selp: A Single-Shot Epistemic Logic Program Solver

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Abstract

Epistemic Logic Programs (ELPs) are an extension of Answer Set Programming (ASP) with epistemic operators that allow for a form of meta-reasoning, that is, reasoning over multiple possible worlds. Existing ELP solving approaches generally rely on making multiple calls to an ASP solver in order to evaluate the ELP. However, in this paper, we show that there also exists a direct translation from ELPs into non-ground ASP with bounded arity. The resulting ASP program can thus be solved in a single shot. We then implement this encoding method, using recently proposed techniques to handle large, non-ground ASP rules, into the prototype ELP solving system “selp”, which we present in this paper. This solver exhibits competitive performance on a set of ELP benchmark instances.

1 Introduction

Epistemic Logic Programs (ELPs), as defined in (Shen and Eiter 2016), are an extension of the well-established formalism of Answer Set Programming (ASP). ASP is a generic, fully declarative logic programming language that allows us to encode problems in such a way that the resulting answers (called answer sets) directly correspond to solutions of the encoded problem (Brewka et al. 2011). Negation in ASP is generally interpreted according to the stable model semantics (Gelfond and Lifschitz 1988), that is, as negation-as-failure or default negation. The default negation ¬a of an atom a is true if there is no justification for a in the same answer set, making it a “local” operator in the sense that it is defined relative to the same answer set. ELPs, on the other hand, extend ASP with the epistemic negation operator not that allows for a form of meta-reasoning, that is, reasoning over multiple answer sets. Intuitively, an epistemically negated atom not a expresses that a cannot be proven true, that is, it is false in at least one answer set. Thus, epistemic negation is defined relative to a collection of answer sets, referred to as a world view. The main reasoning task for ELPs, checking that a world view exists, is \( \Sigma^2_p \)-complete (Shen and Eiter 2016).

Epistemic negation has long been recognized as a desired construct for ASP (Gelfond 1991; Gelfond 1994). In these works, Michael Gelfond introduced the modal operators K (“known” or “provably true”) and M (“possible” or “not provably false”), in order to address this need. Given an atom a, Ka and Ma stand for ¬not a and not ¬a, respectively.

Example 1

A classical example for the use of epistemic negation is the presumption of innocence rule

\[
\text{innocent}(X) \leftrightarrow \text{not guilty}(X),
\]

namely: a person is innocent if they cannot be proven guilty.
Renewed interest in recent years has revealed several flaws in the original semantics, and several approaches (cf. e.g. [Gelfond 2011], [Truszczynski 2011], [Kahl 2014], [del Cerro et al. 2015], [Shen and Eiter 2016]) aimed to refine them in such a way that unintended world views are eliminated. In this work, we will settle on the semantics proposed in [Shen and Eiter 2016]. The flurry of new research also led to the development of ELP solving systems (Kahl et al. 2015, Son et al. 2017). Such solvers employ readily available, highly efficient ASP systems like cling (Gebser et al. 2012), (Gebser et al. 2014) and WASP (Alviano et al. 2013), especially making use of the former solver’s multi-shot solving functionality (Gebser et al. 2019). However, these ELP solving systems rely on ground ASP programs when calling the ASP solver, which, for reasons rooted in complexity theory, generally requires multiple calls in order to check for world view existence. The main aim of our paper is to present techniques and a system for solving ELPs that is able to utilize an ASP solver in such a way that the ELP can be solved in a single shot.

Contributions. Our contributions in this paper are twofold:

- We propose a novel translation from ELPs to ASP programs using large non-ground ASP rules, such that the ELP can be solved by an ASP solving system in a single shot. This is done via a recently proposed encoding technique (Bichler et al. 2016b) that uses large ASP rules to formulate complex checks. This technique builds on a result from (Eiter et al. 2007) that states that evaluating non-ground ASP programs with bounded predicate arity is \( \Sigma_3^p \)-complete, which matches the complexity of evaluating ELPs. Our proposed translation is therefore optimal from a complexity-theoretic point of view. From a practical point of view, such an encoding avoids multiple calls to the ASP solver. State-of-the-art systems use sophisticated heuristics and learning, and multiple calls might result in a loss of knowledge about the problem instance, which the solver has already learned.
- We further discuss how our encoding needs to be constructed in order to be useful in practice. In particular, in current ASP systems, non-ground ASP programs first need to be *grounded*, that is, all variables need to be replaced by all allowed combinations of constants. Since our encoding makes use of large non-ground rules, a naive grounding will often not terminate, since there may be hundreds or thousands of variables in a rule. However, as proposed in (Bichler et al. 2016b), we make use of the *lpopt* rule decomposition tool (Bichler et al. 2016a) that splits such large rules into smaller ones that are more easily grounded, by making use of the concept of *treewidth* and *tree decompositions* (Bodlaender 1993). To use this tool to its full potential, the large rules we use in our encoding must be constructed carefully, in order for *lpopt* to split them up optimally.
- Finally, we present a prototype implementation of our ELP-to-ASP rewriting approach and combine it with the state-of-the-art ASP solving system cling (Gebser et al. 2014) in order to evaluate its performance. We compare our system against EP-ASP (Son et al. 2017) on different benchmarks found in the literature. Our system shows competitive performance on these benchmarks, in particular on instances with good structural properties.

The remainder of the paper is structured as follows: in Section 2 we introduce the formal background of ASP, ELPs, and tree decompositions; Section 3 states our reduction from ELPs to ASP, including practical considerations and a discussion of related work; Section 4 presents how QBF formulas can be encoded as ELP programs; Section 5 introduces our ELP solver; Section 6 presents our benchmark results, making use of results from Section 4, and finally Section 7 closes with some concluding remarks.
This paper is an extended version of (Bichler et al. 2018a; Bichler et al. 2018b). Additional material includes a full correctness proof for our reduction in Section 3 and a formalized and detailed description of the adaptations needed to make the reduction workable in practice. Further, Section 4 describes in detail how our QBF benchmarks, used in Section 6, are constructed.

2 Preliminaries

Answer Set Programming (ASP). A ground logic program (also called answer set program, ASP program, or simply program) is a pair \( \Pi = (\mathcal{A}, \mathcal{R}) \), where \( \mathcal{A} \) is a set of propositional (i.e. ground) atoms and \( \mathcal{R} \) is a set of rules of the form

\[
a_1 \lor \cdots \lor a_l \leftarrow d_{l+1}, \ldots, a_m, \neg a_{m+1}, \ldots, \neg a_n;
\]

where the comma symbol stands for conjunction, \( n \geq m \geq l \geq 0 \) and \( a_i \in \mathcal{A} \) for all \( 1 \leq i \leq n \). Each rule \( r \in \mathcal{R} \) of form (1) consists of a head \( H(r) = \{a_1, \ldots, a_l\} \) and a body given by \( B^+(r) = \{a_{l+1}, \ldots, a_m\} \) and \( B^-(r) = \{a_{m+1}, \ldots, a_n\} \). A literal \( \ell \) is either an atom \( a \) or its (default) negation \( \neg a \). A literal \( \ell \) is true in a set of atoms \( I \subseteq \mathcal{A} \) if \( \ell = a \) and \( a \in I \), or \( \ell = \neg a \) and \( a \notin I \); otherwise \( \ell \) is false in \( I \). A set \( M \subseteq \mathcal{A} \) is a called a model of \( r \) if \( B^+(r) \subseteq M \) together with \( B^-(r) \cap M = \emptyset \) implies that \( H(r) \cap M \neq \emptyset \). We denote the set of models of \( r \) by \( \text{models}(r) \) and the models of a logic program \( \Pi = (\mathcal{A}, \mathcal{R}) \) are given by \( \text{models}(\Pi) = \bigcap_{r \in \mathcal{R}} \text{models}(r) \). The GL-reduct \( \Pi' \) of a ground logic program \( \Pi \) with respect to a set of atoms \( I \subseteq \mathcal{A} \) is the program \( \Pi' = (\mathcal{A}, \{H(r) \leftarrow B^+(r) | r \in \mathcal{R}, B^-(r) \cap I = \emptyset\}) \).

