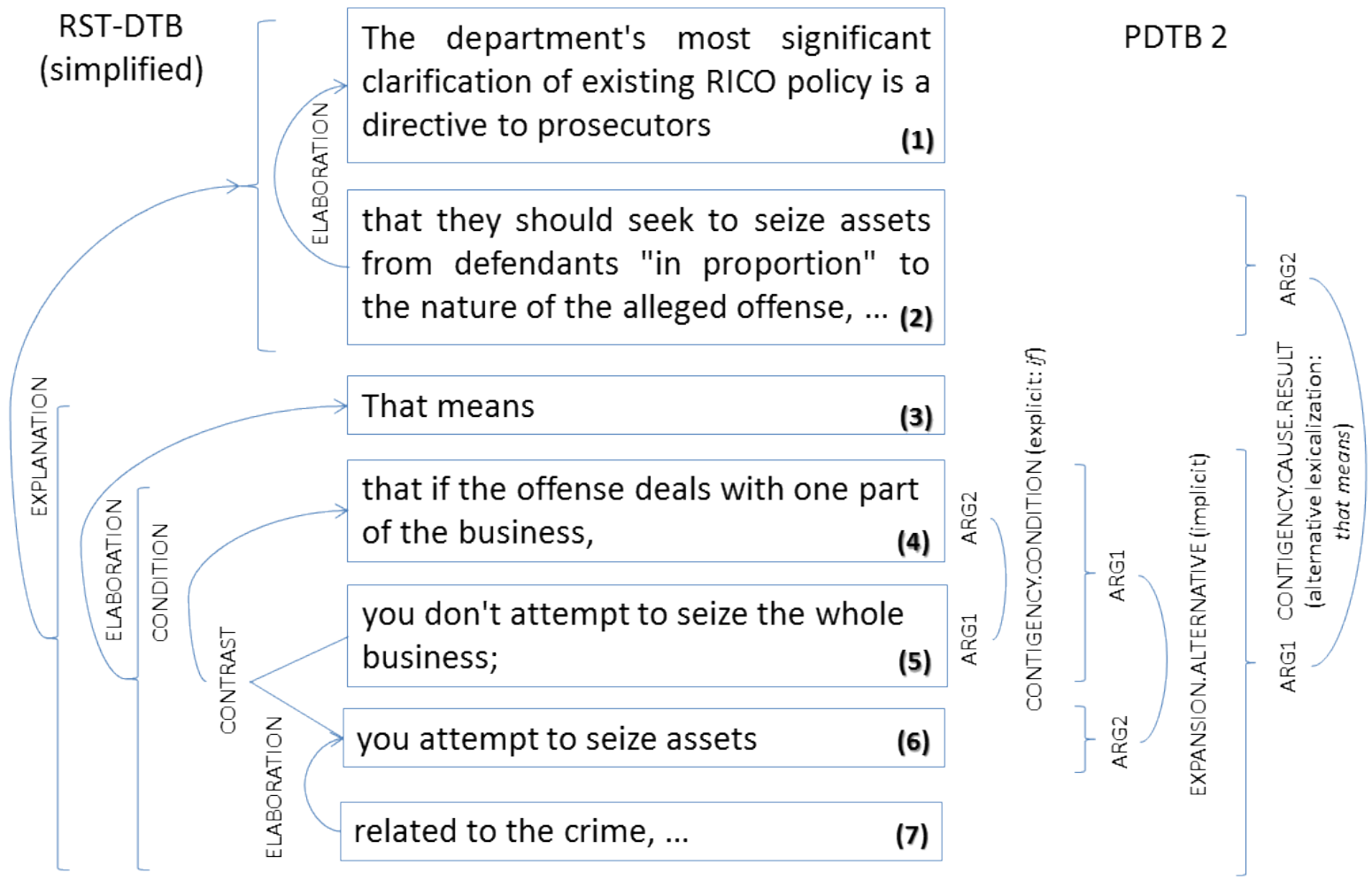
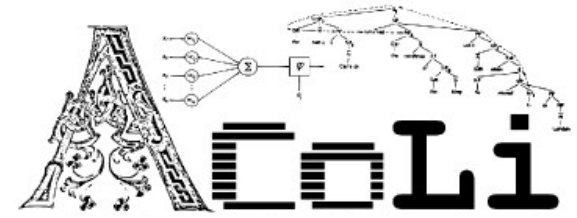
(“deep” discourse parsing)

Penn Discourse Treebank (PDTB)

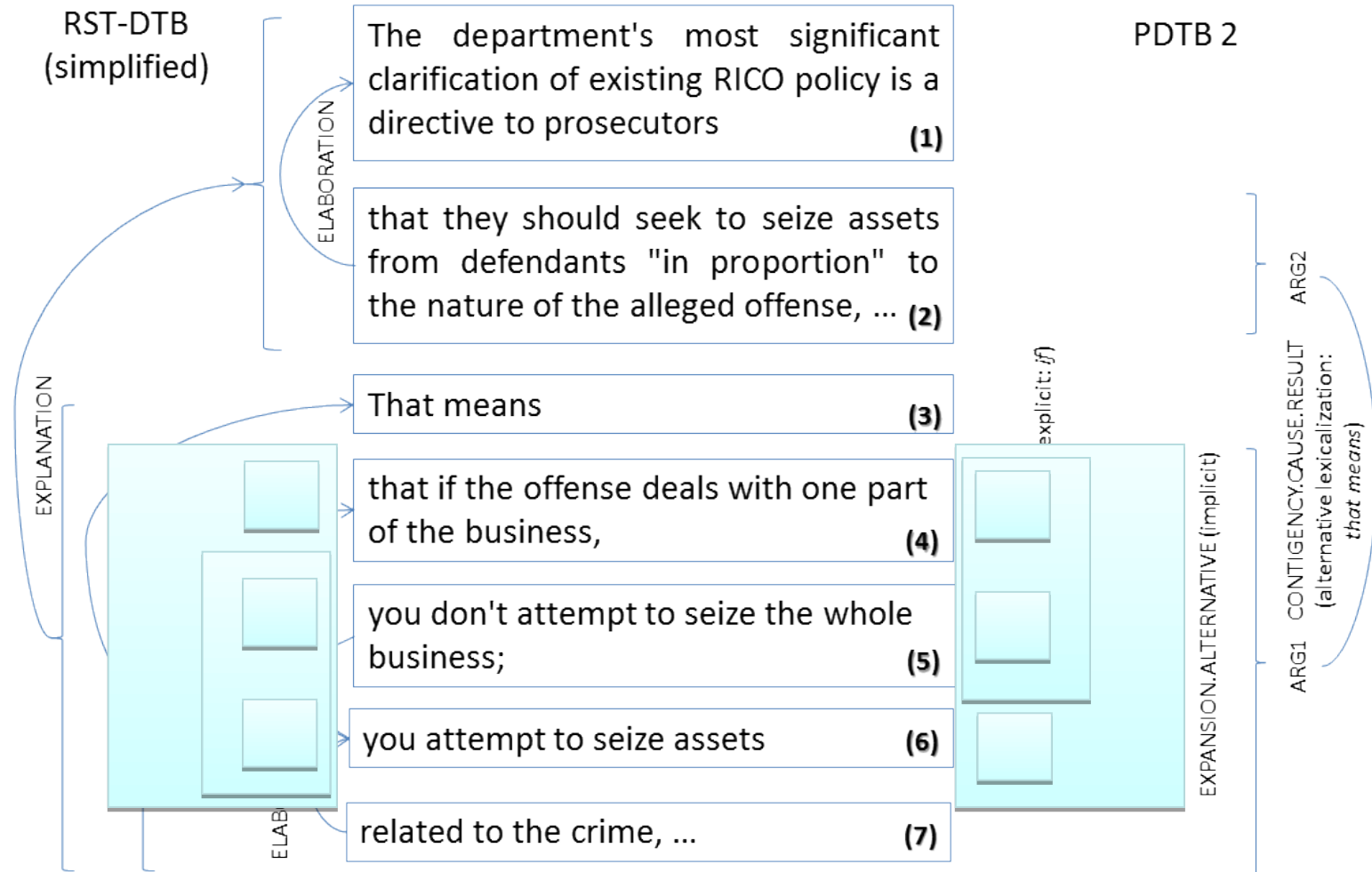
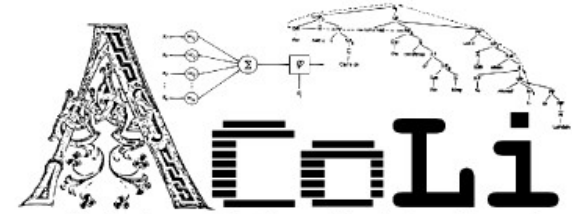
forget about the tree
annotate any discourse relation you see in the local context

(shallow discourse parsing)

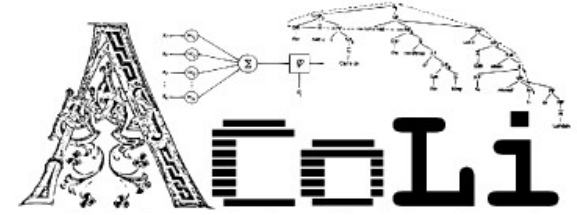
Comparing RST and PDTB Annotations



Comparing RST and PDTB: Structural Differences



Goals: Consolidate and Integrate Existing Data Sources



- across languages
 - multilingual discourse markers
- across frameworks
 - RST, PDTB, etc.
- across formats
 - various CSV, XML and special-purpose formats
- machine-readable semantics
 - knowledge graph(s)



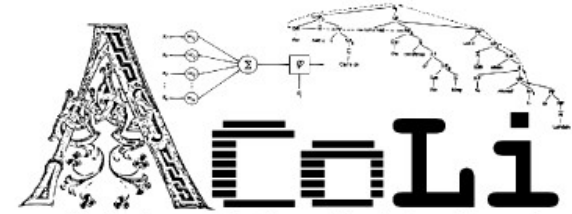
Formalizing Discourse Relations

Discourse in the Ontologies of Linguistic Annotation (OLiA)

(Chiarcos@LREC-2014)

What is an Ontology?

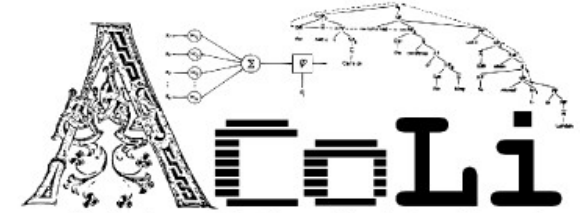
... in Knowledge Representation



An ontology is the formalization of concepts and their relations for a particular domain

- Formalized in terms of the Web Ontology Language (OWL)
 - i.e., an RDF vocabulary for classes (concepts), properties (relations) and axioms
- selected properties
 - *rdf:type* (a) assign a class (type) to an object
 - *rdfs:subClassOf* (\sqsubseteq) subclass relation (cf. logical \rightarrow)
 - *owl:intersectionOf* (\sqcap) intersection between two classes (cf. logical \wedge)
 - *owl:unionOf* (\sqcup) union between two classes (cf. logical \vee)
 - *owl:complementOf* (\neg) complement of a class (cf. logical \neg)

Ontologies of Linguistic Annotation (OLiA)

