

INDUCTIVE SYNTHESIS OF
RECURSIVE LOGIC PROGRAMS

A THESIS
SUBMITTED TO THE DEPARTMENT OF COMPUTER
ENGINEERING AND INFORMATION SCIENCE
AND THE INSTITUTE OF ENGINEERING AND SCIENCE
OF BILKENT UNIVERSITY
IN PARTIAL FULFILLMENT OF THE REQUIREMENTS
FOR THE DEGREE OF
MASTER OF SCIENCE

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August 1997

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ABSTRACT

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M.S. in Computer Engineering and Information Science

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August 1997

The learning of recursive logic programs (i.e. the class of logic programs where at least one clause is recursive) from incomplete information, such as input/output examples, is a challenging subfield both of ILP (Inductive Logic Programming) and of the synthesis (in general) of logic programs from formal specifications. This is an extremely important class of logic programs, as the recent work on constructive induction shows that necessarily invented predicates have recursive programs, and it even turns out that their induction is much harder than the one of non-recursive programs. We call this *inductive program synthesis*. We introduce a system called DIALOGS-II (Dialogue-based Inductive and Abductive LOgic Program Synthesizer-II) whose ancestor is DIALOGS. It is a schema-guided, interactive, and non-incremental synthesizer of recursive logic programs that takes the initiative and queries a (possibly naive) specifier for evidence in her/his conceptual language. It can be used by any learner (including itself) that detects, or merely conjectures, the necessity of invention of a new predicate. Moreover, due to its powerful codification of “recursion-theory” into program schemata and schematic constraints, it needs very little evidence and is very fast.

Keywords: program development, inductive logic programming, automatic program synthesis, schema-guided program synthesis.

ÖZET

ÖZYİNELİ MANTIK PROGRAMLARININ TÜMEVARIMSAL YOLLA SENTEZİ

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Ağustos 1997

Özyineli mantık programlarının (en azından bir yantümcesi özyineli olan) tam olmayan bilgiden yola çıkılarak, mesela, girdi/çıkı örneklerinden, otomatik sentezi oldukça zor bir iştir. Ve bu iş tümevarımsal mantık programlama ile otomatik program sentezinin bir alt çalışma alanıdır. Bu tür programlar mantık programlarının çok önemli bir sınıfını oluştururlar. Yapıcı tümevarım çalışmaları göstermiştir ki özyineli programların sentezi özyineli olmayan programların sentezinden çok daha zordur. Bu çalışma alanı “tümevarımsal program sentezi” diye anılır. DIALOGS-II adıyla geliştirdiğimiz sistem (bu sistemin bir önceki versiyonu DIALOGS adlı sistemdir) taslak-yönetimli, interaktif ve artımsızdır. Sistem inisiyatifi alıp kullanıcıyı kullanıcının dilinde sorgulayarak özyineli mantık programları sentezler. Sistem kendisi tarafından özyineli olarak ya da başka bir sistem tarafından, sistem özyineli bir programın sentezinin gerekliliğini farkettiği zaman kullanılabilir. “Özyineleme Teorisi” sistemin içinde taslaklar tarafından etkili bir şekilde kodlandığı için sistem çok az bilgiye gerek duyar ve çok hızlı çalışır.

Anahtar Sözcükler: program geliştirme, tümevarımsal mantık programlama, otomatik program sentezi, taslak yönetimli program sentezi.

ACKNOWLEDGEMENTS

I am very grateful to my supervisor, Ass't Prof. Pierre Flener, for his invaluable guidance during this study. His instruction will be the closest and most important reference in my future work. I would also like to thank Halime Büyükyıldız and Esra Erdem for sharing their work with me, my family and my dear friend Serap Fırat for their moral and motivating support.

