

An ASP Framework for the Refinement of Authorization and Obligation Policies

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Abstract

This paper introduces a framework for assisting policy authors in refining and improving their policies. In particular, we focus on authorization and obligation policies that can be encoded in Gelfond and Lobo's $\mathcal{AOP}\mathcal{L}$ language for policy specification. We propose a framework that detects the statements that make a policy inconsistent, underspecified, or ambiguous with respect to an action being executed in a given state. We also give attention to issues that arise at the intersection of authorization and obligation policies, for instance when the policy requires an unauthorized action to be executed. The framework is encoded in Answer Set Programming.

KEYWORDS: policy, authorizations and obligations, dynamic domains, ASP

1 Introduction

This paper introduces a framework for assisting policy authors in refining and improving the policies they elaborate. Here, by a *policy* we mean *a collection of statements that describe the permissions and obligations related to an agent's actions*.

In particular, we focus on authorization and obligation policies that can be encoded in the policy specification language $\mathcal{AOP}\mathcal{L}$ by Gelfond and Lobo (2008). $\mathcal{AOP}\mathcal{L}$ allows an author to specify policies for an autonomous agent acting in a changing environment. A description of the dynamic domain in terms of sorts of the domain, relevant fluents, and actions is assumed to be available to the policy writer. Policy rules of $\mathcal{AOP}\mathcal{L}$ may be of two kinds: *authorization* rules specifying what actions are permitted/ not permitted and in which situations, and *obligation* rules indicating what actions an agent must perform or not perform under certain conditions. Rules can either be strict or defeasible, and preferences between defeasible rules can be set by the policy author. The semantics of $\mathcal{AOP}\mathcal{L}$ is defined via a translation into Answer Set Programming (ASP) (Gelfond and Lifschitz 1991). Gelfond and Lobo define policy properties such as consistency and categoricity. However, there is a gap in analyzing what happens at the intersection between authorization and obligation policies, for instance when a policy requires an unauthorized action to be executed, which is called a *modality conflict* by Craven *et al.* (2009).

We propose a framework that detects the rules that make a policy inconsistent, underspecified, or ambiguous with respect to an action and a given state. The goal is to notify the policy author about the natural language statements in the policy that may be causing an issue and explain why that is the case for the particular action and state.

Given rapid advancements in AI in the past years, the importance of setting and enforcing policies on intelligent agents has become paramount. At the same time, policy specifications can become large and intricate. Thus, assisting policy authors and knowledge engineers with policy refinement by automatically detecting issues in a provably correct way and highlighting conflicting policy statements is of great importance.

The contributions of our work are as follows:

- We define a new translation of \mathcal{AOPL} policies into ASP by reifying policy rules.
- We formally define issues that may arise in \mathcal{AOPL} policies and describe how to detect the causing policy statements, using the reified ASP translation.
- We define means for explaining the root causes for issues like inconsistency, underspecification, ambiguity, and modality conflicts.

In what follows, we provide a short description of language \mathcal{AOPL} in Section 2 and give a motivating example in Section 3. We describe our new translation of \mathcal{AOPL} policies into ASP in Section 4 and introduce our framework in Section 5. We discuss related work in Section 6 and end with conclusions and future work.

2 Background: Language \mathcal{AOPL}

Let us now briefly present the \mathcal{AOPL} language. We direct the unfamiliar reader to outside resources on ASP (Gelfond and Lifschitz 1991; Marek and Truszczyński 1999) and action language \mathcal{AL}_d (Gelfond and Incezan 2013; Gelfond and Kahl 2014), which are also relevant to this work.

Gelfond and Lobo (2008)¹ introduced the Authorization and Obligation Policy Language \mathcal{AOPL} for specifying policies for an intelligent agent acting in a dynamic environment. A policy is a collection of authorization and obligation statements, which we simply call authorizations and obligations, respectively. An *authorization* indicates whether an agent's action is permitted or not, and under which conditions. An *obligation* describes whether an agent is obligated or not obligated to perform a specific action under certain conditions. An \mathcal{AOPL} policy works in conjunction with a dynamic system description of the agent's environment written in an action language such as \mathcal{AL}_d . The signature of the dynamic system description includes predicates denoting *sorts* for the elements in the domain; *fluents* (i.e. properties of the domain that may be changed by actions); and *actions*. As in \mathcal{AL}_d , we consider dynamic systems that can be represented by a directed graph, called a *transition diagram*, containing a finite number of nodes representing physically possible states of the dynamic domain. A state is a complete and consistent set of fluent literals. Arcs in the transition diagram are labeled by action atoms (shortly *actions*) that take the system from one state to another. Actions can be elementary or compound, where a compound action is a set of elementary actions executed simultaneously.

¹ Available at <https://www.depts.ttu.edu/cs/research/documents/44.pdf>.

The signature of an $\mathcal{AOP}\mathcal{L}$ policy includes the signature of the associated dynamic system and additional predicates *permitted* for authorizations, *obl* for obligations, and *prefer* for establishing preferences between authorizations or obligations. A *prefer* atom is created from the predicate *prefer*; similarly for *permitted* and *obl* atoms.