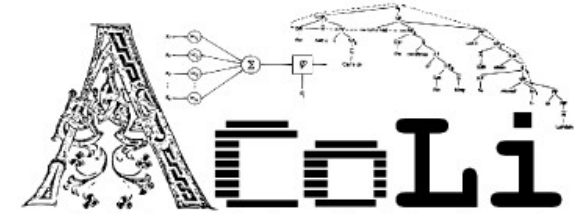


<http://purl.org/olia>

Chiarcos and Sukhareva, SWJ 2015

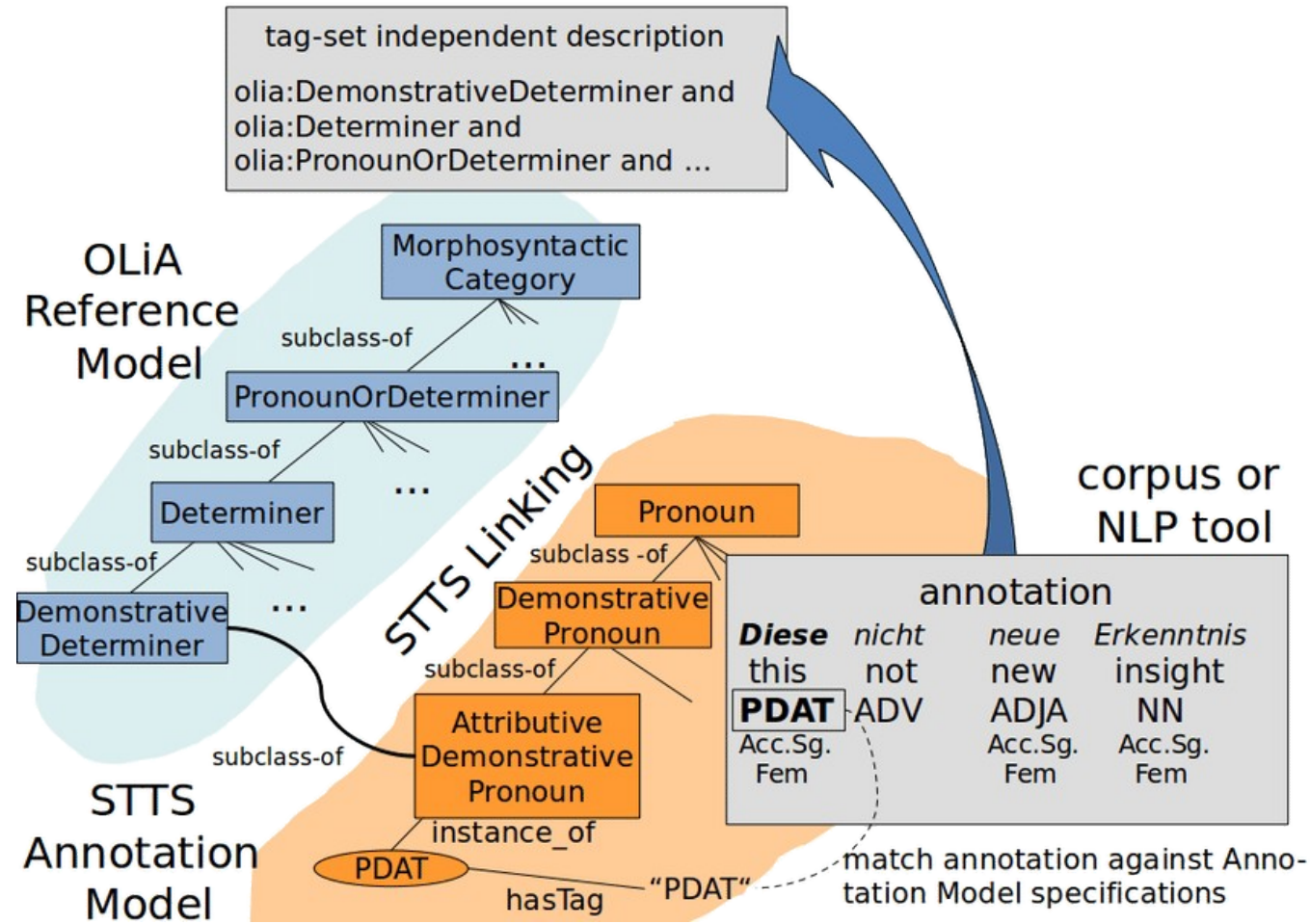
- ❑ one ontology per annotation schema
 - OLiA Annotation Model
- ❑ one ontology that defines common terminology
 - OLiA Reference Model
- ❑ one RDF file with `rdfs:subClassOf` statements
 - OLiA Linking Model: Annotation Model => Reference Model
- ❑ annotation schemas for 100+ languages
 - mostly morphosyntax, inflectional features and syntax

Ontologies of Linguistic Annotation (OLiA)

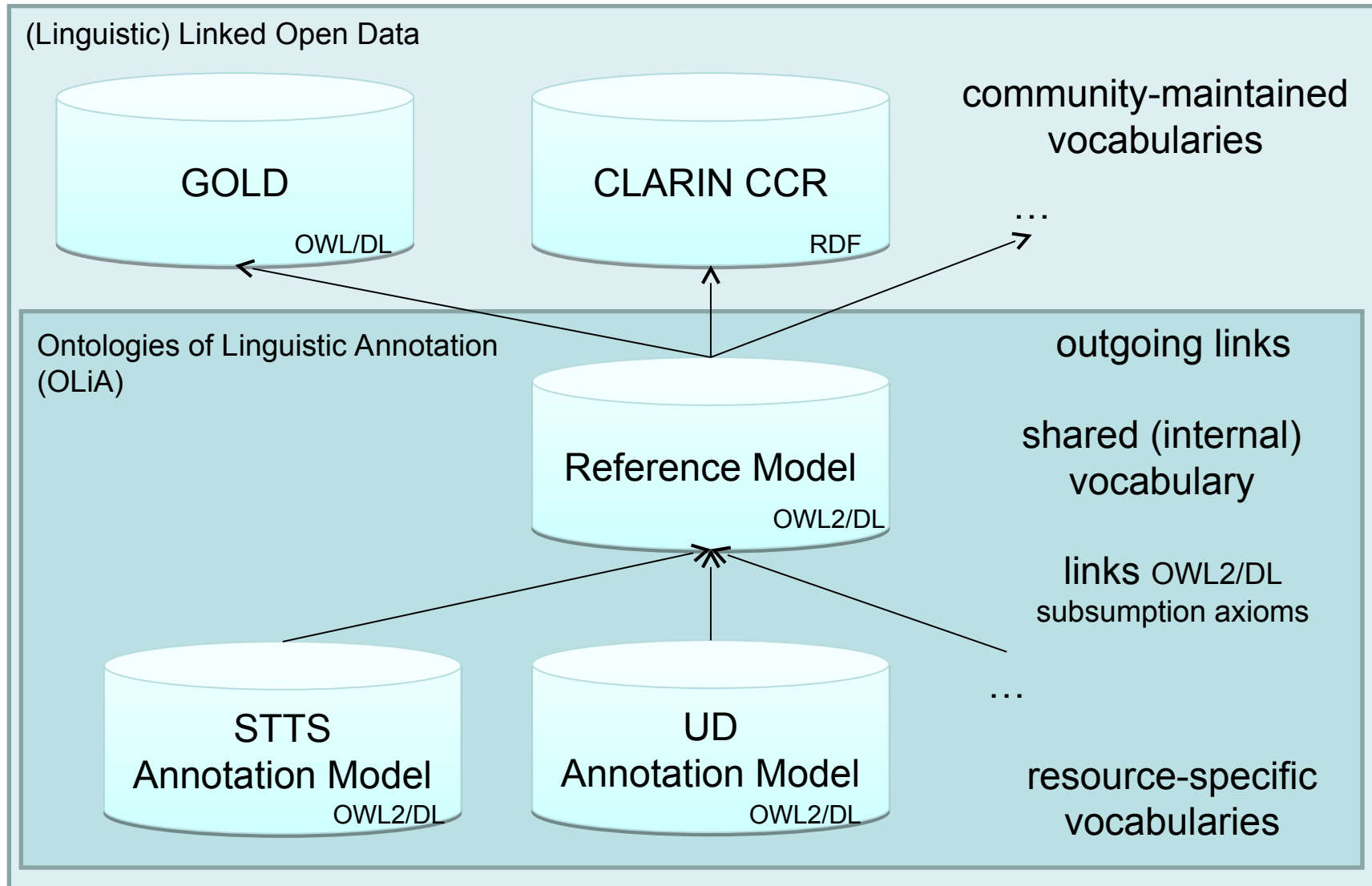
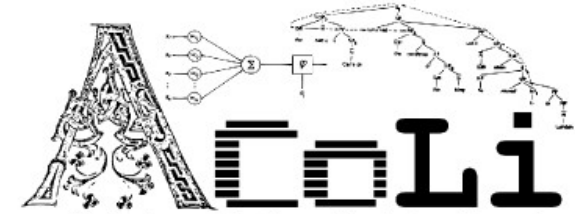


<http://purl.org/olia>

German
parts of
speech

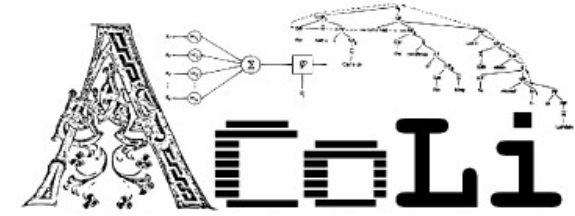


Ontologies of Linguistic Annotation (OLiA)