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Chapter 1

Introduction

In its most general form, the task of Inductive Logic Programming (ILP) is to infer a hypothesis H from assumed-to-be-incomplete information (or: evidence) E and background knowledge B such that $B \wedge H \models E$, where H , E , and B are sets of clauses. We say that H covers E (in B). In practice, B and H are often restricted to sets of Horn clauses (i.e. definite logic programs). Evidence E is usually divided into positive evidence E^+ and negative evidence E^- . Often, the clauses of E^+ are restricted to ground positive literals (or: atoms) and are called positive examples, whereas those of E^- are restricted to ground negative literals and are called negative examples: this yields an extensional description, whereas the hypothesis is an intensional description. In a more traditional machine learning terminology, we would say that a concept description H is to be learned from descriptions E of instances and counter-examples of concepts, whose features are represented by predicate symbols. In general thus, nothing restricts the evidence to be about a single concept, so that multiple (possibly related) concepts may have to be learned at the same time.

For instance, given the positive examples (in the left column) and negative examples (in the right column)

subset([],[])	\neg subset([k],[])
subset([],[a,b])	\neg subset([n,m,m],[m,n])

subset([d,c],[c,e,d])
subset([h,f,g],[f,i,g,h,j])

and given as background knowledge (among others) the logic program

select(X,[X|Xs],Xs) ←
select(X,[H|Ys],[H|Zs]) ← select(X,Ys,Zs)

a possible hypothesis is the logic program

subset([],Xs) ←
subset([X|Xs],Ys) ← select(X,Ys,Zs), subset(Xs,Zs)

though at this point we do not wonder how this could be feasible. The main issue is that we human beings can perform this kind of task, so that the question arises whether a machine can be designed to do it also. The usefulness of such a machine is undeniable as it would be a step towards a form of human/machine communication that more closely models inter-human communication, which usually features a lot of incomplete (and hence ambiguous) information, of course in the presence of background knowledge, and even noisy information. In the following two sub-sections, we will first introduce some terminology and theoretical results (Section 1.1) and next we will present our objective (Section 1.2).

1.1 Terminology and Theoretical Results

We now introduce some terminology (in Section 1.1.1 to Section 1.1.3 and in Section 1.1.6) and mention some theoretical results (in Section 1.1.4 and Section 1.1.5) concerning the induction of recursive clauses.

1.1.1 Approaches and Extensions to ILP (and Inductive Synthesis)

Whether for ILP in general or synthesis in particular, there is additional terminology due to different approaches as well as extensions to the ILP task, all of which we now discuss in a loosely connected fashion.

Often, the agent that provides the inputs to an ILP technique is called the teacher, whereas the ILP technique is called the learner and is said to perform learning. Such a machine learning terminology is misleading [17], and we shall use the more general terminology of *source*, *induction technique*, and *induction* instead.

An *intended relation* is the entire (possibly infinite) relation represented by a predicate symbol. In an ILP task, only *incomplete* information (called evidence) is available, i.e. it does not describe superset(s) of the intended relation(s). We here assume that the evidence has *correct* information, i.e. that it describes subset(s) of the intended relation(s). In this case, one also says that there is no *noise*. Often, the actually described subset(s) are finite. An extreme case of incomplete but correct information is complete and correct information, though this can often only be achieved through some (finite) axiomatization in the hypothesis language, but not in the evidence language.

We partition relations into *semantic manipulation* relations and *syntactic manipulation* relations, depending on whether the actual constants occurring in a ground tuple are relevant or not for deciding whether that tuple belongs to a relation. For instance, **subset** is a syntactic manipulation relation, because it treats constants like variables, whereas **sort** and **insert** would be semantic manipulation relations (see Section 1.1.5).

Induction can be viewed as *search* through a graph (or: search space) where the nodes correspond to hypotheses and the arcs correspond to hypothesis-transforming operators. As usual, the challenge is to efficiently navigate through such a search space, via intelligent control (e.g., by organizing the search space according to a partial order and using pruning techniques).

Induction may be *interactive* or *passive*, depending on whether the technique asks *questions* (or: *queries*) to some *oracle* (or: *informant*) or not. The oracle may or may not be the source. The questions may be of various kinds, such as the request for classification of invented examples as positive or negative ones.