Definition 1 (Policy)

An $\mathcal{AOP}\mathcal{L}$ policy \mathcal{P} is a finite collection of statements of the form:

$$\begin{array}{lll} & \textit{permitted}(e) & \mathbf{if} \textit{ cond} & (1a) \\ & \neg\textit{permitted}(e) & \mathbf{if} \textit{ cond} & (1b) \\ & \textit{obl}(h) & \mathbf{if} \textit{ cond} & (1c) \\ & \neg\textit{obl}(h) & \mathbf{if} \textit{ cond} & (1d) \\ d : \mathbf{normally} & \textit{permitted}(e) & \mathbf{if} \textit{ cond} & (1e) \\ d : \mathbf{normally} & \neg\textit{permitted}(e) & \mathbf{if} \textit{ cond} & (1f) \\ d : \mathbf{normally} & \textit{obl}(h) & \mathbf{if} \textit{ cond} & (1g) \\ d : \mathbf{normally} & \neg\textit{obl}(h) & \mathbf{if} \textit{ cond} & (1h) \\ & \textit{prefer}(d_i, d_j) & & (1i) \end{array}$$

where e is an elementary action; h is a happening (i.e. an elementary action or its negation²); *cond* is a (possibly empty) collection of atoms of the signature, except for atoms containing the predicate *prefer*; d appearing in (1e)–(1h) denotes a defeasible rule label; and d_i, d_j in (1i) refer to two distinct rule labels from \mathcal{P} . Rules (1a)–(1d) encode *strict* policy statements, while rules (1e)–(1h) encode *defeasible* statements (i.e. statements that may have exceptions). Rule (1i) captures *priorities* between defeasible statements.

In deontic terms, rules (1a) and (1e) denote *permissions*; rules (1b) and (1f) denote *prohibitions*; rules (1c) and (1g) denote *obligations*; and rules (1d) and (1h) denote *dispensations*.

The **semantics** of an $\mathcal{AOP}\mathcal{L}$ policy determine a mapping $\mathbf{P}(\sigma)$ from states of a transition diagram \mathcal{T} into a collection of *permitted* and *obl* literals. To formally describe the semantics of $\mathcal{AOP}\mathcal{L}$, a translation of a policy and transition diagram into ASP is defined.

Definition 2 (ASP Translation of a Policy and State)

The translation lp is defined as:

- If x is a fluent literal, action literal, *permitted*, or *obl* literal, then $lp(x) =_{def} x$.
- If L is a set of literals, then $lp(L) =_{def} \{lp(l) : l \in L\}$
- If $r = "l \mathbf{if} \textit{ cond}"$ is a strict rule like the ones in (1a)–(1d), then $lp(r) =_{def} lp(l) \leftarrow lp(\textit{cond})$
- If r is a defeasible rule like (1e) or (1f), or a preference rule "*prefer*(d_i, d_j)" like the one in (1i), then $lp(r)$ is obtained using standard ASP techniques for encoding defaults, as shown in equations (2a), (2b), and (2c) respectively:

$$\textit{permitted}(e) \leftarrow lp(\textit{cond}), \text{not } ab(d), \text{not } \neg\textit{permitted}(e) \quad (2a)$$

$$\neg\textit{permitted}(e) \leftarrow lp(\textit{cond}), \text{not } ab(d), \text{not } \textit{permitted}(e) \quad (2b)$$

$$ab(d_j) \leftarrow lp(\textit{cond}_i) \quad (2c)$$

² If $obl(\neg e)$ is true, then the agent must not execute e .

where $cond_i$ is the condition of d_i . Similarly for defeasible obligations (1g) and (1h).

- If \mathcal{P} is a policy, then $lp(\mathcal{P}) =_{def} \{lp(st) : st \in \mathcal{P}\}$.
- If \mathcal{P} is a policy and σ is a state of the (transition diagram associated with the) dynamic system description \mathcal{T} ,

$$lp(\mathcal{P}, \sigma) =_{def} lp(\mathcal{P}) \cup lp(\sigma)$$

Properties of an $\mathcal{AOP}\mathcal{L}$ policy \mathcal{P} are defined in terms of the answer sets of the logic program $lp(\mathcal{P}, \sigma)$ expanded with appropriate rules.

The following definitions by Gelfond and Lobo are relevant to our work (original definition numbers in parenthesis). In what follows a denotes a (possibly) compound action (i.e. a set of simultaneously executed elementary actions), while e refers to an elementary action. An event $\langle \sigma, a \rangle$ is a pair consisting of a state σ and a (possibly) compound action a executed in σ .³

Definition 3 (Consistency – Def. 3)

A policy \mathcal{P} for \mathcal{T} is called *consistent* if for every state σ of \mathcal{T} , the logic program $lp(\mathcal{P}, \sigma)$ is consistent, that is, it has an answer set.

Definition 4 (Policy Compliance for Authorizations – Defs. 4 and 5)

- An event $\langle \sigma, a \rangle$ is *strongly compliant* with authorization policy \mathcal{P} if for every $e \in a$ we have that $permitted(e) \in \mathbf{P}(\sigma)$ (i.e. the logic program $lp(\mathcal{P}, \sigma)$ entails $permitted(e)$).
- An event $\langle \sigma, a \rangle$ is *weakly compliant* with authorization policy \mathcal{P} if for every $e \in a$ we have that $\neg permitted(e) \notin \mathbf{P}(\sigma)$ (i.e. the logic program $lp(\mathcal{P}, \sigma)$ does not entail $\neg permitted(e)$).
- An event $\langle \sigma, a \rangle$ is *non-compliant* with authorization policy \mathcal{P} if for every $e \in a$ we have that $\neg permitted(e) \in \mathbf{P}(\sigma)$ (i.e. the logic program $lp(\mathcal{P}, \sigma)$ entails $\neg permitted(e)$).

