Towards the Integration of Programming by Demonstration and Programming by Instruction using Golog: Extended Version with Proofs

Christian Fritz and Yolanda Gil

Information Sciences Institute University of Southern California Marina del Rey, California. USA. {fritz,gil}@isi.edu

Abstract

We present a formal approach for combining programming by demonstration (PbD) with programming by instruction (PbI)-a largely unsolved problem. Our solution is based on the integration of two successful formalisms: version space algebras and the logic programming language Golog. Version space algebras have been successfully applied to programming by demonstration. Intuitively, a version space describes a set of candidate procedures and a learner filters this space as necessary to be consistent with all given demonstrations of the target procedure. Golog, on the other hand, is a logical programming language defined in the situation calculus that allows for the specification of non-deterministic programs. While Golog was originally proposed as a means for integrating programming and automated planning, we show that it serves equally well as a formal framework for integrating PbD and PbI. Our approach is the result of two key insights: (a) Golog programs can be used to define version spaces, and (b) with only a minor augmentation, the existing Golog semantics readily provides the update-function for such version spaces, given demonstrations. Moreover, as we will show, two or more programs can be symbolically synchronized, resulting in the intersection of two, possibly infinite, version spaces. The framework thus allows for a rather flexible integration of PbD and PbI, and in addition establishes a new connection between two active research areas, enabling cross-fertilization.

1 Introduction

In large parts, AI is concerned with the problem of making machines behave in particular ways intended by users. The problem is challenging because the intended behaviors can be very complex, it can be difficult to humans, in particular non-programmers, to describe the correct behavior explicitly, and in many cases there are plenty of exceptions or special circumstances to consider. Therefore life-long learning is the norm and there always remains some uncertainty as to what the right behavior should be.

Programming by demonstration (PbD; e.g., (Lieberman 2001)) has shown a lot of promise in end-user programming. In PbD a user demonstrates the intended behavior to the system on a number of specific examples, and the system in turn tries to learn the correct behavior from this. This allows a much larger group of people to "program", as it does not demand specialized programming skills of the user. The

shortcomings of PbD, however, are that it is often hard for the system to generalize from linear examples when the intended behavior has complex structures. Furthermore, in the existing approaches it is difficult for the user to control what is being learned or to recover from errors that were made during demonstration.

On the other hand there are approaches that allow users to program by explicit instruction (PbI). Using, say, natural language a user can provide direct input regarding particular parts of the target program or its high-level structure. It is hence easy to express complex program structure. The disadvantages, however, are that this type of instruction is prone to omissions and imprecision—it is easy for a user to forget a condition or a boundary case— and also, as with explicit programming, the user needs to put more effort into structuring the instructions correctly.

Ideally one would want to combine both, PbD and PbI, as they seem to complement each other well. But this integration is challenging. There have been some promising approaches for making PbD systems more robust to errors and allowing user edits interleaved with demonstration of examples. Oblinger, Castelli, and Bergman (2006) describe the augmentation-based learning (ABL) algorithm that is used by the DocWizards system (Prabaker, Bergman, and Castelli 2006). The algorithm allows users to manually edit procedure hypotheses explicitly interleaved with demonstration and ensures that such edits are not undone by future demonstrations. Chen and Weld (2008) present CHINLE, a system that generates domain specific PbD systems from declarative interface specifications. Via a compelling visualization of procedure hypotheses, the system allows users to discard individual steps of the learned (linear) procedure or explicitly add training examples in support of particular hypotheses. These systems however, limit the types of editing operations the user is allowed to perform and hence only cover part of the spectrum between PbI and PbD. In particular, they do not allow the manual specification of the high-level program structure, which might be most helpful to the PbD part of the system.

The problem is that there is no principled, comprehensive framework for representing, updating, and executing procedure hypotheses. Such a framework should enable any learner to output its acquired knowledge in this form, allow for the combination of (sets of) procedure hypotheses, so that different input/learning methods can be combined, and enable the execution of partially learned procedures if necessary, hence enabling learning from experience or explicit user feedback as well.

Inspired by the recent approaches for combining PbD with user edits, we propose a formal framework that integrates PbD with PbI in a more flexible manner.

The key ideas of our approach are as follows: 1) We can represent (infinite) sets of procedure hypotheses using nondeterministic programs. These are used to represent uncertainty about the target procedure being learned. In particular, we use the logical programming language Golog which readily allows us to represent and reason about such nondeterministic programs. Such a program can be understood to represent the entire set of possible deterministic programs described by it. For the purpose of this paper, we assume that there is a learning component available that generates such programs from user instruction. As such, we believe the language to be well suited for this purpose, as its nondeterministic constructs can be used in places where instructions were ambiguous or omitted. 2) We extend Golog's semantics to implement the update function that removes from a set of hypotheses all those that are inconsistent with newly given demonstrations. This is implemented as refinements to the program representing the version space, resolving some of the uncertainty, i.e., making some non-deterministic parts of the program deterministic. 3) Given Golog's formal semantics, we can define a notion of program synchronization that provably implements a symbolic intersection of (possibly infinite) sets of hypotheses.

By virtue of basing the framework on Golog and its semantics in the situation calculus, we get as a side-effect the ability to execute procedure hypotheses at any time, even while there still remains a lot of uncertainty. This exploits Golog's integration with automated planning and further facilitates learning from experience as well. Perhaps most importantly however, the approach facilitates the reuse of a large body of research regarding situation calculus and Golog that could allows us to extend PbD for other settings, including for environments with only partially observable states, or uncertain action effects.

The contributions of this paper are of mainly theoretic nature rather than describing or evaluating a complete system. A prototype implementation of the framework however exists and will be made available. It could serve as a back-end to systems that focus on the actual user interface and we are actively using and extending this system.

1.1 Motivation

Some possible ways of integrating PbD and PbI are as follows: i) A novice programmer defines the rough, high-level structure of a program using PbI and then starts demonstrating specific examples of the program. She may later go back-and-forth between PbI and PbD to both get the structure but also all specific boundary cases right. ii) An expert programmer designs a template program by instruction that ensures a basic, required behavior. This template is then customized by each end-user individually via demonstration. iii) Two programmers each sketch their (partial) knowledge of the same program using PbI. These two program candidates are then *synchronized*, by which they complement each other and produce a more specific candidate. Similarly, two sets of hypotheses provided by different learning systems could be symbolically combined.

For illustration purposes in this paper, we use examples from a domain in which a robot is being programmed to tidy up a room. One possible, incomplete procedure for this could be stated as follows: *While there remains an item on the floor, we are meant to pick it up and then, if it is a toy, put it in the box, if it is clothing, go to the closet and put it either on a hanger or in the drawer (but we don't yet know which one), and if it is anything else, we don't know what to do.* Then, after receiving a demonstration where a book is put on the shelf, the system may entertain two hypotheses regarding the uncertainty about other objects: (a) that everything else goes on the shelf, or that (b) there are more conditions to consider. We will use this as a running example.

We begin by reviewing the situation calculus and Golog. We then show how version spaces can be represented as Golog programs, can be updated using an extension of Golog's semantics, and can be intersected by *synchronizing* two programs. In Section 4 we describe how the presented framework can be used to flexibly integrate PbD and PbI. We discuss related and future work before we conclude.