Induction may be *incremental* or *non-incremental*, depending on whether evidence is input one-at-a-time with occasional output of (external) intermediate hypotheses, or input all-at-once with output of a unique final hypothesis (though there may be internal intermediate approximations, which are however not considered as hypotheses).

Induction may be *bottom-up* or *top-down*, depending on whether hypotheses (whether internal or external) monotonically evolve from the maximally specific one (namely the empty logic program) or from the maximally general one (namely a logic program succeeding on all possible queries).

In the output hypothesis, some predicate symbols may be recursively defined: the corresponding clauses are partitioned into *base clauses* and *recursive clauses*.

Once a hypothesis is accepted (for whatever reasons), one may want to validate it. Since there is no complete description of the intended relation(s), one can only test the hypothesis, rather than somehow mathematically verifying it. Ideally, a hypothesis covers all the given evidence. One may thus test the hypothesis by measuring its accuracy (expressed in percents) in correctly covering other evidence. The given evidence is thus also called the *training set*, whereas the additional evidence is called the *test set* and is usually in the evidence language. We here assume that the test set is also correct w.r.t. the intended relation(s).

An *identification criterion* defines the moment where an induction technique has been successful in correctly identifying the intended relation(s), whether it “knows” this or not. Sample criteria are finite identification, identification-in-the-limit, probably-approximately-correct (PAC) identification, and so on (see [21] for details). There are limiting theorems stating what hypothesis languages are inducible from what evidence language under what identification criterion.

It seems desirable to achieve some separation of concerns regarding the logic and control components of algorithms (or logic programs): some techniques just induce the *logic* component, assuming that the control can be added later. Adding *control* (such as by clause re-ordering inside programs and literal re-ordering inside clauses so as to ensure safety of negation-by-failure, termination, etc.) is something specific to the (idiosyncrasies of the) execution mechanism of the target language, as well as specific to the desired ways of using the induced program (which are mentioned in additional inputs, see the next sub-section). If an interpreter of the target language is actually used during the induction (say, to verify the coverage of the evidence), such control aspects cannot be entirely ignored while constructing the logic component.

A generalization of the ILP task is known as *theory-guided induction*, or (inductive) *theory revision*, or *declarative debugging*: the idea here is that an additional input is provided, namely an initial hypothesis (or: theory) H_i , under the constraint that the final hypothesis H should be as close a “variant” thereof as possible, in the sense that only the “bugs” of H_i w.r.t. E should be (incrementally) found and corrected (or: “debugged”) in order to produce H . This generalized scheme reduces to the normal one in its extreme cases, that is when H_i is maximally specific or general, depending on whether induction proceeds bottom-up or top-down. In the past, this was also known

as *model-driven* or *approximation-driven* learning, as opposed to *data-driven* learning, where there is no initial theory.

Another variant of the ILP task involves augmenting the inputs with *declarative bias*, which is any form of input information that restricts the search space. There are two complementary approaches to this, and we discuss them separately in the next two sub-sections.

1.1.2 Additional Specification Information

A *specification* of a program contains (i) a description of what problem is (to be) solved by the program, as well as (ii) a description of how to use the program.

The former description should define the intended relation as declaratively as possible. Whether it should be informal or formal is an on-going debate, but we don't have a choice here, since we want it to be processed by a machine. Ideally, it should even be as complete as possible, but, as mentioned earlier, this is rarely achieved in practice. The problem descriptions investigated here (the evidence) are actually even assumed-to-be-incomplete. They are furthermore the most declarative (formal) descriptions that we can imagine (if they are constrained to be non-recursive [16]).

The latter description should give the predicate symbol representing the intended relation, the sequence of names and *types* of its formal parameters, *pre-conditions* (if any) on these parameters, as well as the representation conventions of the formal parameters so that one knows how to interpret their actual values. In logic programming, where we are concerned with relations rather than functions, there should also be an enumeration of the input/output *modes* in which the program may be called (since full reversibility is rarely required or rarely even achieved in practice), as well as optional *multiplicity* (or: *determinism*) information for each mode (stating the minimum and maximum number of correct answers to a query in that mode).