Definition 5 (Policy Compliance for Obligations – Def. 9)

An event $\langle \sigma, a \rangle$ is *compliant* with obligation policy \mathcal{P} if

- For every $obl(e) \in \mathbf{P}(\sigma)$ we have that $e \in a$, and
- For every $obl(\neg e) \in \mathbf{P}(\sigma)$ we have that $e \notin a$.

Definition 6 (Categoricity – Def. 6)

A policy \mathcal{P} for \mathcal{T} is called *categorical* if for every state σ of \mathcal{T} the logic program $lp(\mathcal{P}, \sigma)$ is categorical, that is, has exactly one answer set.

Note that $\mathcal{AOP}\mathcal{L}$ does not discuss interactions between authorizations and obligations referring to the same action, for instance situations when both $obl(e)$ and $\neg permitted(e)$ are part of the answer set of $lp(\mathcal{P}, \sigma)$ for some state σ .

3 Motivating example

To illustrate the policy refinement process that we want to facilitate, let's consider an example provided by Gelfond and Lobo, expanded with an additional rule (4):

³ In policy analysis, we want to encompass all possible events, that is, pairs consisting of a physically possible state σ and physically executable action a in σ .

Example 1 (Authorization Policy Example)

- (1) A military officer is not allowed to command a mission they authorized.
- (2) A colonel is allowed to command a mission they authorized.
- (3) A military observer can never authorize a mission.
- (4) A military officer must command a mission if ordered by their superior to do so.

Before discussing the encoding of this policy, let us assume that the description of this domain includes actions $assume_comm(C, M)$ and $authorize_comm(C, M)$; fluents $authorized(C, M)$ and $ordered_by_superior(C, M)$; and sorts $colonel(C)$ and $observer(C)$, where C is a commander and M is a mission.

In the English description of the policy in Example 1, note that statements (1) and (2) are phrased as strict rules and thus an automated translation process into $\mathcal{AOP}\mathcal{L}$ would produce the policy:

$$\neg permitted(assume_comm(C, M)) \text{ if } authorized(C, M) \quad (3a)$$

$$permitted(assume_comm(C, M)) \text{ if } colonel(C) \quad (3b)$$

$$\neg permitted(authorize_comm(C, M)) \text{ if } observer(C) \quad (3c)$$

$$obl(assume_comm(C, M)) \text{ if } ordered_by_superior(C, M) \quad (3d)$$

Such a policy is inconsistent in a state in which $authorized(c, m)$ and $colonel(c)$ both hold, due to rules (3a) and (3b). Gelfond and Lobo indicate that “[s]ince the nature of the first two authorization policy statements of our example are contradictory we naturally assume them to be defeasible” and replace the encoding in rules (3a) and (3b) with

$$d_1(C, M) : \text{ normally } \neg permitted(assume_comm(C, M)) \text{ if } authorized(C, M)$$

$$d_2(C, M) : \text{ normally } permitted(assume_comm(C, M)) \text{ if } colonel(C)$$

$$prefer(d_2(C, M), d_1(C, M))$$

while leaving rule (3c) as it is. Rule (3d) is unaffected, as it corresponds to the new policy statement (4) that we added to the original example by Gelfond and Lobo, to illustrate obligations.

This approach has several drawbacks: (a) it puts the burden on the knowledge engineer, who may have a more limited knowledge of the domain than the policy author and thus may make false assumptions; (b) it does not scale for large and intricate policies; and (c) it would be difficult to automate. Instead, we propose a framework that detects inconsistencies like the one above, alerts the policy author of the conflicting policy statements and the conditions that cause them, and allows the policy author to refine the policy (with options for refinement possibly suggested, in the future). In particular, for Example 1, we expect our framework to indicate that statements (1) and (2) are in conflict in a state in which both $colonel(c)$ and $authorized(c, m)$ are true. Similarly, our framework should flag the contradiction between the obligation in rule (4) and rule (1) in a state in which $authorized(c, m)$ and $ordered_by_superior(c, m)$ both hold.

To achieve this goal, we modify $\mathcal{AOP}\mathcal{L}$ by introducing labels for all rules (including strict and preference rules)⁴ and connecting rules of $\mathcal{AOP}\mathcal{L}$ with natural language statements of the original policy via a new predicate $text$ as in the following example for the strict policy rule in (3c) where s_1 is the label for the strict authorization rule:

⁴ As in the original $\mathcal{AOP}\mathcal{L}$ language, preferences can be defined only between pairs of defeasible rules.

$s_1 : \neg\text{permitted}(\text{authorize_command}(C, M)) \text{ if } \text{observer}(C)$
 $\text{text}(s_1, \text{"A military observer can never authorize a mission."})$

Additionally, we define a different translation of $\mathcal{AOP}\mathcal{L}$ into ASP, which we will denote by rei_lp as it *reifies* policy rules. We define the rei_lp translation in the next section.

