

A Hypergraph Approach for Logic-based Abduction

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Abstract—Abduction reasoning, which finds possible hypotheses from existing observations, has been studied in many different areas. We consider an abduction problem that takes into account a user’s interest. We propose a new approach to solving such an abduction problem based on a hypergraph representation of an ontology and obtain a linear algorithm for a description logic.

Index Terms—Abduction; Hypergraph; Description Logic

I. INTRODUCTION

Abduction reasoning aims to generate a possible hypothesis for a given observation. Abduction has been applied in many artificial intelligence (AI) areas, such as machine learning, logical programming, and statistical relational AI [7].

We focus on *logical-based abduction* [4] over *description logic ontologies*. Here, ontologies consist of *axioms* that state the relationship of different *concepts* and *relationships* over a specific domain. Then, our abduction problem consists of three parts: (i) a given background knowledge (i.e., an existing ontology \mathcal{O}); (ii) a set of hypotheses (i.e., a set of axioms \mathcal{H}) and (iii) a given conclusion (i.e., a single axiom). There have been many studies of abduction over different ontologies, such as the complexity of abduction over \mathcal{EL} [2] and their application to repairing ontologies [8], abduction over \mathcal{EL} by translation to first-order logic [5], forgetting-based abductive reasoning over expressive ontology \mathcal{ALC} [3], and signature-based abduction over more expressive $\mathcal{ALCOI}\mu$ [6].

We propose a new solution (section IV) of abduction over a special \mathcal{EL} -ontology (free of role restrictions) in Section III, based on a hypergraph representation of ontologies.

II. PRELIMINARIES

An *ontology* \mathcal{O} is a set of *axioms* of the form $A_1 \sqcap \dots \sqcap A_n \sqsubseteq B$, where A_i, B are called *concepts*. An interpretation $\mathcal{I} = \langle \Delta^{\mathcal{I}}, \cdot^{\mathcal{I}} \rangle$ consists of a non-empty domain $\Delta^{\mathcal{I}}$ and a mapping $\cdot^{\mathcal{I}}$ that maps each concept to a subset $A^{\mathcal{I}} \subseteq \Delta^{\mathcal{I}}$. A model of \mathcal{O} is an interpretation that for each $A_1 \sqcap \dots \sqcap A_n \sqsubseteq A \in \mathcal{O}$, we have $A_1^{\mathcal{I}} \cap \dots \cap A_n^{\mathcal{I}} \subseteq A^{\mathcal{I}}$. We say $\mathcal{O} \models A_1 \sqcap \dots \sqcap A_n \sqsubseteq B'$ iff for any models \mathcal{I} of \mathcal{O} , we have $(A_1^{\mathcal{I}})^{\mathcal{I}} \cap \dots \cap (A_n^{\mathcal{I}})^{\mathcal{I}} \subseteq (B')^{\mathcal{I}}$.

A (*directed*) *hypergraph* $\mathcal{H} = \{\mathcal{V}, \mathcal{E}\}$ consists of a node set $\mathcal{V} = \{v_1, v_2, \dots, v_n\}$ and a hyperedge set $\mathcal{E} = \{e_1, e_2, \dots, e_m\}$, where $e_i = \langle T(e_i), f(e_i) \rangle$ with $T(e_i) \subseteq \mathcal{V}$ being a subset and $f(e_i) \in \mathcal{V}$ being a node. Note that a classical hyperedge can have multiple nodes in its head, which we require to be a singleton for computing abduction.

Definition 1 ([1]). Given a hypergraph $\mathcal{H} = \{\mathcal{V}, \mathcal{E}\}$, assume $S \subseteq \mathcal{V}$ and $v \in \mathcal{V}$. A *hyperpath* from S to v is a sequence

$h = [e_1, e_2, \dots, e_n]$ of hyperedges such that (i) $f(e_n) = \{v\}$; (ii) for $i = 1, \dots, n$, $T(e_i) \subseteq S \cup \{f(e_1); \dots, f(e_{i-1})\}$; (iii) for $i = 1, \dots, n$, $f(e_i) \in \bigcup_{i < j \leq n} T(e_j)$.

III. ABDUCTION PROBLEM

We consider an abduction problem that takes into account a user’s interests represented by a set of concepts Σ .

Definition 2. An abduction problem is a tuple

$$\langle \mathcal{O}, \Sigma, A_1 \sqcap \dots \sqcap A_n \sqsubseteq B \rangle,$$

where $\Sigma = \{A', B', \dots\}$ is a set of concept names. A *solution* of this problem is a (minimal) ontology

$$\mathcal{H} = \{A'_1 \sqcap \dots \sqcap A'_n \sqsubseteq B' \mid A'_i, B' \in \Sigma, n \geq 0\}$$

such that $\mathcal{O} \cup \mathcal{H} \models A_1 \sqcap \dots \sqcap A_n \sqsubseteq B$. A solution \mathcal{H} is called a *hypothesis* with respect to Σ .

