## From publications to knowledge graphs

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# The wider aim

A process-oriented approach to supporting research data sharing and open science

## Consider the questions:

how?

Who has worked on a particular topic and what is known about their work?

Which projects has a given method been used in?

work?

How has a particular experiment that uses a specific method been conducted?

Has a particular research question been addressed and

Which are the preferred tools for a certain kind of





## Answering these questions today:



1. Use search engines, consult specialized sources 2. Find relevant publications 3. Read them and find out about • research activities described methodology followed goals, questions, topics addressed results produced • resources and tools used, etc. 4. Find and use other relevant resources (e.g. images, tools, repositories, etc.) 5. Combine all of the above, and continue



## The Scholarly Ontology (SO) :

Captures knowledge about scholarly work so that we can answer questions of the form:

"Who does what, where, when, why and how...?





# The Scholarly Ontology

ontologies.

Ontology (NeMO).

- Framework for documenting research practice.
- Supports leveraging Linked Data.
- Provides a layered architecture that supports interconnection /compatibility with foundational

- Admits domain-specific extensions.
- Models research processes through different perspectives, covering the entire spectrum of scholarly work.
- Extends the -domain specific- NeDiMAH Methods



NeDiMAH: Network for Digital Methods in the Arts and Humanities, ESF Research Network, 2011-15





• Researching digital methods in arts & humanities • A collaborative forum of communities of practice www.nedimah.eu

### **The NeDiMAH Methods Ontology - NeMO:**

- A formal ontology for Digital Humanities, including classification and a shared vocabulary
- Incorporates existing DH taxonomies (e.g. Oxford ICT, TaDiRAH, DHCommons)
- CIDOC CRM compatible
- Contributed to ESF Report: Research Infrastructures in the Arts and Humanities



## Why use an ontólogy?

- answering.

• Provides a formalization of basic concepts. • Provides a conceptual framework for complex query

• Acts as semantic glue between different taxonomies. • Supports the development of an ecosystem of interoperable resources and services for discovering, understanding, selecting, linking and contributing content, tools and methods.





### **Ontology Development:**

Empirical research using semi-structured interviews with scholars from across Europe (earlier work).

Leverage related work: AHDS computational methods taxonomy, TaDiRAH, Scholarly Research Activity Model (Preparing DARIAH, EHRI), ARIADNE, Europeana Cloud, SPAR/CiTO, EXPO/CRM-Sci, etc.

Analysis of the ground evidence, core concepts and relationships of the domain identified. Modelling decisions.

Definitions in textual form, examples and mappings of SO terms to and from terms of other taxonomies.

Encoding in RDFS and SKOS (where needed).

Workshops: validation, collection of use cases and information needs.







Scholarly Ontology (SO):

a 3-layer structure





## Top layer:

most general concepts and properties

frame of reference

basis for linking with reference ontologies



## Middle layer:

### generic aspects of research processes

common across disciplines



### Inter-Discipline Concept / Property Hierarchies





## Bottom layer:

fine-grain aspects of research practices

discipline-specific







## A domain ontology

top layer
+ middle layer
+ discipline specific extension





SO:

## Top Elements

Actor





# Activity Perspective:

Activity: Deliberate acts that have been carried out (e.g. experiments, excavations, evaluations etc.)



## Agency Perspective:

**Goal:** Assertions representing explicit research goals

Actor: Persons Groups / Organizations







# Resource Perspective:

**Information Resource:** Concrete manifestations of conceptual objects (e.g. research article, map, image, dataset,...)



# Procedure Perspective:

Method: Prescribes how to perform a specific act (Activity) (e.g. HDR photography, macro photography)





# Grounding and Evaluation:

researchers.

Examples:

E.g. Perform stylistic analysis

"List the tools used in more than one activity employing methods which concern a particular research topic and come from either Computer Science or Linguistics." E.g. Computational Stylistic Analysis

### Based on about 100 questions gathered from different

## "Given a specific goal, retrieve all research activities that deal with it using machine learning methods."







### Procedure:

- facts and transformed to a SPARQL query. relations of SO.
- Each question was analyzed into the given and requested • Each fact was mapped to corresponding classes /
- Evaluation was based on the % coverage of query concepts by ontology concepts.