4 Reification of policy rules

The new translation of $\mathcal{AOP}\mathcal{L}$ into ASP that we propose follows previous methods for the reification of rules in other domains, such as reasoning about prioritized defaults (Gelfond and Son 1997) or belief states (Balduccini et al. 2020). Similar to the definition of the lp translation function, the signature of $\text{rei_lp}(\mathcal{P})$ for a policy \mathcal{P} applying in a dynamic domain described by \mathcal{T} contains the sorts, fluents, and actions of \mathcal{T} . To simplify the presentation, we limit ourselves to boolean fluents and use the general syntax

$$r : [\text{normally}] \text{hd if } \text{cond} \tag{4}$$

to refer to both strict and defeasible, authorization and obligation rules from \mathcal{P} . We use the term *head* of rule r to refer to the hd part in (4), where $hd \in HD$,

$$HD = \bigcup_{e \in E} \{\text{permitted}(e), \neg\text{permitted}(e), \text{obl}(e), \text{obl}(\neg e), \neg\text{obl}(e), \neg\text{obl}(\neg e)\}$$

and E is the set of all elementary actions in \mathcal{T} . The signature of $\text{rei_lp}(\mathcal{P})$ also includes the elements of HD and the following predicates:

- $\text{rule}(r)$ – where r is a rule label (referred shortly as “rule” below)
- $\text{type}(r, ty)$ – where $ty \in \{\text{strict}, \text{defeasible}, \text{prefer}\}$ is the type of rule r
- $\text{text}(r, t)$ – to denote that rule r corresponds to policy statement t
- $\text{head}(r, hd)$ – to denote the head hd of rule r
- $\text{body}(r, b(r))$ – where $b(r)$ denotes the condition cond in rule r and b is a new function added to the signature of $\text{rei_lp}(\mathcal{P})$
- $\text{mbr}(b(r), l)$ – for every l in the condition cond of rule r (mbr stands for “member”)
- $\text{ab}(r)$ – for every defeasible rule r
- $\text{holds}(x)$ – where x may be a rule r , the head hd of a rule, function $b(r)$ representing the cond of a rule r , literal l of \mathcal{T} , or $\text{ab}(r)$ from above
- $\text{opp}(r, \overline{hd})$ – where r is a defeasible rule and $\overline{hd} \in HD$ (opp stands for “opposite”)
- $\text{prefer}(d_1, d_2)$ – where d_1 and d_2 are defeasible rule labels

The predicate holds helps determine which policy rules are applicable, based on what fluents are true/false in a state and the interactions between policy rules. The predicate $\text{opp}(r, \overline{hd})$ indicates that \overline{hd} is the logical complement of r ’s head hd .

The **translation** $\text{rei_lp}(\mathcal{P})$ consists of facts encoding the rules in \mathcal{P} using the predicates rule , type , head , mbr , and prefer , as well as the set of policy-independent rules below, which define predicates holds and opp , where L is a fluent literal, E an elementary action, and H a happening (i.e. an elementary action or its negation).

$$\begin{aligned} \text{body}(R, b(R)) &\leftarrow \text{rule}(R) \\ \text{holds}(R) &\leftarrow \text{type}(R, \text{strict}), \text{holds}(b(R)) \\ \text{holds}(R) &\leftarrow \text{type}(R, \text{defeasible}), \text{holds}(b(R)), \\ &\quad \text{opp}(R, O), \text{not } \text{holds}(O), \text{not } \text{holds}(\text{ab}(R)) \end{aligned}$$

$holds(B)$	\leftarrow	$body(R, B), N = \#count\{L : mbr(B, L)\},$ $N = \#count\{L : mbr(B, L), holds(L)\}$
$holds(ab(R2))$	\leftarrow	$prefer(R1, R2), holds(b(R1))$
$holds(Hd)$	\leftarrow	$rule(R), holds(R), head(R, Hd)$
$opp(R, permitted(E))$	\leftarrow	$head(R, \neg permitted(E))$
$opp(R, \neg permitted(E))$	\leftarrow	$head(R, permitted(E))$
$opp(R, obl(H))$	\leftarrow	$head(R, \neg obl(H))$
$opp(R, \neg obl(H))$	\leftarrow	$head(R, obl(H))$

Definition 7 (Reified ASP Translation of a Policy and State)

Given a state σ of \mathcal{T} , $rei_lp(\mathcal{P}, \sigma) =_{def} rei_lp(\mathcal{P}) \cup \{holds(l) : l \in \sigma\}$.

This definition will be used in conducting various policy analysis tasks in Section 5.

Proposition 1 (Relationship between the Original and Reified ASP Translations)

Given a state σ of \mathcal{T} , there is a one-to-one correspondence $map : \mathcal{A} \rightarrow \mathcal{B}$ between the collection of answer sets \mathcal{A} of $lp(\mathcal{P}, \sigma)$ and the collection of answer sets \mathcal{B} of $rei_lp(\mathcal{P}, \sigma)$ such that if $map(A) = B$ then $\forall hd \in HD \cap A, \exists holds(hd) \in B$.

5 Policy analysis

In what follows, we assume that the *cond* part of a policy rule cannot include atoms from the set HD (i.e. atoms obtained from predicates *permitted* and *obl*). We plan to consider more general policies in future work. Lifting this restriction complicates the task of finding explanations beyond the goal of the current work.