2 Preliminaries

The situation calculus is a sorted logic for specifying and reasoning about dynamical systems (Reiter 2001). In the situation calculus, the state of the world is expressed in terms of *fluents*, functions and relations relativized to a *situation s*, e.g., $F(\vec{x}, s)$. A situation is a history of the primitive actions a performed from a distinguished initial situation S_0 . The function do(a, s) maps an action and a situation into a new situation thus inducing a tree of situations rooted in S_0 . The relation \Box provides an ordering on situations of the same branch, and $s \sqsubseteq s'$ abbreviates $s = s' \lor s \sqsubset s'$. We abbreviate $do(a_n, do(a_{n-1}, \dots, do(a_1, s)))$ to $do([a_1, \dots, a_n], s)$ or $do(\vec{a}, s)$. We denote the set of actions by A. In the situation calculus, all actions have deterministic effects and this is what we are assuming in this paper. Nevertheless, there exist extensions that account for actions with uncertain outcomes (e.g., (Mateus et al. 2001)).

In the situation calculus, background knowledge of a domain (e.g., our cleaning robot domain) is encoded as a basic action theory, \mathcal{D} . It comprises four domain-independent foundational axioms, and a set of domain-dependent axioms. Details of the form of these axioms can be found in (Reiter 2001). We write $s \sqsubset s'$ to say that situation s precedes s' in the tree of situations. This is axiomatized in the foundational axioms. Included in the domain-dependent axioms are the following sets:

Initial state axioms, \mathcal{D}_{S_0} : a set of first-order sentences relativized to situation S_0 , specifying what is true in the initial state, e.g., $OnFloor(Book_1, S_0)$ and $Book(Book_1)$ state that initially there is a book on the floor.

Successor state axioms: provide a parsimonious representation of frame and effect axioms under an assumption of the completeness of the axiomatization. There is one successor state axiom for each fluent, F, of the form $F(\vec{x}, do(a, s)) \equiv \Phi_F(\vec{x}, a, s)$, where $\Phi_F(\vec{x}, a, s)$ is a formula with free variables among \vec{x}, a, s . $\Phi_F(\vec{x}, a, s)$ characterizes the truth value of the fluent $F(\vec{x})$ in the situation do(a, s) in terms of what is true in situation s.

Action precondition axioms: specify the conditions under which an action is possible. There is one axiom for each action a of the form $Poss(a(\vec{x}), s) \equiv \prod_a(\vec{x}, s)$ where $\prod_a(\vec{x}, s)$ is a formula with free variables among \vec{x}, s . For instance, $Poss(PickNextItem, s) \equiv (\exists x).NextItem(s) = x \land OnFloor(x, s).$

Given the semantics, situations compactly describe *traces* of state-action pairs and can hence be used to represent demonstrations of procedures.

2.1 Golog

Golog (Levesque et al. 1997) is a programming language defined in the situation calculus. It allows a user to specify programs whose set of legal executions defines a subtree of the tree of situations of a basic action theory. Golog has an Algol-inspired syntax extended with flexible *non-deterministic constructs*. Golog programs are created inductively using the following constructs:

nil	empty program
$a \in \mathcal{A}$	primitive action
$\phi?$	test condition ϕ
$[\delta_1; \delta_2]$	sequence
if ϕ then δ_1 els	se δ_2 conditional
while ϕ do δ'	loops
$(\delta_1 \mid \delta_2)$	non-deterministic choice
$(\pi v)\delta(v)$	non-deterministic choice of argument
δ^*	non-deterministic iteration

In addition, Golog allows the definition of procedures. The semantics of a Golog program δ is defined in terms of macro expansion into formulae of the situation calculus. $Do(\delta, s, s')$ is understood to denote a formula expressing that executing δ in situation s is possible and may result in situation s'. This is defined inductively over the program structure. For instance for primitive actions: $Do(a, s, s') \stackrel{\text{def}}{=} Poss(a[s], s) \land s' = do(a[s], s),$ where a[s] denotes the action a with all its arguments instantiated in situation s. For simple non-determinism: $Do(\delta_1 \mid \delta_2, s, s') \stackrel{\text{def}}{=} Do(\delta_1, s, s') \lor Do(\delta_2, s, s')$. The complete semantics can be found in (Levesque et al. 1997). While deterministic constructs enforce the occurrence of particular actions, non-deterministic constructs define "open parts" that allow for several ways of interpretation/execution. We call a Golog program deterministic if it contains neither $(\delta_1|\delta_2)$ nor (πv) nor δ^* .

3 Version spaces as Golog programs

In this section we illustrate how naturally Golog programs can be used to define version spaces and how the existing semantics can be augmented in order to refine a program given an example, hence realizing the update function for the version space. We further define a notion of *program* synchronization that allows us to symbolically intersect version spaces.

Intuitively, in the context of this paper, non-deterministic parts of a Golog program express uncertainty about the precise target procedure being learned and will be filled-in via demonstration. Roughly, non-deterministic constructs can be compared to the "union" operator and all other constructs to "join" operators often used in version space algebra based approaches (e.g., (Lau et al. 2003)). Hence, just like version space algebras hierarchically combine simpler spaces into more complex ones, we can use the inductive definition of the Golog language to combine several sub-programs into more complex ones. The availability of the concept of procedures is very helpful in this respect: it allows us to define and reuse sub-version spaces and give them names. For instance, our cleaning robot example could be specified as:

$$P_1 \stackrel{\text{def}}{=} \textbf{while} (\exists x).OnFloor(x) \textbf{do} [PickNextItem; \\ \textbf{if } Toy(Item) \textbf{ then } Put(Item, Box) \\ \textbf{else if } Clothing(Item) \textbf{ then } [goto(Closet); \\ (Put(Item, Hanger) | Put(Item, Drawer))] \\ \textbf{else } Process(Item)]$$

1.0

We use the (non-deterministic) procedure Process(x), defined as follows, to describe the space of possible (deterministic) procedures of how to stow away any other item:

$$Process(x) \stackrel{\text{def}}{=} \\ (Stow(x) \mid \\ ((\pi c)[cond(c)?; \text{if } c \text{ then } Stow(x) \text{ else } nil])^*) \\ Stow(x) \stackrel{\text{def}}{=} [((\pi l)goto(l))^*; (\pi v)Put(x,v)]$$

where $cond(\varphi)$ is a user-defined predicate that defines the set (version space) of conditions to consider. This is common practice in version space algebra based approaches and could, for instance, limit conditions to be simple fluent literals or conjunctions of such of up to a certain length. The system could then conjecture conditions that distinguish the place certain types of items are meant to be stowed at. The auxiliary procedure Stow(x) non-deterministically chooses such a place and this non-deterministic gets resolved as more examples are incorporated.

For instance, given the described demonstration about the book being put on the shelf, the above described hypothesis (a) would replace Process(Item) by Put(Item, Shelf), which is one possible option of realizing the procedures, choosing the Stow(x) branch, not doing any *goto*'s, and then choosing *Shelf* for v). Hypothesis (b), on the other hand, would replace it with

$$\begin{bmatrix} cond(Book(Item))?; \\ if Book(Item) then Put(Item, Shelf) else nil; \\ ((\pi c)[cond(c)?; if c then Stow(x) else nil])^* \end{bmatrix}$$

which is obtained by choosing the second branch in the nondeterministic choice of the Process(x) procedure.

We are now ready to define the problem of programming by demonstration and what counts as a solution. We assume the target procedure to be deterministic. However, in the absence of sufficient training data, we may not be able to specify all the details of this target procedure and hence still end up with a non-deterministic procedure, after considering a number of examples. This is reflected in the following definition.