Since such information is part of a (useful) specification anyway, it is only natural to provide (some of) it as an additional input to an ILP task, especially for a program synthesis task. In the ILP literature, such information is usually called *semantic bias* (a kind of declarative bias that restricts the behavior of hypotheses), but we find this terminology insufficient, as it fails to establish the link with (good) specification practice. Type and mode information are the most commonly used, and, not surprisingly, they reduce search spaces drastically. Some techniques efficiently exploit a particular case

of multiplicity information, namely that the intended relation is a total function in a given mode (i.e. its multiplicity is 1–1). Of course, such statements should ideally also be provided for all the predicates defined in the background knowledge.

1.1.3 Syntactic Bias

Syntactic bias is another, complementary form of declarative bias. It restricts the language of hypotheses. Ideally, it is a parameter of an induction technique, rather than hardwired into it. As a parameter, it can be provided either by the source as an additional input, or made available to the technique by its designers.

One particularly useful and common approach is to bias induction by a schema. A *program schema* contains a template program abstracting a class of actual programs (called *instances*), in the sense that it represents their dataflow and control-flow by means of parameterized place-holders, but does not contain (all) their actual computations nor (all) their actual data structures, together with a set of *constraints* that the place-holders of the schema should satisfy.

One could for instance design a template program capturing the class of divide-and-conquer programs, or a sub-class thereof, e.g. those featuring two parameters, with division of the first parameter into two components that are somehow smaller than it:

$$\begin{aligned} r(X,Y) &\leftarrow \text{primitive}(X), \text{solve}(X,Y) \\ r(X,Y) &\leftarrow \text{nonPrimitive}(X), \text{decompose}(X,HX, TX_1, TX_2), \\ &\quad r(TX_1, TY_1), r(TX_2, TY_2), \text{compose}(HX, TY_1, TY_2, Y) \end{aligned}$$

The intended semantics (data-flow constraints) of this template can be informally described as follows. For an arbitrary relation r over formal parameters X and Y , an instance is to determine the value(s) of Y corresponding to a given value of X . Two cases arise: either X has a value (when the `primitive` test succeeds) for which Y can be easily directly computed (through `solve`), or X has a value (when the `nonPrimitive` test succeeds) for which Y cannot be so easily directly computed.¹ In the latter case, the divide-and-conquer principle is applied by (i) division (through `decompose`) of X into a term HX and two terms TX_1 and TX_2 that are both of the same type as X but smaller

1. Note that *both* cases may apply, as there may be values of Y that it is easy to directly compute from a given X , as well as other values of Y that it is not so easy to directly compute from that X .

than X according to some well-founded relation, (ii) conquering (through r) in order to determine the value(s) of TY_1 and TY_2 corresponding to TX_1 and TX_2 , respectively, and (iii) combining (through `compose`) terms HX , TY_1 , TY_2 in order to build Y .

Enforcing this intended semantics must be done “manually,” as the template by itself has no semantics, in the sense that many programs can be seen as an instance of it, not just divide-and-conquer ones. One way of doing this is to attach to the template the set of specifications of its predicate place-holders: these specifications are in terms of each other, including the one of r , and are thus generic (because even the specification of r is unknown), but can be abduced once and for all according to the informal semantics of the schema [15]. Such a schema (i.e. template plus specification set) constitutes an extremely powerful syntactic bias, because it encodes algorithm design knowledge that would otherwise have to be hardwired or rediscovered the “hard way” during each synthesis.