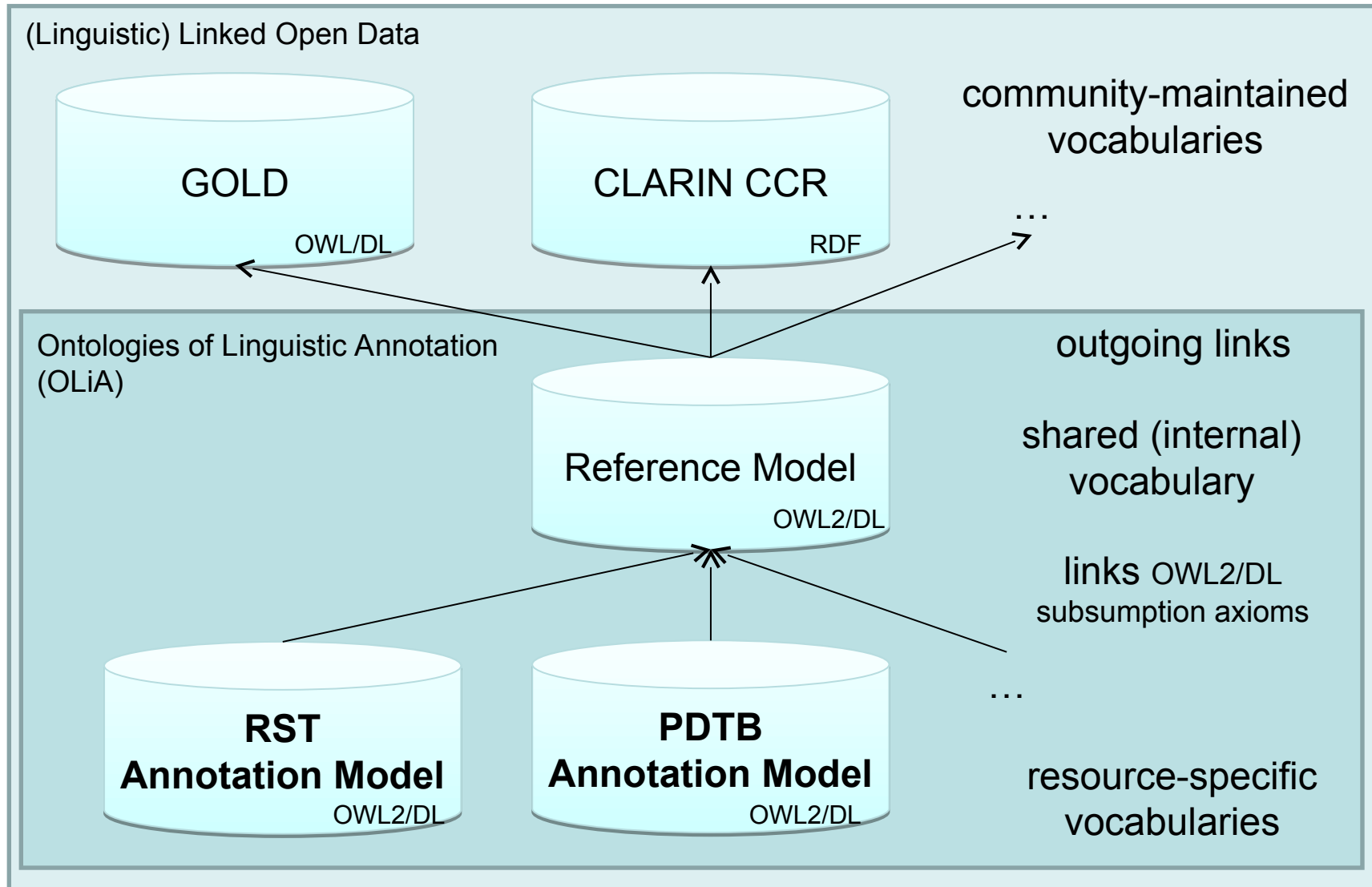


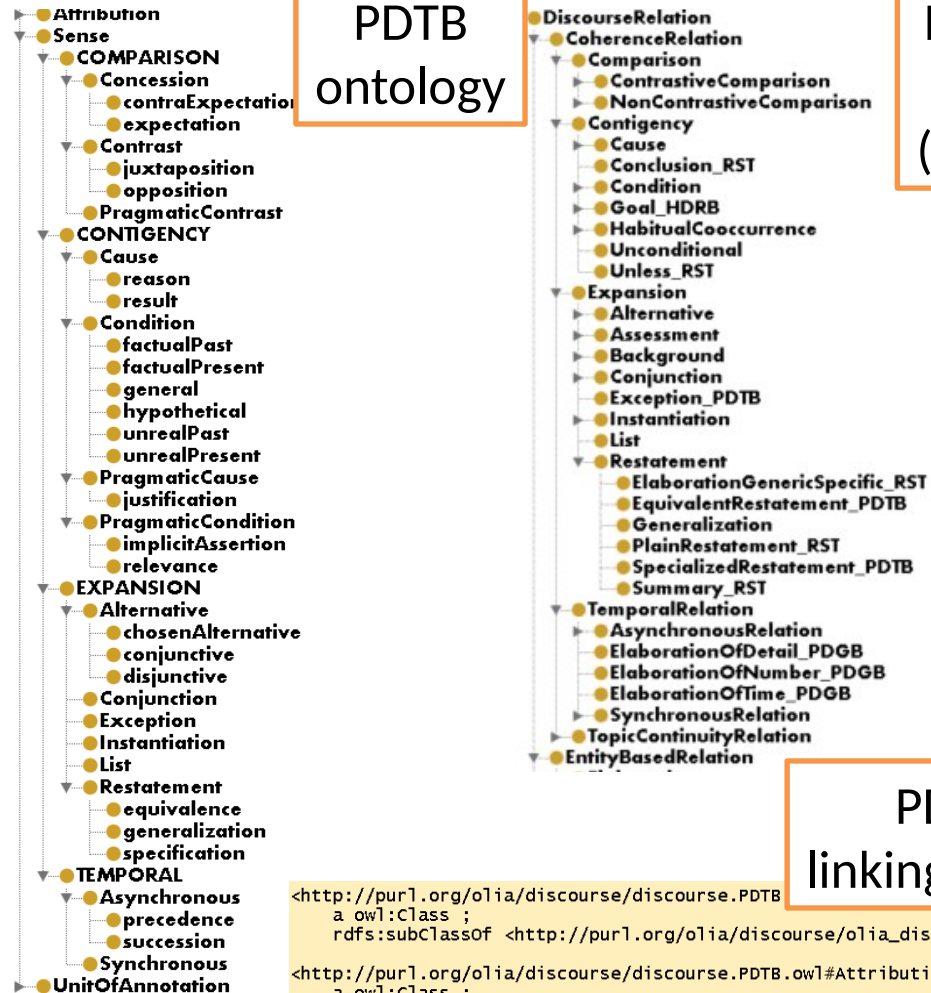
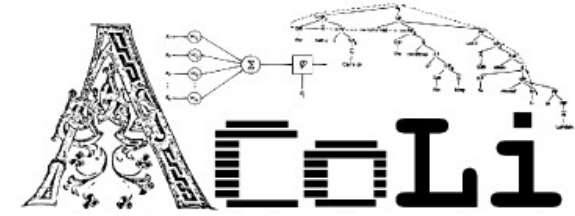
OLiA Discourse Extension

(Chiarcos@LREC-2014)



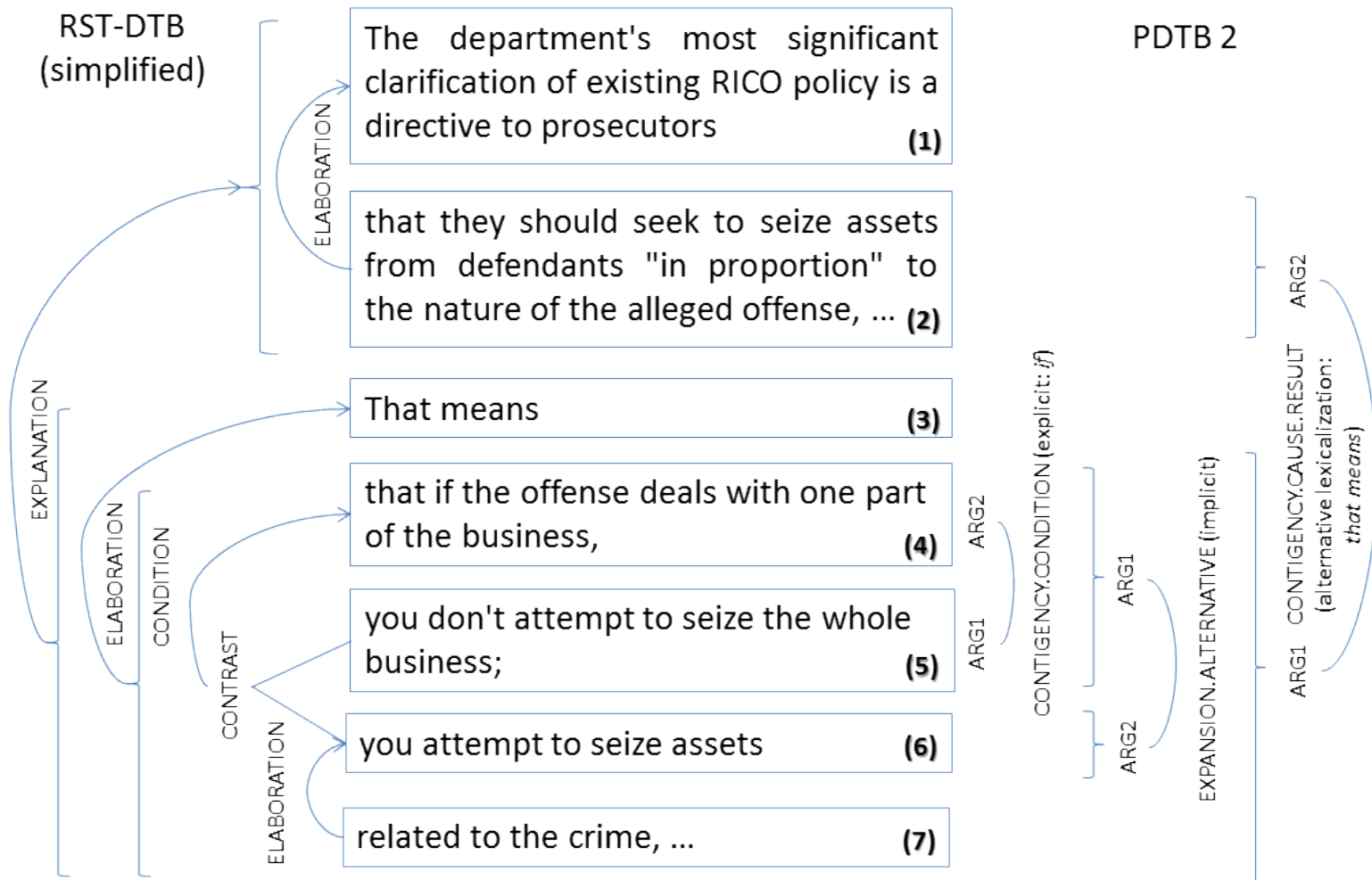
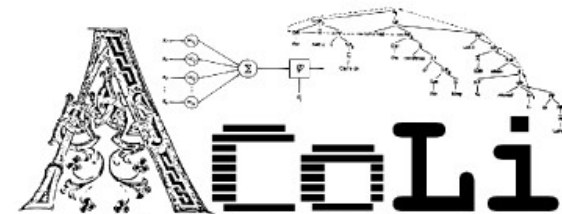
Annotation models and reference definitions for RST, PDTB, etc.



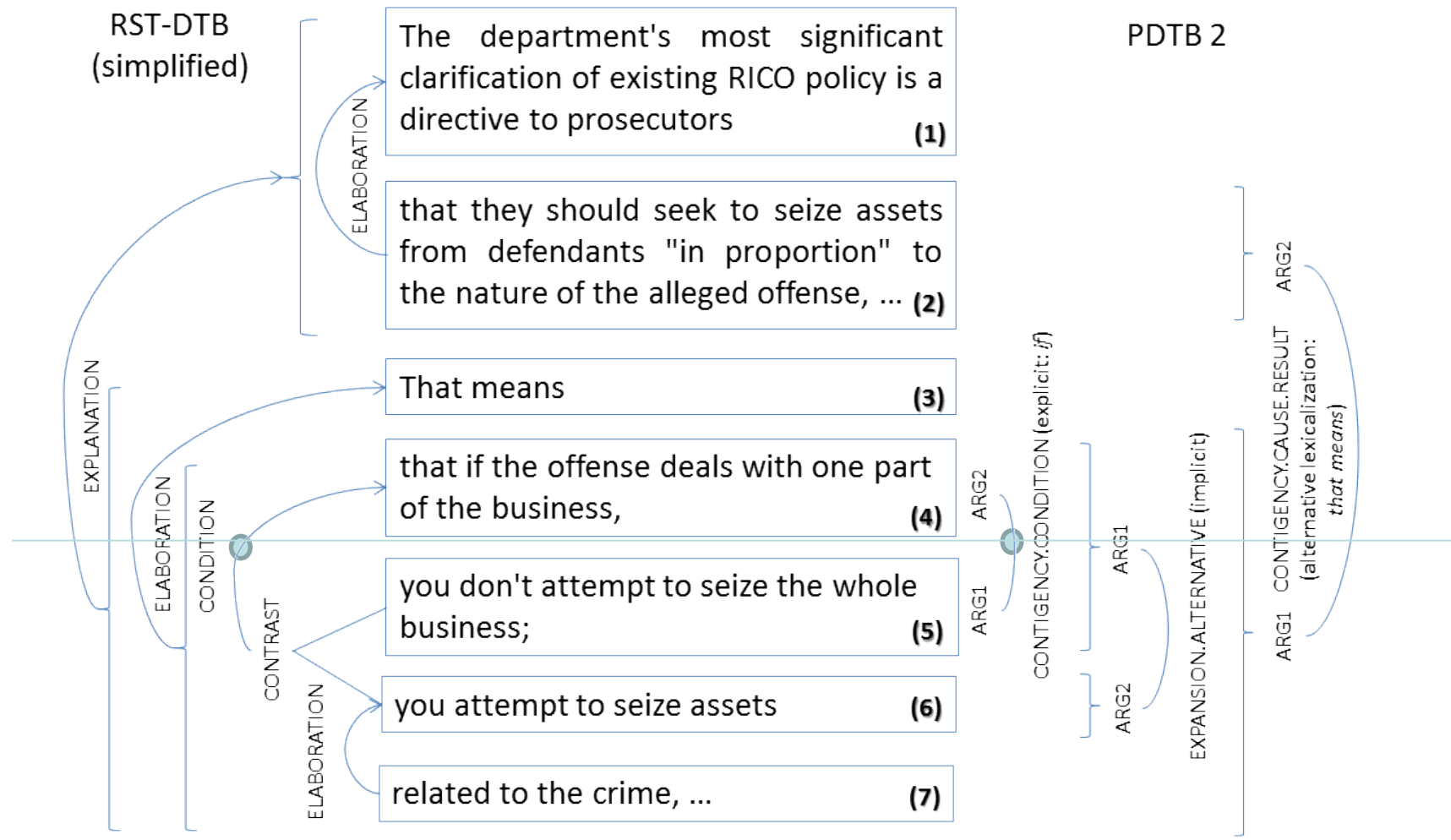
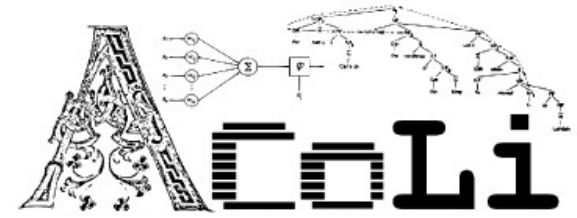


- top-level structure based on PDTB
- enriched to cover RST and other corpora
- linked with meta-vocabularies (CCR, ISO SemAF Core DRs)

Comparing Across Frameworks



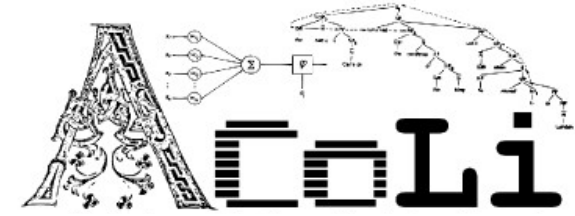
Comparing Across Frameworks: What links (4) and (5)?