Definition 2 (Gelfond and Lifschitz 1988; Gelfond and Lifschitz 1991) \( M \subseteq \mathcal{A} \) is an answer set of a program \( \Pi \) if (1) \( M \in \text{models}(\Pi) \) and (2) there is no subset \( N \subset M \) such that \( N \in \text{models}(\Pi^M) \).

The set of answer sets of a program \( \Pi \) is denoted \( \text{AS}(\Pi) \). The consistency problem of ASP (decide whether, given \( \Pi \), \( \text{AS}(\Pi) \neq \emptyset \)) is \( \Sigma^2_p \)-complete (Eiter and Gottlob 1995).

General non-ground logic programs differ from ground logic programs in that variables may occur in rules. Such rules are \( \forall \)-quantified first-order implications of the form \( H_1 \lor \cdots \lor H_k \leftarrow P_1, \ldots, P_n, \neg N_1, \ldots, \neg N_m \) where \( H_i \), \( P_i \) and \( N_i \) are (non-ground) atoms. A non-ground atom \( A \) is of the form \( p(t) \) and consists of a predicate name \( p \), as well as a sequence of terms \( t \), where each term \( t \in I \) is either a variable or a constant from a domain \( \Delta \), with \( |t| \) being the arity of \( p \). Let \( \text{var}(A) \) denote the set of variables \( X \) in a non-ground atom \( A \). This notation naturally extends to sets. We will denote variables by capital letters, constants and predicates by lower-case words. A non-ground rule can be seen as an abbreviation for all possible instantiations of the variables with domain elements from \( \Delta \). This step is usually explicitly performed by a grounder that transforms a (non-ground) logic program into a set of ground rules of the form (1). Note that, in general, such a ground program may be exponential in the size of the non-ground program. For non-ground programs of bounded arity, the consistency problem is \( \Sigma^1_p \)-complete (Eiter et al. 2007).

Epistemic Logic Programs. A ground epistemic logic program (ELP) is a pair \( \Pi = (\mathcal{A}, \mathcal{R}) \), where \( \mathcal{A} \) is a set or propositional atoms and \( \mathcal{R} \) is a set of rules of the following form:

\[
a_1 \lor \cdots \lor a_k \leftarrow \ell_1, \ldots, \ell_m, \xi_1, \ldots, \xi_j, \neg \xi_{j+1}, \ldots, \neg \xi_n,
\]

where each \( a_i \) is an atom, each \( \ell_i \) is a literal, and each \( \xi_j \) is an epistemic literal, that is, a formula \( \text{not} \ell \), where \( \text{not} \) is the epistemic negation operator, and \( \ell \) is a literal. W.l.o.g. we assume that
In this section, we provide our novel translation for solving ELPs via a single call to an ASP with consistency minimum width over all tree decompositions of $G$ (Shen and Eiter 2016). Let $\Phi$ be the following ELP, with all epistemic literals not $q$.

In order to define the main reasoning tasks for ELPs, we recall the notion of the epistemic reduct (Shen and Eiter 2016). Let $\Phi \subseteq \text{elit}(\Pi)$ (called a guess). The epistemic reduct $\Pi^\Phi$ of the program $\Pi = (\mathcal{A}, \mathcal{R})$ w.r.t. $\Phi$ consists of the rules $\{r^- | r \in \mathcal{R}\}$, where $r^-$ is defined as the rule $r$ with all epistemic literals not $\ell$ in $\Phi$ (resp. in $\text{elit}(\Pi) \setminus \Phi$) replaced by $\top$ (resp. $\neg \ell$). Note that $\Pi^\Phi$ is a logic program without epistemic negation. This leads to the following, central definition.

**Definition 3**

Let $\Phi$ be a guess. The set $\mathcal{M} = \text{AS}(\Pi^\Phi)$ is called a candidate world view of $\Pi$ iff

1. $\mathcal{M} \neq \emptyset$,
2. for each epistemic literal not $\ell \in \Phi$, there exists an answer set $M \in \mathcal{M}$ wherein $\ell$ is false, and
3. for each epistemic literal not $\ell \in \text{elit}(\Pi) \setminus \Phi$, it holds that $\ell$ is true in each answer set $M \in \mathcal{M}$.

**Example 4**

Let $\Pi$ be the following ELP, with $\mathcal{R} = \{r_1, r_2\}$:

$$
\begin{align*}
  r_1 &: p \leftarrow \text{not } q \\
  r_2 &: q \leftarrow \text{not } p
\end{align*}
$$

ELP $\Pi$ has two candidate world views: (1) $\Phi = \{\text{not } q\}$ with $\text{AS}(\Pi^\Phi) = \{\{p\}\}$; (2) $\Phi = \{\text{not } p\}$ with $\text{AS}(\Pi^\Phi) = \{\{q\}\}$.

The main reasoning task we treat in this paper is the world view existence problem (or ELP consistency), that is, given an ELP $\Pi$, decide whether a candidate world view exists. This problem is known to be $\Sigma^P_2$-complete (Shen and Eiter 2016).

**Tree Decompositions.** A tree decomposition of a graph $G = (V, E)$ is a pair $\mathcal{T} = (T, \chi)$, where $T$ is a rooted tree and $\chi$ is a labelling function over nodes $t$, with $\chi(t) \subseteq V$, such that the following holds: (i) for each $v \in V$ there is a node $t$ in $T$ such that $v \in \chi(t)$; (ii) for each $(v, w) \in E$ there is a node $t$ in $T$ such that $\{v, w\} \subseteq \chi(t)$; and (iii) for all nodes $r, s$, and $t$ in $T$, where $s$ lies on the path from $r$ to $t$, $\chi(r) \cap \chi(t) \subseteq \chi(s)$. The width of a tree decomposition $\mathcal{T}$ is defined as the maximum cardinality of $\chi(t)$ minus one, over all nodes $t$ of $\mathcal{T}$. The treewidth of a graph $G$ is the minimum width over all tree decompositions of $G$. Trees have treewidth 1, cliques of size $k$ have treewidth $k$. Finding a tree decomposition of minimal width is NP-hard in general.

### 3 Single-Shot ELP Solving

In this section, we provide our novel translation for solving ELPs via a single call to an ASP solving system. The goal is to transform a given ELP $\Pi$ to a non-ground ASP program $\Pi'$ with

\footnote{This can be achieved by introducing auxiliary atoms whenever an atom appears twice in a rule, and add two rules that ensure that the original and auxiliary atom must be equivalent.}

\footnote{We interpret double negation according to (Faber et al. 2011).}
predicates of bounded arity, such that \( \Pi \) is consistent (i.e. it has a candidate world view) iff \( \Pi' \) has at least one answer set. A standard ASP solver can then decide the consistency problem for the ELP \( \Pi \) in a single call, by solving \( \Pi' \).

### 3.1 Reducing ELPs to ASP Programs

The reduction is based on an encoding technique proposed in \cite{Bichler2016}, which uses large, non-ground rules. Given an ELP \( \Pi \), the ASP program \( \Pi' \) will roughly be constructed as follows. \( \Pi' \) contains a guess part that chooses a set of epistemic literals from \( elit(\Pi) \), representing a guess \( \Phi \) for \( \Pi \). Then, the check part verifies that, for \( \Phi \), a candidate world exists. In all, the ASP program \( \Pi' \) consists of five parts:

\[
\Pi' = \Pi'_\text{facts} \cup \Pi'_\text{guess} \cup \Pi'_\text{check}_1 \cup \Pi'_\text{check}_2 \cup \Pi'_\text{check}_3
\]

where the sub-program \( \Pi'_\text{facts} \) is a set of facts representing the ELP \( \Pi \), and \( \Pi'_\text{check}_i \) represents the part of the program that checks Condition \( i \) of Definition 3. We now proceed to the construction of the program \( \Pi' \). Let \( \Pi = (\mathcal{A}, \mathcal{R}) \) be the ELP to reduce from. To ease notation, let \( \mathcal{A} = \{a_1, \ldots, a_n\} \).

