Efficient Identification of Implicit Facts in Incomplete OWL2-EL Knowledge Bases

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Web Ontology Language (OWL)

- Extends the RDF Schema (RDFS)
 rdfs:Class, rdfs:Property
- Complex class expressions
 ₀ Woman ≡ Person ⊓ Female
- Complex property expressions
 o hasMother ∘ hasSister ⊑ hasAunt
- Property characteristics
 - We can define a RDF property like "isSiblingWith" as symmetric and transitive

Web Ontology Language (OWL)

 The semantics of OWL imply additional knowledge, i.e., new RDF triples

<mary rdf:type Woman> + "Woman ≡ Person ⊓ Female"

<mary rdf:type Person> <mary rdf:type Female>

- Such "hidden" triples cannot be queried directly with traditional SPARQL engines
- Reasoning is needed

Complexity of Reasoning in OWL

- Exponential for OWL in general
- Three tractable fragments:
 - o OWL2-QL
 - OWL2-EL
 - OWL2-RL
- Each fragment poses different restrictions in the syntax of OWL

Problem

- We are given a large collection of OWL2-EL axioms and a set of inference rules
- <u>Goal</u>: Infer all axioms that are implied by the rules
- <u>How</u>: Apply all rules to the collection of axioms exhaustively till no new axioms are produced (fix-point)

We are given a collection of OWL2-EL axioms of the form $X \sqsubseteq Y$:

- **1.** InfectedWithVirusA \sqcap NotVaccinated \sqsubseteq III
- 2. ∃Vaccinated.VaccineTypeX ⊑ NotVaccinated
- **3**. Vaccinated1994 ⊑ ∃Vaccinated.{va}
- **4**. {va} ⊑ VaccineTypeX
- 5. {john} ⊑ Vaccinated1994
- 6. {john} ⊑ InfectedWithVirusA

- 2. ∃Vaccinated.VaccineTypeX ⊑ NotVaccinated
- **3**. Vaccinated1994 ⊑ ∃Vaccinated.{va}
- **4**. {va} ⊑ VaccineTypeX

From axioms 3, 4 and $2 \rightarrow 7$. Vaccinated 1994 \sqsubseteq NotVaccinated

5. {john} \sqsubseteq Vaccinated1994

From axioms 3, 4 and $2 \rightarrow 7$. Vaccinated 1994 \sqsubseteq NotVaccinated

From axioms 5 and 7 \rightarrow 8. {john} \sqsubseteq NotVaccinated

1. InfectedWithVirusA ⊓ NotVaccinated ⊑ III

6. {john} ⊑ InfectedWithVirusA

From axioms 3, 4 and $2 \rightarrow 7$. Vaccinated 1994 \sqsubseteq NotVaccinated

From axioms 7 and 5 \rightarrow 8. {john} \sqsubseteq NotVaccinated

From axioms 6, 8 and $1 \rightarrow 9$. {john} \sqsubseteq III

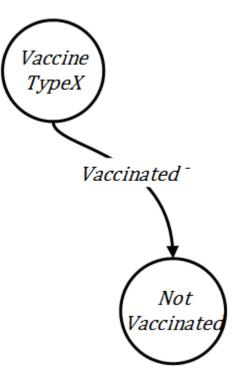
Challenges

- Inference rules for OWL2-EL are complex and mutually recursive (each one affects the other)
- The collection of axioms does not always fit in main-memory
- The inference requires repetitive scans of the axioms
 - $_{\odot}$ The problem becomes I/O bounded.

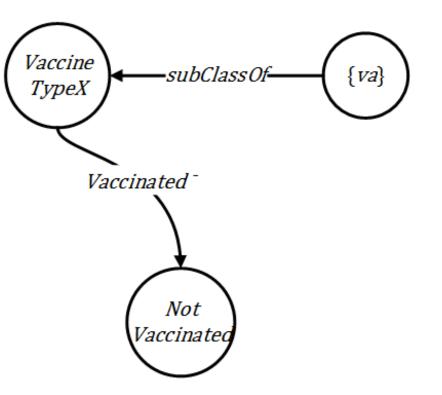
Our Contribution

- All existing rule engines apply the inference rules sequentially
- They scan the ontology on a per-rule basis
- We define a uniform access pattern which allows for the in-bulk application of many rules within the same scan