CONDITION

CONTINGENCY.CONDITION.GENERAL

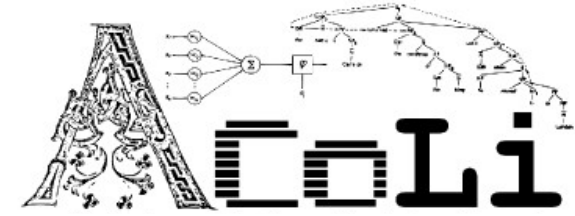
Comparing Across Frameworks: What links (4) and (5)?



PDTB Annotation & Linking	<code>pdtb:contingency.condition.general a pdtb:GeneralCondition</code>
	<code>pdtb:GeneralCondition ⊆ olia:GeneralCondition_PDTB</code>
OLiA Reference Model	<code>olia:GeneralCondition_PDTB ⊆ olia:SemanticCondition</code>
	<code>olia:SemanticCondition ⊆ olia:Condition</code>
	<code>olia:Condition ⊆ olia:Contingency</code>
	<code>olia:Contingency ⊆ olia:CoherenceRelation</code>
	<code>olia:CoherenceRelation ⊆ olia:DiscourseRelation</code>
RST Linking & Annotation	<code>olia:ParatacticDiscourseRelation ⊆ olia:DiscourseStructuralPattern</code>
	<code>olia:DiscourseStructuralPattern ⊆ olia:DiscourseRelation</code>
	<code>rst:Condition ⊆ olia:SemanticCondition ∩ olia:ParatacticDiscourseRelation</code>
	<code>rst:condition a rst:Condition</code>

- The annotations are not quite alike, but for the utterances under consideration, they agree on the features **in bold**

Comparing Across Frameworks: What links (4) and (5)?



PDTB Annotation & Linking	<code>pdtdb:contingency.condition.general a pdtdb:GeneralCondition</code>
	<code>pdtdb:GeneralCondition \sqsubseteq olia:GeneralCondition_PDTB</code>
OLiA Reference Model	<code>olia:GeneralCondition_PDTB \sqsubseteq olia:SemanticCondition</code>
	<code>olia:SemanticCondition \sqsubseteq olia:Condition</code>
	<code>olia:Condition \sqsubseteq olia:Contingency</code>
	<code>olia:Contingency \sqsubseteq olia:CoherenceRelation</code>
RST Linking & Annotation	<code>olia:CoherenceRelation \sqsubseteq olia:DiscourseRelation</code>
	<code>olia:ParatacticDiscourseRelation \sqsubseteq olia:DiscourseStructuralPattern</code>
RST Linking & Annotation	<code>olia:DiscourseStructuralPattern \sqsubseteq olia:DiscourseRelation</code>
	<code>rst:Condition \sqsubseteq olia:SemanticCondition \sqcap olia:ParatacticDiscourseRelation</code>
	<code>rst:condition a rst:Condition</code>

- The annotations are not quite alike, but for the utterances under consideration, they agree on the features **in bold**
- We can now compare across frameworks
 - and we can derive a mapping between them
 - SPARQL \Rightarrow the shortest path of `rdf:type (a)` and `rdfs:subClassOf (\sqsubseteq)` statements

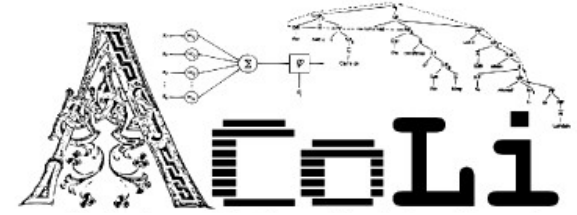
A background network graph with light blue nodes and thin grey lines connecting them, set against a light blue gradient background. The nodes are scattered across the frame, with a higher density in the lower right quadrant.

Linking Discourse Marker Inventories

From Discourse Marker Inventories to OntoLex (RDF)

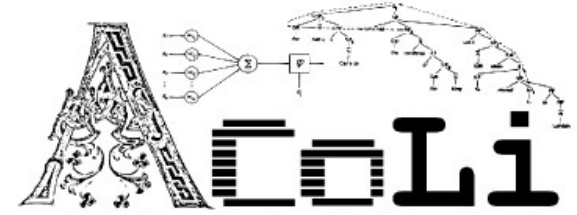
(Chiarcos & Ionov@LDK-2021)

Discourse Marker Inventories



- The most elementary step in discourse annotation is to identify discourse markers and their respective relations
- For a considerable number of languages, discourse marker inventories have been developed
 - to facilitate discourse parsing and downstream tasks
 - map discourse markers to (possible) discourse relations
- Different formats, different theoretical frameworks
 - ⇒ Our contribution: consolidation and integration

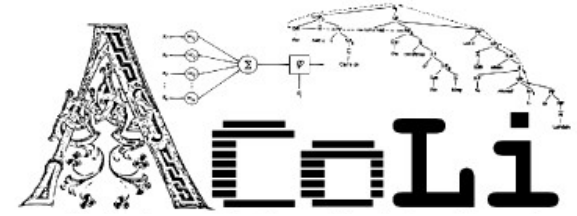
Related Research: TextLink



■ TextLink

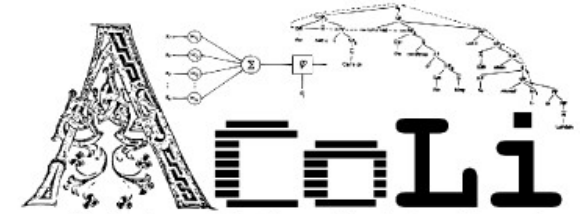
- *Cost Action Structuring Discourse in Multilingual Europe (2014-2018)*
- multilingual discourse marker inventories
 - (mostly) providing PDTB relations as senses
 - (mostly) following a consistent XML format (DimLex, Stede & Umbach 1998)
- <http://connective-lex.info/>

Beyond TextLink: We aimed to improve ...



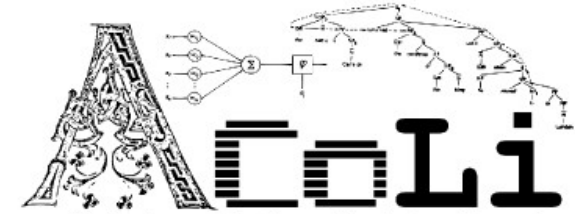
- coverage
 - extend the range of languages and resources
- semantics
 - provide machine-readable semantics
 - preserve the original sense definitions
- usability
 - enable cross-framework comparison and search
 - link and query across languages

Beyond TextLink



- An RDF edition of TextLink and other discourse marker inventories
 - using the RDF vocabulary OntoLex for machine-readable dictionaries
- Maintain original sense definitions (discourse relations)
 - link with OLiA annotation models (PDTB, RST, ...)
- Map flexibly between frameworks
 - traverse with SPARQL
 - PDTB -> OLiA reference model -> RST (or ISO SemAF, CCR, etc.)

Example: German DimLex-XML

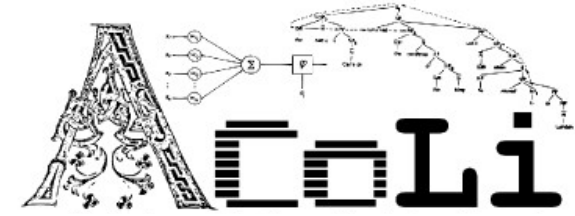


■ Format: DimLex-XML

```
<dimlex>
  <entry id="k1" word="aber">
    <orths>
      <orth type="cont" canonical="1" onr="k1o1">
        <part type="single">aber</part>
      </orth>
    </orths>
    <non_conn_reading>
      <example type="ADV" tfreq="940">aber und abermals</example>
      <example type="ADV">Du bist aber fies!</example>
    </non_conn_reading>
    <syn>
      <cat>konnadv</cat>
      <ordering>
        <ante>0</ante>
        <post>1</post>
        <insert>0</insert>
      </ordering>
      <sem>
        <pdtb3_relation sense="concession-arg2-as-denier" freq="7" anno_N="18"/>
      </sem>
    </syn>
  </entry>
  ...
</dimlex>
```

- Scheffler & Stede (2016)
 - CC-BY-NC-SA 4.0
 - <https://github.com/discourse-lab/dimlex>
- 274 entries
 - 763 forms
 - 432 sense links (28 PDTB 3.0 relations)

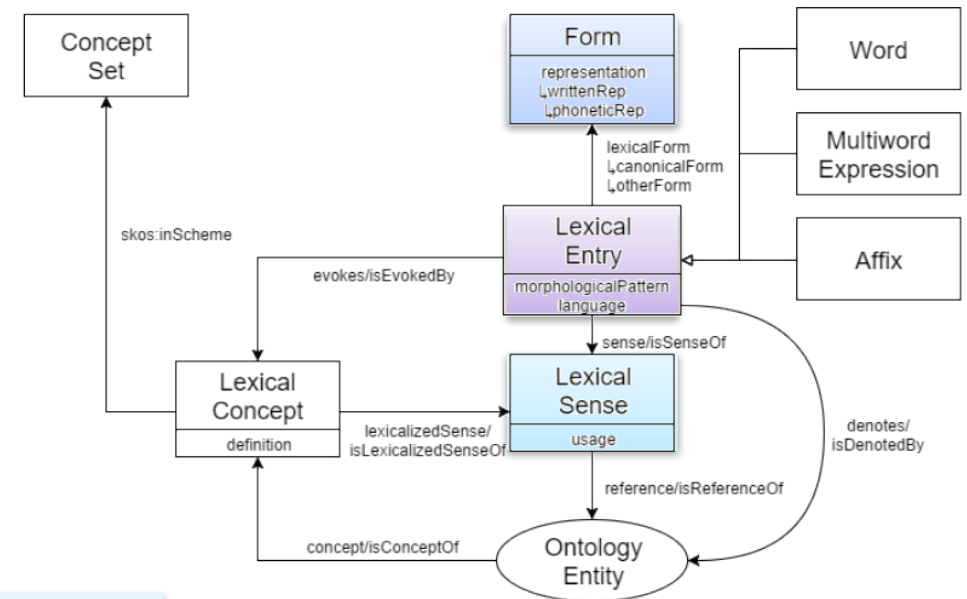
Example: German DimLex → RDF (OntoLex)



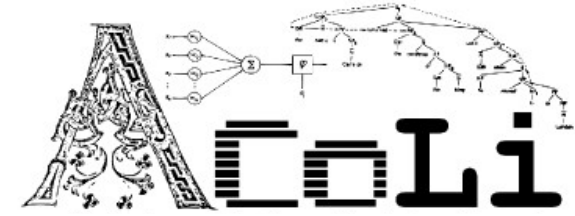
Format: DimLex-XML

```
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  <entry id="k1" word="aber">
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    <syn>
      <cat>konnadv</cat>
      <ordering>
        <ante>0</ante>
        <post>1</post>
        <insert>0</insert>
      </ordering>
      <sem>
        <pdtb3_relation sense="concession-arg2-as-denier" freq="7" anno_N="18"/>
      </sem>
    </syn>
  </entry>
  ...
</dimlex>
```

OntoLex vocabulary



Example: German DimLex → RDF (OntoLex)