### **Results:**

- 97% coverage of the questions (errors were mainly due to unclearly formulated questions) • 82% of the questions correspond to direct link queries





### Activity

# Activity type

## Method

### Activity

Translating Herodo Histories

Creating a digital collecters from WW1

Creating an annotation of poems from WW

Photographing Lour sculptures

	Activity type	Method
otus	Translating	Semantic translation Idiomatic translation
ollection of	Collecting	Crowdsourcing
ted corpus V1	Annotating	POS tagging
Ivre	Photographing	HDR photographin Macro photography









## Activity types:

- Denote the nature of activities
- Organized as a taxonomy
- Provide semantic context for relations
- Serve as index terms for retrieval
- Function as a "gateway" through which other taxonomic structures can be imported/mapped





**Research Spotlight** 

## Approach:

- Harvest repositories and websites
- Extract metadata
- Extract information from text
- Populate SO Classes
- Publish as linked data





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![](_page_23_Picture_11.jpeg)

## Information extraction from publications

## Challenges:

Information from publication metadata needs to be exploited.

semantics.

- Named entities of non-common type (such as research methods) need to be recognized from plain text.
- Non-named entities (such as activities, goals, propositions) need to be identified and extracted from plain text.
- Extracted entities need to be interrelated according to their

All extracted information needs to be aligned in a semantic framework for comparison or integration with other existing knowledge published as linked data.

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![](_page_24_Figure_12.jpeg)

![](_page_24_Figure_13.jpeg)

![](_page_24_Figure_14.jpeg)

![](_page_24_Picture_15.jpeg)

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Stage 1

![](_page_25_Figure_2.jpeg)

Persons, Organizations, Content Items (Images, Tables, Bibliographic References)

![](_page_25_Picture_4.jpeg)

![](_page_25_Picture_5.jpeg)

Stage 2

![](_page_26_Figure_1.jpeg)

Activities, Methods, Goals, Propositions

Persons, Organizations, Content Items (Images, Tables, Bibliographic References)

![](_page_26_Figure_4.jpeg)

![](_page_26_Picture_5.jpeg)

Stage 3

![](_page_27_Figure_1.jpeg)

follows, hasPart, hasObjective, employs, resultsIn, hasParticipant, hasTopic, etc.

Activities, Methods, Goals, Propositions

Persons, Organizations, Content Items (Images, Tables, Bibliographic References)

![](_page_27_Figure_5.jpeg)

![](_page_27_Picture_6.jpeg)

![](_page_28_Figure_1.jpeg)

![](_page_28_Figure_2.jpeg)

owl:sameAs, owl:equivalentProperty, rdfs:Label, skos:altLabel

follows, hasPart, hasObjective, employs, resultsIn, hasParticipant, hasTopic, etc.

Activities, Methods, Goals, Propositions

Persons, Organizations, Content Items (Images, Tables, Bibliographic References)

![](_page_28_Figure_7.jpeg)

## The process:

![](_page_29_Figure_2.jpeg)

### **Preprocessing:**

- Use DBpedia for creating lists of NE (Methods, Topics)
- Harvest research articles and use the NE lists for distant supervision

![](_page_30_Figure_4.jpeg)

### Main processing:

- Harvest research articles for IE
- Extract Metadata
- Extract Named Entities
- Extract Non-Named Entities
- Extract Relations
- Create URIs
- Link with other Linked Data
- Access to Knowledge Base through Web Interface / SPARQL Endpoint

![](_page_31_Figure_10.jpeg)

### **Example:**

To generate our topic model, we created a 'bag of words' for each person in our dataset, comprised of all words that appear before and after the person's name in the ONDB. Specifically, for each person in the network, we located all mentions in the ONDB, and used the previous fifteen words and next 25 words -excluding named entities- as their bag of words. We then removed all named-entity mentions in these biographies and converted the remaining words into lowercase. Next we applied the Porter stemmer, in order to strip away standard English suffixes in a specific order. For example, the Porter stemmer turns the word 'publisher' into 'publish', and does same to the word 'published'. We then dropped words that are in a standard stop list - which includes words like 'and', 'the', etc. -provided in the text-mining R package tm.

