

Optimizing the Computation of a Possibilistic Heuristic to Test OWL SubClassOf Axioms Against RDF Data

Evolutionary Axiom Discovery from Knowledge Graphs

Rémi FELIN; Olivier CORBY; Catherine FARON; Andrea G. B.
TETTAMANZI

Université Côte d'Azur, Inria, CNRS, I3S
WIMMICS Team

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

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

- **RDF** to represent knowledge graphs as a set of **triples**:
`dbr:Queen_(band) dbo:genre dbr:Rock_music .`
- **OWL** to represent an **ontology**:
 $\mathcal{O} = \langle \mathcal{C}, \mathcal{R}, \mathcal{I}, \mathcal{A} \rangle$,
- **SPARQL** to query RDF data.

- 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.
 - $\text{SubClassOf}(\langle C \rangle \langle D \rangle)$ 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!

Possibility theory

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

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

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

Possibility theory applied on subsumption axioms evaluation ¹

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

$$\Pi(\phi) = 1 - \sqrt{1 - \left(\frac{v_\phi - v_\phi^-}{v_\phi}\right)^2}, N(\phi) = \begin{cases} \sqrt{1 - \left(\frac{v_\phi - v_\phi^+}{v_\phi}\right)^2}, & \text{if } \Pi(\phi) = 1 \\ 0 & \text{otherwise} \end{cases}$$

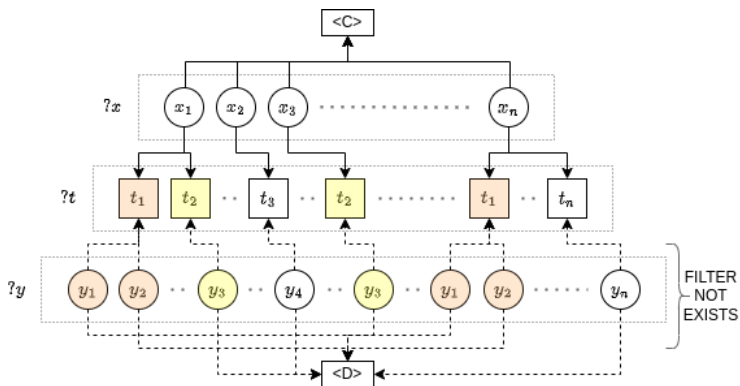
¹Tettamanzi, A., Faron Zucker, C., and Gandon, F. - International Journal of Approximate Reasoning (2017).

ARI (Acceptance/Rejection Index) ²

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

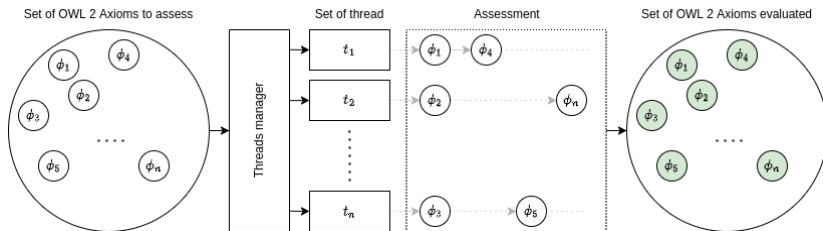
$$\text{ARI}(\phi) = N(\phi) + \Pi(\phi) - 1 \in [-1, 1].$$

²Tettamanzi, A., Faron Zucker, C., and Gandon, F. - International Journal of Approximate Reasoning (2017).

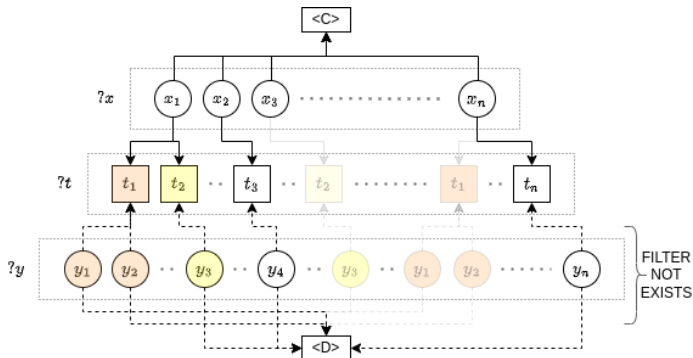


```
SELECT ?class (count(distinct ?x) as ?n)
WHERE {
    ?x a ?class .
    FILTER(
        STRSTARTS(
            STR(?class),
            "http://dbpedia.org/ontology/"
        )
    )
}
GROUP BY ?class
ORDER BY DESC(?n)
LIMIT 10
```

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



- **Aim:** avoid the computation of types t already involved.