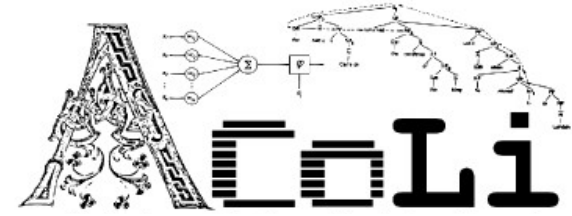


- Format: DimLex-XML OntoLex + custom properties (namespace *dimlex*:)

```
<dimlex>
  <entry id="k1" word="aber">
    <orths>
      <orth type="cont" canonical="1" onr="k1o1">
        <part type="single">aber</part>
      </orth>
    </orths>
    <non_conn_reading>
      <example type="ADV" tfreq="940">aber und abermals</example>
      <example type="ADV">Du bist aber fies!</example>
    </non_conn_reading>
    <syn>
      <cat>konnadv</cat>
      <ordering>
        <ante>0</ante>
        <post>1</post>
        <insert>0</insert>
      </ordering>
      <sem>
        <pdtb3_relation sense="concession-arg2-as-denier" freq="7" anno_N="18"/>
      </sem>
    </syn>
  </entry>
  ...
</dimlex>
```

custom properties
correspond 1:1 to
XML elements
and attributes
⇒ different dialects
represented in a
lossless fashion

Example: German DimLex RDF



XSLT

```
<dimlex>
  <entry id="k1" word="aber">
    <orths>
      <orth type="cont" canonical="1" onr="k1o1">
        <part type="single">aber</part>
      </orth>
    </orths>
    <non_conn_reading>
      <example type="ADV" tfreq="940">aber und abermals</example>
      <example type="ADV">Du bist aber fies!</example>
    </non_conn_reading>
    <syn>
      <cat>konnadv</cat>
      <ordering>
        <ante>0</ante>
        <post>1</post>
        <insert>0</insert>
      </ordering>
      <sem>
        <ptdb3_relation sense="concession-arg2-as-denier" freq="7" anno_N="18"/>
      </sem>
    </syn>
  </entry>
  ...
</dimlex>
```

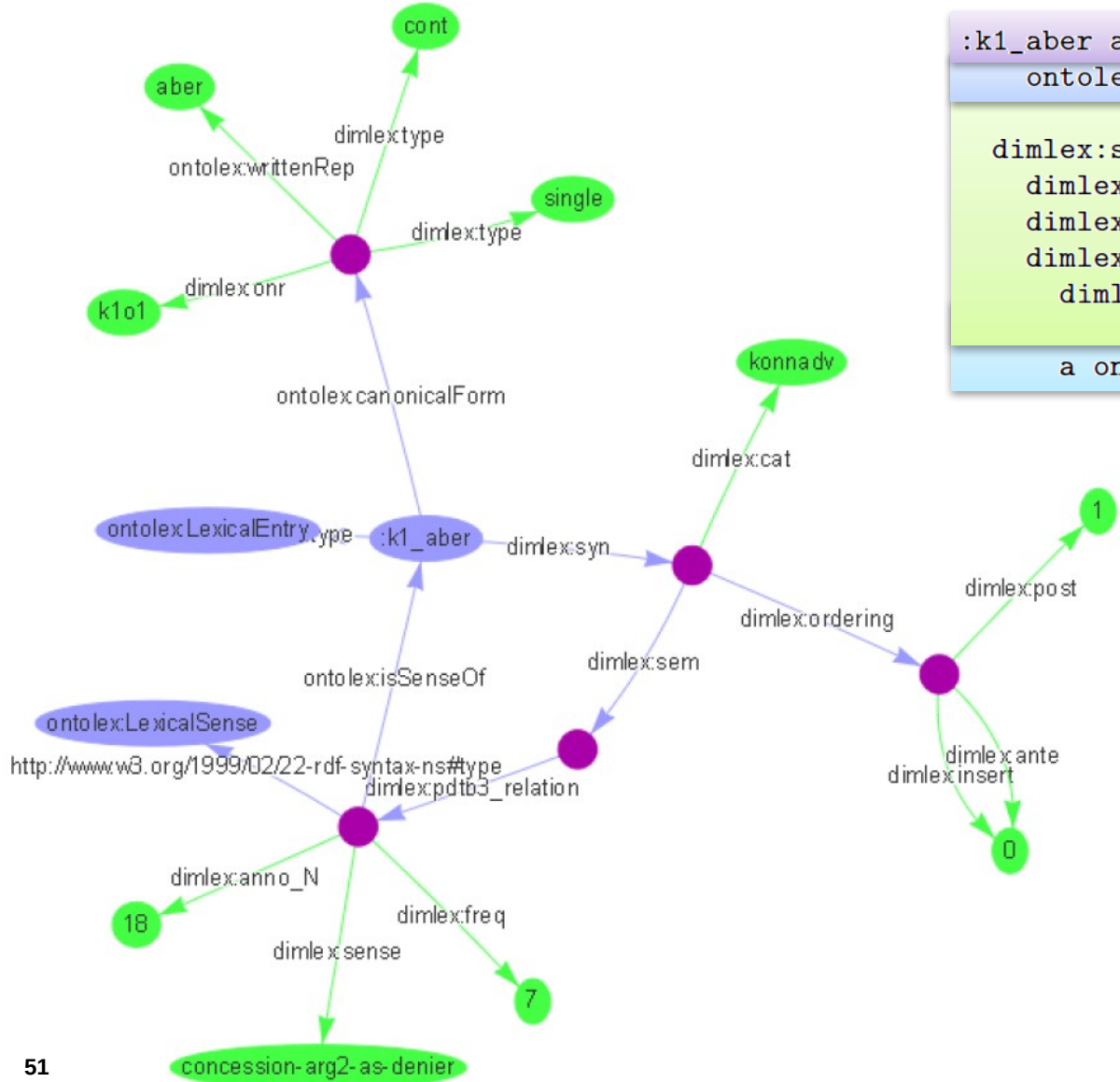
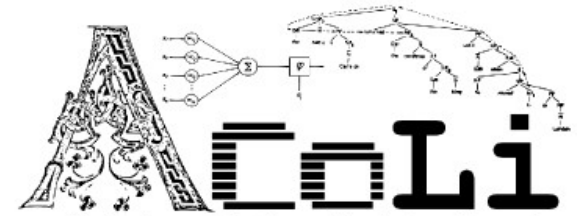
```
:k1_aber a ontolex:LexicalEntry;
  ontolex:canonicalForm [ ontolex:writtenRep "aber"@de; dimlex:type "cont";
    dimlex:onr "k1o1"; dimlex:type "single"];

dimlex:syn [
  dimlex:cat "konnadv";
  dimlex:ordering [ dimlex:ante "0"; dimlex:post "1"; dimlex:insert "0" ];
  dimlex:sem [
    dimlex:ptdb3_relation [ dimlex:sense "concession-arg2-as-denier";
      dimlex:freq "7"; dimlex:anno_N "18";
    a ontolex:LexicalSense; ontolex:isSenseOf :k1_aber ] ] .
```

Full DimLex-RDF

- OntoLex concepts
- original structure
- lossless encoding

Example: German DimLex RDF



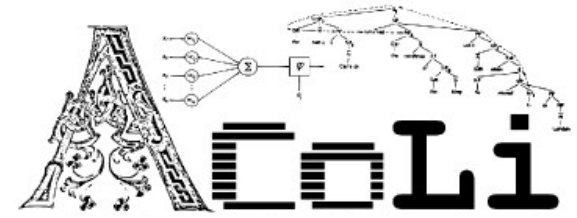
```

:k1_aber a ontolex:LexicalEntry;
  ontolex:canonicalForm [ ontolex:writtenRep "aber"@de; dimlex:type "cont";
                          dimlex:onr "k1o1"; dimlex:type "single"];

  dimlex:syn [
    dimlex:cat "konnadv";
    dimlex:ordering [ dimlex:ante "0"; dimlex:post "1"; dimlex:insert "0" ];
    dimlex:sem [
      dimlex:pdtb3_relation [ dimlex:sense "concession-arg2-as-denier";
                              dimlex:freq "7"; dimlex:anno_N "18";
                              a ontolex:LexicalSense; ontolex:isSenseOf :k1_aber ] ] .
  
```

Full DimLex-RDF

- OntoLex concepts
- original structure
- lossless encoding

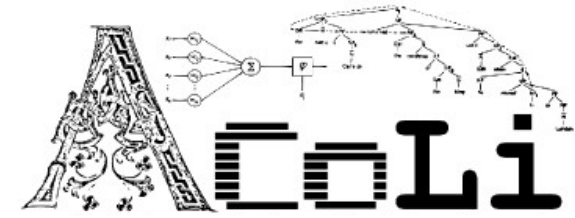


- Just use one trivial SPARQL Update

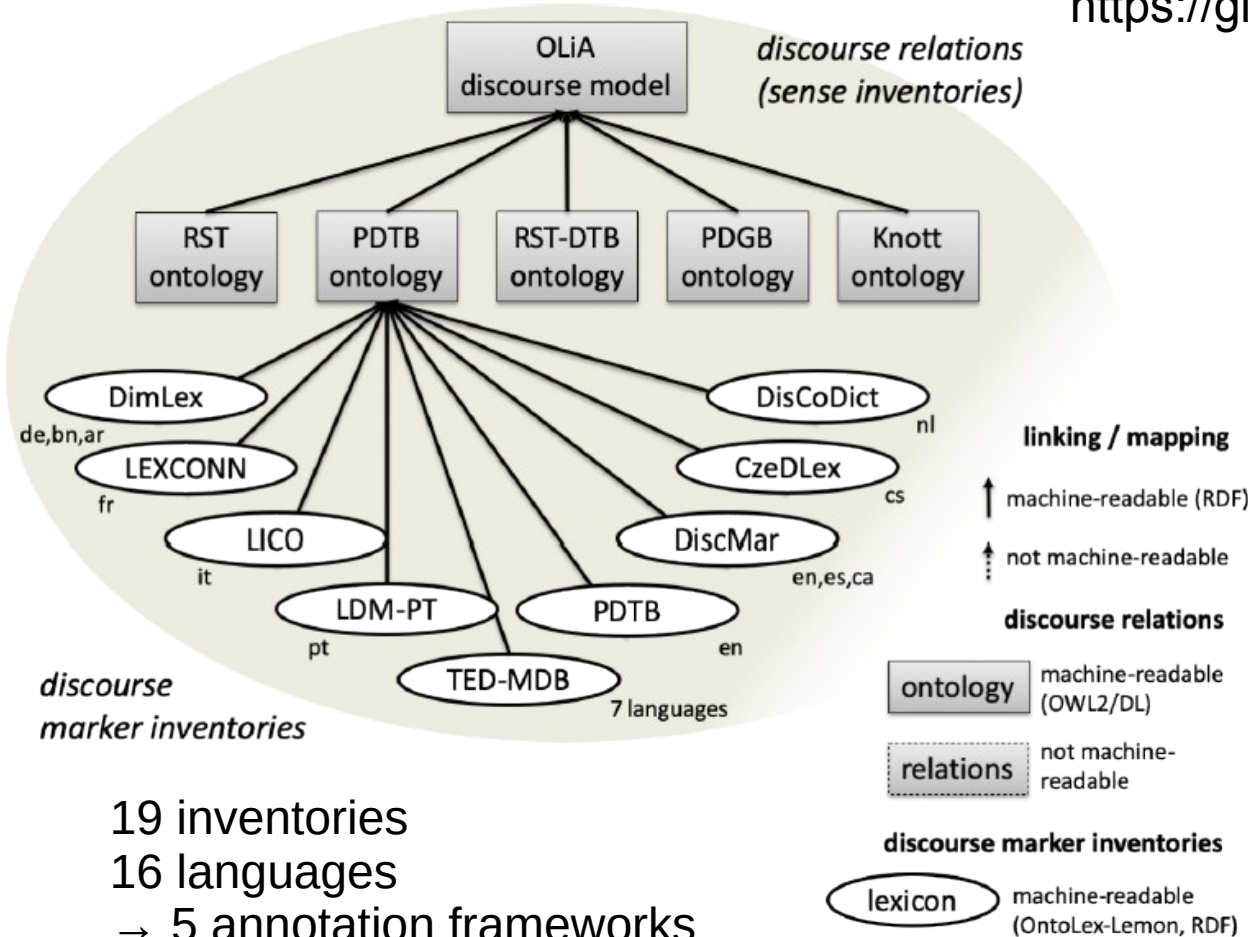
```
PREFIX dimlex: <https://github.com/discourse-lab/dimlex/blob/master/DimLex.dtd#>
LOAD <http://purl.org/olia/discourse/discourse.PDTB.owl>;
INSERT {
  ?dimlex_relation ontolex:reference ?pdtb_sense.
} WHERE {
  ?dimlex_relation dimlex:sense ?label.
  ?pdtb_sense (rdfs:label|skos:altLabel) ?sense_label.
  FILTER(!case(?label)=!case(?sense_label))
};
```