Definition 1. A *PbD problem* is a tuple $\langle \delta, \{(S_1, S'_1), \dots, (S_n, S'_n)\} \rangle$, where δ is a Golog program and each (S_i, S'_i) is a tuple of situation terms such that $S_i \sqsubset S'_i$. A solution to this problem is any δ' such that:

- 1. for any pair of situations S, S', if $\mathcal{D} \models Do(\delta', S, S')$, then also $\mathcal{D} \models Do(\delta, S, S')$; and
- 2. for $(S_i, S'_i) \in \{(S_1, S'_1), \dots, (S_n, S'_n)\}$:

 $\mathcal{D} \models Do(\delta', S_i, S_i') \land (\forall s). Do(\delta', S_i, s) \supset s = S_i'$

That is, a solution to a PbD problem is a program that does not admit any executions not admitted by the original program, but does (at least) admit the given set of demonstrated executions. Intuitively, in the definition, δ is a (usually non-deterministic) Golog program describing a space of possible procedures, and δ' is another Golog program that is a specialization of δ and in particular ensures that the program behaves according to the given examples. We show how such a solution can be obtained via simulation of the given program over the demonstrations while keeping track of the non-deterministic choices made during the simulation. Note that even though we assume the target procedure to be deterministic, a solution to a PbD problem is not necessarily deterministic. This is in particular the case when not enough demonstrations were given to rule out any remaining uncertainty. Nevertheless, Golog's semantics readily provides the means for executing such non-deterministic programs, if necessary.

3.1 Resolving uncertainty in Golog programs

Given a starting situation S and a program δ , the originally intended use of Golog was to create a constructive proof for the query $\mathcal{D} \models \exists s'.Do(\delta, S, s')$, hence obtaining as a sideeffect a situation term s' that is a sequential *plan* for how the program can be executed successfully. We, however, will use the semantics in a different way: given *two* situation terms S, S' and a program δ with non-determinism, verify that executing δ starting in S is possible and can result in S'. In doing so, we heavily exploit the logical underpinnings of the presented framework, in order to actively exploit the content of S' to infer how decisions need to be made, rather than following a trial-and-error approach. Recall that situations are sequences of actions and also, given a basic action theory \mathcal{D} , completely describe the state of the world.

Hence, without any modification necessary, the existing Golog semantics can be used to *verify* that a given program is consistent with a given demonstration. This verification can be done efficiently, as the sequence of actions in S' guides the interpretation of the program, hence limiting the amount of search necessary to one-step look-ahead. However, the program can actually also be *refined* during this process such that all choices regarding non-determinism in the program that were necessary in order to explain the given

 $Do'(nil, s, s', \delta') \equiv s' = s \land \delta' = nil$ $Do'(a, s, s', \delta') \equiv Poss(a[s], s) \land s' = do(a[s], s) \land \delta' = a$ $Do'(\varphi?, s, s', \delta') \equiv \varphi[s]? \land s = s' \land \delta' = \varphi?$ $Do'([\delta_1; \delta_2], s, s', \delta') \equiv (\exists s^*, \delta_1', \delta_2') Do'(\delta_1, s, s^*, \delta_1') \wedge$ $Do'(\delta_2, s^*, s', \delta_2') \wedge \delta' = [\delta_1'; \delta_2']$ Do'(**if** φ **then** δ_1 **else** $\delta_2, s, s', \delta') \equiv (\exists \delta'').$ $(\varphi[s] \wedge Do'(\delta_1, s, s', \delta'') \wedge \delta' = \text{if } \varphi \text{ then } \delta'' \text{ else } \delta_2) \vee$ $(\neg \varphi[s] \land Do'(\delta_2, s, s', \delta'') \land \delta' = \mathbf{if} \varphi \mathbf{then} \, \delta_1 \mathbf{else} \, \delta'')$ Do'(while φ do $\delta, s, s', \delta') \equiv \neg \varphi[s'] \land$ $(\forall P).\{(\forall s_1, s_2, s_3, \delta_1, \delta_2, \delta_3)[\varphi[s_1] \land Do'(\delta_1, s_1, s_2, \delta_2) \land$ P(while φ do $\delta_2, s_2, s_3,$ while φ do $\delta_3) \supset$ P(while φ do $\delta_1, s_1, s_3,$ while φ do $\delta_3)] \land$ $(\forall s_1, \delta_1) \neg \varphi[s_1] \supset P($ while φ do $\delta_1, s_1, s_1,$ while φ do $\delta_1) \}$ $\supset P(\delta, s, s', \delta')$ $Do'((\delta_1|\delta_2), s, s', \delta') \equiv$ $Do'(\delta_1, s, s', \delta') \vee Do'(\delta_2, s, s', \delta')$ $Do'((\pi v)\delta(v), s, s', \delta') \equiv (\exists x)Do'(\delta(x), s, \prime, \delta')$ $Do'(\delta^*, s, s', \delta') \equiv (\forall P).\{(\forall s_1)P(s_1, s_1, nil) \land$ $(\forall s_1, s_2, s_3, \delta_1, \delta_2)[Do'(\delta, s_1, s_2, \delta_1) \land P(s_2, s_3, \delta_2)]$ $\supset P(s_1, s_3, \delta_1; \delta_2)] \} \supset P(s, s', \delta')$ $Do'(P(\vec{x}), s, s', \delta') \equiv Proc(P(\vec{x}), \delta) \wedge Do'(\delta, s, s', \delta')$



demonstration are recorded. The refined program represents the version space updated with the considered example. We accomplish this by extending the original Golog semantics as shown in Figure 1, where we use an additional forth argument that "returns" the refined program. Intuitively, the refined program is the same as the original program for all deterministic constructs, and for non-deterministic constructs it contains the specific choice that was made in order to "execute" the program. The latter, however, is only the case for those non-deterministic choices that are actually visited. For instance, choices occurring in the 'then' or the 'else' branch of an if-then-else, are only resolved if that branch applied to the considered example. Choices not visited during program execution are left as is. In the next section we will see that this refinement can be used to identify those specializations of the original program that are consistent with a given program execution. This is accomplished by fixing the second situation term (s'), which forces the program to execute in compliance with the actions in that situation term. Hence, by keeping track of the choices made during execution, we obtain a program that could be used to reproduce the demonstration described by the given s'. Further, since only those choices are made that are actually required to execute the program, no candidate programs are ruled out.

There are a few things to note: First, note that the cases for if-then-else and while-loops are still in accordance with the original Golog semantics. We here have merely unwound the macro-definition (in terms of $(\delta_1 | \delta_2)$ and δ^*). The reason for this is that we want to keep the if-then-else/while-

loop in the refined program, rather than replacing them with the specific sequence of actions resulting from resolving the non-deterministic choices for the specific case considered.

Second, since we want to explicitly refer to programs as objects, in order to produce the refined program, we require the reification of programs as objects in the language. This property is shared with the so called transition semantics for Golog, as described by (De Giacomo, Lespérance, and Levesque 2000), and we assume programs are reified analogously. This also allows us to define the semantics in terms of a predicate rather than macro-expansion. Third, we do not consider recursive procedures. Also, the reification of programs allows us to assume that procedures are defined in the background knowledge using the relation $Proc(P(\vec{x}), \delta)$, where P denotes the procedure name, \vec{x} the formal arguments, and δ the body, which may mention elements from \vec{x} . Finally, the following property follows from analogy to the original semantics.

Proposition 1. For any two situations S, S' and a Golog program δ without procedures:

$$\mathcal{D} \models Do(\delta, S, S')$$
 iff $\mathcal{D} \models (\exists \delta').Do'(\delta, S, S', \delta')$

3.2 Updating version spaces

Intuitively, the new, forth argument in Do' "returns" the refined program that results from making the necessary choices during execution of the program in order to reach s' from s. Hence, in order to refine a Golog program δ using a given demonstration (S, S') we constructively prove the query:

$$\mathcal{D} \models (\exists \delta').Do'(\delta, S, S', \delta')$$

Note that there may be several such programs δ' . As with the original Golog, the provided definition of Do' lends itself to a rather straightforward Prolog implementation, casting the problem of constructing a proof into a search problem. We will make such an implementation available on our web site. It can be used to obtain all possible refined programs δ' .