There are two approaches for representing schemata. The first approach is representing the schemata as higher-order expressions, sometimes augmented by extra-logical annotations and features, where the actual programs are obtained by applying higher-order substitutions to the schema. The reason why some researchers prefer this approach is that they find this approach suitable for some applications such as schema-guided program transformation [6], where a schematic program transformation could begin only if one can find some form of higher-order matching between actual programs and schemata. In the second approach, the schemata are represented as first-order programs, where actual programs are obtained by an interpretation of the relations and the functions of the schema. In other words, the actual programs are obtained by adding programs for its *open* relations, where openness means that an arbitrary interpretation can apply to the relation and the function. This kind of schemata is called *open programs* [15]. A *synthesis strategy* determines a way in which the open relations of the schema are instantiated. There could be more than one strategy for a given schema, depending on which open relation(s) to instantiate first (e.g. instantiation of `decompose`, `primitive`, and `nonPrimitive`), and which open relations to instantiate next (e.g. `solve` and `compose`).

There are two ways of biasing synthesis by a schema. *Schema-based* synthesis infers a program guaranteed to fit the template of a pre-determined schema and to satisfy its specification set, but the schema itself is to a certain degree hardwired into the tech-

nique. A useful variant is *schema-guided* synthesis, where the schema is a parameter to the technique (which is thus schema-independent) and thus actively guides the synthesis. As a parameter, it can be provided either by the source as an additional input, or made available to the technique by its designers.

Less common approaches to syntactic bias are the clause description language of [1], antecedent description grammars [7], argument dependency graphs [27], etc., and are surveyed in [26].

1.1.4 Generality

Given the formula $G \Rightarrow S$, we say that G is *more general* than S , and that S is *more specific* than G . In ILP, the aim is to compute a hypothesis H given background knowledge B and evidence E , such that $B \wedge H \Rightarrow E$. The generality relation \Rightarrow is a partial order, but doesn't induce a lattice on the set of formulas. Indeed, there is not always a unique least generalization under implication of an arbitrary pair of clauses. For instance, the clauses $p(f(X)) \leftarrow p(X)$ and $p(f(f(X))) \leftarrow p(X)$ have both $p(f(f(X))) \leftarrow p(X)$ and $p(f(X)) \leftarrow p(Y)$ as least generalizations. In [22], the existence and computability of a least generalization under implication for any finite set of clauses that contains at least one non-tautologous function-free clause is proven. Since implication between Horn clauses is undecidable, there are a number of different models of inductive inference.

θ -subsumption. In the model called θ -subsumption [23], the background knowledge B is empty. The model is defined for clauses, which are viewed as sets of literals.

Definition 1.1: A clause g θ -subsumes a clause s iff there exists a substitution σ such that $g\sigma \subseteq s$. Two clauses are θ -subsumption-equivalent iff they θ -subsume each other. A clause is said to be *reduced* iff it is not θ -subsumption-equivalent to any proper subset of itself.

For instance, The clause $p(X,Y) \leftarrow q(X,Y), r(X)$ θ -subsumes $p(V,Z) \leftarrow q(V,Z), q(V,T), r(V), s(Z)$ with the substitution $\{X/V, Y/Z\}$.

If a clause g θ -subsumes a clause s , then $g \Rightarrow s$, but the reverse is not true for self-recursive clauses [21]. For instance, for the recursive clauses $p(f(X)) \leftarrow p(X)$ and $p(f(f(X))) \leftarrow p(X)$ (called g and s respectively), although $g \Rightarrow s$ (note that s is simply g

self-resolved), g does not θ -subsume s . Therefore, θ -subsumption is not equivalent to implication among clauses. Hence, it is not adequate for handling recursive clauses.