If a Dimlex relation has a *dimlex:sense* that matches the label of an OLiA PDTB relation, link them by *ontolex:reference*

Results: A Knowledge Graph for Discourse Markers



<https://github.com/acoli-repo/rdf4discourse>



discourse marker inventories

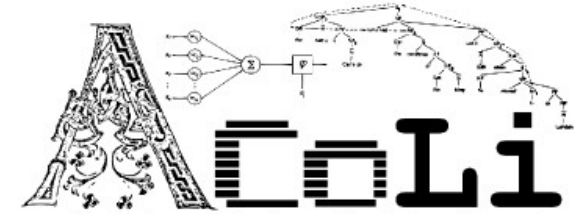
19 inventories

16 languages

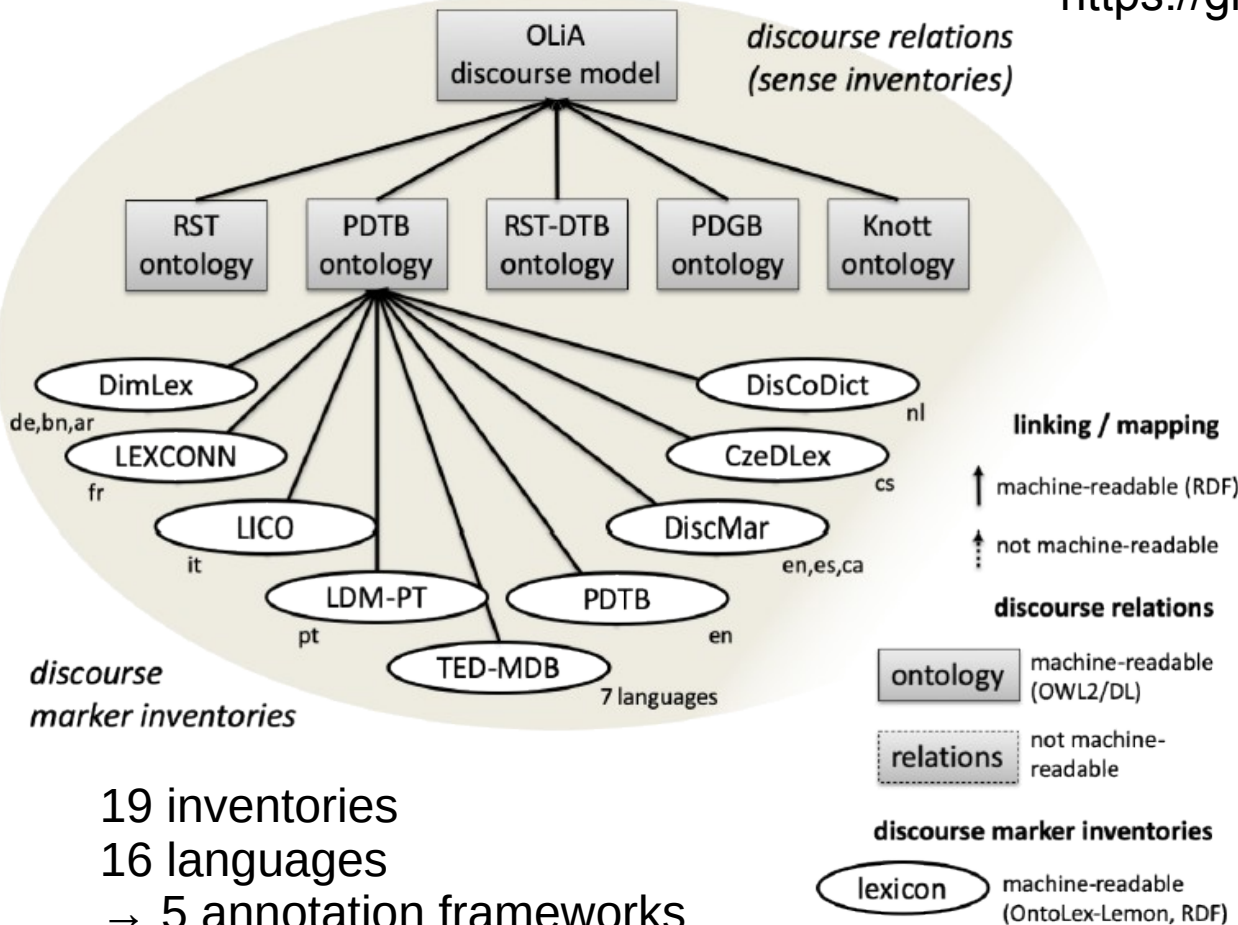
→ 5 annotation frameworks

→ 2 meta frameworks (CCR, ISO SemAF)

Results: A Knowledge Graph for Discourse Markers



<https://github.com/acoli-repo/rdf4discourse>



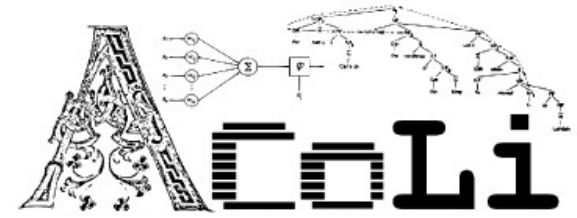
Related research

- ❑ <http://connective-lex.info/>
- ❑ outcome of TextLink
- ❑ designed for human consumption
 - no machine-readable semantics
 - based on structured XML data
- ❑ PDTB senses only
- ❑ no crosslingual integration

19 inventories
 16 languages
 → 5 annotation frameworks
 → 2 meta frameworks (CCR, ISO SemAF)

USP: We can now much more easily retrieve information from the discourse marker inventories

Querying it with SPARQL



```
SELECT distinct ?pdtb ?olia ?rst
# OntoLex and PDTB data
FROM <http://purl.org/acoli/dimlex/en/pdtb2.ttl>
FROM <http://purl.org/olia/discourse/discourse.PDTB.owl>
# OLiA Discourse Extensions
FROM <http://purl.org/olia/discourse/discourse.PDTB-link.rdf>
FROM <http://purl.org/olia/discourse/olia_discourse.owl>
FROM <http://purl.org/olia/discourse/discourse.RST-link.rdf>
FROM <http://purl.org/olia/discourse/discourse.RST.owl>
WHERE {
  ?pdtb rdfs:subClassOf*/^ontolex:reference/ontolex:isSenseOf/
    (ontolex:lexicalForm|ontolex:canonicalForm)/
    ontolex:writtenRep "because"@en.

  # the directly assigned olia senses
  ?pdtb rdfs:subClassOf ?olia.
  FILTER(contains(str(?olia),"olia_discourse"))

  # RST subsenses
  ?rst rdfs:subClassOf+ ?olia.
  FILTER(contains(str(?rst),"discourse.RST"))
} ORDER BY ?pdtb ?rst
```

(1)
load the relevant
knowledge graphs
with FROM

Not a trivial query,
but not that hard
to adapt

(2)
"because" → PDTB

(from discourse
marker inventory)

(3)
PDTB → OLiA

(from OLiA PDTB
model)

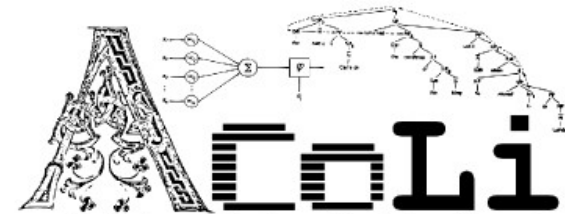
(4)
OLiA → RST

(from OLiA RST +
reference model)

Given the English PDTB 2.0 discourse marker lexicon, retrieve all possible RST relations for "because"



Querying it with SPARQL



```

SELECT distinct ?pdtb ?olia ?rst
# OntoLex and PDTB data
FROM <http://purl.org/acoli/dimlex/en/pdtb2.ttl>
FROM
# OLi
FROM
FROM
FROM
WHERE
  ?pd
# t
?pd
FIL
# R
?rs
FIL
} ORDER BY ?pdtb ?rst
    
```

(1)

Not a trivial query,
that hard
apt

pdtb	olia	rst
pdtb:Cause	olia_discourse:Cause	rst:Evidence
pdtb:Cause	olia_discourse:Cause	rst:Justify
pdtb:Cause	olia_discourse:Cause	rst:Motivation
pdtb:Cause	olia_discourse:Cause	rst:NonVolitionalCause
pdtb:Cause	olia_discourse:Cause	rst:NonVolitionalResult
pdtb:Cause	olia_discourse:Cause	rst:Purpose
pdtb:Cause	olia_discourse:Cause	rst:VolitionalCause
pdtb:Cause	olia_discourse:Cause	rst:VolitionalResult
pdtb:Condition	olia_discourse:Condition	rst:Condition
pdtb:Condition	olia_discourse:Condition	rst:Enablement
pdtb:Condition	olia_discourse:Condition	rst:Means

Given the English PDTB 2.0 discourse marker lexicon, retrieve all possible RST relations for “because”



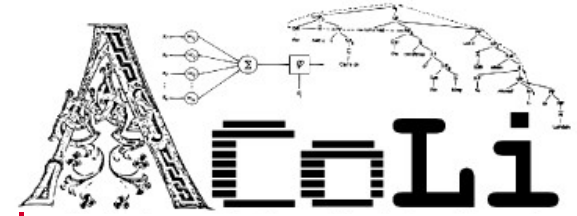
A background network graph with light blue nodes and thin grey lines connecting them, set against a light blue gradient background. The graph is dense and spans the entire width of the slide.

Inducing Discourse Marker Inventories

from machine-readable dictionaries

(Chiarcos@LREC-2022)

Lexical Induction with the ACoLi Dictionary Graph



Discourse Marker Inventories + interlinked dictionaries → induction for other languages

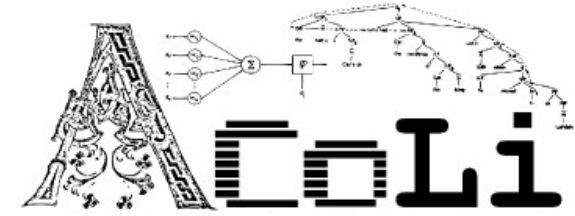
Machine-readable dictionaries

<http://github.com/acoli-repo/acoli-dicts>

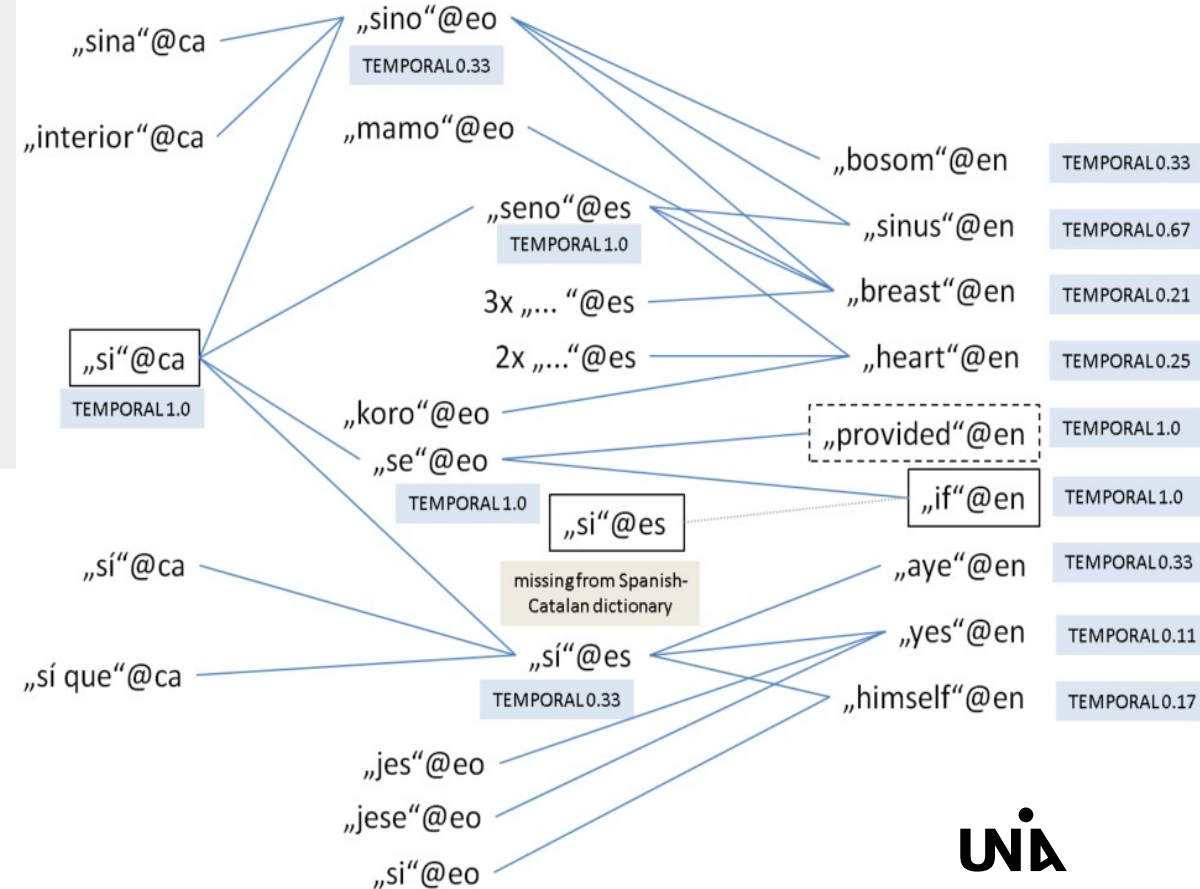
- 430+ languages, 3000+ bilingual dictionaries (Chiarcos et al. 2020)
- RDF layer over PanLex, Apertium, FreeDict, MUSE, etc.
- **Data model:** OntoLex
- **Formats:** RDF (=> TSV, with SPARQL)
- **Selected subsets**
 - **Apertium** 53 dictionaries for MT, mostly Romance
 - **FreeDict** 145 dictionaries, heterogeneous
 - **MUSE** 108 dictionaries, machine-generated

