

Knowledge Graphs

Open Data

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1 Introduction

The project is publicly available at Github.

2 TBOX definition

In order to create our ontology we decided to use *OWL* since our ontology required restrictions which cannot be expressed with *RDFS*.

To model our ontology, we represented the problem statement as a UML diagram (see fig. 2) and then we translated this UML to OWL.

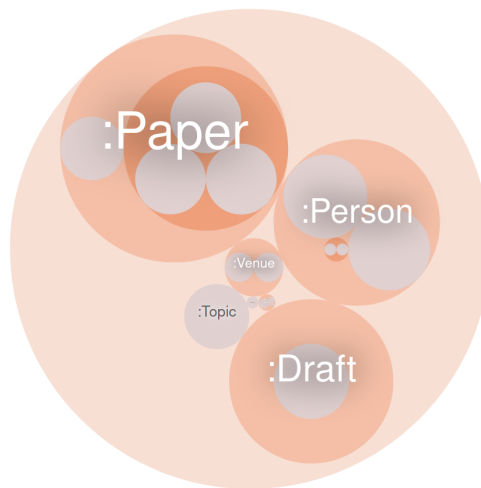


Figure 1: TBOX Graph Representation

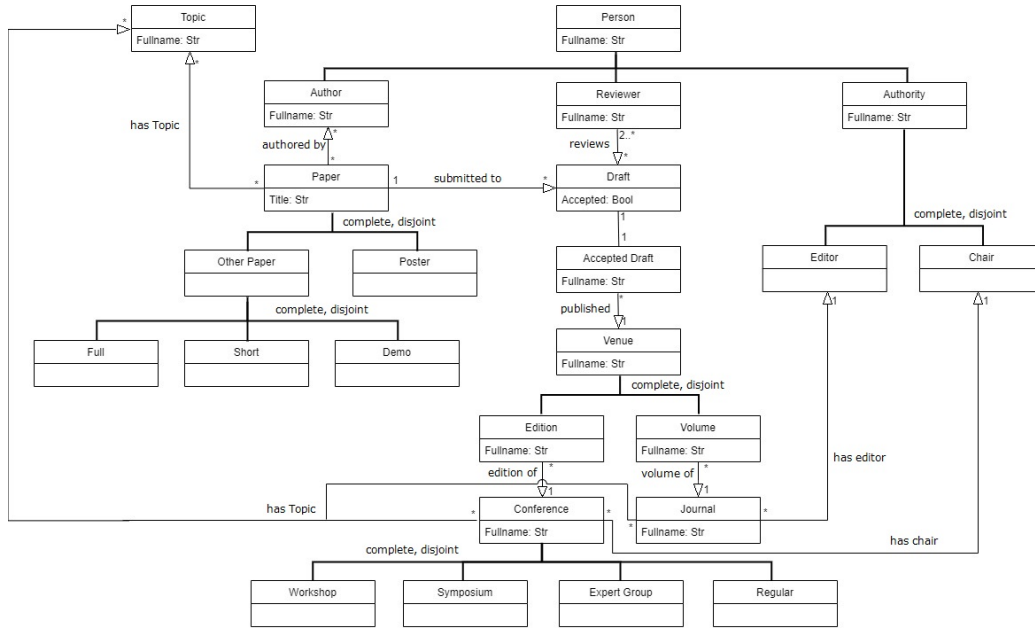


Figure 2: UML Class Diagram

For this project, we have used the programming language *Python* and Python's library *Owlready2* [1]. Owlready2 is an eDSL for Python which translates Python's syntax to OWL's definitions using the following rules:

- `owl:Class` and `rdf:Property` is defined using a python's class definition.
- `rdfs:subClassOf` and `rdfs:subPropertyOf` is defined using Python's inheritance.
- `owl:ObjectProperty`, `owl:DataProperty` and `owl:FunctionalProperty` are defined using class inheritance from Python's classes `ObjectProperty`, `DataProperty` and `FunctionalProperty`, respectively.
- `rdfs:domain` and `rdfs:range` is defined using a class attribute.
- `owl:Restriction` is also defined using a class attribute.