As an example of how this might look in practice, consider our example program P_1 from the previous section, a situation S where there is only one book on the floor, and the action sequence *PickNextItem*, *Put(Item, Shelf)*. Then one can verify using above definitions that the following holds:

$$\mathcal{D} \models Do'(P_1, S, do([PickNextItem, Put(Item, Shelf)], S), P_2)$$

for P_2 being like P_1 but with Process(Item) replaced according to any one of the two previously described hypotheses (a) or (b).

The resulting programs can be further refined through demonstration or manual editing. Regarding the former, one can show that this leads to solutions of PbD problems.

Theorem 1. Let $M = \langle \delta, \{(S_1, S'_1), \dots, (S_n, S'_n)\} \rangle$ be a PbD problem. Any program δ' such that:

$$\mathcal{D} \models (\exists \delta_1, \dots, \delta_{n-1}) . Do'(\delta, S_1, S'_1, \delta_1) \land Do'(\delta_1, S_2, S'_2, \delta_2) \land \dots \land Do'(\delta_{n-1}, S_n, S'_n, \delta')$$

is a solution to M.

Proof: See Appendix A.

The theorem gives rise to a host of possible algorithms for finding solutions. This includes the common filtering algorithm for updating version spaces: Given the first demonstration generate the entire set of consistent version spaces (i.e., refined programs δ_1). Then, given subsequent examples, remove from this set all those spaces (programs) that are inconsistent with any of the examples. Note that our use of logic readily realizes the "lazy evaluation" approach that is popular in many version space algebra based approaches to PbD in order to handle infinite version spaces.

Other possible algorithms could follow a depth-first or a best-first search approach. The latter could, for instance, be realized by devising an evaluation function that ranks refined programs by some understanding of likelihood. For example, shorter and/or simpler programs could be explored before more complicated ones are considered. There is a host of related research on situation calculus and Golog from which such specifications and search strategies could be drawn, e.g., (Bacchus, Halpern, and Levesque 1999; Bienvenu, Fritz, and McIlraith 2006; Grosskreutz and Lakemeyer 2000).

3.3 Intersecting Version Spaces

Thus far we have made the assumption that demonstrations are given directly to our framework, and that only within our framework the learning from examples takes place. In practice, it may, however, be interesting to consider the input from other learners. For instance, for a certain domain a particular learning algorithm may be known to provide better generalization than the version space approach underlying our framework. To also enable the combination of PbI with such external algorithms, and hence loosen our dependence on the specific approach for learning from examples assumed in the previous section, we here consider the problem of *intersecting* version spaces. The problem can be stated as follows: Given two version spaces for the same target procedure, determine a new version space that contains all and only those program executions that were contained by each of the two given spaces. In terms of our program based representation of version spaces, we can define the problem more precisely:

Definition 2. Let δ_1, δ_2 be two Golog programs over action theory \mathcal{D} . The Golog program δ' is called an *intersection* of δ_1, δ_2 if for any two situations S, S' with $S \sqsubseteq S'$: $\mathcal{D} \models$ $Do(\delta', S, S')$ iff $\mathcal{D} \models Do(\delta_1, S, S')$ and $\mathcal{D} \models Do(\delta_2, S, S')$, and there is at least one such pair S, S' such that $\mathcal{D} \models$ $Do(\delta_1, S, S')$ and $\mathcal{D} \models Do(\delta_2, S, S')$.

To understand the intuitive use of this, consider two programs, both describing some partial knowledge of the same target procedure of going to another room x:

> [*leaveCurrentRoom*; $((\pi v)v)^*$] [$((\pi z)z)^*$; *enterRoom*(x)]

One possible, and in fact the most general, intersection of these is [*leaveCurrentRoom*; $((\pi v)v)^*$; *enterRoom*(x)].

We conjecture that the problem of (constructively) proving the existence of an intersection is undecidable for general Golog programs and do not consider it any further.¹ Instead we only consider a restricted form of Golog programs without while-loops and recursive procedures, and where (bounded) non-deterministic iteration is only allowed over primitive actions and expressions of the form $(\pi v)a(v)$ or $(\pi v)v$, i.e., non-deterministic choice of action arguments or actions themselves. For clarity, we refer to this language as Golog⁻. For this class of programs we can define the set of axioms shown in Figure 2, denoted Σ_{sync} , regarding a new predicate sync that can be used to constructively prove the existence of an intersection of two programs, i.e., generate a program that represents the intersection of the version spaces. This representation is very compact, as it uses the non-deterministic constructs of Golog to represent the space of procedures still considered to be a candidate for the target procedure. For ease of presentation we assume-without loss of generality-that all programs are sequences (e.g., [A; nil] instead of A). We use the notation nondet(Z), where Z is a set, to denote the nondeterministic choice between the elements of the set, i.e., for $Z = \{z_1, \ldots, z_n\}$, nondet $(Z) \stackrel{\text{def}}{=} (z_1 \mid \cdots \mid z_n)$. Our restrictions on non-deterministic iteration ensure that this set always turns out to be finite². This set of axioms again can be implemented and reasoned about efficiently in Prolog as in contained in our implementation.

Theorem 2. Let δ_1, δ_2 be two Golog⁻ programs over some action theory. If there exists a Golog⁻ program δ' such that $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$, then it is an intersection of δ_1, δ_2 . Further, if there exists an intersection of δ_1, δ_2 , then there is a δ' s.t. $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$.

Note that the theorem does not state that all possible intersections can be identified using *sync*. This is because there may be infinitely many possible intersections, all of which characterizing the same sub-set of the version space. Instead, only one such representation is found. This, however, does not cause practical limitations.

4 Example

To demonstrate how the presented framework can be used to flexibly combine PbD and PbI we consider an extended example. We assume that the correct, intended target procedure is as follows (or equivalent to):

$$P_{1} = \text{while} (\exists x).OnFloor(x) \text{ do } [PickNextItem; \\ \text{ if } Toy(Item) \text{ then } Put(Item, Box) \\ \text{ else if } Clothing(Item) \text{ then} \\ [goto(Closet); Put(Item, Hanger)] \\ \text{ else if } Book(Item) \text{ then } Put(Item, Shelf) \\ \text{ else } [goto(Bin); Put(Item, Bin)]]$$