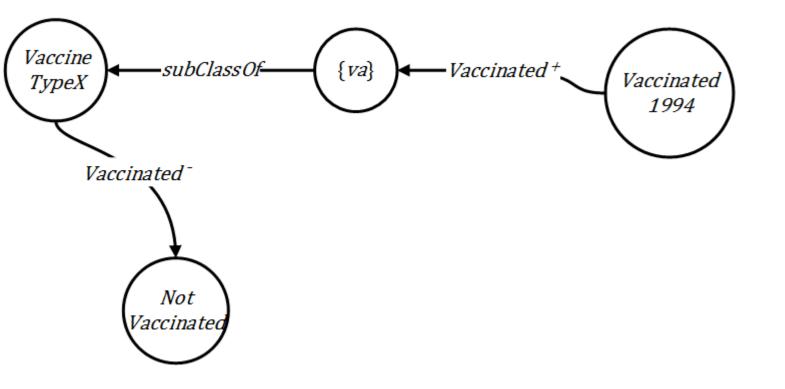
2. ∃Vaccinated.VaccineTypeX ⊑ NotVaccinated



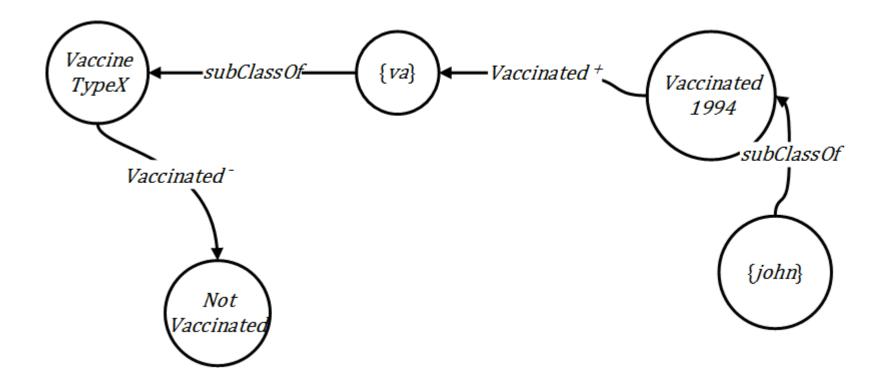
4. {va} ⊑ VaccineTypeX



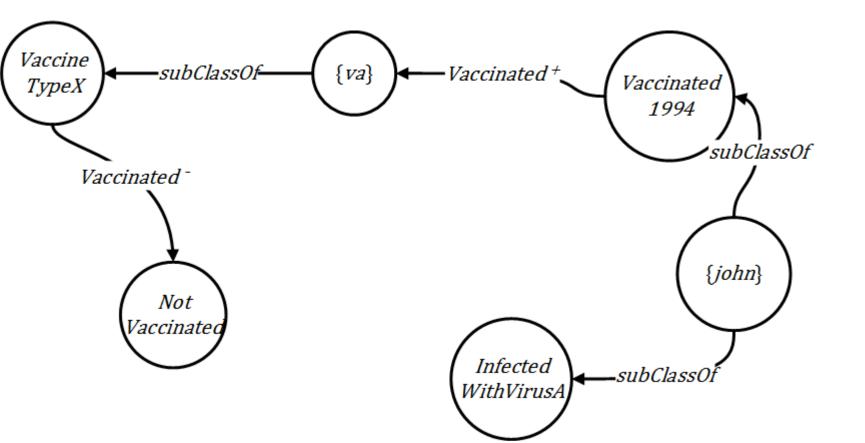
3. Vaccinated1994 ⊑ ∃Vaccinated.{va}

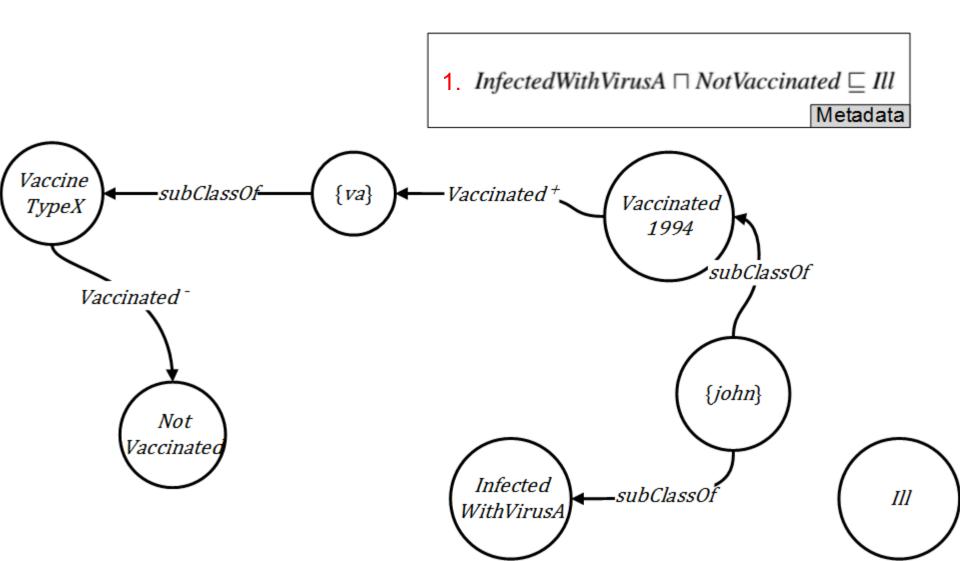


5. {john} ⊑ Vaccinated1994

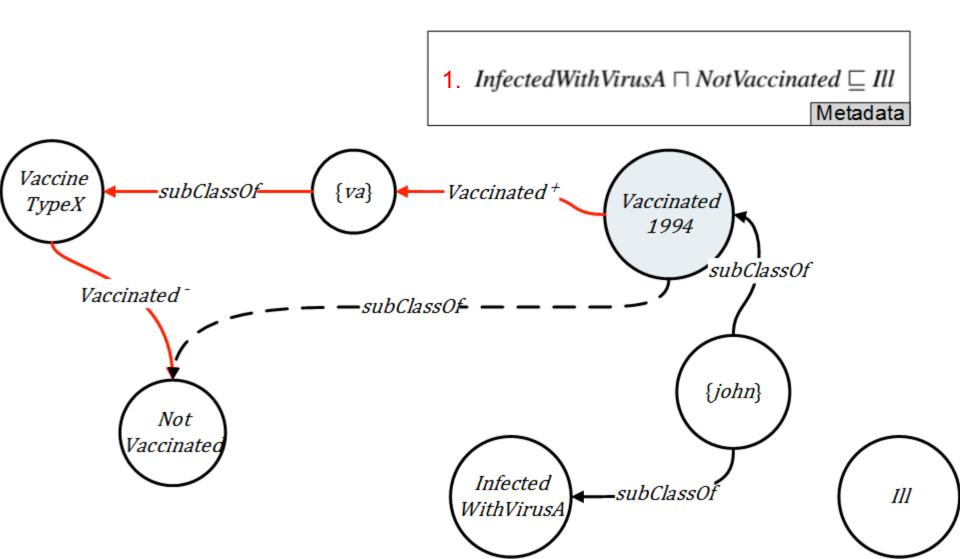


6. {john} ⊑ InfectedWithVirusA