As an example of our ontology, the class *AcceptedDraft* is defined as follows

```
<owl:Class rdf:about="#AcceptedDraft">
  <rdfs:subClassOf rdf:resource="#Draft"/>
  <rdfs:subClassOf>
    <owl:Restriction>
      <owl:onProperty rdf:resource="#published"/>
      <owl:onClass rdf:resource="#Venue"/>
      <owl:qualifiedCardinality
        ↪ rdf:datatype="http://www.w3.org/2001/XMLSchema#nonNe_
        ↪ gativeInteger">1</owl:qualifiedCardinality>
    </owl:Restriction>
  </rdfs:subClassOf>
  <owl:equivalentClass>
    <owl:Class>
      <owl:intersectionOf rdf:parseType="Collection">
        <rdf:Description rdf:about="#Draft"/>
        <owl:Restriction>
          <owl:onProperty rdf:resource="#accepted"/>
          <owl:hasValue rdf:datatype="http://www.w3.org/2001/X_
            ↪ MLSchema#boolean">true</owl:hasValue>
        </owl:Restriction>
      </owl:intersectionOf>
    </owl:Class>
  </owl:equivalentClass>
</owl:Class>
```

Listing 1: AcceptedDraft Class definition

The code to generate the ontology and the ontology itself can be found at OD-E_B1_AbellaGrifols.py and OD-E_B1_AbellaGrifols.rdf, respectively.

3 ABOX definition

For the ontology’s ABOX, we have used data extracted from DBLP [2]. The missing classes such as paper and workshop subtypes were assigned following a normal distribution from a synthetic dataset.

For this part, we have also used Owlready2. Owlready2 has a unique approach to ABOX generation. Once an ontology is loaded into Python’s runtime, all elements from the ontology are available as Python’s constructs such

as classes or class attributes. The user only has to instantiate those classes and attributes in order to generate the corresponding ABOx. For example, a `publications:Draft` is instantiated as follows:

```

1 draft = Draft("draft_1")
2 draft.isReviewedBy = [reviewer1, reviewer2]
3 draft.supervisedBy = [chair]
4 draft.accepted = True
5 draft.published = [volume]
```

Listing 2: ABOX programatically definition

The code and ontology instances for this section can be found at `OD-E_B2_AbellaGrifols.py` and `OD-E_B2_AbellaGrifols.rdf`, respectively.

4 Create the final ontology

An OWL DL reasoner was capable of inferring the following knowledge from our ontology:

- A Person is inferred to be an Author/Reviewer/Editor/Chair depending on the relationship domain/range.
- A Venue is inferred to be a Conference/Journal depending on the relationship domain/range.
- A Draft is inferred to be AcceptedDraft depending on the Datatype property 'accepted'.
- A Paper is inferred to be a Poster if it is presented in a Conference.
- Inverse properties such as AuthorOf/AuthoredBy are inferred.

As an extra, Owlready2 is equipped with a powerful reasoner engine with allowed us to check the correctness of our ontology and to verify if our instances were following the restrictions of our ontology. Owlready2 also provides an SPARQL engine, which combined with the provided reasoner, the user can test if the ontology is sound and the inference is working as expected.

Number of Classes	115
Number of Properties	90

Table 1: Ontology statistics

We provide a summary table of the resulting knowledge graph (see tables 1, 2, 3)

Paper	OtherPaper	Full	4335
		Demo	4198
		Short	4292
		Poster	1815
			14640
Person	Authority	Chair	3
		Editor	3
	Author		20317
		Poster	15846
Venue	Edition		30
	Volume		30
	Total Venue		60
Draft	AcceptedDraft		7283
			14640
Conference	Regular Conference		2
	Expert Group		1
	Symposium		0
	Workshop		0
			3
Journal			3
Topic			2139

Table 2: Classes breakdown

authorOf	38765
chairOf	5
editorOf	4
draftOf	14640
responsibleFor	21960
editionOf	30
volumeOf	30
hasTopic	36629
published	7283
reviews	24808

