

Top Program Construction and Reduction for Polynomial-Time Meta-Interpretive Learning*

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Top Program Construction (simplified)

Input: Positive and negative examples E^+, E^- ; Background knowledge B ; Metarules \mathcal{M} .

Output: The Top Program, T , a set of clauses each of which entails at least one positive and exactly 0 negative examples.

Construction: Clauses in T are instances of the metarules derived by SLD-refutation of examples in E^+ with clauses in B and \mathcal{M} .

Algorithm:

1. Generalisation step: Construct each clause that entails one or more examples in E^+ and add that clause to T .
2. Specialisation step: Remove from T each clause that entails one or more examples in E^- .
3. Return T .

Motivation

Top Program Construction avoids an exponential-time search of a possibly large Hypothesis Space. Instead, the Top Program is *constructed* in time polynomial to the set of constructible clauses.

Implementation

Top Program Construction is implemented in Prolog in the new Meta-interpretive Learning system Louise:
<https://github.com/stassa/louise/>

Louise constructs a Top Program and then reduces it to remove logically redundant clauses.

Top program construction

MIL problem

$$E^+ = \{path(a, b) \leftarrow, path(a, c) \leftarrow\}$$

$$E^- = \{\leftarrow path(1, 2), \leftarrow path(1, 3)\}$$

$$B = \{edge_alpha(a, b), edge_alpha(b, c), edge_alnum(a, b), edge_alnum(b, c), edge_alnum(1, 2), edge_alnum(2, 3)\}$$

$$\mathcal{M} = \{P(x, y) \leftarrow Q(x, y), P(x, y) \leftarrow Q(x, z), R(z, y)\}$$

Generalisation step

$$path(x, y) \leftarrow edge_alnum(x, y)*$$

$$path(x, y) \leftarrow edge_alpha(x, y)$$

$$path(x, y) \leftarrow path(x, y)$$

$$path(x, y) \leftarrow edge_alnum(x, z), edge_alnum(z, y)*$$

$$path(x, y) \leftarrow edge_alnum(x, z), edge_alpha(z, y)$$

$$path(x, y) \leftarrow edge_alpha(x, z), edge_alnum(z, y)$$

$$path(x, y) \leftarrow edge_alpha(x, z), edge_alpha(z, y)$$

$$path(x, y) \leftarrow path(x, z), edge_alnum(z, y)$$

$$path(x, y) \leftarrow path(x, z), edge_alpha(z, y)$$

Specialisation step

$$path(x, y) \leftarrow edge_alpha(x, y)$$

$$path(x, y) \leftarrow path(x, y)$$

$$path(x, y) \leftarrow edge_alnum(x, z), edge_alpha(z, y)$$

$$path(x, y) \leftarrow edge_alpha(x, z), edge_alnum(z, y)$$

$$path(x, y) \leftarrow path(x, z), edge_alnum(z, y)$$

$$path(x, y) \leftarrow path(x, z), edge_alpha(z, y)$$

Top program reduction

$$path(x, y) \leftarrow edge_alpha(x, y)$$

$$path(x, y) \leftarrow edge_alnum(x, z), edge_alpha(z, y)$$

$$path(x, y) \leftarrow path(x, z), edge_alnum(z, y)$$

$$path(x, y) \leftarrow path(x, z), edge_alpha(z, y)$$

A Meta-Interpretive Learning problem consists of examples, background definitions and second-order metarules.

In the Generalisation step clauses are constructed by refutation of examples by SLD-resolution with clauses in B and \mathcal{M} .

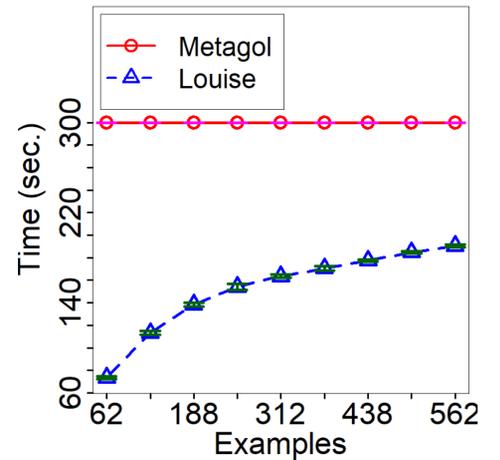
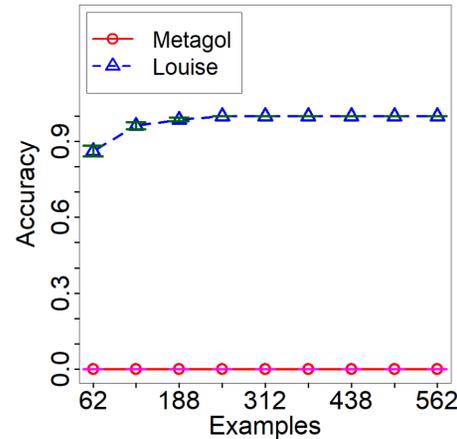
Left, clauses marked with * entail negative examples.