![](_page_32_Figure_3.jpeg)

![](_page_32_Picture_4.jpeg)

### **Example:**

![](_page_33_Figure_1.jpeg)

![](_page_33_Picture_2.jpeg)

### **SO-driven knowledge extraction from text**

We used Random Forests in order to perform the classification experiment and then we evaluated the results . In addition, we conducted two more experiments using SVM and Logistic Regression respectively -

![](_page_34_Picture_3.jpeg)

### identify textual chunks

We used Random Forests in order to perform the classification experiment and then we evaluated the results . In addition, we conducted two more experiments using SVM and Logistic Regression respectively -

![](_page_35_Picture_2.jpeg)

![](_page_35_Picture_3.jpeg)

### extract entities

![](_page_36_Figure_1.jpeg)

Activity

![](_page_36_Figure_4.jpeg)

![](_page_36_Picture_5.jpeg)

### extract relations

![](_page_37_Figure_1.jpeg)

![](_page_37_Picture_5.jpeg)

### add actor and resource metadata

![](_page_38_Figure_1.jpeg)

![](_page_38_Picture_5.jpeg)

### assign activity types

![](_page_39_Figure_1.jpeg)

![](_page_39_Picture_5.jpeg)

### link with existing knowledge

![](_page_40_Figure_1.jpeg)

![](_page_41_Figure_1.jpeg)

conducted two more experiments using SVM and Logistic Regression respectively

### semantic paths through common Actors

![](_page_42_Figure_1.jpeg)

## Evaluation:

## Rule-based extraction

**Source:** 50 research articles from Digital Humanities, Geology, Medicine, Bioinformatics, Biology, Computer Science, Sociology and Anthropology.

### **Reference standard:**

- agreement: 81% kappa)

### **Dataset:**

- 400 employs()

### **Methodology:**

- micro- & macro-averaging

Annotations by two human annotators (inter-annotator)

• Duration for manual annotation: 3.5 - 4 hrs per article!

 1700 Activities, 300 Goals, 700 Propositions, 1000 follows(), 100 hasPart(), 250 hasObjective(), 200 resultsIn(),

token-based, entity-based evaluation (threshold: 86%)

![](_page_43_Picture_16.jpeg)

## Evaluation

### Rule-based extraction

Results:

### **Entity Evaluation:**

	Entity- based	Token- based
Entity Type	F1	F1
Activity	0,72	0,81
Goal	0,76	0,80
Proposition	0,79	0,82
Method	0,91	0,85

**Error sources:** 

- Human errors (author / editor misspellings) • External modules • Rules and constraints

### **Relation Evaluation:**

![](_page_44_Picture_11.jpeg)

### Using machine learning techniques:

- Algorithms used: Logistic regression, SVM, Random forests
- Combinations of handcrafted features and word embeddings
- Pipeline proposed splitting sentence from token classification

### Extracting Activity and follows(Activity, Activity)

### Activity

![](_page_45_Figure_6.jpeg)

### follows(Activity, Activity)

![](_page_45_Picture_8.jpeg)

res

Test sets					
DH	BIOINF	MED	ALL		
0,82	0,88	0,92	0,88		
0,73	0,72	0,71	0,72		

### F1 scores

Test se		
BIOINF	MED	ALL
0,86	0,92	0,89

![](_page_45_Picture_13.jpeg)

![](_page_45_Picture_14.jpeg)

![](_page_45_Picture_15.jpeg)

## Use cases:

- research
- planning
- methods
- scholarship

• Find information on earlier work relevant to one's own

• Goal-oriented organization of research work and project

• Discovery of connections between resources, tools and

• Gathering evidence of the use of digital resources for

• Critical evaluation of digital humanities

![](_page_46_Picture_11.jpeg)

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![](_page_47_Figure_9.jpeg)

![](_page_47_Figure_10.jpeg)

![](_page_47_Picture_11.jpeg)

## Thank you!

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