











# Optimizing the Computation of a Possibilistic Heuristic to Test OWL SubClassOf Axioms Against RDF Data Evolutionary Axiom Discovery from Knowledge Graphs

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Introduction

- A research area focused on Ontology Enrichment of the Semantic Web.
- An evolutionary approach for OWL axiom extraction based or grammatical evolution combined with possibility theory.
- We use W3C standards such as RDF, OWL and SPARQL to extract meaningful axioms.













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#### A set of W3C standards for linked data on the Web :

- RDF to represent knowledge graphs as a set of triples dbr:Queen\_(band) dbo:genre dbr:Rock\_music .
- **OWL** to represent an **ontology**:  $\mathcal{O} = \langle \mathcal{C}, \mathcal{R}, \mathcal{T}, A \rangle$
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Preliminaries 🌼 An example of a SPARQL Query

- "I would like to know the number of tracks on each of Queen's albums."
- Querying DBpedia SPARQL Endpoint: https://dbpedia.org/sparql

```
SELECT distinct ?album

(count(distinct ?title) as ?n_title) WHERE {
    ?album a dbo:Album .
    ?album dbp:artist dbr:Queen_(band) .
    ?album dbp:title ?title .
} GROUP BY ?album
```













Preliminaries > A look on our approach

#### ■ We focus on 2 research questions:

- How to automatically discover new OWL axioms over a knowledge graph ?
- How to assess OWL axioms against a given RDF graph ?
- We focus on the assessment of OWL Subsumption axioms
  - SubClassOf(<C> <D>) where C and D are concepts from a given ontology.
  - $C \sqsubseteq D$  highlight the perfect inclusion of the instances of C within the instances of D, i.e. the instances of C are also instances of D
  - verify the compatibility of a given subsumption axiom with the known facts: a possibilistic heuristic to test it!













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Background > Possibility theory

## Possibility theory

Possibility theory is a mathematical theory of epistemic uncertainty which considers events denoted  $\omega$  of a universe of discourse  $\Omega$  ( $\omega \in \Omega$ ) where each  $\omega$  has a degree of possibility  $\pi(\omega) \in [0,1]$ . The theory includes a measure of possibility denoted by  $\Pi$  and a measure of necessity denoted by N such that:

$$\Pi(A) = \max_{\omega \in A} \pi(\omega), \ N(A) = 1 - \Pi(\bar{A}) = \min_{\omega \in \bar{A}} \{1 - \pi(\omega)\},$$

where  $A \sqsubseteq \Omega$  or  $A = \{\omega : \omega \models \phi\}$ .











Related work > Possibility and Necessity

## Possibility theory applied on subsumption axioms evaluation <sup>1</sup>

Let us consider  $v_{\phi}^{+}$  the *confirmations* and  $v_{\phi}^{-}$  the *exceptions* observed among the elements of  $v_{\phi}$ , the **support** for an axiom  $\phi$ . We define the **possibility**  $\Pi(\phi)$  and **necessity**  $N(\phi)$  of an axiom as follows:

$$\Pi(\phi) = 1 - \sqrt{1 - \left(rac{\upsilon_\phi - \upsilon_\phi^-}{\upsilon_\phi}
ight)^2}, N(\phi) = \left\{egin{array}{c} \sqrt{1 - \left(rac{\upsilon_\phi - \upsilon_\phi^+}{\upsilon_\phi}
ight)^2}, & \textit{if} \, \Pi(\phi) = 1 \ 0 & \textit{otherwise} \end{array}
ight.$$

Tettamanzi, A., Faron Zucker, C., and Gandon, F. - International Journal of Approximate Reasoning (2017).













Related work > ARI

## ARI (Acceptance/Rejection Index) <sup>2</sup>

In order to decide about a given axiom  $\phi$ , we define an ARI value using  $\Pi(\phi)$  and  $N(\phi)$ :

$$ARI(\phi) = N(\phi) + \Pi(\phi) - 1 \in [-1,1].$$

<sup>&</sup>lt;sup>2</sup>Tettamanzi, A., Faron Zucker, C., and Gandon, F. - International Journal of Approximate Reasoning (2017).











Related work > SPARQL Queries

SPARQL queries are built to extract the **number of instances** implied by a given axiom  $(v_{\phi})$ , the **number of confirmations**  $(v_{\phi}^+)$  and the **number of exceptions**  $(v_{\phi}^-)$ .

Let's consider the given axiom: SubClassOf(<C> <D>)

## <code>SPARQL</code> Query to extract $v_{\epsilon}$











Related work > SPARQL Queries

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Let's consider the given axiom: SubClassOf(<C> <D>)

## SPARQL Query to extract $v_{\phi}$

```
SELECT (count(distinct ?x) as ?u_phi) WHERE {
     ?x a <C> .
}
```











Related work > SPARQL Queries

```
SPARQL Query to extract v_{\phi}^{+}
```

```
SELECT (count(distinct ?x) as ?conf) WHERE {
    ?x a <C> , <D> .
}
```

## SPARQL Query to extract $v_{\phi}^-$ : a first heuristic

```
SELECT (count(distinct ?x) AS ?exc) WHERE {
    ?x a <C> , ?T .
    FILTER NOT EXISTS {
         ?y a ?T , <D> .
    }
}
```













Related work > Time-capping the computation of exceptions

- lacksquare a second heuristic: adding a **time cap** for the computation of  $v_{\phi}^-$ :  $^3$ 
  - if the limit is reached, a quick but approximate computation of exceptions is adopted.