Example 1. Let an ontology \mathcal{O}_0 be:

$$peopleWithDiploma \sqsubseteq doctor$$

$$peopleHasPaper \sqsubseteq researcher$$

$$doctor \sqcap employeeWithUniversityChair \sqsubseteq professor$$

\mathcal{O}_0 can not derive the following axiom α_0 :

$$\alpha_0 : doctor \sqcap employeeWithUniversityChair \sqsubseteq researcher$$

although it should be true. Consider $\Sigma_0 = \{professor, peopleHasPaper\}$. If we add a hypothesis $\mathcal{H}_0 = \{professor \sqsubseteq peopleHasPaper\}$, we have $\mathcal{O}_0 \cup \mathcal{H}_0 \models \alpha_0$. Therefore, \mathcal{H}_0 is a solution of the abduction problem $\mathcal{A}_0 = \langle \mathcal{O}_0, \Sigma_0, \alpha_0 \rangle$. It is clear that \mathcal{H}_0 is also a minimal solution to the abduction problem. But there is no solution to \mathcal{A}_0 if $\Sigma_0 = \{professor, peopleWithDiploma\}$.

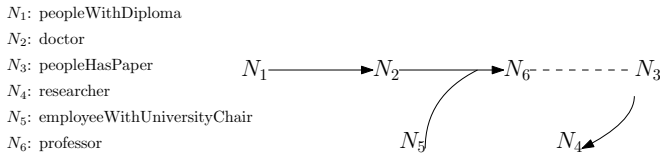
IV. A HYPERGRAPH-BASED ALGORITHM

We now present a method of finding a (minimal) solution to the abduction problem using hypergraphs.

Definition 3. For each set \mathcal{O} of axioms, we define a hypergraph $H_{\mathcal{O}} = (\mathcal{N}_h, \mathcal{E}_h)$, where $\mathcal{N}_h := \{N_{A'} \mid A' \in \mathbf{N}_c\}$ and

$$\mathcal{E}_h := \{\{N_{A'_1}, \dots, N_{A'_n}\} \rightarrow N_{A'} \mid A'_1 \sqcap \dots \sqcap A'_n \sqsubseteq A' \in \mathcal{O}\}$$

Example 2 (Example 1 cont’d). By definition, the hypergraph $H_{\mathcal{O}_0}$ of \mathcal{O}_0 is shown in Figure. 1. Now, we add an edge


 Fig. 1: The hypergraph representation $H_{\mathcal{O}_0}$ of \mathcal{O}_0 in Example 1

$\{N_6\} \rightarrow N_3$ to the hypergraph $H_{\mathcal{O}_0}$. Then, we can find a hyperpath h from $\{N_2, N_5\}$ to N_4 :

$$h = [\{N_2, N_5\} \rightarrow N_6, \{N_6\} \rightarrow N_3, \{N_3\} \rightarrow N_4]$$

Theorem 1. Given an ontology \mathcal{O} and its associated hypergraph $H_{\mathcal{O}}$, an ontology \mathcal{H} is a (minimal) solution to the abduction problem $\langle \mathcal{O}, \Sigma, A_1 \sqcap \dots \sqcap A_n \sqsubseteq B \rangle$ iff $H_{\mathcal{H}}$ is a (minimal) hypergraph such that (i) All nodes in $H_{\mathcal{H}}$ are of the form N_A , $A \in \Sigma$, and (ii) There exists a hyperpath from N_{A_1}, \dots, N_{A_n} to N_B in $H_{\mathcal{O}} \cup H_{\mathcal{H}}$.

Example 3 (Example 1 cont'd). By Theorem 1, to solve the abduction problem \mathcal{A}_0 , it is enough to find an $H_{\mathcal{H}}$ such that there exists a hyperpath from $\{N_2, N_5\}$ to N_4 in $H_{\mathcal{O}_0} \cup H_{\mathcal{H}}$. The hypergraph $H_{\mathcal{H}}$ consists of a single edge $\{N_6\} \rightarrow N_3$ satisfying the requirement, leading to the hyperpath given in Example 2 as the minimal solution of the problem.

Before stating our main Algorithm 2, we define a property of *saturation* for a hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ and $V \subset \mathcal{V}$. We define $U \subset V$ to be *saturated* (under V) if there exists $e \in \mathcal{E}$ such that $T(e) = U$ and $f(e) \not\subseteq V$. For example, in Fig. 1, if we have $V = \{N_1, N_2, N_5\}$, then $\{N_1\}$ and $\{N_2, N_5\}$ are saturated under V , while other subsets of V are not. Algorithm 1 finds all vertices approachable from V in run-time $O(|\mathcal{E}|)$.

Proposition 1. For a hypergraph $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ and $V \subset \mathcal{V}$, $v \in \text{Span}(\mathcal{V}, \mathcal{E}, V)$ iff. there is a hyperpath from V to v .