The set of facts \( \Pi'_\text{facts} \) represents basic knowledge about the ELP \( \Pi \), plus some auxiliary facts, and is given as:

- \( \text{atom}(a) \), for each atom \( a \in \mathcal{A} \);
- \( elit(\ell) \), for each epistemic literal \( \not\ell \in elit(\Pi) \);
- \( \text{leq}(0, 0), \text{leq}(0, 1), \text{and} \text{leq}(1, 1) \), representing the less or equal relation for Boolean values; and
- \( \text{or}(0, 0, 0), \text{or}(0, 1, 1), \text{or}(1, 0, 1), \text{and} \text{or}(1, 1, 1) \), representing the Boolean relation \( \text{or} \).

**Sub-Program \( \Pi'_\text{guess} \).** This part of the program consists of a single, non-ground rule that guesses a subset of the epistemic literals (stored in predicate \( g \)) as follows:

\[
g(L, 1) \lor g(L, 0) \leftarrow elit(L).
\]

**Shorthands.** Before defining the three check parts of the program, we will introduce some useful shorthands which will be used at several occasions. To this end, we use a context identifier \( \mathcal{C} \).

We first define the following:

\[
H^\mathcal{C}_{val}(A) \equiv v_{\mathcal{C}}(A, 1) \lor v_{\mathcal{C}}(A, 0),
\]

that is, \( H^\mathcal{C}_{val}(A) \) guesses a truth assignment for some variable \( A \) and stores it in relation \( v_{\mathcal{C}} \). We will often use variables \( X = \{X_1, \ldots, X_n\} \) or \( Y = \{Y_1, \ldots, Y_n\} \) to represent a subset \( M \) of \( \mathcal{A} \), where assigning \( X_i \) to 1 characterizes \( a_i \in M \), and \( X_i = 0 \) otherwise. Let

\[
B^\mathcal{C}_{val}(X) \equiv \bigwedge_{a_i \in \mathcal{A}} v_{\mathcal{C}}(a_i, X_i),
\]

that is, \( B^\mathcal{C}_{val}(X) \) extracts the truth assignment from relation \( v_{\mathcal{C}} \) into the variables \( X \) as described above. Finally, for some rule \( r \) in \( \Pi \), we define a formula \( B'_{val}(X, Y, S) \) that checks whether the

\footnote{Note that we use the literal \( \ell \) as an ASP constant.}
rule $r$ is satisfied in the epistemic reduct $\Pi^\Phi$ w.r.t. the guess $\Phi$ encoded in the relation $g$, when the negative body (resp. positive body and head) is evaluated over the set of atoms encoded by $X$ (resp. $Y$). If the rule is satisfied, $B'_{sat}(X,Y,1)$ should hold, and $B'_{sat}(X,Y,0)$ otherwise. This is done as follows. Let $r$ contain the atoms $\{a_{i1}, \ldots, a_{im}\}$ (recall that no atom appears twice in a rule), where $i_1, \ldots, i_m \in \{1, \ldots, n\}$. For ease of notation, we will use a four-ary or relation, which can easily be split into two of our three-ary or atoms using a helper variable $T$:

\[
\text{or}(W,X,Y,Z) \leftarrow \text{or}(W,X,T), \text{or}(T,Y,Z).
\]

The following is the central building block of our reduction:

\[
B'_{sat}(X,Y,R_m) \equiv R_0 = 0, \bigwedge_{a_j \in H(r)} \text{or}(R_{j-1},Y_j,R_j),
\]

\[
\bigwedge_{a_j \in B(r)} \text{or}(R_{j-1},1-Y_j,R_j), \bigwedge_{\neg a_j \in B(r)} \text{or}(R_{j-1},X_j,R_j),
\]

\[
\bigwedge_{\text{not} a_j \in B(r)} g(a_j,N_j), \text{or}(N_j,1-X_j,T_j), \text{or}(R_{j-1},1-T_j,R_j),
\]

\[
\bigwedge_{\text{not} \neg a_j \in B(r)} g(\neg a_j,N_j), \text{or}(N_j,Y_j,T_j), \text{or}(R_{j-1},1-T_j,R_j),
\]

\[
\bigwedge_{\text{not} \neg a_j \in B(r)} g(a_j,N_j), \text{or}(R_{j-1},1-Y_j,R_j),
\]

\[
\bigwedge_{\text{not} \neg a_j \in B(r)} g(\neg a_j,N_j), \text{or}(R_{j-1},N_j,X_j,R_j).
\]

For a rule $r$, each big conjunction in the above formula encodes a reason for $r$ to be satisfied. For example, the fifth line encodes the fact that rule $r$ is true if the disjunct $\neg \text{not} a_j$ is not satisfied, that is, if the epistemic literal $\text{not} a_j$ is part of the guess $\Phi$, or the atom $a_j$ is false (represented by $1-Y_j$). Each disjunct of rule $r$ is evaluated in this way, and the results are connected via the or relation (with the result of the first $i$ disjuncts stored in variable $R_i$). Therefore, $R_m$ will be 1 if $r$ is satisfied, and 0 otherwise, as desired (recall that $r$ has $m$ disjuncts). The following example illustrates how this shorthand is constructed for a concrete input program.

**Example 5**

Recall program $\Pi = (\mathcal{A}, \mathcal{R})$ from Example 4. Let $\mathcal{A} = \{a_1, a_2\}$, where $a_1 = p$ and $a_2 = q$. Let rule $r_2 \in \mathcal{R}$ contain the atoms $\{a_{i1}, a_{i2}\}$, where $i_1 = 2$ and $i_2 = 1$. We give the core construct, $B'_{sat}(\cdot, \cdot, \cdot, \cdot)$ for rule $r_2$:

\[
B'_{sat}(X_1,X_2,Y_1,Y_2,R_2) \equiv \text{or}(0,Y_2,R_1), g(p,N_2), \text{or}(N_2,1-X_1,T_2), \text{or}(R_1,1-T_2,R_2).
\]

Finally, we define $B_{sat}(X,Y)$, which makes sure that the variables $Y$ identify a strict subset of the atoms identified by $X$. Let $B_{sat}(X,Y) \equiv$

\[
N_0 = 0, N_n = 1, \bigwedge_{a_j \in \mathcal{A}} \text{leq}(Y_j,X_j), \text{or}(N_{i-1},X_i-Y_i,N_i).
\]

We can now proceed with the remainder of our reduction.
Sub-Program $\Pi'_{\text{check}}$ This part of the program needs to check that, given the guess $\Phi$ made in $\Pi'_{\text{guess}}$, there exists at least one answer set of the epistemic reduct $\Pi^\Phi$, as per Definition 3(2). Therefore, according to Definition 3(2), we need to find a set $M \subseteq \mathcal{A}$, such that (1) $M$ is a model of $\Pi^\Phi$, and (2) there is no proper subset of $M$ that is a model of the GL-reduct $(\Pi^\Phi)^M$. $\Pi'_{\text{check}}$ contains the following rules:

- $H_{\text{val}}^{\text{check}}(A) \leftarrow \text{atom}(A)$;
- $\bot \leftarrow B_{\text{red}}^{\text{check}}(X), B_{\text{sat}}^r(X, X, 0)$, for each $r \in \mathcal{R}$; and
- $\bot \leftarrow B_{\text{red}}^{\text{check}}(X), B_{\text{sat}}^r(X, X, 0)$, for each $r \in \mathcal{R}$.