Table 3: Properties breakdown

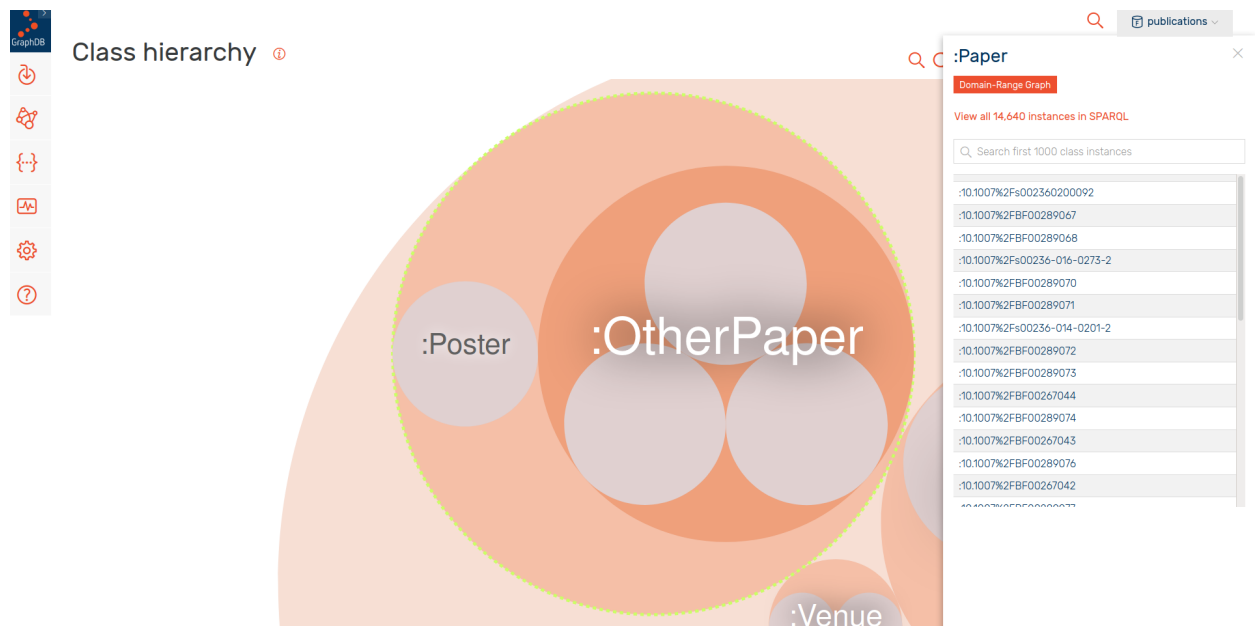


Figure 3: GraphDB instances

5 Querying the ontology

5.1 Find all Authors

PREFIX : <http://localhost:7200/publications#>

```
select ?name where {
  ?author a :Author .
  ?author :fullname ?name
} limit 100
```

5.2 Find all properties whose domain is Author

PREFIX : <http://localhost:7200/publications#>

PREFIX rdf: <http://www.w3.org/2000/01/rdf-schema#>

```
select ?p where {
  ?p rdf:domain :Author .
} limit 100
```

5.3 Find all properties whose domain is either Conference or Journal

```
PREFIX : <http://localhost:7200/publications#>
```

```
PREFIX rdf: <http://www.w3.org/2000/01/rdf-schema#>
```

```
select distinct ?p where {  
  {?p rdf:domain :Conference}  
  UNION  
  {?p rdf:domain :Journal}  
} limit 100
```

5.4 Find all the papers written by a given author that where published in database conferences

```
PREFIX : <http://localhost:7200/publications#>
```

```
PREFIX rdf: <http://www.w3.org/2000/01/rdf-schema#>
```

```
select ?author (group_concat(?title) as ?titles) where {  
  ?author :authorOf ?paper .  
  ?paper :title ?title .  
  ?paper :submittedTo ?draft .  
  ?draft a :AcceptedDraft .  
  ?draft :published ?venue .  
  ?venue a :Edition .  
  ?venue :hasTopic ?topic .  
  ?topic :fullname ?label .  
  filter contains(?label, "big data")  
}  
group by ?author  
limit 100
```

Notice, we changed “*database*” for “*big data*” because our dataset does not include “*database*” as a topic.