Constrained Induction

- Operate over confidence scores for discourse relations
- Initialize word w with $1/(\text{number of senses})$
- Propagate relation score to word v :
average over relation scores for translations (w. score)



- Constraints: (optionally) filter by
 - min result score
 - min pivots (translations)
 - min pivot languages (of translations)
 - max senses (top k relations, only)



Experimental Setup

<http://github.com/acoli-repo/rdf4discourse/lexical-induction>

- 11 inventories, 9 languages
- mapped to PDTB and CCR
- evaluate prec, rec, f against target inventories
- **Publish 10 induced inventories**
(Bulgarian, Greek, Esperanto, Finnish, Japanese, Norwegian, Polish, Russian, Swedish and Turkish)



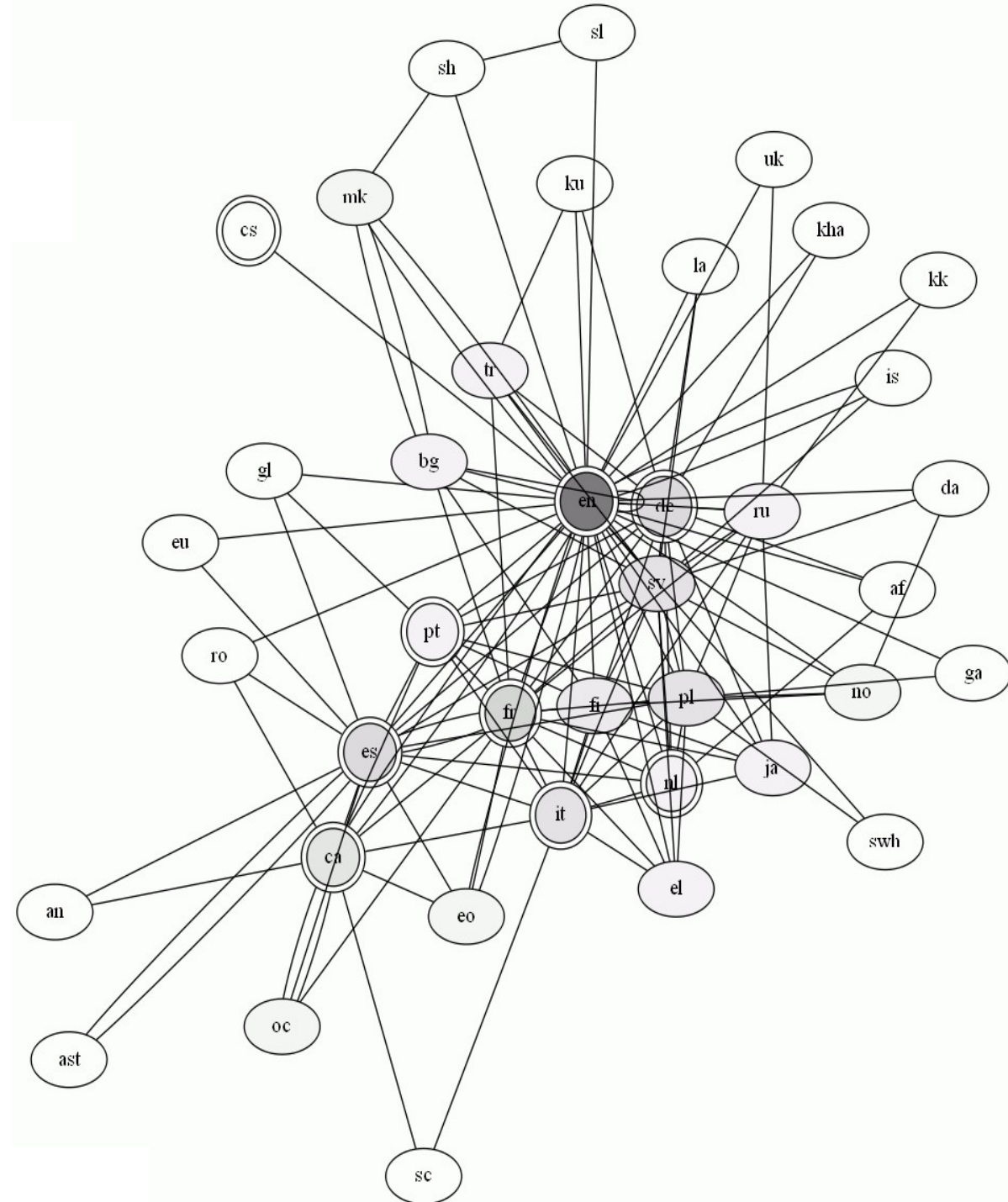
language



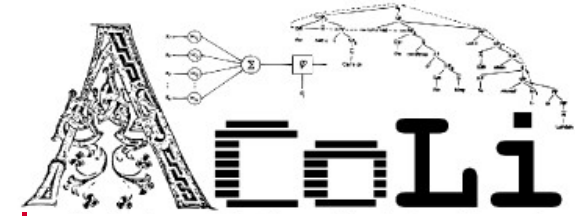
language with discourse marker inventory



≥ 1 dictionary



Lexical Induction with the ACoLi Dictionary Graph



Discourse Marker Inventories + interlinked dictionaries → induction for other languages

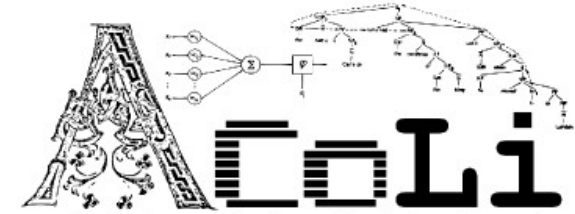
- Direct induction (e.g., from/to English) yields best results, but depends on dictionary quality (Apertium > FreeDict/MUSE)
- Constrained indirect induction is a feasible fallback-strategy

dicts	level	min score	min pivot languages	max senses	prediction	<i>p</i>	<i>r</i>	<i>f</i>
best-performing direct induction (over aggregated/all dictionaries, cs,de,fr,it,nl,pt)								
all:pt-en	2				535	0.164	0.815	0.274
all:pt-en	3				707	0.127	0.804	0.220
average scores for direct induction (cs,de,fr,it,nl,pt)								
all	2				604	0.154	0.682	0.242
all	3				645	0.106	0.403	0.164
best-performing pivot language restriction								
all	2	0.50	6	unrestricted	441	0.222	0.632	0.329
all	3	0.75	6	unrestricted	251	0.247	0.369	0.296
best-performing restriction on projected senses								
all	2	0.45	5	4	250	0.364	0.669	0.472
all	3	0.45	5	4	256	0.309	0.622	0.413

Precision is dissatisfying, but recall is reasonable
=> Baseline

Generated inventories can be a basis for **manual pruning** (note that discourse marker inventories are small, < 1000 entries)

Lexical Induction with the ACoLi Dictionary Graph



DM discourse marker	DM score	PDTB relation	relation score
"dlatego"@pl	0.96	CONTIGENCY	0.958
"dlatego"@pl	0.96	CONTIGENCY.Cause	0.949
"dlatego"@pl	0.96	CONTIGENCY.Cause.result	0.931
"dlatego"@pl	0.96	CONTIGENCY.Cause.reason	0.019
"dlatego"@pl	0.96	CONTIGENCY.Condition	0.009
"dlatego"@pl	0.96	TEMPORAL.Asynchronous.precedence	0.005
"dlatego"@pl	0.96	TEMPORAL.Asynchronous	0.005
"dlatego"@pl	0.96	TEMPORAL	0.005
"zatem"@pl	0.95	CONTIGENCY	0.699
"zatem"@pl	0.95	CONTIGENCY.Cause	0.671
"zatem"@pl	0.95	CONTIGENCY.Cause.result	0.435
"zatem"@pl	0.95	CONTIGENCY.Cause.reason	0.256
"zatem"@pl	0.95	TEMPORAL	0.199
"zatem"@pl	0.95	TEMPORAL.Asynchronous	0.180
"zatem"@pl	0.95	TEMPORAL.Asynchronous.precedence	0.176
"zatem"@pl	0.95	EXPANSION	0.048
"zatem"@pl	0.95	CONTIGENCY.Condition	0.038
"zatem"@pl	0.95	TEMPORAL.Synchronous	0.029
"zatem"@pl	0.95	EXPANSION.Alternative	0.029
"zatem"@pl	0.95	EXPANSION.Alternative.disjunctive	0.024
"gdy"@pl	0.84	TEMPORAL	0.461
"gdy"@pl	0.84	CONTIGENCY	0.361
"gdy"@pl	0.84	TEMPORAL.Synchronous	0.310

PDTB induction, Polish inventory, 4 pivot languages


538 potential discourse markers

not evaluated, but ranked according to confidence scores for being a discourse marker and for each possible relation

can be a seed for a discourse marker inventory, requires manual pruning

limitations:

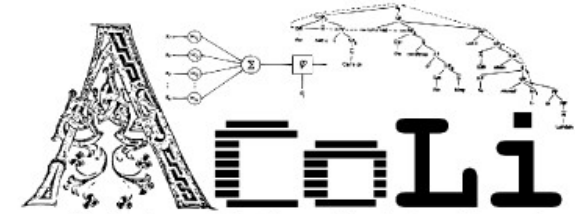
- only if in dictionary, mostly single word translations, no phrasal expressions
- some potential discourse markers might actually not be discourse markers after all

A background network diagram consisting of numerous grey dots (nodes) connected by thin grey lines (edges). The nodes are distributed across the entire frame, with a higher density in the lower right quadrant. The overall appearance is that of a complex, interconnected web or graph.