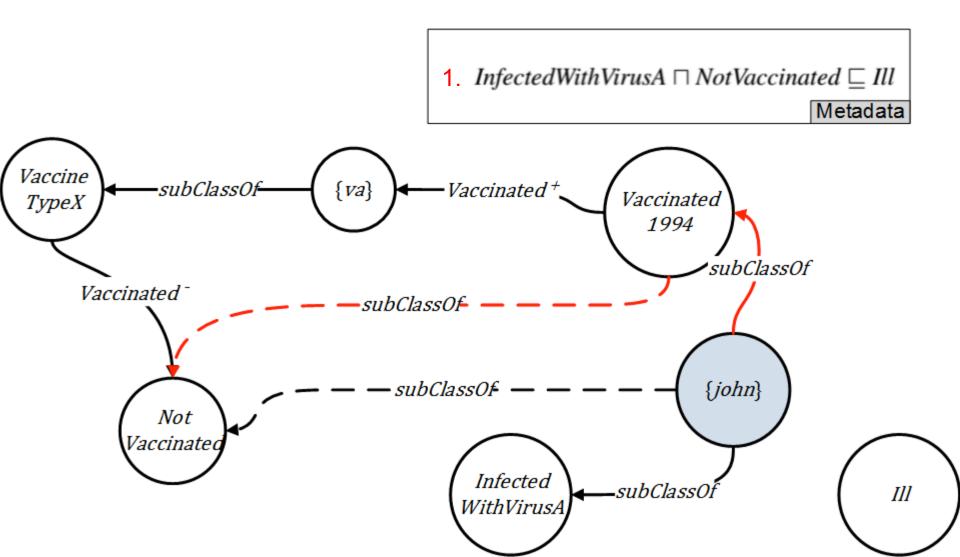




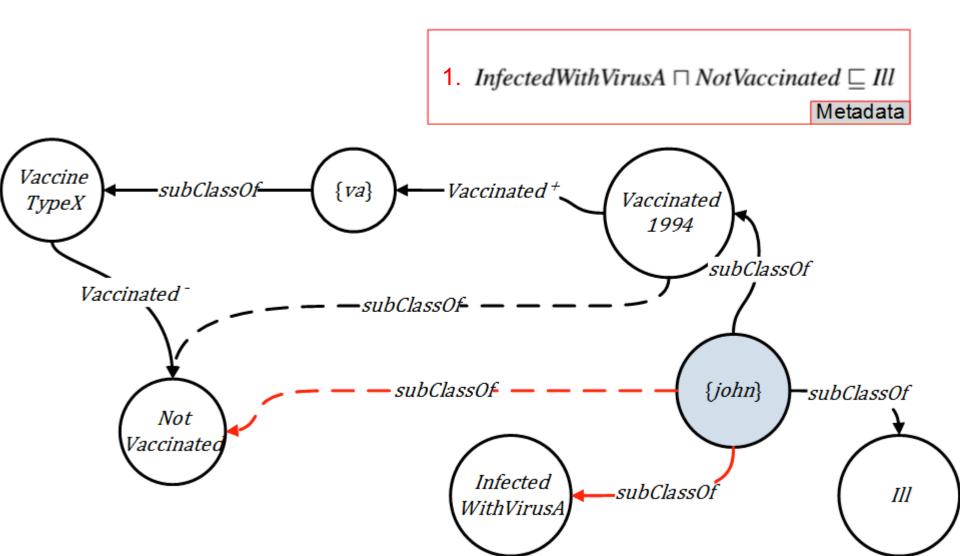
Inference on the Graph



Inference on the Graph



Inference on the Graph



Idea of the Algorithm

- Store the graph in a way that allows efficient lookups in the neighbourhood of each node
- Keep track of the changes made in the graph
- At each subsequent step check only the neighbourhoods affected from the previous step