 $sync(\delta_1, \delta_2, \delta') \equiv sync'(\delta_1, \delta_2, \delta') \lor sync'(\delta_2, \delta_1, \delta')$ $sync'(nil, nil, y) \equiv y = nil$ $sync'([a; b_1], [a; b_2], y) \equiv (\exists z).sync(b_1, b_2, z) \land y = [a; z]$ $sync'([\varphi?;x], \delta_2, y) \equiv (\exists z).sync(x, \delta_2, z) \land y = [\varphi?;z]$ $sync'([\mathbf{if} \varphi \mathbf{then} \delta_a \mathbf{else} \delta_b; x], \delta_2, y) \equiv$ $((\exists z_a).sync([\delta_a; x], \delta_2, z_a) \land$ $((\exists z_b).sync([\delta_b; x], \delta_2, z_b) \land y =$ if φ then z_a else $z_b) \lor$ $((\exists z_b).sync([\delta_b; x], \delta_2, z_b) \land y = [\varphi?; z_a])) \lor$ $((\not\exists z_a).sync([\delta_a; x], \delta_2, z_a) \land$ $(\exists z_b).sync([\delta_b; x], \delta_2, z_b) \land y = [\neg \varphi?; z_b])$ $sync'([(\delta_a \mid \delta_b); x], \delta_2, y) \equiv$ $((\exists z_a).sync([\delta_a; x], \delta_2, z_a) \land$ $((\exists z_b).sync([\delta_b; x], \delta_2, z_b) \land y = (z_a \mid z_b)) \lor$ $((\exists z_b).sync([\delta_b; x], \delta_2, z_b) \land y = z_a)) \lor$ $((\not\exists z_a).sync([\delta_a; x], \delta_2, z_a) \land$ $(\exists z_b)$.sync $([\delta_b; x], \delta_2, z_b) \land y = z_b)$ $sync'([(\pi v)\delta(v);x],\delta_2,y) \equiv ((\exists v',x').\delta_2 = [(\pi v')\delta(v');x'] \land$ $(\exists \delta').sync(x, x', \delta') \land y = [(\pi v)\delta(v); \delta']) \lor$ $((\exists v', x').\delta_2 = [(\pi v')\delta(v'); x'] \land (\exists Z).y = nondet(Z) \land$ $Z \neq \emptyset \land (\forall z). z \in Z \equiv (\exists q). sync([\delta(q); x], \delta_2, z))$ $sync'([a(v)^*; x], \delta_2, y) \equiv ((\exists x').\delta_2 = [a(v)^*; x'] \land$ $(\exists \delta').sync([a(v)^*; x], x', \delta') \land y = [a(v)^*; \delta']) \lor$ $(\exists x').\delta_2 = [((\pi v')a(v'))^*; x'] \land$ $(\exists \delta').sync([a(v)^*; x], x', \delta') \land y = [a(v)^*; \delta']) \lor$ $\left((\exists x')\delta_2 = [a(v)^*; x'] \land (\exists x')\delta_2 = [(\pi v)a(v)^*; x'] \land\right)$ $(\exists Z).y = nondet(Z) \land Z \neq \emptyset \land$ $(\forall z).z \in Z \equiv (sync([a(v); a(v)^*; x], \delta_2, z) \lor sync(x, \delta_2, z)))$ $sync'([((\pi v')a(v'))^*;x], \delta_2, y) \equiv ((\exists x').\delta_2 = [a(v)^*;x'] \land$ $(\exists \delta').sync([a(v)^*; x], x', \delta') \land y = [a(v)^*; \delta']) \lor$ $(\exists x').\delta_2 = [((\pi v)a(v))^*; x'] \land y = [((\pi v')a(v'))^*; \delta'] \land$ $(\exists \delta').sync([((\pi v')a(v'))^*;x],x',\delta')) \lor$ $\left((\not\exists x')\delta_2 = [a(v)^*; x'] \land (\not\exists x')\delta_2 = [(\pi v)a(v)^*; x'] \land\right)$ $(\exists Z).y = nondet(Z) \land Z \neq \emptyset \land$ $(\forall z).z \in Z \equiv (sync([((\pi v')a(v')); ((\pi v')a(v'))^*; x], \delta_2, z) \lor$ $sync(x, \delta_2, z))$



Following the approach (i) of Section 1.1 the user may first provide the rough structure of the program,

$$P_2 = \text{while} (\exists x).OnFloor(x) \text{ do}$$

[PickNextItem; $Process(x)$]

where Process(x) is the procedure defined in Section 3.2. To further limit the scope of conditions, she may define as part of the background theory $cond(c) \equiv (\exists f, o).Fluent(F) \land c = f(o)$ and ensure that fluents only hold for finitely many objects. Given

¹Our intuition stems from the problem's similarity to the difficult problem of determining program equivalence.

²Intuitively this is because δ^* 's—which are the only source of infinity—are unified when they "meet" in *sync*.

this, she can proceed by demonstrating what to do for individual items, by showing situations where only one item is lying on the floor. Consider a situation S_1 where only one shirt is on the floor, and the demonstration $S'_1 =$ $do([PickNextItem; goto(Closet); Put(Item, Hanger)], S_1)$. Using our Prolog implementation we can prove that $\mathcal{D} \models$ $(\exists \delta').Do'(P_1, S_1, S'_1, \delta')$ holds for a (finite) number of programs δ' , all of which are like P_1 but replace the Process(x)by different sub-programs: one program replaces it by [PickNextItem; goto(Closet); Put(Item, Hanger)], all others replace it by

 $[\mathbf{if} F(O)]$

then [PickNextItem; goto(Closet); Put(Item, Hanger)]
else nil;

 $((\pi c)[cond(c)?; if c then Stow(x) else nil])^*]$

for some fluent F and some object O, such that $\mathcal{D} \models F(O)[do(PickNextItem, S)]$. That is, any fluent literal that is true in the demonstration after picking up the item is a candidate for the condition. Many of those will be found inconsistent as soon as a second demonstration, regarding a different type of item, is considered. Any such program can be readily executed, randomly picking a place to store any item not covered by the hypothesized condition (if the chosen condition at all regards the current item).

5 Discussion

We have described a formal framework that implements a version space algebra approach to PbD as an extension of the logic programming language Golog. This enables the integration of programming by demonstration with explicit programming by instruction and further allows for the symbolic intersection of version spaces. Our use of Golog also facilitates extending PbD in other ways. For instance, one might use planning in order to resolve some of the non-determinism remaining in a procedure hypothesis if it needs to be executed before a deterministic target procedure has been learned. Also learning from experience is facilitated.

Finally, also Blythe et al. (2007) aim to integrate PbD and PbI, though, focusing on a particular domain. Unfortunately, the exposition does not provide sufficient technical detail to understand the extend to which the goal of integration has been achieved in that work in generality and to compare the described system to our framework.

Other possible directions for future work include: the study of other algorithms for searching the space of hypotheses in different ways, e.g., guided by some specification of most likely choices in the program, or based on heuristics; considering the feasibility of PbD under incomplete knowledge of the state, or in the face of uncertain action effects; investigating the effects of the availability of detailed background knowledge about the domain and its actions; and considering environments with continuous state and action domains and possibly durative actions. In all of these areas it might be possible to readily build on existing work to extend the scope of PbD accordingly.

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A Appendix

A.1 Proof of Theorem 1

We prove the theorem in two steps. First we show the theorem for single demonstrations, by induction over the structure of the program. The result then trivially extends to sequences of demonstrations via induction over the length of the sequence, where the induction step is provided by the lemma of the first part.

Proof for single demonstrations

Lemma 1. Let $M = \langle \delta, \{(S_1, S'_1)\} \rangle$ be a PbD problem. Any program δ' such that $\mathcal{D} \models Do'(\delta, S_1, S'_1, \delta')$ is a solution to M.

Proof: The proof proceeds by induction over the structure of the program, where for each item we show the two items (1.+2.) of Definition 1 separately. Note that showing the second item boils down to proving deterministic behavior of the program along the action sequence described by the situation in question.

 $\delta = nil, \delta = a, \text{ or } \delta = \varphi?$

By definition, $\delta' = \delta$, and hence, both items follow trivially: the programs are equal and deterministic.

 $\delta = [\delta_1; \delta_2]$

By definition, $\delta' = [\delta'_1; \delta'_2]$. Item 1.: By definition and Proposition 1 there is a situation S^* such that $\mathcal{D} \models Do(\delta'_1, S, S^*)$ and $\mathcal{D} \models Do(\delta'_2, S^*, S')$. Hence, by induction hypothesis, also $\mathcal{D} \models Do(\delta_1, S, S^*)$ and $\mathcal{D} \models Do(\delta_2, S, S^*)$. Therefore, by definition of Do, $\mathcal{D} \models Do(\delta, S, S')$.