5.1 Analyzing authorization policies

Our goal in analyzing policies is to assist a policy author in refining their policies by indicating to them the rules that cause concern. Thus, when analyzing an authorization policy \mathcal{P} with respect to an elementary action e in a state σ , we focus on the tasks:

- **Explain the causes of inconsistencies** – determining the rules that cause a policy to derive both $holds(permitted(e))$ and $holds(\neg permitted(e))$ when using the *rei_lp* translation
- **Detect and explain underspecification** – determining whether rules about e exist or not, and, if they exist, explain why they do not fire. (Craven *et al.* 2009) call this situation *coverage gaps*.)
- **Detect and explain ambiguity** – determining whether there are conflicting defeasible rules that produce $holds(permitted(e))$ in some answer sets and $holds(\neg permitted(e))$ in others, and indicating which rules these are.

5.1.1 Inconsistency

To detect and explain inconsistencies with respect to an elementary action e and state σ we introduce the following predicates:

- $inconsistency(e, r_1, r_2)$ – indicates that the pair of rules r_1 and r_2 both fire and cause the inconsistency; r_1 produces $permitted(e)$ and r_2 produces $\neg permitted(e)$

- $inconsistency_expl(e, t_1, t_2)$ – does the same but indicates the natural language texts of the corresponding policy statements
- $inconsistency_expl_pos(e, l)$ – indicates that l is a fluent/static that holds in σ and contributes to the inconsistency in a rule that produces $permitted(e)$
- $inconsistency_expl_neg(e, l)$ – similar to the previous predicate, but for rules that produce $\neg permitted(e)$

We define a logic program \mathbf{I} consisting of the rules below:

$$\begin{aligned}
 inconsistency(E, R1, R2) &\leftarrow holds(permitted(E)), holds(\neg permitted(E)), \\
 &\quad holds(R1), head(R1, permitted(E)), \\
 &\quad holds(R2), head(R2, \neg permitted(E)) \\
 inconsistency_expl(E, T1, T2) &\leftarrow inconsistency(E, R1, R2), \\
 &\quad text(R1, T1), text(R2, T2) \\
 inconsistency_expl_pos(E, L) &\leftarrow inconsistency(E, R1, _), head(R1, permitted(E)), \\
 &\quad mbr(b(R1), L), holds(L) \\
 inconsistency_expl_neg(E, L) &\leftarrow inconsistency(E, _, R2), head(R2, \neg permitted(E)), \\
 &\quad mbr(b(R2), L), holds(L)
 \end{aligned}$$

We restate Definition 3 in terms of the $rei_lp(\mathcal{P}, \sigma)$ translation.

Definition 8 (Inconsistency Redefined)

An authorization policy \mathcal{P} is *inconsistent* with respect to an elementary action e and a state σ if the answer set of $rei_lp(\mathcal{P}, \sigma) \cup \mathbf{I}$ contains $inconsistency(e, r_1, r_2)$ for a pair of rules r_1 and r_2 .

Definition 9 (Explaining the Causes of Inconsistency)

- An explanation for the inconsistency of e in σ is the set of pairs of strings $\{(t_1, t_2) : inconsistency_expl(e, t_1, t_2) \in rei_lp(\mathcal{P}, \sigma) \cup \mathbf{I}\}$.
- A fluent literal l contributes positively (or negatively) to the inconsistency of e in σ if the answer set of $rei_lp(\mathcal{P}, \sigma) \cup \mathbf{I}$ contains $inconsistency_expl_pos(e, l)$ (or $inconsistency_expl_neg(e, l)$, respectively).

The collection of atoms identified in Definition 9 can be collected from the answer set of $rei_lp(\mathcal{P}, \sigma) \cup \mathbf{I}$ and post-processed to provide more human-friendly output.

5.1.2 Underspecification

We define the notion of *underspecification* of an elementary action e in a state σ as the lack of *explicit* information as to whether e is permitted or not permitted in that state, similar to the concept of *coverage gap* defined by Craven *et al.* (2009). Note that underspecification is different from non-categoricity, where a policy may be ambiguous because e is deemed permitted in some answer sets and not permitted in others.

Definition 10 (Categoricity and Underspecification of an Action in a State)

A consistent authorization policy \mathcal{P} is *categorical* with respect to an elementary action e and state σ if one of the following cases is true:

1. $rei_lp(\mathcal{P}, \sigma)$ entails $holds(permitted(e))$, or
2. $rei_lp(\mathcal{P}, \sigma)$ entails $holds(\neg permitted(e))$, or

3. For every answer set S of $rei_lp(\mathcal{P}, \sigma)$,

$$\{holds(permitted(e)), holds(\neg permitted(e))\} \cap S = \emptyset$$

In this last case, we say that e is *underspecified* in state σ .

Underspecification is important because it may reflect an oversight from the policy author. If it's unintended, it can have negative consequences in planning domains for example, when an agent may want to choose the most compliant plan and thus actions that are mistakenly underspecified may never be selected. To test whether an action e is underspecified in a state, we define the set of rules $Check_{und}(e)$ consisting of the set of constraints:

$$\left\{ \begin{array}{l} \leftarrow holds(permitted(e)), \\ \leftarrow holds(\neg permitted(e)) \end{array} \right\}$$

Definition 11 (Detecting Underspecification)

Action e is *underspecified* in σ if the logic program $lp(\mathcal{P}, \sigma) \cup Check_{und}(e)$ is consistent.