Algorithm 1: $\text{Span}(\mathcal{V}, \mathcal{E}, V)$

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input : hypergraph  $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ , set  $V \subset \mathcal{V}$ 
output:  $W \subset \mathcal{V}$  of all vertices spanned from  $V$ 
1  $W = V$ .
2  $\mathcal{U} = \{U, U \text{ is saturated under } V\}$ .
3 while  $\mathcal{U} \neq \emptyset$  do
4   choose  $U \in \mathcal{U}$ ,
5   while there exists  $v \in \mathcal{V} \setminus W$  such that  $(W, v) \in \mathcal{E}$ 
6     do
7       put  $v$  into  $W$ ;
8       put all saturated sets containing  $v$  into  $\mathcal{U}$ .
9   end
10 end
11 return  $W$ 
    
```

In Algorithm 2, we first check if v can be directly reached by V (Line 1-3). Then, we check if the aiming hypergraph exists (Line 4-11). These 2 steps have run-time $O(|\mathcal{E}|)$. In the minimizing step, for each $e \in \mathcal{E}'$, we check only once if e can be deleted. Hence, the total run-time is $O(|\Sigma||\mathcal{E}|)$.

Algorithm 2:

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input : hypergraph  $H = (\mathcal{V}, \mathcal{E}), \Sigma \subset \mathcal{V}, S \subset \mathcal{V}, v \in \mathcal{V}$ 
output: hypergraph  $\mathcal{H}$  on  $\Sigma$ 
1  $V = \text{Span}(\mathcal{V}, \mathcal{E}, S)$ .
2 if  $v \in V$  then
3   return empty graph.
4 else
5   if  $\Sigma \subset V$  or  $\Sigma \cup V = \emptyset$  return non-existence.
6    $\Sigma \setminus V = \{v_1, \dots, v_m\}$ ,
7   choose  $m$  hyper-edges  $\mathcal{E}' = \{e_1, \dots, e_m\}$  where
8      $T(e_i) \subset \Sigma \cap V$  and  $f(e_i) = v_i$  for  $1 \leq i \leq m$ .
9    $V = V \cup \Sigma$ .
10 if  $v \notin \text{Span}(\mathcal{V}, \mathcal{E} \cup \mathcal{E}', V)$  then
11   return non-existence.
12 else
13   minimize  $\mathcal{E}'$  (check if there exists  $e \in \mathcal{E}'$  such that
14      $\mathcal{E}' - e$  satisfies until we get a minimal size).
15   return  $\mathcal{H} = (\Sigma, \mathcal{E}')$ .
    
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We explain Algorithm 2 via the following example.

Solution of Example 3. (via Algorithm 2)

- 1) $H = H_{\mathcal{O}}, \Sigma = \{N_3, N_6\}, S = \{N_2, N_5\}, v = N_4$.
- 2) Line 1: $V = \text{Span}(\mathcal{V}, \mathcal{E}, S) = \{N_2, N_5, N_6\}$.
- 3) Line 7: $\mathcal{E}' = \{\{N_6\} \rightarrow N_3\}$.
- 4) Line 10: $v \in \text{Span}(\mathcal{V}, \mathcal{E} \cup \mathcal{E}', V = \mathcal{V})$.
- 5) Line 13: we see $\{\{N_6\} \rightarrow N_3\}$ cannot be deleted. It returns $\mathcal{H} = (\Sigma, \mathcal{E}')$.

Theorem 2. For Algorithm 2, the output \mathcal{H} is a minimal hypergraph satisfying the conditions (i) and (ii) in Theorem 1.

V. CONCLUSION

In this work, we introduce a hypergraph-based algorithm for solving abduction problems over \mathcal{EL} -ontologies that do not have role restrictions, which have a linear time complexity w.r.t. the size of the input ontology. As for future work, we plan to implement our algorithm and extend it to handle general \mathcal{EL} -ontologies with role restrictions, as well as more expressive ontologies such as \mathcal{ALC} .

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REFERENCES

- [1] A. Giorgio and L. Luigi, "Directed hypergraphs: Introduction and fundamental algorithms-a survey", *Theoretical Computer Science*, vol. 658, pp. 293–306, 2017.
- [2] M. Bienvenu, "Complexity of abduction in the EL family of lightweight description logics", *Proc. of KR'08*, 2008, pp. 220–230.
- [3] W. Del-Pinto and R. A. Schmidt, "Abox abduction via forgetting in ALC", *Proc. of AAAI'19*, 2019, pp. 2768–2775.
- [4] T. Eiter and G. Gottlob, "The complexity of logic-based abduction", *J. ACM*, vol. 42 (1), 1995, pp. 3–42.
- [5] F. Haifani, P. Koopmann, S. Tournet, and C. Weidenbach, "Connection-minimal abduction in EL via translation to FOL", *Proc. of IJCAR'22*, 2022, pp. 188–207.

- [6] P. Koopmann, W. Del-Pinto, S. Tourret, and R. A. Schmidt “Signature-based abduction for expressive description logics”, Proc. of KR’20, 2020, pp. 592–602.
- [7] Sindhu V. Raghavan, “Bayesian Abductive Logic Programs: A Probabilistic Logic for Abductive Reasoning”, Statistical Relational Artificial Intelligence, Proc. of IJCAI’11, 2011, pp. 2840–2841.
- [8] F. Wei-Kleiner, Z. Dragisic, and P. Lambrix, “Abduction framework for repairing incomplete EL ontologies: Complexity results and algorithms”, Proc. of AAAI’14, 2014, pp. 1120–1127.