Item 2.: By induction hypothesis both subprograms δ'_1 and δ'_2 are deterministic, hence, also δ' .

$\delta = \mathbf{if} \varphi \mathbf{fhen} \, \delta_1 \, \mathbf{else} \, \delta_2$

There are two cases: either $\mathcal{D} \models \varphi[S_1]$ or not. Since these two cases are quite analogous, we only consider the first, hence, $\delta' = \mathbf{if} \varphi$ then δ'' else δ_2 where δ'' is such that $\mathcal{D} \models Do'(\delta_1, s, s', \delta'')$.

For Item 1 we need to distinguish two more cases: If $\mathcal{D} \models \varphi[S]$ then δ'' is executed, otherwise, δ_2 is executed. In the first case Item 1 follows by induction hypothesis. In the second it follows trivially, since the program being executed is the same.

For Item 2, since we assume $\mathcal{D} \models \varphi[S_1]$, we only need to show that δ'' is deterministic along the S_1, S'_1 trajectory, which follows by induction hypothesis.

$\delta =$ while φ do δ_1

From the induction principle of the definition of Do' for this case, we get that $\delta' =$ while φ do δ'_1 where δ'_1 is such that there is a finite sequence of situations S^1, \ldots, S^n and a corresponding sequence of programs $\delta^1, \ldots, \delta^n$ such that $\delta^1 = \delta_1, \, \delta^n = \delta'_1$, and $S_1 = S^1 \sqsubset \cdots \sqsubset S^n = S'_1$ and for all $1 \le i < n, \mathcal{D} \models Do'(\delta^i, S^i, S^{i+1}, \delta^{i+1})$ and $\mathcal{D} \models \varphi[S^i]$, and, of course, $\mathcal{D} \models \neg \varphi[S'_1]$.

We show Items 1 and 2 by two separate inductions over the length of this sequence (once backward, once forward). We refer to this induction as the inner induction to distinguish it from the outer induction (over the structure of the program).

Item 1: The base case for Item 1 is when n = 0 which corresponds to the case where $\mathcal{D} \models \neg \varphi[S_1]$. Then the claim follows trivially since the refined program δ' is the same as the original (δ) , by definition.

For the induction step we assume the property holds for n = j. By outer induction hypothesis and Proposition 1 we have that $\mathcal{D} \models Do(\delta^j, \bar{S}, \bar{S}')$ implies $\mathcal{D} \models Do(\delta^{j-1}, \bar{S}, \bar{S}')$ for any pair of situations \bar{S}, \bar{S}' . And since inner induction hypothesis gives us that $\mathcal{D} \models Do(\delta^{j-1}, \bar{S}, \bar{S}')$ implies $\mathcal{D} \models$ $Do(\delta, \bar{S}, \bar{S}')$, again for all such situation pairs, it follows that $\mathcal{D} \models Do(\mathbf{while } \varphi \mathbf{do } \delta_j, \bar{S}, \bar{S}')$ implies $\mathcal{D} \models$ $Do(\mathbf{while } \varphi \mathbf{do } \delta, \bar{S}, \bar{S}')$. The claim now follows by induction for any n.

Item 2: The base case for Item 2 is when $\mathcal{D} \models \neg \varphi[S_1]$. The claim follows trivially since there is only one possible execution, namely the empty one: it must be true that $S'_1 = S_1$.

For the induction step, again we assume the property holds for n = j and hence, by applying the inductive definition of Do(j times):

$$\mathcal{D} \models [Do(\delta^{j}, S^{1}, S^{2}) \land \dots \land Do(\delta^{j}, S^{j-1}, S^{j})] \land \\ [((\forall s) Do(\delta^{j}, S^{1}, s) \supset s = S^{2}) \land \dots \land \\ ((\forall s) Do(\delta^{j}, S^{j-1}, s) \supset s = S^{j})].$$

From outer induction hypothesis we get that:

 $\mathcal{D} \models Do(\delta^{j+1}, S^j, S^{j+1}) \land$ $(\forall s).Do(\delta^{j+1}, S^j, s) \supset s = S^{j+1}$

and since by assumption $\mathcal{D} \models \neg \varphi[S^n]$ we get, by definition of Do (unwinding the string of j+1 sub-executions):

$$\begin{split} \mathcal{D} &\models Do(\textbf{while } \varphi \textbf{ do } \delta', S_1, S_1') \land \\ (\forall s). Do(\textbf{while } \varphi \textbf{ do } \delta', S_1, s) \supset s = S_1' \end{split}$$

i.e., the thesis of the inner induction. The claim now follows by induction for any n.

$$\delta = (\delta_1 | \delta_2)$$

Item 1: Without loss of generality we assume $\mathcal{D} \models Do'(\delta_1, S_1, S'_1, \delta')$. Then, by induction hypothesis: $\mathcal{D} \models Do(\delta', \overline{S}, \overline{S}')$ implies $\mathcal{D} \models Do(\delta, \overline{S}, \overline{S}')$ for any pair of situations $\overline{S}, \overline{S}'$. Hence, since of course $X \supset X \lor Y$, the claim.

Item 2: Again, Without loss of generality we assume $\mathcal{D} \models Do'(\delta_1, S_1, S'_1, \delta')$. Then, by induction hypothesis δ' is deterministic.

$$\delta = (\pi v)\delta(v)$$

There are two cases to consider: either the introduced program variable is bound to some term T during refinement, or it is still variable in δ' .

Item 1: In the first case we have from induction hypothesis that $\mathcal{D} \models Do(\delta', \overline{S}, \overline{S}')$ implies $\mathcal{D} \models Do(\delta(T), \overline{S}, \overline{S}')$ for any pair of situations $\overline{S}, \overline{S}'$. Hence, also $\mathcal{D} \models$ $Do(\delta', S, S')$ implies $\mathcal{D} \models (\exists x) Do(\delta(x), \overline{S}, \overline{S}')$.

In the second case, Item 1 follows directly from induction hypothesis.

Item 2: follows directly by induction hypothesis.

 $\delta=\delta^*$

This case is quite analogous to that for while-loops.

From the induction principle of the definition of Do' for this case, we get that $\delta' = \delta^1; \ldots; \delta^n$ for some n > 0such that there is a corresponding sequence of situations S^1, \ldots, S^n such that $S_1 = S^1 \sqsubset \cdots \sqsubset S^n = S'_1$ and for all $1 \le i < n, \mathcal{D} \models Do'(\delta, S^i, S^{i+1}, \delta^{i+1})$.

We show Items 1 and 2 by two separate inductions over the length of this sequence (once backward, once forward). We refer to this induction as the inner induction to distinguish it from the outer induction (over the structure of the program).

Item 1: The base case for Item 1 is when n = 0. Then δ' is the empty program *nil*, which does not admit any executions other than the empty one. Hence, S' = S and by definition of Do for δ^* we have that indeed $\mathcal{D} \models Do(\delta^*, S, S)$.

For the induction step we assume the property holds for n = j. By outer induction hypothesis and Proposition 1 we have that $\mathcal{D} \models Do(\delta^j, \bar{S}, \bar{S}')$ implies $\mathcal{D} \models Do(\delta, \bar{S}, \bar{S}')$ for any pair of situations \bar{S}, \bar{S}' . And since inner induction hypothesis gives us that $\mathcal{D} \models$ $Do(\delta^1; \ldots; \delta^{j-1}, \bar{S}, \bar{S}')$ implies $\mathcal{D} \models Do(\delta^*, \bar{S}, \bar{S}')$, again for all such situation pairs, it follows that $\mathcal{D} \models$ $Do(\delta^1; \ldots; \delta^j, \bar{S}, \bar{S}')$ implies $\mathcal{D} \models Do(\delta^*, \bar{S}, \bar{S}')$ as well. The claim now follows by induction for any n.