Whenever an elementary action is underspecified in a state, there may be two explanations: (Case 1) the authorization policy contains no rules about e , or (Case 2) rules about e exist in the policy but none of them apply in state σ . Once we establish that an elementary action is underspecified, we want to explain to the policy author why that is the case. For the first case, we just want to inform the policy author about the situation. In the second case, we want to report, for each authorization rule about e , which fluents make the rule non-applicable. Note that a defeasible rule r with head hd (see (4)) cannot be made unapplicable by the complement \overline{hd} , as the complement is underivable as well in an underspecified policy. Similarly, a preference rule cannot disable a defeasible rule either, as this can only be the case when the complement \overline{hd} can be inferred.

Let \mathbf{U} be the logic program

$$\begin{array}{ll} rules_exist(E) & \leftarrow head(R, permitted(E); \neg permitted(E)) \\ underspec.1(E) & \leftarrow not\ rules_exist(E) \\ underspec.1_expl(\text{"Case 1," } E) & \leftarrow underspec.1(E) \\ underspec.2(E) & \leftarrow rules_exist(E) \\ underspec.2(E, R, L) & \leftarrow underspec.2(E), rule(R), \\ & \quad head(R, permitted(E); \neg permitted(E)), \\ & \quad mbr(b(R), L), \\ & \quad not\ holds(L) \\ underspec.2_expl(\text{"Case 2," } E, R, L, T) & \leftarrow underspec.2(E, R, L), \\ & \quad text(R, T) \end{array}$$

Definition 12 (Explaining the Causes of Underspecification)

An explanation for the underspecification of e in σ is the set of atoms formed by predicates $underspec.1_expl$ and $underspec.2_expl$ found in the answer set of $rei_lp(\mathcal{P}, \sigma) \cup \mathbf{U}$.

For a more human-friendly explanation, an atom $underspec.1_expl(\text{"Case 1," } e)$ in the answer set can be replaced with an explanation of the form "There are no authorization rules about e " in the post-processing phase. A collection of atoms of the form

$$\{underspec.2_expl(\text{"Case 2," } e, r, l_1, t), \dots, underspec.2_expl(\text{"Case 2," } e, r, l_n, t)\}$$

can be replaced with the explanation “Rule r about action e (stating that ‘ t ’) is rendered inapplicable by the fact that fluent(s) l_1, \dots, l_n do not hold in this state.”

5.1.3 Ambiguity

We define ambiguity as the case when the policy allows a choice between $permitted(e)$ and $\neg permitted(e)$. This notion of ambiguity overlaps with that of a non-categorical policy. However, given our assumption that $permitted$ atoms are not included in the condition $cond$ of policy rules, ambiguity is a much more specific case. We claim that, if \mathcal{P} is a consistent, non-categorical policy with respect to e in σ (see Definition 10), then $holds(permitted(e))$ will be in some answer sets of $rei_lp(\mathcal{P}, \sigma)$ and $holds(\neg permitted(e))$ will be in others, but it cannot be the case that an answer set does not contain either.⁵

The justification is that the body $cond$ of policy rules is fully determined by the *unique* values of fluents in σ . Hence, strict rules either fire or do not. If a strict rule fired, it would automatically override the defeasible rules with the complementary head, and thus lead either to inconsistency (depending on which other strict rules fire) or categoricity. The only source of non-categoricity can be the presence of defeasible rules with complementary heads and satisfied conditions, and which are not overridden by preference rules.

Definition 13 (Ambiguity of an Action in a State)

Let \mathcal{P} be a policy that is consistent and non-categorical with respect to elementary action e and state σ . Let $rei_lp(\mathcal{P}, \sigma)$ have n answer sets, out of which n_p answer sets contain $holds(permitted(e))$ and n_{np} contain $holds(\neg permitted(e))$.

\mathcal{P} is *ambiguous* with respect to e and σ if $n \neq n_p$, $n \neq n_{np}$ and $n = n_p + n_{np}$.

Next, let’s describe how we detect ambiguity.

Definition 14 (Detecting Ambiguity)

Action e is *ambiguous* in σ if $holds(permitted(e))$ and $holds(\neg permitted(e))$ are not entailed by $rei_lp(\mathcal{P}, \sigma)$ and e is not underspecified in σ .

Once ambiguity is established, an explanation for ambiguity is needed. To produce it, we define the logic program **A** consisting of the rules:

$$\begin{aligned}
 ambiguous(E, R1, R2) &\leftarrow defeasible_rule(R1), head(R1, permitted(E)), \\
 &\quad defeasible_rule(R2), head(R2, \neg permitted(E)), \\
 &\quad holds(b(R1)), holds(b(R2)), \\
 &\quad not\ holds(ab(R1)), not\ holds(ab(R2)) \\
 ambiguity_expl(E, T1, T2) &\leftarrow ambiguous(E, R1, R2), text(R1, T1), text(R2, T2)
 \end{aligned}$$

Definition 15 (Explaining the Causes of Ambiguity)

An explanation for the ambiguity of e in σ is the set of pairs of strings:

$$\{(t_1, t_2) : ambiguity_expl(e, t_1, t_2) \in rei_lp(\mathcal{P}, \sigma) \cup \mathbf{A}\}$$

⁵ Note that, if we lift our restriction and allow $cond$ to contain $permitted$ (or obl) atoms, then for a weakly compliant action e , there can be a combination of answer sets containing $holds(permitted(e))$ and answer sets not containing neither $holds(permitted(e))$ nor $holds(\neg permitted(e))$, if $cond$ contains a $permitted(e_1)$ atom such that e_1 is an action that is ambiguous in σ .