<sup>&</sup>lt;sup>3</sup>Tettamanzi, A., Faron Zucker, C., and Gandon, F. - K-CAP (2015)



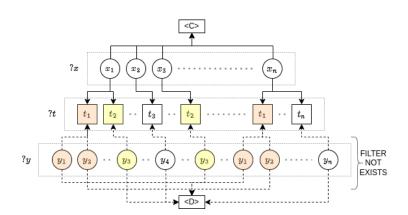








Contributions > Computational problem















Contributions > An idea of the possible impact

dbo concept	# instances
dbo:Agent	1,472,369
dbo:Person	1,124,388
dbo:Place	754,415
dbo:CareerStation	577,196
dbo:PopulatedPlace	531,228
dbo:Settlement	466,791
dbo:Work	409,594
dbo:Organisation	329,500
dbo:Athlete	313,730







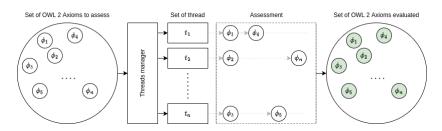






Contributions ▷ A. Multi-threading

## A **multi-threading system** for the assessment of OWL axioms has been implemented.









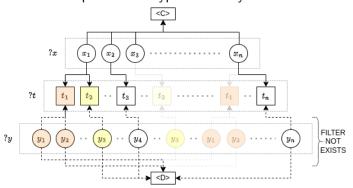






Contributions > B. Avoiding redundant computation of types

■ Aim: avoid the computation of types t already involved.















Contributions ▷ B. Avoiding redundant computation of types

Recovering distinct types (i.e., classes) being evaluated as potentially containing exceptions.

## Step 1: Recovering distinct types













Contributions ▷ B. Avoiding redundant computation of types

Retrieving the instances that belong to <C> and at least one of the types <t> just retrieved, which will count as exceptions.

### Step 2: Retrieving the instances

```
SELECT distinct ?x WHERE {
    ?x a <C> .
    ?x a ?t values (?t) {
        (<t1>) (<t2>) ... (<tn>)
    }
} LIMIT $limit OFFSET $offset
```











#### Contributions ▷ C. Extending SPARQL 1.1 Federated Query

- Adding a new operator as a URI parameter in federated query services.<sup>4</sup>
- Allows you to automatically iterate a service call.
- Implemented in the Corese semantic web factory.

<sup>&</sup>lt;sup>4</sup>Corby, O., Faron, C., Gandon, F., Graux, D., and Michel, F - WEBIST (2021) □ → 《 □ → 《 ≧ → 《 ≧ → ② ○ ○













Contributions ▷ C. Extending SPARQL 1.1 Federated Query

■ Sub-processing of the pagination: more efficient and less code writing.













#### Experiments

- 1 Benchmarking our approach with previous results.<sup>5</sup>
  - a set of 722 subsumption axioms to assess.
  - checking the equality of the ARIs between both approaches.
  - comparison of the computation times (CPU) for each axiom.
- 2 Assessing the usage of the loop operator.
  - what benefit do we get from using this operator ?
  - comparison of the computation times (CPU) for each axiom assessed in Experiment 1.

Tettamanzi, A., Faron Zucker, C., and Gandon, F. - K-CAP (2015)





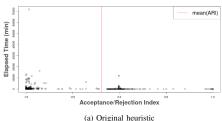


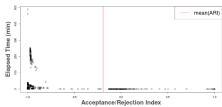




Results > Comparison of ARIs

- The integrity of the ARI values is respected
  - $mean(ARI) \simeq -0.1936$  in both settings.
- A significant reduction of the elapsed time
  - mean time spent to assess axioms ...
    - with the original heuristic: 577.9 minutes
    - with contributions A+B: **30.14** minutes
  - maximum computation time goes down from 71,699 to 489 minutes.





(b) Results obtained with contributions A+B







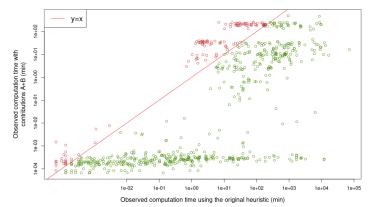






#### Results > Comparison of computation times (CPU)

- 593 axioms (82%, in green) are faster to assess with contrib. A+B
  - lacktriangle average computation time saved:  $\sim$  679 minutes.
- 129 axioms (18%, in red) are slower to assess with contrib. A+B
  - average computation time increase: 57 minutes.
  - maximum loss: 244 minutes.









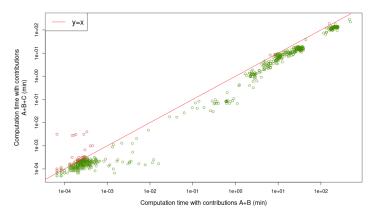






#### Results > Comparison of computation times (CPU)

- lacksquare a reduction of  $\sim 12$  minutes of the average computation time.
- 683 axioms (95%, in green) are faster to assess with contrib. A+B+C.
- 39 axioms (5%, in red) are slower to assess with contrib. A+B+C.















- A significant reduction of computation time (CPU) without any side effects.
- Makes it possible to lead huge experiments on OWL subsumption axioms mining in a reasonable time, without a time cap.
- Opens up new perspectives for the processing of the largest knowledge graphs in the Linked Open Data.













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## Thank you!