The first rule guesses a truth assignment for all atoms. The second rule verifies that there is no rule in $\Pi^\Phi$ that is violated by the candidate answer set $M$, represented by the variables $X$, guessed by the first rule. $B_{\text{red}}^{\text{check}}$ checks whether a subset of $M$ is a model of the GL-reduct $(\Pi^\Phi)^M$. To this end, let

$$B_{\text{red}}^{\text{check}} = B_{\text{val}}^{\text{check}}(X), B_{\text{sat}}(X, Y), \bigwedge_{r \in \mathcal{R}} B_{\text{sat}}^r(X, Y, 1).$$

The last big conjunction in $B_{\text{red}}^{\text{check}}$ makes sure that the subset $N \subset M$ identified by the variables $Y$ is indeed a model of every rule in the GL-reduct $(\Pi^\Phi)^M$. This completes $\Pi'_{\text{check}}$.

Sub-Program $\Pi'_{\text{check}}$ This part needs to check that, for every epistemic literal $\text{not}\ell \in \Phi$, the epistemic reduct $\Pi^\Phi$ has some answer set wherein $\ell$ is false. $\Pi'_{\text{check}}$ contains the following rules and facts, for each epistemic literal $\text{not}\ell \in \text{elit}(\Pi)$ (used as the context $\mathcal{C}$ so guesses are independent):

- $H_{\text{val}}^{\ell}(A) \leftarrow \text{atom}(A), g(\ell, 1)$;
- $v_r(a, \eta)$, where $\eta = 1$ if $\ell = \neg a$, or $\eta = 0$ if $\ell = a$;
- $\bot \leftarrow B_{\text{val}}^{\ell}(X), B_{\text{sat}}(X, X, 0)$, for each $r \in \mathcal{R}$; and
- $\bot \leftarrow B_{\text{red}}^{\ell}$.

These rules guess, for each epistemic literal $\text{not}\ell \in \Phi$, a candidate answer set $M$ wherein $\ell$ is false, and then verify that $M$ is indeed an answer set, using the same technique as in $\Pi'_{\text{check}}$. This ensures Condition 2 of Definition 3.

Sub-Program $\Pi'_{\text{check}}$ Finally, this part needs to check that, for every epistemic literal $\text{not}\ell \in \text{elit}(\Pi) \setminus \Phi$, every answer set of $\Phi$ satisfies $\ell$. The construction makes use of the technique of saturation (Eiter and Gottlob 1995):

- $H_{\text{val}}^{\ell}(A) \leftarrow \text{atom}(A)$;
- $v_{\text{check}}^{\ell}(A, 0) \leftarrow \text{sat}, \text{atom}(A)$;
- $v_{\text{check}}^{\ell}(A, 1) \leftarrow \text{sat}, \text{atom}(A)$; and
- $\bot \leftarrow \text{sat}$.

This setup checks that, for every candidate answer set $M$ guessed in the first rule, the atom $\text{sat}$ is derived. Since we are only interested in answer sets, we first check that $M$ is indeed one, using the following rules, similarly to $\Pi'_{\text{check}}$. 

sat ← \[ B^\text{\text{check}}_{\text{val}}(X), B^\text{\text{check}}_{\text{red}}(X,X,0), \] for each \( r \in \mathcal{R} \); and

sat ← \[ B^\text{\text{check}}_{\text{red}}. \]

It now remains to check that in each answer set \( M \) (that is, where \( sat \) has not been derived yet) all epistemic literals \( \text{not} \ell \) are either in \( \Phi \), or otherwise \( \ell \) is true in \( M \). This is done by adding the following rule to \( \Pi'_\text{check} \):

sat ← \[ \bigwedge_{\text{not} \ni a \in \text{elit}(\Pi)} g(a,N_a), v_{\text{check}}(a,X_a), \text{or}(N_a,X_a,1), \]

\[ \bigwedge_{\text{not} \ni \neg a \in \text{elit}(\Pi)} g(\neg a,N_a), v_{\text{check}}(\neg a,X_a^\neg), \text{or}(N_a^\neg,1-X_a^\neg,1). \]

This completes the reduction. We will now show that this reduction indeed accomplishes our goals. The correctness of our reduction can be intuitively seen from the observation that each of the three check parts of the constructed ASP program \( \Pi' \) ensures precisely one of the three conditions that define a candidate world view. Each answer set \( A \) of \( \Pi' \) is a witness for the fact that a guess \( \Phi \subseteq \text{elit}(\Pi) \) encoded in \( A \) indeed gives rise to a candidate world view. The next theorem formally states that our reduction is correct.

**Theorem 6**

Let \( \Pi = (\mathcal{A}, \mathcal{R}) \) be an ELP and let \( \Pi' \) be the ASP program obtained from \( \Pi \) via the above reduction. Then, \( \Pi \) has a candidate world view if and only if \( \Pi' \) has an answer set.

**Proof**

We will begin with the “if” direction. To this end, assume that there is a guess \( \Phi \subseteq \text{elit}(\Pi) \) for \( \Pi \) that gives rise to a candidate world view \( \mathcal{M} = AS(\Pi^\Phi) \). We will show that \( \Pi' \) has an answer set \( M \). Clearly, \( M \) contains all the facts from \( \Pi'_{\text{facts}} \). Furthermore, let \( M \) contain the fact \( g(\ell,1) \) for each epistemic literal \( \text{not} \ell \in \Phi \) and the fact \( g(\ell,0) \) for each epistemic literal \( \text{not} \ell \in \text{elit}(\Pi) \setminus \Phi \). This clearly satisfies sub-program \( \Pi'_{\text{guess}} \).

Now, let \( M' \in \mathcal{M} \) be any answer set of \( \Pi^\Phi \) (such an answer set exists, since, by assumption, \( \mathcal{M} \) is a candidate world view for \( \Phi \) and by Definition 5, \( \mathcal{M} \) is non-empty). Let \( M \) contain the fact \( v_{\text{check}}(a,1) \) for each \( a \in M' \) and the fact \( v_{\text{check}}(a,0) \) for each \( a \in \mathcal{A} \setminus M' \). This satisfies sub-program \( \Pi'_{\text{check}} \) as follows. Clearly, \( M \) satisfies the first line of the sub-program. Since the atoms with relation \( v_{\text{check}} \) encode precisely the answer set \( M' \) of \( \Pi^\Phi \), and since \( M' \) is a model of \( \Pi^\Phi \), also the second line of the sub-program is satisfied, which, by construction, checks that the assignment encoded in relation \( v_{\text{check}} \) satisfies all the rules of \( \Pi^\Phi \). Finally, the third line, by construction, checks that the same assignment is also minimal w.r.t. the GL-reduct \([\Pi^\Phi]^R\). Since \( M' \) is an answer set of \( \Pi^\Phi \), also this line of the sub-program \( \Pi'_{\text{check}} \) is satisfied.

The argument for satisfaction of \( \Pi'_{\text{check}} \) is similar to the one for \( \Pi'_{\text{guess}} \). Since \( \mathcal{M} \) is a candidate world view for guess \( \Phi \), it contains, for each epistemic literal \( \text{not} \ell \in \Phi \), an answer set \( M_\ell \in \mathcal{M} \) such that \( \ell \) is false in \( M_\ell \). Thus, the argument for sub-program \( \Pi'_{\text{check}} \) can be analogously applied for each \( \text{not} \ell \in \Phi \), taking the answer set \( M_\ell \) instead of \( M' \).