5.1.4 Observation about strongly vs weakly compliant policies

Gelfond and Lobo distinguish between actions that are strongly compliant in a state versus weakly compliant (see Definition 4). In planning, as shown by Meyer and Incelean (2021), it seems reasonable to prefer strongly compliant actions over weakly compliant ones. However, in the theorem below we show that the class of weakly compliant actions includes strongly compliant ones. What we really need for planning purposes is distinguishing between strongly compliant and underspecified actions in a state, so that we can create a preference order between actions.

Theorem 1 (Strongly vs Weakly Compliant Actions)

All elementary actions that are strongly compliant in a state σ are also weakly compliant.

Proof

Note that, in this proof, we consider the original lp translation of $\mathcal{AOP}\mathcal{L}$, which is equivalent with, but more convenient to use here than, the reified translation rei_lp as stated in Proposition 1. According to Definition 4 borrowed from Gelfond and Lobo's work, elementary action e is strongly compliant with authorization policy \mathcal{P} if $lp(\mathcal{P}, \sigma)$ entails $permitted(e)$, and it is weakly compliant if $lp(\mathcal{P}, \sigma)$ does not entail $\neg permitted(e)$. For consistent policies, the latter condition is obviously true if e is strongly compliant in σ , as having $permitted(e)$ in every answer set of $lp(\mathcal{P}, \sigma)$ implies that $\neg permitted(e)$ must be absent from each such answer set. If the policy is inconsistent, the theorem is vacuously true. \square

Given our assumption that *permitted* (and *obl*) atoms cannot appear in the *cond* part of policy rules, an elementary action e can only be either strongly compliant in σ or underspecified.⁶

We formulate the following proposition, which is useful in creating an ordering of actions based on compliance (relevant in planning).

Proposition 2 (Properties of Authorization Policies)

- If condition *cond* of authorization rules is not allowed to contain *permitted* (or *obl*) atoms and \mathcal{P} is *categorical* with respect to e and σ , then e is either strongly compliant, non-compliant, or underspecified in σ .
- If condition *cond* of authorization rules is not allowed to contain *permitted* (or *obl*) atoms and \mathcal{P} is *non-categorical* with respect to e and σ , then e is neither strongly compliant nor non-compliant; it may be either underspecified or ambiguous.

5.2 Obligation policy analysis

The techniques from Section 5.1 for determining rules that create inconsistencies, underspecification, and ambiguity with respect to an elementary action and a state can be easily adapted to obligation policies as well. Obligation policies apply to *happenings*, which are actions or their negations. For instance, given an elementary action e , the

⁶ If this restriction is lifted and the condition *cond* of a policy rule for e contains a *permitted*(e_1) atom, such that e_1 is *ambiguous*, then e may be a weakly compliant action because it will be permitted in some answer sets and unknown in others.

following literals are part of the signature of the policy: $obl(e)$, $obl(\neg e)$, $\neg obl(e)$, and $\neg obl(\neg e)$. Inconsistencies between $obl(e)$ and $\neg obl(e)$ on one hand, or between $obl(\neg e)$ and $\neg obl(\neg e)$ are easy to detect. However, there are additional incongruencies that may decrease the quality of a policy, and we may want to alert the policy writer about them as well.

For instance, if a policy \mathcal{P} entails both $obl(e)$ and $obl(\neg e)$ in a state σ , then any event $\langle \sigma, a \rangle$ will be non-compliant, no matter whether $e \in a$ or $e \notin a$. In actuality, it means that the policy does not allow for the agent to be compliant with respect to obligations in state σ . Thus the notions of inconsistency and ambiguity should be adapted or expanded to include this situation. We propose the following definition:

Definition 16 (Conflicting Policy)

Given a consistent obligation policy \mathcal{P} , a state σ and an elementary action e , we call \mathcal{P} a *conflicting obligation policy* with respect to σ and e if the logic program $rei_lp(\mathcal{P}, \sigma)$ entails both $holds(obl(e))$ and $holds(obl(\neg e))$.

Explanations for conflicting obligation policies can be found using techniques similar to the ones in Section 5.1.

5.3 The intersection between authorization and obligation policies

When combining an authorization policy with an obligation policy, there are a few cases that, while not necessarily inconsistent, certainly seem to require non-compliant behavior from the agent. This is especially the case when an event $\langle \sigma, e \rangle$ is strongly compliant with the obligation policy but non-compliant with the authorization policy (i.e. in terms of the original translation lp of $\mathcal{AOP}\mathcal{L}$ into ASP, $lp(\mathcal{P}, \sigma)$ entails both $obl(e)$ and $\neg permitted(e)$ according to Definitions 4 and 5). Other situations that may require the policy authors' attention, though to a lesser degree, are when an action is permitted but the agent is obligated not to execute it (i.e. $lp(\mathcal{P}, \sigma)$ entails both $permitted(e)$ and $obl(\neg e)$) or when the agent is obligated to execute an action that is underspecified in that state. We indicate the level of urgency of each of these situations by adding a number from 1 to 3, with 1 being the most needing of re-consideration and 3 being the least urgent.