Clauses entailing negative examples are removed in the Specialisation step.

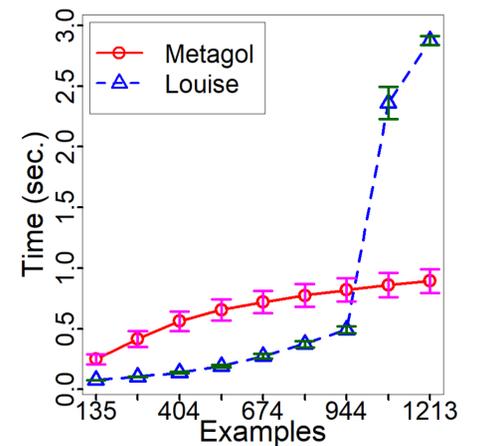
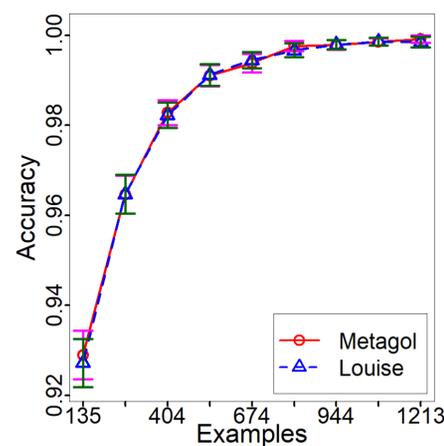
Logically redundant clauses are removed from the Top program by Plotkin's program reduction algorithm.

Comparing Louise to Metagol

Experiment 1: Grid world navigation

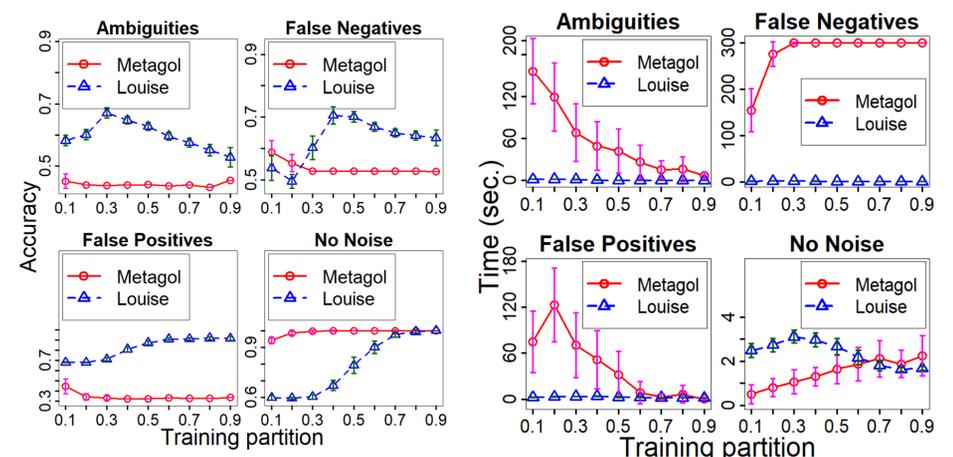


Experiment 2: Learning a CNL grammar (M:tG Fragment)

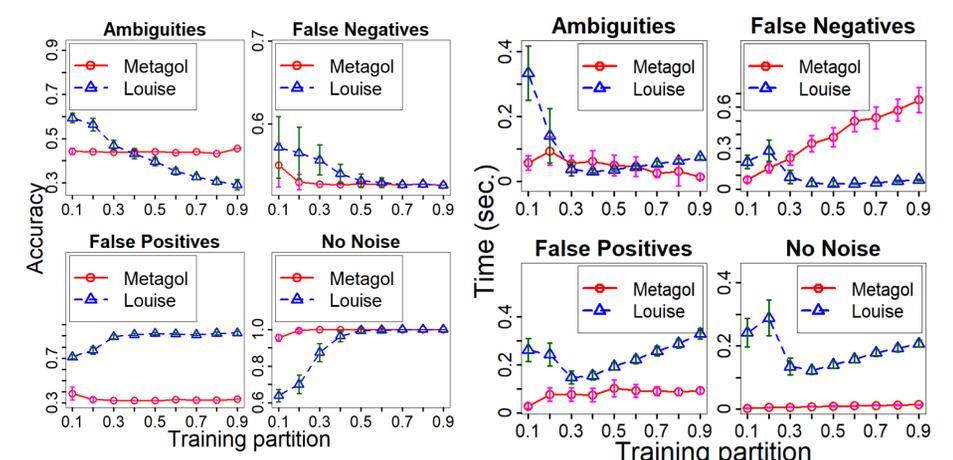


Experiment 3: Learning with mislabelled examples

Coloured graph connectedness – Some irrelevant background knowledge.



Coloured graph connectedness – No Irrelevant background knowledge.



* Recipient of the best ILP student paper award (journal track)