Experiments

- Real ontologies
 - SNOMED CT
 - o GALEN8
- Synthetic ontologies
 - Ontologies of different sizes whose graphs are isomorphic to the graphs of SNOMED CT and GALEN
 - Ontologies of different sizes by increasing the number of labeled edges per node

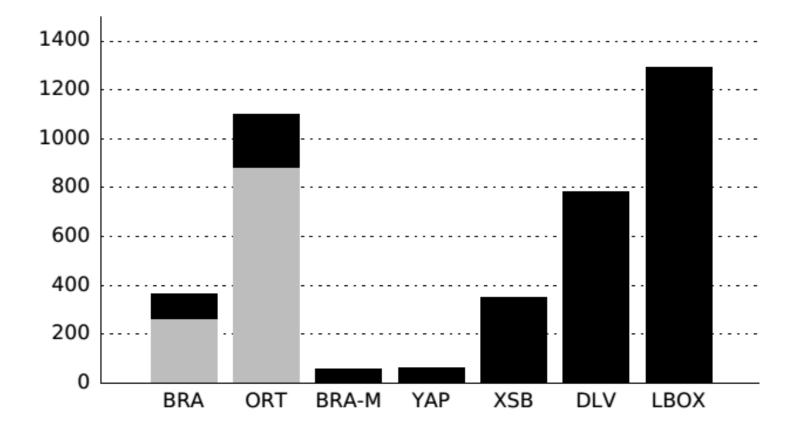
Experiments on Real Data (SNOMED CT)