Item 2: In the base case, n = 0, δ' is again the empty program *nil*, which is trivially deterministic.

For the induction step, again we assume the property holds for n = j and hence, by definition of Do for sequences that:

$$\mathcal{D} \models Do(\delta^1; \dots; \delta^j, S_1, S^j) \land$$
$$((\forall s) Do(\delta^1; \dots, \delta^j, S_1, s) \supset s = S^j)$$

From outer induction hypothesis we get that:

$$\begin{aligned} \mathcal{D} &\models Do(\delta^{j+1}, S^j, S^{j+1}) \land \\ (\forall s). Do(\delta^{j+1}, S^j, s) \supset s = S^{j+1} \end{aligned}$$

Hence:

$$\begin{aligned} \mathcal{D} &\models Do(\delta^1; \dots; \delta^{j+1}, S_1, S_1') \land \\ (\forall s). Do(\delta^1; \dots; \delta^{j+1}, S_1, s) \supset s = S_1' \end{aligned}$$

i.e., the thesis of the inner induction. The claim now follows by induction for any n.

Proof for the general case For the general case, i.e., for a PbD problem with a sequence of demonstrations, the two items of the definition of solution follow by induction over the length of the sequence of demonstrations.

Item 1: The base case is for the empty sequence, in which case the refined program δ' is the same as the original δ . The induction step is provided by Lemma 1.

Item 2: The base case is provided my Lemma 1. For the induction step, we need to show that after refining program δ_i using the demonstration described by S_{i+1}, S'_{i+1} , obtaining δ_{i+1} , it is still the case that

$$\mathcal{D} \models Do(\delta_i, S_i, S'_i) \land (\forall s). Do(\delta_i, S_i, s) \supset s = S'_i$$

i.e., that further refinement of a program never disables previously considered demonstrations (first conjunct) and never introduces additional non-determinism (second conjunct).

The first conjunct follows from Lemma 1: the program δ_i is deterministic over the trajectory S_i, S'_i , and by assumption $\mathcal{D} \models (\exists \delta^*) Do'(\delta_i, S_{i+1}, S'_{i+1}, \delta^*)$, namely $\delta^* = \delta_{i+1}$, and Do' only modifies the non-deterministic parts of a program.

The second conjunct follows by inspection of the axioms for Do': we observe that in none of the cases the refined program δ' mentions any non-deterministic sub-programs that were not already part of the original program.

A.2 Proof of Theorem 2

Applying the definition of intersection, we can rewrite the theorem as follows:

Theorem 3. Let δ_1, δ_2 be two Golog⁻ programs over some action theory, and S, S' two situations from that theory such that $S \sqsubseteq S'$. (i) If there exists a Golog⁻ program δ' such that $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$, then $\mathcal{D} \models Do(\delta', S, S')$ iff $\mathcal{D} \models$ $Do(\delta_1, S, S')$ and $\mathcal{D} \models Do(\delta_2, S, S')$. Further, (ii) if $\mathcal{D} \models$ $Do(\delta_1, S, S')$ and $\mathcal{D} \models Do(\delta_2, S, S')$, then there is a δ' s.t. $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$.

The theorem states that synchronization is a sound and complete means of determining intersections of programs. We prove these two pieces separately. The theorem follows directly from these two lemmata.

Proof of Completeness

Lemma 2 (Completeness). Let δ_1, δ_2 be two Golog⁻ programs over some action theory, and S, S' two situations from that theory such that $S \sqsubseteq S'$. If $\mathcal{D} \models Do(\delta_1, S, S')$ and $\mathcal{D} \models Do(\delta_2, S, S')$, then there is a δ' s.t. $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$.

Proof: By contradiction: If there is no δ' s.t. $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$ then either $\mathcal{D} \not\models Do(\delta_1, S, S')$ or $\mathcal{D} \not\models Do(\delta_2, S, S')$.

This is shown by induction over the structure of δ_1 .

The base case is given by $\delta_1 = nil$. Then, the only way to "fail" is when δ_2 is a non-empty sequence (not nil), starting with a primitive action, since in all other cases, that program could be further expanded upon (using the second disjunct of *sync*). Hence, we have $\delta_2 = [a; b_2]$ for some action aand some sub-program b_2 . By definition of Do, δ_1 only accepts pairs of equal situations, whereas δ_2 only accepts pairs of situations S, S', where $S' = do(\vec{a}, do(a, S))$, for some, possibly empty sequence of actions \vec{a} . Hence the thesis.

The cases for the induction step are as follows.

$\delta_1 = [a; b_1]$ with a a primitive action

Then, there doesn't exists a program δ' such that $\sum_{sync} \models sync(\delta_1, \delta_2, \delta')$ iff δ_2 doesn't start with a, or $\delta_2 = [a; b_2]$ and $\sum_{sync} \nvDash sync(b_1, \delta_2, \delta')$. In the former case, the thesis follows as in the base case, by definition of Do, and in the later by induction hypothesis for b_1 .

 $\delta_1 = [\varphi?; x]$

Since $sync'([\varphi?; x], \delta_2, y)$ is defined for any δ_2 in terms of $sync(x, \delta_2, z)$, the thesis follows by induction hypothesis for the case for sub-program x.

 $\delta_1 = [\mathbf{if} \varphi \mathbf{then} \ \delta_a \ \mathbf{else} \ \delta_b; x]$

There are four possible cases:

1.
$$\Sigma_{sync} \models (\exists z) sync([\delta_a; x], \delta_2, z)$$
 and $\Sigma_{sync} \models (\exists z) sync([\delta_b; x], \delta_2, z),$

- 2. only $\Sigma_{sync} \models (\exists z) sync([\delta_a; x], \delta_2, z)$, or
- 3. only $\Sigma_{sync} \models (\exists z) sync([\delta_b; x], \delta_2, z)$, or

4. neither.

Only in the latter case, there is no synchronized program. The thesis follows from induction hypothesis and definition of Do for **if**.

 $\delta_1 = \left[(\delta_a \mid \delta_b); x \right]$

This case is analogous to the previous: the only possibility for this case to "fail", i.e., there not being a synchronized program, is when neither $\sum_{sync} \models (\exists z) sync([\delta_a; x], \delta_2, z)$ nor $\sum_{sync} \models (\exists z) sync([\delta_b; x], \delta_2, z)$. The thesis hence again follows from induction hypothesis for these cases and definition of Do for the simple non-deterministic choice (at least one of the programs needs to be executable).

 $\delta_1 = [(\pi v)\delta(v); x]$

Again, since *sync'* for this case is recursively defined for any program δ_2 , the thesis follows by induction hypothesis for *x*.

 $\delta_1 = [\delta^*; x]$ See the previous case.

Proof of Soundness

Lemma 3 (Soundness). Let δ_1, δ_2 be two Golog⁻ programs over some action theory. If there exists a Golog⁻ program δ' such that $\Sigma_{sync} \models sync(\delta_1, \delta_2, \delta')$, then for any two situations S, S' with $S \sqsubset S' \colon \mathcal{D} \models Do(\delta', S, S')$ iff $\mathcal{D} \models Do(\delta_1, S, S')$ and $\mathcal{D} \models Do(\delta_2, S, S')$.

Proof: By induction over the structure of δ_1 . We here assume without loss of generality that for δ_1 always the first disjunct of the axiom for *sync* applied, i.e., that $\Sigma_{sync} \models sync'(\delta_1, \delta_2, \delta')$.