Finally, we need to verify that \( M \) also satisfies the rules in \( \Pi'_{\text{check}} \). To this end, let \( M \) contain the facts \( M_{\text{check}} \) consisting of the fact \( sat \), as well as the fact \( v_{\text{check}}(a,b) \) for each \( a \in \mathcal{A} \) and \( b \in \{0,1\} \). It is easy to verify that all the rules in \( \Pi'_{\text{check}} \) are classically satisfied. However, since
the negative literal \(-\text{sat}\) appears in line 4, in order to verify that \(M\) is indeed an answer set, we also need to look at minimality w.r.t. GL-reduct. Since line 4 is removed in the GL-reduct of \(\Pi'_{\text{check}}\), it may be the case that some subset of \(M_{\text{check}}\setminus\{\text{sat}\}\) may indeed satisfy the GL-reduct. However, we will show that every such subset requires \(\text{sat}\) to be true via lines 5, 6, or 7 of \(\Pi'_{\text{check}}\) (i.e., those rules of \(\Pi'_{\text{check}}\) with \(\text{sat}\) in the head), and can therefore not exist. Indeed, every subset of \(M_{\text{check}}\setminus\{\text{sat}\}\) that does not encode an answer set of \(\Pi^{\Phi}\) in relation \(v_{\text{check}}\) derives \(\text{sat}\) via lines 5 or 6, by construction (the argument to see this is analogous to the one for the previous two sub-programs). It remains to show that all other remaining subsets (i.e. the subsets of \(M_{\text{check}}\setminus\{\text{sat}\}\) that encode answer sets of \(\Pi^{\Phi}\)) also derive \(\text{sat}\). However, since every answer set \(M' \in \mathcal{M}\), by Definition 3, has to satisfy precisely the condition encoded by line 7 of \(\Pi'_{\text{check}}\), this is easy to see. We thus have that \(M\), as constructed above, is indeed an answer set of \(\Pi'\).

The “only if” direction can be seen via similar arguments to the above. By construction, any answer set \(M\) of \(\Pi'\) will encode a guess \(\Phi\) for \(\Pi\). Since any such answer set \(M\), to be an answer set, must satisfy the three check sub-programs of \(\Pi'\) in the way described above, and these three check sub-programs, by construction, correspond directly to the three conditions of Definition 3, we have that \(M\) encodes a guess \(\Phi\) for \(\Pi\) that leads to a candidate world view.

As we have seen, our reduction works as intended: the ASP program \(\Pi'\) derived from the input ELP \(\Pi\) has an answer set precisely when \(\Pi\) has a candidate world view. The next interesting observation is that our reduction is, in fact, a polynomial-time reduction, as stated below.

**Theorem 7**
Given an ELP \(\Pi\), the reduction above runs in time \(O(e \cdot n)\), where \(n\) is the size of \(\Pi\) and \(e = |\text{elit}(\Pi)|\), and uses predicates of arity at most three.

**Proof**
Predicates of arity at most three are used if the four-ary \(\text{or}\) relation is not materialized as an actual relation in ASP, but viewed as a shorthand for two connected ternary \(\text{or}\) relations (cf. the paragraph on shorthands of our reduction). The reduction’s runtime (and output size) can be seen to be in \(O(e \cdot n)\) by noting the fact that the construct \(B_{\text{red}}\) is of size linear in \(n\) (it precisely encodes each rule using the \(\text{or}\) predicates). \(B_{\text{red}}\) is then used once for each epistemic literal in \(\Pi\) (cf. \(\Pi'_{\text{check}}\)).

Note that the above theorem shows that our reduction is indeed worst-case optimal as claimed in Section 1: checking consistency of non-ground, fixed-arity ASP programs is \(\Sigma^p_3\)-complete, as is checking world view existence for ELPs.

### 3.2 Using the Reduction in Practice

As we have seen, using the construction in the previous subsection, we can solve the consistency problem for a given ELP via a single call to an ASP solving system. However, when trying this in practice, the performance is less than optimal, mainly for the following reason. At several points in the construction, large non-ground rules are used (i.e. where \(B'_{\text{red}}\) appears in a rule body). As noted in Section 2 these rules need to be grounded, but may contain hundreds or thousands of variables, which need to be replaced by all possible combinations of constants; a hopeless task for ASP groundsers, as the resulting ground program is exponential in the number of variables.
However, as noted in (Bichler et al. 2016b), such large rules can often be decomposed into smaller, more manageable rules, using the lpopt tool (Bichler et al. 2016a). This tool roughly works as follows: (1) compute a rule graph \( G_r \) for each non-ground rule \( r \), where there is a vertex for each variable \( V \) in \( r \), and there is an edge between \( V_1 \) and \( V_2 \), if the two variables appear together in an atom of \( r \); then (2) compute a tree decomposition of \( G_r \) of minimal width; and finally, (3) in a bottom-up manner, output a rule for each node in the tree decomposition. The resulting rules each contain only as many variables as the tree width of \( G_r \).

After this rule decomposition step, grounding now becomes much easier, since the number of variables in each rule is reduced. Note that, since finding optimal tree decompositions is NP-hard, lpopt employs heuristics to find good decompositions.

In our construction, \( B^\text{red}_r \) stands for a long rule body that effectively encodes the entire input ELP \( \Pi \). Each atom \( a_i \) in \( \Pi \) is represented by the two variables \( X_i \) and \( Y_i \). If we represent \( \Pi \) as a graph \( G_{\Pi} \), where each atom \( a_i \) is a vertex, and there is an edge between two atoms if they appear together in a rule in \( \Pi \), then this graph structure can be found (as a minor) in the rule graph of \( B^\text{red}_r \). However, in addition, \( B^\text{red}_r \) also adds a series of \( \text{or}(\cdot, \cdot, \cdot) \) atoms (via \( B_{ss}(X, Y) \)), that introduce additional connections in the rule graph of \( B^\text{red}_r \). These connections may increase the treewidth substantially. In fact, even if \( G_{\Pi} \) has a treewidth of 1, by introducing the additional connections in a bad way, the treewidth may increase arbitrarily: imagine that \( G_{\Pi} \) is a chain, depicted in black in Figure 1 and imagine the \( \text{or}(\cdot, \cdot, \cdot) \)-chain from \( B_{ss}(X, Y) \) is inserted into \( G_{\Pi} \), illustrated in pink. The treewidth now depends on the chain’s length (and thereby on the size of \( \Pi \)), and lpopt can no longer split the rule well.

In the following, we will formalize the problem described above and present an extension to our reduction presented in the previous subsection that will alleviate the problem. First, we define the primal graph of an ELP \( \Pi \), a standard notion in topics of satisfiability, constraint programming, and logic programming; cf. standard textbooks, e.g. (Ebbinghaus and Flum 1995).

**Definition 8**
The primal graph of an ELP \( \Pi = (\mathcal{A}, \mathcal{R}) \) is the graph \( G_{\Pi} = (V, E) \), where \( V = \mathcal{A} \) and there is an edge \((a_i, a_j) \in E \) iff the atoms \( a_i \) and \( a_j \) occur together in a rule in \( \mathcal{R} \).

Then, we define the rule graph for a non-ground ASP rule \( r \):

**Definition 9**
The rule graph of a non-ground ASP rule \( r \) is the graph \( G_r = (V, E) \), such that \( V = \text{var}(r) \), and there is an edge between two variables \( X \) and \( Y \) in \( E \) iff \( X \) and \( Y \) occur together in an atom in \( r \).