BRA: Batch Rule Application

ORT: One Rule at a Time

BRA-M: Main-memory version of BRA

YAP, XSB: Prolog-based systems **DLV**, **LogicBlox**: Datalog engines



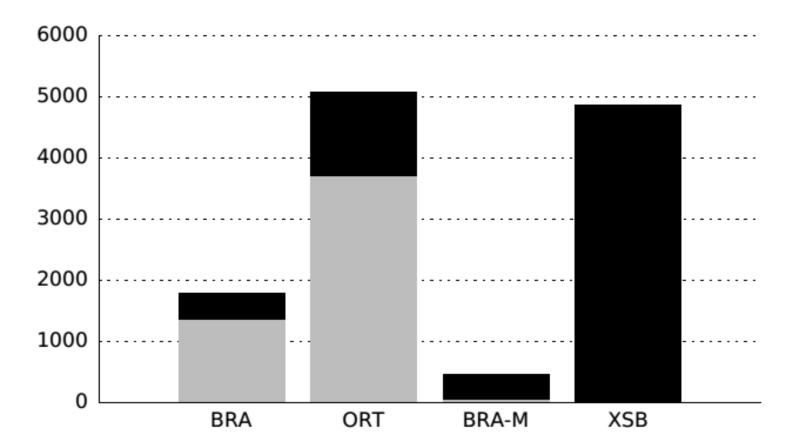
Experiments on Real Data (GALEN8)

BRA: Batch Rule Application

ORT: One Rule at a Time

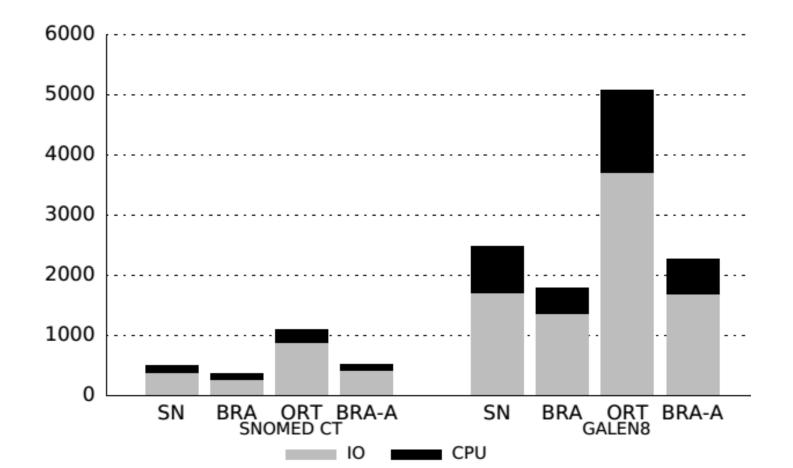
BRA-M: Main-memory version of BRA

YAP, XSB: Prolog-based systems **DLV**, **LogicBlox**: Datalog engines

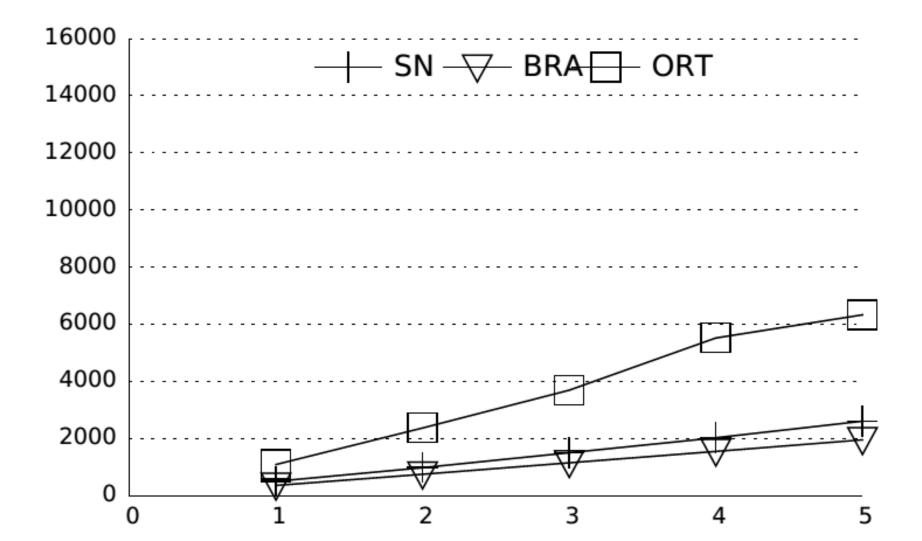


Experiments on Real Data (Optimizations)