Once it has been established that the policies are strongly compliant, non-compliant, or underspecified with respect to the state and elementary action, the following ASP rules determine which policy rules need to be re-visited.

$$\begin{aligned}
 require_cons(E, R1, R2, 1) & \leftarrow holds(R1), head(R1, obl(E)), \\
 & \quad holds(R2), head(R2, \neg permitted(E)) \\
 require_cons(E, R1, R2, 2) & \leftarrow holds(R1), head(R1, obl(\neg E)), \\
 & \quad holds(R2), head(R2, permitted(E)) \\
 require_cons(E, R1, R2, 3) & \leftarrow holds(R1), head(R1, obl(E)), \\
 & \quad not\ holds(permitted(E)), \\
 & \quad not\ holds(\neg permitted(E)) \\
 require_cons_expl(E, T1, T2, N) & \leftarrow require_cons(E, R1, R2, N), \\
 & \quad text(R1, T1), text(R2, T2).
 \end{aligned}$$

6 Related work

Meyer and Incezan (2021) developed an architecture for policy-aware intentional agents (*APJA*) by leveraging Blount *et al.*'s theory of intentions (2015). An agent's behavior was ensured to be compliant with authorization and obligation policies specified in *AOPL* and translated into ASP. Meyer and Incezan's work first highlighted the issues that may arise at the intersection between *AOPL* authorization and obligation and policies. In the *APJA* architecture, conflicts of this nature were resolved by modifying the policy's ASP encoding to state that such conflicts render a policy inconsistent. In the current work our intention is to alert policy authors about such situations and provide them with the opportunity to decide which policy statements to modify in order to restore consistency. Additionally, in the current work we delve deeper into the tasks associated with policy analysis and look at underspecification and ambiguity as well. We also focus on providing explanations as to why such issues arise.

Craven *et al.*'s work (2009) is the closest to ours in its intent. The authors define language \mathcal{L} for policy specification and include both authorization and obligation policies. They define a solid set of tasks that an automated analysis of a policy should accomplish, such as discovering *modality conflicts* and *coverage gaps*, which we target in our work as well. Their research assumes that the underlying dynamic domain is specified in Event Calculus (Kowalski and Sergot 1989). Explanations are found via an abductive constraint logic programming proof procedure. Given the absence of a comparison between languages \mathcal{L} and *AOPL*, it is important to study the problem of policy analysis with respect to language *AOPL* as well. *AOPL* has clear advantages, including its ability to express defeasible policies and preferences between policies. Moreover, *AOPL* can be seamlessly integrated with ASP-based dynamic system descriptions, as different properties of *AOPL* policies can be checked by finding the answer sets of an ASP program. This would allow coupling policies with system descriptions specified in higher-level action languages that translate into ASP, such as the modular action language *ALM* (Incezan and Gelfond 2016), and associated libraries about action and change (Incezan 2016; 2019).

Other research on policy modeling or analysis using ASP exists, but the goals tend to be different from ours. Corapi *et al.* (2011) use inductive logic programming and software engineering-inspired processes to assist policy authors with policy refinement. In their work, refinement suggestions are provided, but this process is driven by use cases that need to be manually created. As a result, the quality of the resulting policy depends on the quality and coverage of the use cases that are provided as an input. In turn, our approach is meant to be more comprehensive and transparent, as it is guided by the policy rules themselves. Another work that uses ASP for policy modeling is that of De Vos *et al.* (2019). Their work encompasses the same types of policies as *AOPL*, but their focus is on compliance checking and providing explanations for the compliance or non-compliance of events. In contrast, we focus on policy analysis, not compliance checking; our explanations highlight potential problems with a policy and indicate statements that need to be refined. Havur *et al.* (2021) present a framework called DALICC for comparing and resolving compatibility issues with licenses. The goal of their framework is more narrow than ours in the sense that it only focuses on licences and not normative

statements in general. For a survey on other policy analysis methods and tools, not necessarily ASP-related, we direct the reader to the paper by Jabal *et al.* (2019).

In general, in the policy specification and analysis community, there is an intense focus on access control policies, which may involve the Role-Based Access Control (RBAC) model outlined by Ferraiolo *et al.* (2001); the Attribute-Based Access Control Model (ABAC) explored for instance by Davari and Zulkernine (2021) and Xu *et al.* (2016); or the Category-based Access Control Model explored by Alves and Fernández (2014). A secondary focus falls on policies for the management of computer systems. In contrast, $\mathcal{AOP}\mathcal{L}$ is more general and could be used to represent social norms, for example.

Finally, our work touches upon explainability and finding the causes of issues encountered in $\mathcal{AOP}\mathcal{L}$ policies. To find even deeper causes that reside in the inner-workings of the dynamic system, we can leverage existing work on explainability in reasoning about action and change, such as the research by LeBlanc *et al.* (2019); planning domains, as in the work by Vasileiou *et al.* (2022); or logic programming in general, including research by Fandinno and Schultz (2019) or Cabalar *et al.* (2020).

7 Conclusions

In this paper we introduced a framework for analyzing policies described in the language $\mathcal{AOP}\mathcal{L}$ with respect to inconsistencies, underspecification, ambiguity, and modality conflict. We reified policy rules in order to detect which policy statements cause the particular issue and (if relevant), which fluents of the domain contribute to such problems. In doing so, we defined new properties of $\mathcal{AOP}\mathcal{L}$ policies and took a special look at what happens at the intersection of authorization and obligation policies.

As part of future work, we plan to create a system that implements this framework in a way that is user-friendly for policy writers and knowledge engineers. We also intend to extend the framework by lifting some of the simplifying restrictions that we imposed here, for instance by studying the case when there is incomplete information about a state or allowing *permitted* and *obl* atoms in the conditions of policy rules.

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