From the construction, it is not difficult to see that any rule \( r \) containing \( B^\text{red}_r \) reflects the structure of the input ELP \( \Pi \), or, more formally, the graph \( G_{\Pi} \) is contained (as a minor) in the graph \( G_r \). Thus, by well-known graph-theoretic results, the treewidth of \( G_r \) is at least the treewidth of \( G_{\Pi} \). Since this is an integral part of our construction, we cannot hope for lpopt to split up rule \( r \) any better than the structure of \( \Pi \) allows. However, as noted in the intuitive problem description above, \( G_r \) contains additional connections between variables. These are introduced by the subformula \( B_{ss}(\cdot, \cdot) \) that effectively links all the variables in a rule \( r \) into a chain in \( G_r \).
In the worst case, as illustrated by Figure 1, these additional connections in $G_r$ may increase the treewidth arbitrarily, making it almost impossible for $lpopt$ to split up the rule well. It is therefore important to introduce these additional connections carefully. We will now introduce a more involved construction of $B_{ss}(\cdot,\cdot)$ that preserves the treewidth of $G_{\Pi}$ in $G_r$ (i.e. does not arbitrarily increase it). In this modified version, $B_{ss}(\cdot,\cdot)$ is constructed as follows:

1. First, compute a tree decomposition $\mathcal{T}_{\Pi}$ of $G_{\Pi}$ with minimal width.
2. Secondly, construct $B_{ss}(\cdot,\cdot)$ in a bottom-up (i.e. post-order traversal) fashion along this tree decomposition in the following way, for each node type. To this end, let $\mathcal{A} = \{a_1,\ldots,a_n\}$, and, for a node $t$ of $\mathcal{T}_{\Pi}$, let $\mathcal{X}(t)$ contain the set of atoms $\{a_{i_1},\ldots,a_{i_m}\}$, with $i_j \in \{1,\ldots,n\}$.

   **Leaf Node $t$:** For a leaf node $t$ of $\mathcal{T}_{\Pi}$, let $B_{ss}(\mathbf{X},\mathbf{Y})$ contain the following conjunction of atoms:

   $$N^t_0 = 0\,\land\,N^t = N^t_m\,\land\,\bigwedge_{i_j \in \mathcal{X}(t)} leq(Y_{i_j},X_{i_j}), or(N^t_{i_j-1},X_{i_j} - Y_{i_j},N^t_{j}),$$

   that is, $N^t$ contains 1, if the proper subset condition between $\mathbf{X}$ and $\mathbf{Y}$ is already fulfilled in node $t$, and 0 otherwise.

   **Inner Node $t$:** For an inner node $t$ of $\mathcal{T}_{\Pi}$ with children $t_1,\ldots,t_k$, let $B_{ss}(\mathbf{X},\mathbf{Y})$ contain the same conjunction as for a leaf node, but where the equality atom $N^t = N^t_m$ is replaced by the following disjunction:

   $$or(N^t_m,N^t_1,\ldots,N^t_k,N^t),$$

   where the $k+2$-ary $or$ atom can be split into 3-ary $or$ atoms in the same way as with the 4-ary $or$ atom in our main construction. Intuitively, we now have that $N^t$ is set to 1 if the proper subset condition is already fulfilled somewhere in the subtree rooted at $t$.

   **Root Node $t_{root}$:** Finally, for the root node $t_{root}$, we add the same conjunction of atoms to $B_{ss}(\mathbf{X},\mathbf{Y})$ as for an inner node, but, in addition, we add the final condition $N^{t_{root}} = 1$, that makes sure that, at the root node, the proper subset condition is fulfilled.

If constructed in the way described above, it is not difficult to see that $B_{ss}(\mathbf{X},\mathbf{Y})$ still ensures the same condition as in our original construction from Section 3.1, namely, that the variables $\mathbf{Y}$ identify a proper subset of the atoms identified by the variables $\mathbf{X}$. However, the treewidth of a rule containing $B_{ss}(\cdot,\cdot)$ is now not increased arbitrarily. In fact, it can be verified that the treewidth of $G_{\Pi}$ is preserved up to a constant additive factor, for any rule containing $B_{ss}(\cdot,\cdot)$ when using the alternative construction for $B_{ss}(\cdot,\cdot)$ provided above. In practice, this means that $lpopt$ is able to split the rule up as well as possible; that is, as well as the structure of $\Pi$ allows.

### 3.3 Discussion and Related Work

As we have seen, the reduction proposed above allows us to solve ELPs via a single call to an ASP solving system. However, our encoding also has several other interesting practical properties, which make it very flexible for use with, for example, different ASP semantics, or harder problems. A brief discussion follows.

**Other ASP Semantics.** Apart from the original semantics for ASP (called stable model semantics, (Gelfond and Lifschitz 1988; Gelfond and Lifschitz 1991)), several different semantics have been proposed that investigate how to interpret more advanced constructs in ASP, like double negation, aggregates, optimization, etc (Lifschitz et al. 1999; Pearce 2006; Pelov et al. 2007).
Epistemic reducts may contain double negation, and we have opted to use the FLP semantics by Faber et al. (2011), as used by Shen and Eiter (2011), to interpret this. The actual interpretation of double negation is encoded in the $B^r_{sat}(\cdot,\cdot)$ shorthand defined in our reduction. This construction is very flexible and can easily be modified to use different ASP semantics (e.g. (Lifschitz et al. 1999)).

**Enumeration of World Views.** Modern ASP systems like clasp (Gebser et al. 2012) contain several useful features not included in the ASP base language. One such feature is an advanced implementation of projection, as presented in (Gebser et al. 2009): given a set of atoms (or relations), the solver will output answer sets where all other atoms are projected away, and will also guarantee that there are no repetitions (even if multiple answer sets with the same assignment on the projected atoms exist), while still maintaining efficiency. This can be used to enumerate candidate world views by projecting away all relations in our encoding, except for $g(\cdot)$ and $\nu_{check}^{\Pi}(\cdot)$. When enumerating all projected answer sets in this way, our encoding yields all guesses together with their candidate world views (when grouped by $g(\cdot)$).

**Comparison to Related Work.** Classic ELP solvers generally work by first establishing a candidate epistemic guess $\Phi$ and then use an answer set solver to verify that the epistemic guess indeed yields an epistemic reduct whose answer sets form a candidate world view of the original ELP w.r.t. $\Phi$. Different approaches are used to find promising epistemic guesses, and also to verify that they lead to candidate world views, but, generally, these systems have in common that an underlying ASP solver is used, and called multiple times, to solve the ELP. Notable recent ELP solvers include that follow this approach include EP-ASP (Son et al. 2017), GISolver (Zhang et al. 2015) and a later, probabilistic, variant called PelpSolver, and ELPSolve (Kahl et al. 2016). A comprehensive survey of recent ELP solving systems (including the one presented in the present paper) can be found in (Leclerc and Kahl 2018).

We are not aware of another single-shot ELP solver that only needs to call an underlying ASP system once. However, the idea of our approach is similar to the one used by Bichler et al. (2016b), where a single-shot ASP encoding for disjunctive ASP, which is rewritten into non-ground normal ASP with fixed arity, is presented. That is, a solving system for normal ASP would be able to solve a disjunctive ASP program in a single call. However, this approach was not implemented, and only presented as an example to show how long non-ground rules with fixed arity can be used to solve hard problems. In order to use such encodings (including our own presented herein), (Bichler et al. 2016b) make use of rule decomposition, where large non-ground ASP rules are split up into smaller parts based on tree decompositions (Morak and Woltran 2012). This rule decomposition approach has been implemented as a stand-alone tool called lpopt (Bichler et al. 2016a), but has recently also been integrated into ASP solving systems like I-DLV (Calimeri et al. 2017).

### 4 Application: QBF Solving

In this section, we illustrate the power of ELPs by illustrating a way to solve QBF formulas with at most three quantifier alternations (3-QBF) by encoding them as ELPs. This provides an alternative way to show the $\Sigma^P_3$ lower bound for ELP consistency, but relies on the existence of a reduction from so-called restricted 3-QBF formulas. (Shen and Eiter 2016) present such a reduction from restricted 3-QBF formulas to ELP world view existence. Our aim is to generalize...
this by presenting a reduction from (general) 3-QBF formulas to restricted 3-QBF formulas. We will use this result to benchmark our ELP solver presented in Section 5. Let us begin by first recalling the definition of such 3-QBF formulas.