The base case is when $\delta = nil$. Then, by definition of sync', both $\delta_2 = nil$ and $\delta' = nil$. The thesis for this case hence follows trivially (X iff X and X).

For the induction step, the cases are as follows.

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\delta_1 = [a; b_1] with a primitive action
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Then, by definition, $\delta_2 = [a; b_2]$ for some subprogram b_2 , and $\delta' = [a; z]$ for some z such that $\Sigma_{sync} \models sync(b_1, b_2, z)$. From the definition of Doit follows that $do(a, S) \sqsubset S'$, i.e., there is a (possibly empty) sequence of actions \vec{b} such that S' = $do(\vec{b}, do(a, S))$. By induction hypothesis we know that $\mathcal{D} \models Do(z, do(a, S), S')$ iff $\mathcal{D} \models Do(b_1, do(a, S), S')$ and $\mathcal{D} \models Do(b_2, do(a, S), S')$. Hence, from definition of Do we get that also $\mathcal{D} \models Do([a; z], S, S')$ iff $\mathcal{D} \models Do([a; b_1], S, S')$ and $\mathcal{D} \models Do([a; b_2], S, S')$, i.e., the thesis for this case.

$$\delta_1 = [\varphi?; x]$$

By definition of Do, $\mathcal{D} \models \varphi[S]$. By definition of sync', $\delta' = [\varphi?; z]$ and by induction hypothesis, $\mathcal{D} \models Do(z, S, S')$ iff $\mathcal{D} \models Do(x, S, S')$ and $\mathcal{D} \models$ $Do(\delta_2, S, S')$. From definition of Do hence also $\mathcal{D} \models$ $Do([\varphi?; z], S, S')$ iff $\mathcal{D} \models Do([\varphi?; x], S, S')$ and $\mathcal{D} \models$ $Do(\delta_2, S, S')$, i.e., the thesis.

 $\delta_1 = [\mathbf{if} \varphi \mathbf{then} \delta_a \mathbf{else} \delta_b; x]$ There are three cases to consider:

1. $y = \mathbf{if} \varphi$ then z_a else z_b , then if $\mathcal{D} \models \varphi[S]$ the thesis follows by induction hypothesis for $[\delta_a; x]$ and definition of Do, otherwise it follows by induction hypothe-

sis for $[\delta_b; x]$ and again definition of Do.

- y = [φ?; z_a]. For the negation in this case we need the completeness lemma. For the ⇒ direction: it has to be that D ⊨ φ[S], and hence the thesis follows by induction hypothesis for [δ_a; x]. For the ⇐ direction: By Lemma 2 we know that since there is no z such that Σ_{sync} ⊨ sync([δ_b; x], δ₂, z) that also it cannot be the case that D ⊨ Do([δ_b; x], S, S') ∧ Do(δ₂, S, S'). Hence, if D ⊨ ¬φ[S], then neither side of the "iff" hold. Otherwise, if D ⊨ φ[S], the thesis follows from induction hypothesis for [δ_a; x].
- 3. $y = [\neg \varphi?; z_b]$. This case follows in complete analogy to the previous.

$$\delta_1 = \left[(\delta_a \mid \delta_b); x \right]$$

This case follows in analogy to the previous.

 $\delta_1 = [(\pi v)\delta(v); x]$ By definition of *sync'*, there are two possible cases:

1.
$$\delta_2 = [(\pi v')\delta(v'); x']$$

2. or not.

In the first case, $y = [(\pi v)\delta(v); \delta']$. Then by definition of Do there is a situation S^* such that $S \sqsubseteq S^* \sqsubseteq S'$ and $\mathcal{D} \models Do((\pi v)\delta(v), S, S^*)$, or else neither $\mathcal{D} \models$ Do(y, S, S') nor $\mathcal{D} \models Do(\delta_1, S, S')$. If the said situation S^* exists, then, since $\Sigma_{sync} \models sync(x, x', \delta')$, we have by induction hypothesis that $\mathcal{D} \models Do(\delta', S^*, S')$ iff $\mathcal{D} \models Do(x, S^*, S')$ and $\mathcal{D} \models Do(x', S^*, S')$. Hence, once again by definition of Do for sequences the thesis.

In the second case, y = nondet(Z) where Z is a set of programs z all of which satisfy $\sum_{sync} \models (\exists q) sync([\delta(q); x], \delta_2, z)$, and furthermore by definition of sync' for this case, there is no q' such that at the same time, $\sum_{sync} \models sync([\delta(q'); x], \delta_2, z)$ and $z \notin Z$. Hence, for every choice of q in δ_1 there is a corresponding program z in Z that was generated by synchronizing that instance of δ_1 with δ_2 and vice versa. The thesis now follows by definition of Do for nondet (simple non-deterministic choice of any program in the set), and induction hypothesis for each possible value for q.

 $\delta_1 = [a(v)^*; x]$

By definition of *sync'*, there are three possible cases: 1. $\delta_2 = [a(v)^*; x'],$ 2. $\delta_2 = [((\pi v')a(v'))^*; x'],$ 3. or neither.

In the first case, $y = [a(v)^*; \delta']$. Then by definition of Do there is a situation S^* such that $S \sqsubseteq S^* \sqsubseteq S'$ and $\mathcal{D} \models Do(a(v)^*, S, S^*)$, or else neither $\mathcal{D} \models Do(y, S, S')$ nor $\mathcal{D} \models Do(\delta_1, S, S')$. If the said situation S^* exists, then, since $\Sigma_{sync} \models sync([a(v)^*; x], x', \delta')$, we have by induction hypothesis that $\mathcal{D} \models Do(\delta', S^*, S')$ iff $\mathcal{D} \models Do([a(v)^*; x], S^*, S')$ and $\mathcal{D} \models Do(x', S^*, S')$. Hence, once again by definition of Do for sequences the thesis. (Note that the induction is well defined, since the second program δ_2 gets strictly monotonically smaller.)

In the second case, $y = [a(v)^*; \delta']$. Then by definition of Do there is a situation S^* such that $S \sqsubseteq S^* \sqsubseteq S'$ and $\mathcal{D} \models Do(a(v)^*, S, S^*)$, or else neither $\mathcal{D} \models Do(y, S, S')$ nor $\mathcal{D} \models Do(\delta_1, S, S')$. If the said situation S^* exists, then, since $\Sigma_{sync} \models sync([a(v)^*; x], x', \delta')$, we have by induction hypothesis that $\mathcal{D} \models Do(\delta', S^*, S')$ iff $\mathcal{D} \models Do([a(v)^*; x], S^*, S')$ and $\mathcal{D} \models Do(x', S^*, S')$. Hence, once again by definition of Do for sequences and for π the thesis.

In the third case, y = nondet(Z) where Z is a set of programs z all of which satisfy $\sum_{sync} \models$ $sync([a(v); a(v)^*; x], \delta_2, z) \lor sync(x, \delta_2, z)$. Hence, for each of the two choices of executing a(v) at least once or not, there is a corresponding program z in Z that was generated by synchronizing that choice with δ_2 and vice versa. The thesis now follows by definition of Do for *nondet* (simple non-deterministic choice of any program in the set), and induction hypothesis for the programs resulting from these choices.

It is important to note our restriction that δ can only be a primitive action (possibly with a non-deterministic choice of argument). Otherwise, the induction would not be well defined in the third case, since the first program δ_1 can increase in size. However, given the restriction and the fact that, by assumption for this case, δ_2 does not start with δ^* or a non-deterministic iteration over a term of the form $(\pi v)a(v)$ that could match δ , the second program strictly monotonically decreases in size.

 $\delta_1 = [((\pi v)a(v))^*; x]$ In analogy to the previous case.