Towards a Multilingual Corpus of Discourse and Reference

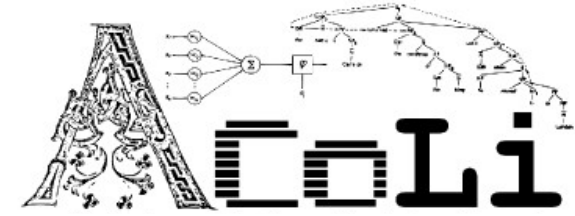
Bringing it all together

Interim Summary



- We now have
 - a number of multilingual discourse marker inventories
 - a technology to induce discourse marker inventories for hundreds of languages
 - and we can flexibly switch between theory-specific relation inventories
- This can be used to
 - create discourse marker pre-annotation for a novel languages
 - convert an RST corpus into a PDTB or ISO SemAF corpus, say, to increase the amount of training data
(if the framework-specific data structures can be transformed, as well)
- This has not been done yet, **but we have the right technology in place**
 - Take a corpus, transform it into an (RDF) graph
 - Apply SPARQL updates for enrichment and transformation
 - Serialize into target format

Interim Summary



Doing that with off-the-shelf RDF technology sounds like a performance nightmare
But we provide special tooling

The Flexible Integrated Transformation and Annotation eEngineering platform

- NLP formats \leftrightarrow RDF graphs
- one sentence (and its local context) at a time
- parallel processing
- streaming



- This has not been done yet, **but we have the right technology in place**

- Take a corpus, transform it into an (RDF) graph
- Apply SPARQL updates for enrichment and transformation
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FINTAN: Transforming heterogeneous data in a unified way



Fäth et al. @LREC-2020

- Convert *any* kind of language resource to RDF graphs.
- Manipulate/link/transform graphs with SPARQL.
- Serialize as RDF or in conventional NLP formats

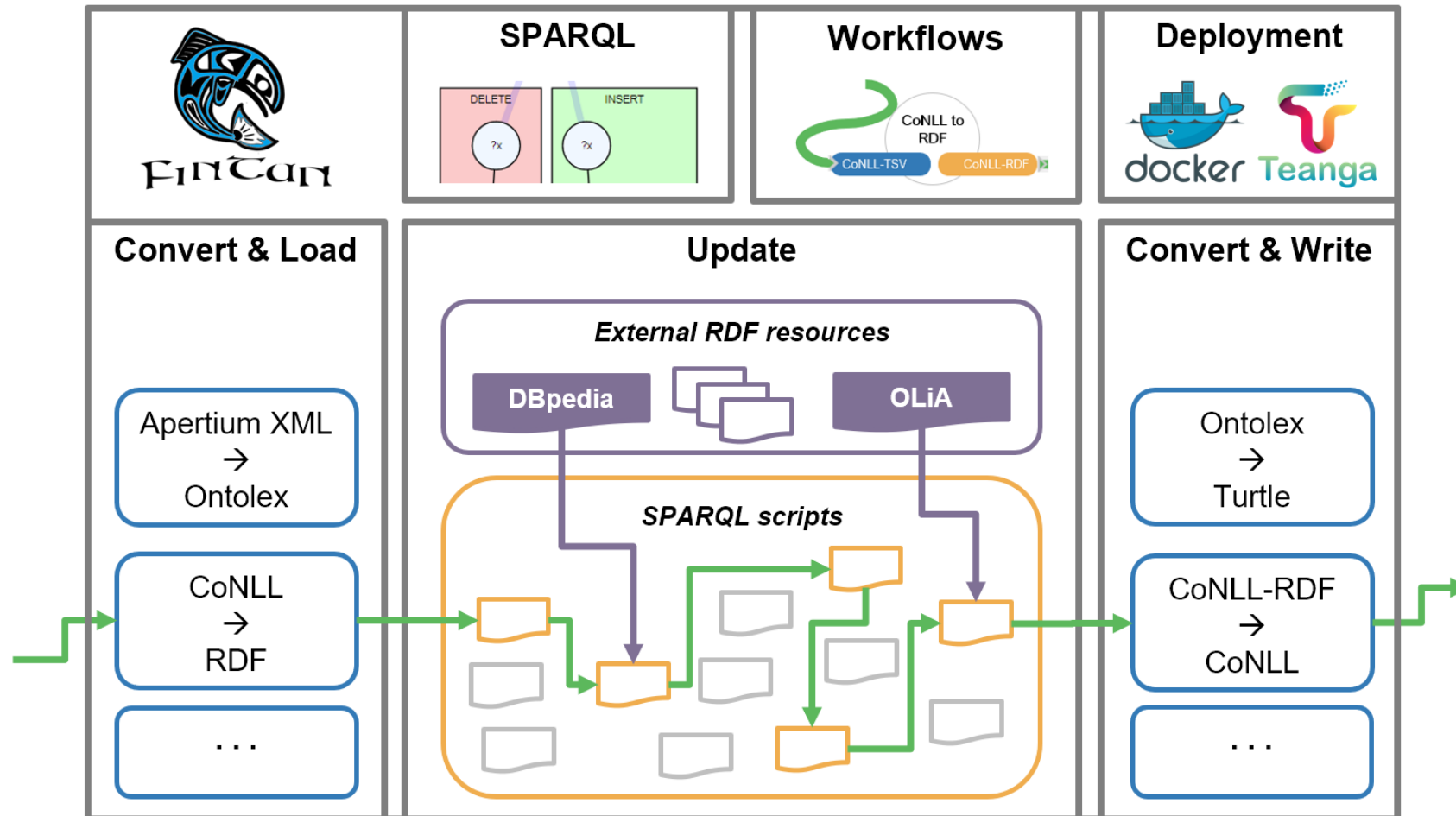
Modular: Pipelines composed of small, reusable pieces
Reusable: Same RDF vocabulary => same modules
Extensible: Add your own (SPARQL, Docker, Java, ...)
Scalable: Stream processing & parallelization

<https://github.com/Pret-a-LLOD/Fintan> (wrapper repo)

<https://github.com/acoli-repo/conll-rdf> (CoNLL customization)

FINTAN: Transforming heterogeneous data in a unified way

Fäth et al. @LREC-2020



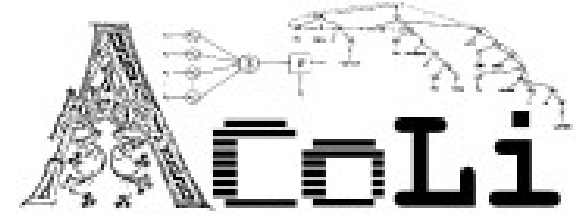
FINTAN: Transforming heterogeneous data in a unified way



Fäth et al. @LREC-2020

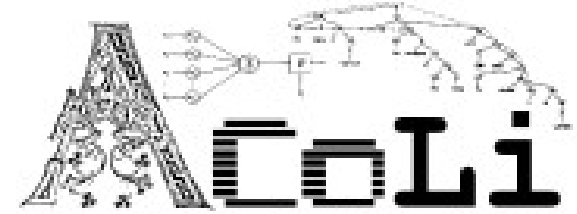
- previously, FINTAN has been used for
 - various conversion and enrichment/linking tasks
 - rule-based post-processing of automated annotation tasks
 - unified querying of heterogeneously annotated corpora
 - creating a semantically annotated treebank by transforming, decomposing and combining information from PropBank and UD
 - for Role and Reference Grammar
 - pre-annotation of the Augsburg Corpus for Reference and Information Structure
 - automated pre-annotations for discourse markers and the language-specific classification of referring expressions
 - converting existing annotations for coreference (Disco-MT) and discourse (TED-MDB) to the AURIS schema

Augsburg Corpus for Reference and Information Structure (AURIS)



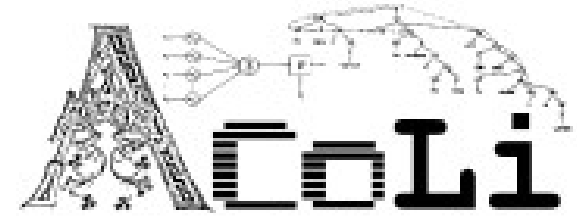
- general lack of language resources for discourse, pragmatics and semantics beyond the sentence
- build such resources together with students (department of philology and history)
 - seminars in German, English & Romance studies, translation science since 2023
- requires / benefits from
 - minimal technical entry barrier
 - offline and online editing
 - multilingual data (students must be able to work on the language of their studies)
 - existing annotations to evaluate students

Augsburg Corpus for Reference and Information Structure (AURIS)



- general lack of language resources for discourse, pragmatics and semantics beyond the sentence
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 - seminars in German, English & Romance studies, translation science since 2023
- requires / benefits from
 - minimal technical entry barrier (pre-annotation => spreadsheets)
 - offline and online editing
 - multilingual data (parallel text in 5-750 languages, literature, religious, news, TED)
 - existing annotations to evaluate students (bootstrapping AURIS annotations from TED MDB, DiscoMT, OntoNotes, FrameNet)

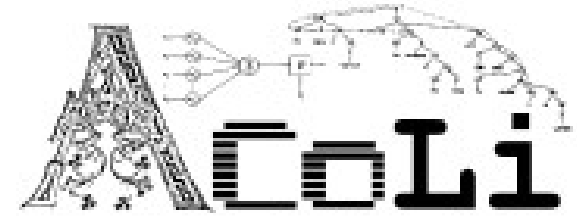
Augsburg Corpus for Reference and Information Structure (AURIS)



- general lack of language resources for discourse, pragmatics and semantics beyond the sentence
- UD parser + FINTAN => spreadsheet
- **discourse-level sheet**
 - segmented by sentence
 - pre-annotation for discourse markers
 - annotate discourse relations
 - target / external argument
 - relation
 - formulas for dynamic pre-annotation
 - sheet protection

ID	PREDICATE	TEXT	MARKER	TARGET	RELATION	COMMENT
1	grow	I grew up on a steady diet of science fiction .	???	-	-	
2	take	In high school , I took a bus to school an hour each way every day .	???	-	-	
3	absorb	And I was always absorbed in a book , science fiction book , which took my mind to other worlds , and satisfied , in a narrative form , this insatiable sense of curiosity that I had .	and?	-	-	
4	know	And you know , that curiosity also manifested itself in the fact that whenever I was n't in school	also?	-	-	
5	be	I was out in the woods , hiking and taking " samples " frogs and snakes and bugs and pond water and bringing it back , looking at it under the microscope .	???	-	-	
6	know	You know , I was a real science geek .	???	-	-	

Augsburg Corpus for Reference and Information Structure (AURIS)



- general lack of language resources for discourse, pragmatics and semantics beyond the sentence
- UD parser + FINTAN => spreadsheet
- **word-level sheet**
 - segmented by token and sentence
 - grammatical roles and syntactic embedding
 - automatically classify referring expressions
 - COREF: annotate referent ID, manually
 - REF: referentiality, predicted from COREF
 - IS: information status, -“-
 - CB: topic annotation, -“-

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A	B	C	D	E	F	G	H	L
WORD	GR	NP_TYPE	REF_AUTO	COREF	REF	IS	CB	COMMENT
# text = I grew up on a steady diet of science fiction.								
I	SBJ	pron.pper	?OLD	!!!				
grew								
up								
on								
a								
steady								
diet	other	indef-np.a						
of								
science								
fiction	other_2	indef-np.bare						
.								
# text = In high school, I took a bus to school an hour each way every day.								
In								
high								
school	other	indef-np.bare						
,								
I	SBJ	pron.pper	?OLD	!!!				
took								
a								
bus	OBJ	indef-np.a						
to								
school	other_2	indef-np.bare						
an								
hour	OBJ	indef-np.a						
each								

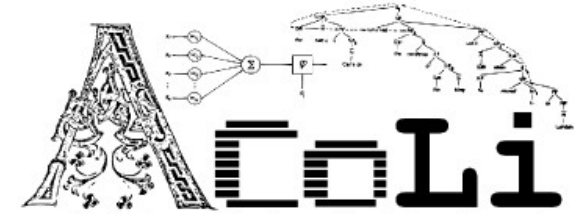
++ sentence-level annotation word-level annotation

A background network diagram consisting of numerous grey nodes connected by thin grey lines, forming a complex web-like structure. The nodes are distributed across the entire frame, with a higher density in the lower right quadrant. A large, solid red rectangle is positioned in the lower-left area, partially overlapping the network diagram.

Epilogue

wrap up ;)

Summing up



I presented a number of technologies and resources designed to support aspects of discourse processing and discourse annotation, in particular

- RDF technologies and Linked Open Data, and their application to
 - establish a level of interoperability over theory-specific inventories of discourse relations
 - access discourse marker inventories as a knowledge graph,
 - link them them with these inventories and map their,
 - link them with a lexical knowledge graph in order to induce discourse marker inventories in other languages, and
 - convert (or, pre-annotate) annotations for discourse and co-reference

Overall, the main contribution of this technology is its versatility, in discourse studies, in proving training data for NLU, or beyond

If you want to learn more, please consider to participate in our MOOCs on Linguistic Linked Data (see QR Codes)



Thank you very much!