**Definition 10**
A 3,∃-QBF in CNF form (or QBF, for short) is a formula of the form

\[ \exists X \forall Y \exists Z \varphi \]

where \( X, Y, \) and \( Z \) are sets (or sequences) of distinct (propositional) atoms (also called variables), and \( \varphi = \bigwedge_{i=1}^{k} C_i \) is a CNF over the atoms \( X \cup Y \cup Z, \) i.e. \( C_i = \bigvee_{j=0}^{k_i} L_{i,j} \) is a clause of size \( k_i \) and \( L_{i,j} \) is either an atom \( a \) or its negation \( \neg a \).

W.l.o.g. we can assume that the clause size \( k_i = 3 \) for each \( 0 < i \leq k \), that is, that \( \varphi \) is given in 3-CNF form, where each clause has at most three elements. In (Shen and Eiter 2016), the authors make use of a version of QBFs called restricted QBFs. These are QBFs that evaluate to true under all interpretations of the existentially quantified variables if all universally quantified variables are replaced by \( \top \) (i.e. if they are set to true).

**Definition 11**
A restricted QBF is a QBF where \( \varphi[y/\top \mid y \in Y] \) is a tautology.

The hardness proof of Theorem 5 of (Shen and Eiter 2016) is a reduction from the validity problem of restricted QBFs to the consistency problem of epistemic logic programs. While the actual construction of the reduction is not needed for our purposes in this section, we nevertheless report it here, for completeness sake.

**Proposition 12**
(Shen and Eiter 2016) Proof of Theorem 5) Let \( \Theta = \exists X \forall Y \exists Z \varphi \) be a restricted QBF. Then, there exists an ELP \( \Pi \) such that \( \Pi \) has a (candidate) world view iff \( \Theta \) is satisfiable.

**Proof**
The ELP \( \Pi \) consists of the following rules:

- For each variable \( X \in X \):
  \[ X \leftarrow \neg X, \]

- For each variable \( Y \in Y \):
  \[ Y \leftarrow \neg Y, \]

- For each variable \( Z \in Z \):
  \[ Z \lor \neg Z. \]

- For each clause \( C_i, 0 < i \leq k \):
  \[ U \leftarrow L_{i,1}^+, L_{i,2}^+, L_{i,3}^+, \]

where \( ^* \) is an operator that converts a positive literal \( W \) into \( \overline{W} \), and a negative literal \( \neg W \) into \( W \).

- And, finally, the rule
  \[ V \leftarrow \neg V, \neg U. \]

ELP \( \Pi \) has a (candidate) world view iff \( \exists X \forall Y \exists Z \varphi \) is satisfiable (Shen and Eiter 2016).
We now show a more general reduction that also works for the non-restricted case. To this end, we will combine the (Shen and Eiter 2016) reduction with our own reduction of QBF formulas to restricted QBF formulas. To achieve our goal, we are going to introduce one new atom \( v_i \) in each clause \( C_i \) and \( \forall \)-quantify these new atoms together with the \( Y \) atoms.

**Definition 13**

Given a QBF \( \Theta = \exists X \forall Y \exists Z \varphi \) with \( \varphi \) being constructed as in Definition 10 let its extension, denoted \( \Theta^\uparrow \), be the QBF

\[
\Theta^\uparrow = \exists X \forall (Y \cup V) \exists Z \varphi'
\]

where

\[
\varphi' = \bigwedge_{i=1}^k \left( v_i \lor \bigvee_{j=0}^{k_i} L_{i,j} \right)
\]

and \( V = \{v_1, \ldots, v_k\} \) is a list of fresh atoms.

It is easy to see that any extension \( \Theta^\uparrow \) of a QBF \( \Theta \) is a restricted QBF.

**Proposition 14**

Let \( \Theta \) be a QBF. Its extension \( \Theta^\uparrow \) is a restricted QBF.

We will now show that validity-equivalence between a QBF and its extension is preserved. For the proof, we establish the following terminology: given a subset of atoms \( \sigma \subseteq S \), we define its out-set \( \overline{\sigma} = \{-a \mid a \in S \setminus \sigma\} \) and its literal-set \( \overline{\sigma} = \sigma \cup \overline{\sigma} \).

**Proposition 15**

\( \Theta \) and \( \Theta^\uparrow \) are validity-equivalent.

**Proof**

(\( \Rightarrow \)) Assume \( \Theta \) is valid, i.e. there exists an interpretation \( \sigma_X \subseteq X \), such that for any interpretation \( \sigma_Y \subseteq Y \) there exists an interpretation \( \sigma_Z \subseteq Z \) such that \((\sigma_X \cup \sigma_Y \cup \sigma_Z) \cap C_i \neq \emptyset \) for all \( i \in \{1, \ldots, k\} \). By monotonicity of non-emptiness of set intersections, also \((\sigma_X \cup \sigma_Y \cup \sigma_V \cup \sigma_Z) \cap \{C_i \cup \{v_i\}\} \neq \emptyset \) for any interpretation \( \sigma_Y \subseteq V \) of a list of new atoms \( V = \{v_1, \ldots, v_k\} \). This is proof of the validity of \( \Theta^\uparrow \).

(\( \Leftarrow \)) For the other direction, assume \( \Theta^\uparrow \) is valid, i.e. there exists an interpretation \( \sigma_X \subseteq X \) such that for any interpretations \( \sigma_Y \subseteq Y \) and \( \sigma_V \subseteq V \) there exists an interpretation \( \sigma_Z \subseteq Z \) such that \((\sigma_X \cup \sigma_Y \cup \sigma_V \cup \sigma_Z) \cap \{C_i \cup \{v_i\}\} \neq \emptyset \) for all \( i \in \{1, \ldots, k\} \). By setting \( \sigma_V = \emptyset \), we especially get that there exists an interpretation \( \sigma_X \subseteq X \) such that for any interpretation \( \sigma_Y \subseteq Y \) there exists an interpretation \( \sigma_Z \subseteq Z \) such that \((\sigma_X \cup \sigma_Y \cup \{-v_1, \ldots, -v_k\} \cup \sigma_Z) \cap \{C_i \cup \{v_i\}\} \neq \emptyset \) for all \( i \in \{1, \ldots, k\} \). Since the only literals containing a \( v_i \) variable on the left-hand side of the \( \cap \) are negative and the only ones on the right-hand side are positive, we get \((\sigma_X \cup \sigma_Y \cup \sigma_Z) \cap C_i \neq \emptyset \) for all \( i \in \{1, \ldots, k\} \), which establishes validity of \( \Theta \).

Now it is straightforward to generalize the reduction from (Shen and Eiter 2016): let \( \Theta \) be a QBF and apply the reduction from (Shen and Eiter 2016) to the restricted QBF \( \Theta^\uparrow \).

**Theorem 16**

Let \( \Theta \) be a QBF. Let the ELP \( \Pi_\Theta \) be obtained by applying the reduction by Shen and Eiter (2016) to the restricted QBF \( \Theta^\uparrow \). It holds that \( \Theta \) is valid iff \( \Pi_\Theta \) is consistent, that is, \( \Pi_\Theta \) has at least one candidate world view.

Correctness of this theorem follows from immediately from Proposition 12 and Proposition 15.
5 The selp System

We implemented the reduction in Section 3 as part of the single shot ELP solving toolbox selp, available at https://dbai.tuwien.ac.at/proj/selp. In addition, the toolbox features a grounder for ELPs and a grouping script which groups answer sets of the reduction into candidate world views (allowing for enumeration). The tools are implemented in python and depend on the parser generator LARK 4, the rule decomposition tool lpopt (Bichler et al. 2016a), the tree decomposition tool htd_main (Abseher et al. 2017), and the plain ASP grounder gringo (Gebser et al. 2011).