θ -subsumption induces a lattice on the set of reduced clauses: any two clauses have a unique least upper bound (lub) and a unique greatest lower bound (glb). The least generalization under θ -subsumption (abbreviated $\text{lg}\theta$) of two clauses c and d , denoted $\text{lg}\theta(c,d)$, is the lub of c and d in the θ -subsumption lattice. The $\text{lg}\theta$ of two terms $f(s_1, \dots, s_n)$ and $f(t_1, \dots, t_n)$, denoted $\text{lg}\theta(f(s_1, \dots, s_n), f(t_1, \dots, t_n))$, is $f(\text{lg}\theta(s_1, t_1), \dots, \text{lg}\theta(s_n, t_n))$, whereas the $\text{lg}\theta$ of the terms $f(s_1, \dots, s_n)$ and $g(t_1, \dots, t_m)$, where $f \neq g$ or $n \neq m$, is a variable V , where V represents this pair of terms throughout. The $\text{lg}\theta$ of two atoms (similarly for two negative literals) $p(s_1, \dots, s_n)$ and $p(t_1, \dots, t_n)$, denoted $\text{lg}\theta(p(s_1, \dots, s_n), p(t_1, \dots, t_n))$, is $p(\text{lg}\theta(s_1, t_1), \dots, \text{lg}\theta(s_n, t_n))$, whereas the $\text{lg}\theta$ of the atoms $p(s_1, \dots, s_n)$ and $q(t_1, \dots, t_m)$, where $p \neq q$ or $n \neq m$, is \top , where \top denotes the “most general literal”. Finally, the $\text{lg}\theta$ of two clauses c and d , denoted $\text{lg}\theta(c,d)$, is $\{\text{lg}\theta(l_1, l_2) \mid l_1 \in c \text{ and } l_2 \in d\}$.

For instance, the $\text{lg}\theta$ of the clauses $p(V,W) \leftarrow q(V,W), r(V), s(W)$ and $p(T,N) \leftarrow q(T,N), r(T), r(N)$ is the clause $p(X,Y) \leftarrow q(X,Y), r(X), r(Z)$.

Relative θ -subsumption. An extension of θ -subsumption that uses background knowledge B is called relative subsumption [23].

Definition 1.2: If the background knowledge B consists of a conjunction of ground facts, then the *relative least generalization under θ -subsumption* (abbreviated $\text{rlg}\theta$) of two ground atoms E_1 and E_2 relative to background knowledge B is $\text{rlg}\theta((E_1 \leftarrow B), (E_2 \leftarrow B))$.

The $\text{rlg}\theta$ of two clauses is not necessarily finite. However, it is possible [21] to construct finite $\text{rlg}\theta$ s under the syntactic bias of *ij*-determinacy.

Definition 1.3: If L_i is a literal in the ordered Horn clause $A \leftarrow L_1, \dots, L_n$, then the *input variables* of the literal L_i are those variables appearing in L_i that also appear in the clause $A \leftarrow L_1, \dots, L_{i-1}$; all other variables in L_i are called *output variables*. A literal L_i is *determinate* iff its output variables have at most one possible binding, given the binding of the input variables. If a variable V appears in the head of a clause, then the *depth* of V is zero, and otherwise, if F is the first literal containing the variable V and d is the maximal depth of the input variables of F , the depth of V is $d + 1$. A clause is *ij*-deter-

minate iff it is determinate and its body contains only variables of depth at most i and predicate symbols that have arity at most j [8].

Inverse Resolution. Another model of generality is inverse resolution. There are four inductive inference rules of inverse resolution: *absorption*, *identification*, *intra-construction*, and *inter-construction* [21]:

$$\frac{(q \leftarrow A) (p \leftarrow A, B)}{(q \leftarrow A) (p \leftarrow q, B)} \quad \frac{(p \leftarrow A, B) (p \leftarrow A, q)}{(q \leftarrow B) (p \leftarrow A, q)}$$

$$\frac{(p \leftarrow A, B) (p \leftarrow A, C)}{(q \leftarrow B) (p \leftarrow A, q) (q \leftarrow C)} \quad \frac{(p \leftarrow A, B) (q \leftarrow A, C)}{(p \leftarrow r, B) (r \leftarrow A) (q \leftarrow r, C)}$$

In the rules above, lower-case letters represent atoms and upper-case letters represent conjunctions of atoms. The absorption and identification rules invert only one resolution step. The intra-construction and inter-construction rules introduce new predicate symbols (predicate invention, see the next subsection).

