

# mOWL: Python library for machine learning with biomedical ontologies

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# Learning Objectives

- Review about different components of ontologies
- Understand the use of different components of ontologies in a machine learning setting
- Learn to use mOWL in order to work with ontologies in machine learning.

# Ontologies

Ontologies contain information on different axes:

- Classes and relations
- Domain vocabulary
- Descriptions
- Axioms

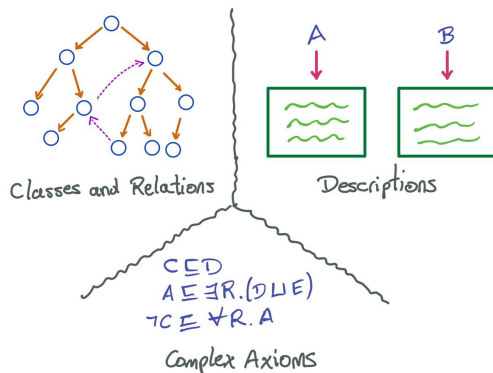


Figure: Components of ontologies

## Ontology axioms

Ontology axioms are defined over a signature  $\Sigma = (C, R, I)$  where  $C$  is a set of concept names,  $R$  is a set of role names and  $I$  is a set of individual names.

### Concepts:

- Person
- Mother
- Child

### Roles:

- hasChild
- hasSibling
- hasFriend

### Individuals:

- Mary
- John
- Bob

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  - Embedding generation
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  - Embedding generation
  - $f : \Sigma \rightarrow \mathbb{R}^n$
- In recent years, many machine learning models have been developed to *embed* ontologies.

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- Usually, the target structure is more suitable to perform operations on the entities.
- Embedding ontologies into  $\mathbb{R}^n$  might be useful to perform operations such as:
  - Similarity computation
  - Inferences

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- Embedding Post-processing



mOWL provides methods to manipulate ontologies:

- creation
- modification
- reasoning

mOWL interfaces the OWLAPI

# Ontology Transformation

- mOWL provides functionalities to transform ontologies and/or extract information in different ways:

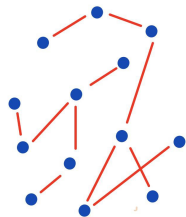


Figure: To graphs

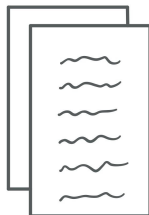


Figure: To text

Axioms

$C \sqsubseteq D$   
 $C \sqsubseteq \exists R.E$   
 $\exists R.F \sqsubseteq G$

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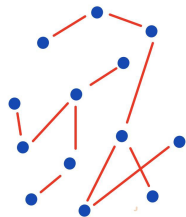


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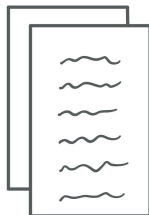


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- Each transformation result can be used as input of a machine learning model.

# Ontology Transformation: Graphs

- We call *projection* to the process of transforming an ontology into a graph.
- There are many methods to project an ontology.
- In general, every projection undergoes some kind of *loss of information*

# Graph projections in mOWL

mOWL provides several projection methods:

- Taxonomy
- Taxonomy + existential relations
- DL2Vec
- OWL2Vec\*

# Ontology Transformation: Graphs

## mOWL code

```
from mowl.projection import DL2VecProjector  
projector = DL2VecProjector(bidirectional_taxonomy=True)
```

# Ontology Transformation: Text

This approach uses the syntactic information of the axioms and generates text sentences out of them.

- Axioms are defined over a syntax (symbols, operators, ...)
- Syntactic elements can be represented as words
- Onto2Vec, OPA2Vec

# Ontology Transformation: Axioms

Some methods would require preprocessing of axioms:

- Normalization (ELEmbeddings, ELBoxEmbeddings)
- Grouping into common structural patterns (FALCON)



- Accessible from:
  - GitHub: <https://github.com/bio-ontology-research-group/mowl/tree/main/mowl>
  - PyPi: `pip install mowl-borg`