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

Step 1: Recovering distinct types

```
SELECT distinct ?t WHERE {
  {
    SELECT ?t WHERE {
      {
        SELECT distinct ?t WHERE {
          ?x a <C> , ?t .
        } ORDER BY ?t
      }
    } LIMIT $limit OFFSET $offset
  }
  FILTER NOT EXISTS {
    ?y a ?t , <D> .
  }
}
```


- 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.

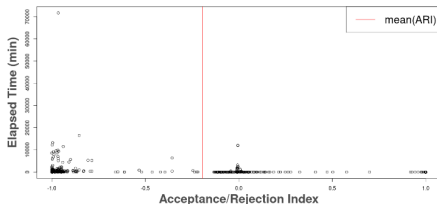
2 Assessing the usage of the loop operator.

- *what benefit do we get from using this operator ?*
- comparison of the computation times (CPU) for each axiom assessed in Experiment 1.

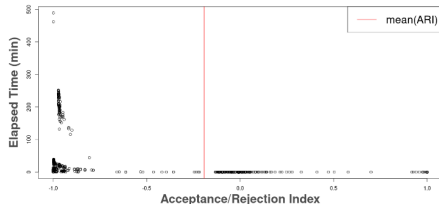
⁵Tettamanzi, A., Faron Zucker, C., and Gandon, F. - K-CAP (2015)

Results ▷ Comparison of ARIs

- The integrity of the ARI values is respected
 - $\text{mean}(\text{ARI}) \simeq -0.1936$ 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.



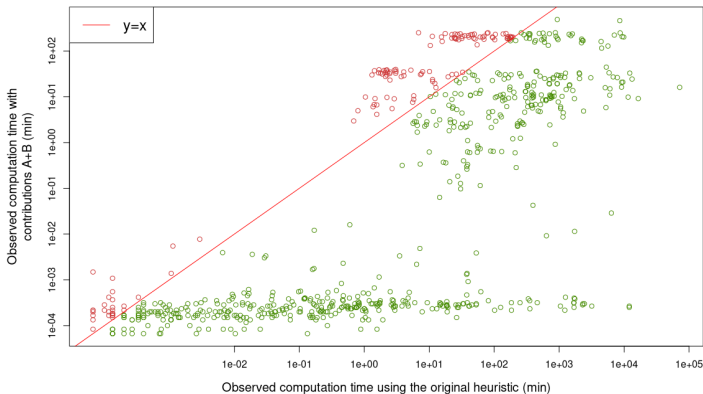
(a) Original heuristic



(b) Results obtained with contributions A+B

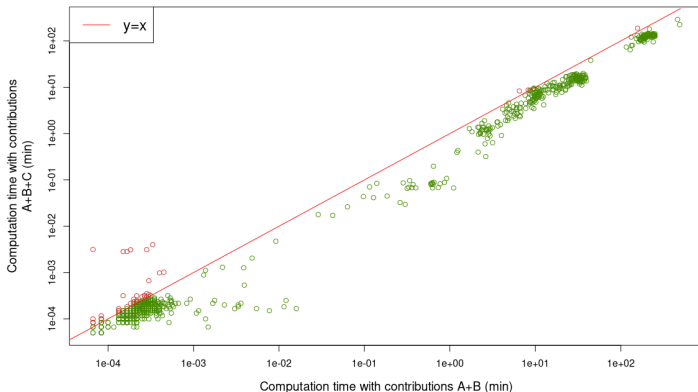
Results ▷ Comparison of computation times (CPU)

- 593 axioms (82%, in green) are faster to assess with contrib. A+B
 - average computation time saved: ~ **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)

- a reduction of ~ 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.



Conclusion

- 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.

Conclusion

- 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.