Input Formats. The selp solver reads the EASP-not file format, which is a restriction of the ASP input language of gringo to plain ground logic programs as defined in Section 2, extended with the $\text{not}$ operator for epistemic negation. This allows us to encode ELPs as defined in Section 2. selp also supports EASP-KM, defined by adding the operators $\text{K}$ and $\text{M}$ instead of $\text{not}$. By allowing variables in body elements, both formats also have a non-ground version. The toolbox offers scripts to translate between the two formats.

Toolbox. We briefly present the main building blocks of selp.

easpGrounder.py takes as input a non-ground EASP-not program and outputs its equivalent ground form by rewriting it into an ASP program that the gringo grounder can understand and ground. This is done by encoding epistemic negation as predicate names and, after grounding, re-introducing epistemic negation where a placeholder predicate appears. Our grounding component, easpGrounder.py, supports arithmetics and the sorts format (Kahl et al. 2015) as input.

easp2asp.py is selp’s key component. It takes a ground EASP-not program, performs the reduction given in Section 3.1 (with some modifications to account for the extended language of ASP used by today’s ASP systems and some straightforward optimizations), also adhering to the practical considerations presented in Section 3.2, and it finally outputs the resulting non-ground logic program in the syntax of gringo. Optionally, additional clasp directives are generated to allow for enumeration; cf. Section 3.3. For concrete implementation details, please consult the freely available source code at https://dbai.tuwien.ac.at/proj/selp.

groupWorldViews.py takes clasp’s output in JSON format, groups the answer sets into candidate world views according to their $g(\cdot)$ atoms, and outputs them in a human-readable format.

Usage. As a typical use case, suppose the file problem.easp contains a non-ground ELP encoding of a problem of interest and the file instance.easp contains a problem instance. In order to output all candidate world views, one would use the following command (flags --pas and --project enable projection of answer sets onto relevant predicates only. --n0 tells clasp to compute all answer sets, and --outf=2 to print in JSON format. lpopt is used to decompose long rule bodies. The --sat-prepro=3 flag is recommended by lpopt):

```bash
cat problem.easp instance.easp |
easpGrounder.py -sELP | easp2asp.py -pas |
lpopt | gringo | clasp --n0 --outf=2 --project --sat-prepro=3 |
groupWorldViews.py
```

4 Available here: https://github.com/erezsh/lark
6 Experimental Evaluation

We tested our system _selp_ against the state-of-the-art ELP solver, _EP-ASP_ (Son et al. 2017), using three test sets. For every test set, we measured the time it took to solve the consistency problem. For _selp_, _clasp_ was stopped after finding the first answer set. For _EP-ASP_, search was terminated after finding the first candidate world view.\(^5\) Note that a single answer set of the _selp_ system is enough to establish consistency of an input ELP. _EP-ASP_ needs to compute a full candidate world view to be able to prove consistency.

Experiments were run on a 2.1GHz AMD Opteron 6272 system with 224 GB of memory. Each process was assigned a maximum of 14 GB of RAM. For _EP-ASP_, we used the required _clingo_ 4.5.3, since newer versions are incompatible with the solver. For _selp_, we used _clingo_ 5.2.2, _htd_main_ 1.2.0, and _lpopt_ 2.2. The time it took _EP-ASP_ to rewrite the input to its own internal format was not measured. _EP-ASP_ was called with the preprocessing option for brave and cautious consequences on, since it always ran faster this way. The _selp_ time is the sum of running times of its components.

Benchmark Instances. We used three types of benchmarks, two coming from the ELP literature and one from the QSAT domain that contains structures of low treewidth.\(^6\)

Scholarship Eligibility (SE). This set of non-ground ELP programs is shipped together with _EP-ASP_. Its instances encode the scholarship eligibility problem for 1 to 25 students.

Yale Shooting (YS). This test set consists of 25 non-ground ELP programs encoding a simple version of the Yale Shooting Problem, a conformant planning problem: the only uncertainty is whether the gun is initially loaded or not, and the only fluents are the gun’s load state and whether the turkey is alive. Instances differ in the time horizon. We follow the ELP encoding from (Kahl et al. 2015).

\(^5\) Note that to have a fair comparison we disabled the subset-maximality check on the guess that _EP-ASP_ performs by default.

\(^6\) Benchmark archive: [https://dbai.tuwien.ac.at/proj/selp](https://dbai.tuwien.ac.at/proj/selp)
Tree QBFs (TQ). The hardness proof for ELP consistency (Shen and Eiter 2016) relies on a reduction from the validity problem for restricted quantified boolean formulas with three quantifier blocks (i.e. 3-QBFs), which can be generalized to arbitrary 3-QBFs as discussed in Section 3.5. We apply this extended reduction to the 14 “Tree” instances of QBFEVAL’16 (Pulina 2016), available at [link](http://www.qbflib.org/family_detail.php?idFamily=56) splitting each instance’s variables into three random quantifier blocks.

Results. The results for the first two sets are shown in Figure 2. selp solves all instances from (SE) within 30 seconds, while EP-ASP only solves 17 within the time limit of 8 hours. For (YS), on the other hand, selp is able to solve only 6 instances within the time limit of 30 minutes, whereas EP-ASP can solve 17. Finally, for (TQ), selp can solve 6 of the 14 instances within the time limit of 12 hours, whereas EP-ASP was unable to solve any instances at all.

These results confirm that selp is highly competitive on well-structured problems: in the (SE) instances, knowledge about students is not interrelated, and hence the graph $G_E$ of the ground ELP consists of one component for each student, thus having constant treewidth. The (TQ) instances keep their constant treewidth thanks to the fact that both the reduction from QBF to ELP (cf. Section 4) and from ELP to non-ground ASP (cf. Section 3.2) preserve the low treewidth of the original QBF instance. Different from selp, EP-ASP is not designed to exploit such structural information of ELPs and, consequently, performs worse than selp in these benchmarks. On the other hand, (YS) contains instances of high treewidth, even though it does not depend on the horizon. EP-ASP is therefore able to outperform selp on such instances. A similar observation can be made for the “Bomb in the Toilet” problem, as benchmarked in (Son et al. 2017), which inherently contains a huge clique structure. selp is not designed to solve such instances, and is therefore most suited to solve ELPs of low treewidth, where it is able to efficiently exploit the problem structure.

7 Conclusions

In this paper, we have seen that ELPs can be encoded into ASP programs using long non-ground rules, such that a single call to an ASP solver is sufficient to evaluate them. A prototype ELP solver implementation, selp, performs particularly well on problems whose internal structure is of low treewidth. A combined solver that either calls selp or another state-of-the-art solver based on the treewidth of the input may therefore lead to even better overall performance.

Another topic for future work is that, under the FLP semantics, checking whether a given atom $a$ is true in all candidate world views with a subset-maximal guess $\Phi$ is known to be $\Sigma_2^P$-complete (Shen and Eiter 2016). To solve this problem, advanced optimization features of state-of-the-art ASP solvers could allow us to encode this subset-maximality condition, while leaving the core of our encoding unchanged.

Finally, an interesting question is program optimization. Recently, a practical, easily applicable notion of strong equivalence for ELPs has been defined (Faber et al. 2019). It would be interesting to investigate if and how parts of ELPs can be replaced in such a way that the solving performance of selp improves, seeing that selp is sensitive to treewidth. This could lead to an encoding technique for ELPs that tries to minimize the treewidth, similar to the class of connection-guarded ASP (Bliem et al. 2017), which was recently proposed in order to write ASP programs in such a way as to keep the treewidth of the resulting ground program low.
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