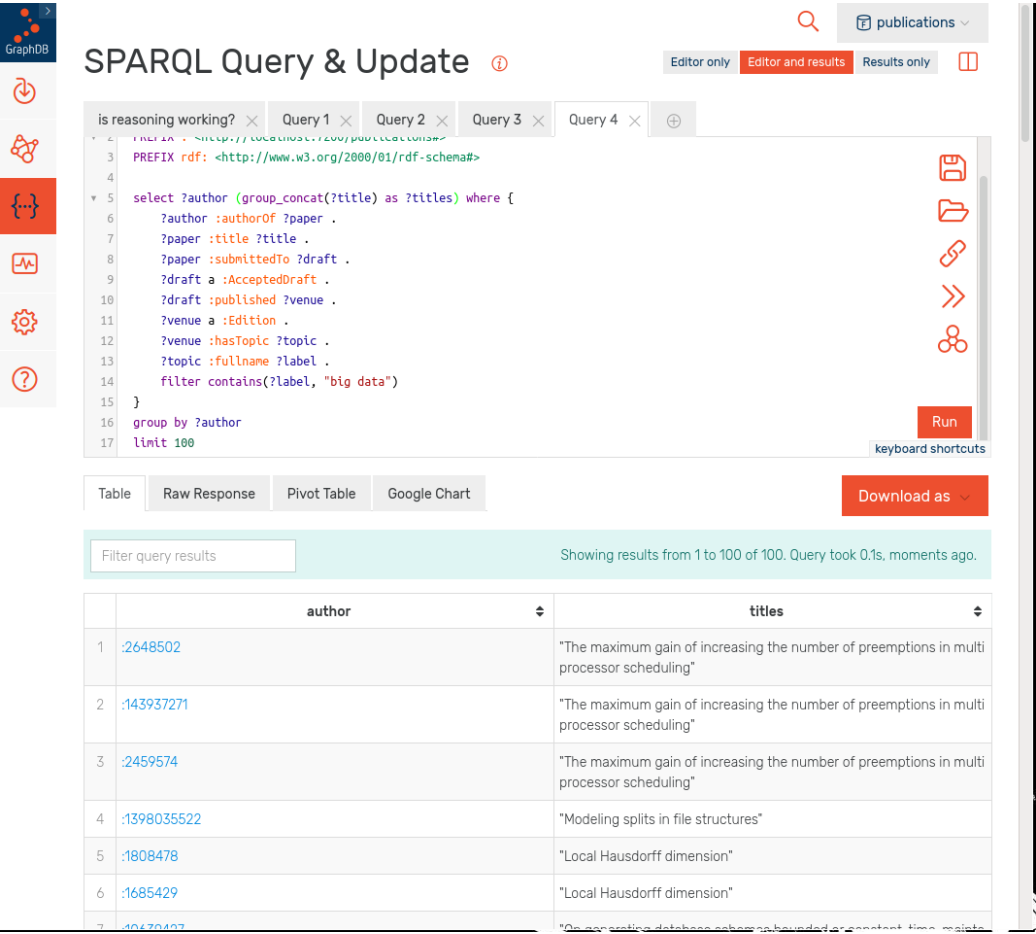


Figure 4: SPARQL: Query 4

References

[1] Lamy JB. *Owlready: Ontology-oriented programming in Python with automatic classification and high level constructs for biomedical ontologies. Artificial Intelligence In Medicine* 2017;80:11-28

[2] The dblp team: dblp computer science bibliography. Monthly snapshot release of March 2021. <https://dblp.org/xml/release/dblp-2021-03-01.xml.gz>