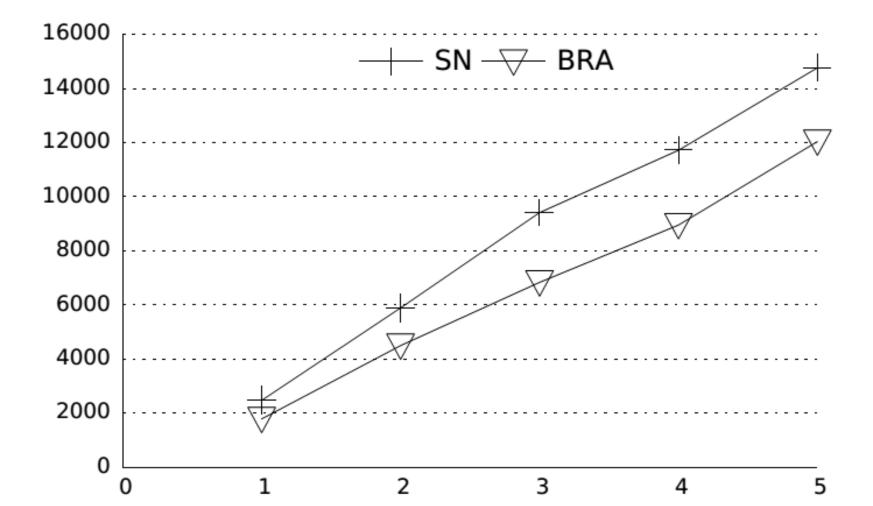
BRA: Batch Rule Application **BRA-A**: BRA on the schema of ORT **SN**: BRA without optimizations **ORT**: One Rule at a Time



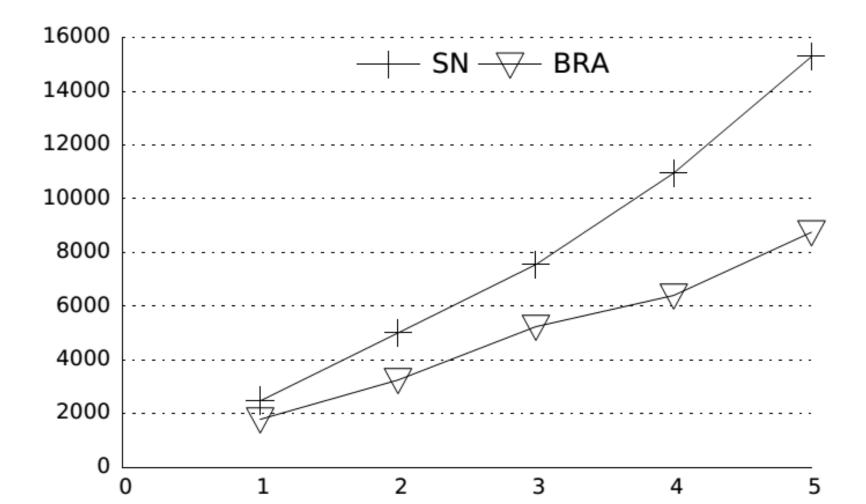
Experiments on Synthetic Data (Isomorphic Graphs of SNOMED CT)



Experiments on Synthetic Data (Isomorphic Graphs of GALEN8)



Experiments on Synthetic Data (Increasing Node Degree for GALEN8)



OWL2-EL Syntax and Semantics

- Intersection of Classes
 - Father \equiv Male \sqcap Parent
 - "Father is the class of all individuals which are of type Male and also of type Parent"

• Existential Restrictions

- Grandparent $\equiv \exists$ hasChild.Parent
- "Grandparent is the class of all individuals which are linked through property "hasChild" with an individual of type Parent"

OWL2-EL Syntax and Semantics

• Reflexivity

- Narcissus $\equiv \exists likes.Self$
- "Narcissus is the class of all individuals which are linked through property 'likes' with themselves"

• Property axioms

- \circ hasSister ⊑ siblingWith
- hasMother ₀ hasSister ⊑ hasAunt

OWL2-EL Syntax and Semantics

- Singleton Nominals (individuals/instances)
 - \circ {mary} ⊑ Woman ↔ <mary rdfs:type Woman>
 - \circ {mary} ⊑ ∃siblingWith.{tom} ↔ <mary siblingWith tom>
 - {mary} \equiv {maria} \leftrightarrow <mary owl:sameAs maria>
- These are the actual data modeled with the ontology

Why OWL2-EL??

- It is widely used in Life Sciences
- Large OWL2-EL ontologies like SNOMED CT have become vital parts of the Health Information Systems in many countries
- It supports the definition of Tuple-Generating Dependencies (TGDs)
 - Suitable for data integration scenarios