1.1.5 Predicate Invention

Predicate invention can be defined as follows: (i) introducing into the hypothesis some predicate(s) that are not in the evidence, nor in the background knowledge (this is called shifting the bias by extending the hypothesis language [25]), and (ii) inducing programs of these new predicates. This requires the usage of constructive rules of inductive inference (where the inductive consequent may involve symbol(s) that are not in the antecedent), as opposed to selective ones. Such constructive induction thus doesn't (simplistically) assume that the preliminary induction tasks of representation and vocabulary choice have already been solved, and represents thus a crucial field in induction.

One can distinguish two types of predicate invention: *necessary predicate invention* and *non-necessary predicate invention*.

Necessary Predicate Invention. We'll first give an example of necessary predicate invention, and then define it.

Example 1: In the absence of background knowledge, the induction from positive and negative examples of the following logic program for the `sort` predicate (where

$\text{sort}(L,S)$ holds iff S is a non-descendingly ordered permutation of L , where L, S are integer-lists):

$$\begin{aligned} \text{sort}([],[]) &\leftarrow \\ \text{sort}([H|T],S) &\leftarrow \text{sort}(T,Y), \text{insert}(H,Y,S) \end{aligned}$$

involved the invention of the `insert` predicate (where $\text{insert}(E,L,R)$ holds iff integer-list R is non-descendingly ordered integer-list L with integer E inserted), whose logic program hereafter is a by-product:

$$\begin{aligned} \text{insert}(E,[],[E]) &\leftarrow \\ \text{insert}(E,[H|T],[E,H|T]) &\leftarrow E \leq H \\ \text{insert}(E,[H|T],[H|R]) &\leftarrow \neg(E \leq H), \text{insert}(E,T,R) \end{aligned}$$

Note that the invention of the `insert` predicate required in turn the invention of the \leq predicate (whose obvious specification and program are omitted here).

Definition 1.4: Predicate invention is *necessary* iff there is no finite logic program for the observational concepts in the evidence that uses only the fixed vocabulary of predicate symbols from the evidence and the background knowledge.

In Example 1, once synthesis was committed to the recursive call $\text{sort}(T,Y)$, where T is the tail of L (i.e. $L=[H|T]$), the predicate `insert` *had to* be invented, especially that its recursive program cannot be unfolded into the program for `sort`. If committed to some other recursive call(s), another predicate would have had to be invented. Otherwise, the background knowledge being empty, `sort` would have to be implemented at most in terms of itself only, which is impossible without generating the non-terminating program $\text{sort}(L,S) \leftarrow \text{sort}(L,S)$, or without generating an infinite program (which extensionally encodes the model).

Non-necessary Predicate Invention. One can distinguish two types of non-necessary predicate invention: *useful predicate invention* and *pragmatic predicate invention* [12].

First, we discuss useful predicate invention. If there were `permutation` and `ordered` predicates in the background knowledge of Example 1, the invention of `insert` such that it is recursively defined (e.g. as above) would be useful. Indeed, otherwise the `insert` predicate would not have to be invented as its unfoldable (because non-recursive) program would involve the `permutation` and `ordered` predicates:

$\text{insert}(E,L,R) \leftarrow \text{permutation}([E|L],R), \text{ordered}(R)$

and would have a complexity of $O(n!)$, where n is the length of the list L , and would thus be inefficient compared to the recursive `insert` program above, which is $O(n)$. Hence, the use of a recursive `insert` program would decrease the complexity of the overall `sort` program. The invention of a recursive `insert` program is thus considered useful although non-necessary.

Definition 1.5: Given a partially constructed logic program for the observational concepts in the evidence, predicate invention is *useful* iff there is a way to complete the program by inventing a predicate whose logic program is recursive.

Let's now give an example of pragmatic predicate invention.

Example 2: Given evidence of the `grandDaughter` relation (where `grandDaughter(G,P)` holds iff person G is a grand-daughter of person P), and background knowledge of the `parent`, `female`, and `male` relations (where `parent(P,Q)` holds iff person P is a parent of person Q), the induction of the following logic program for `grandDaughter`:

$\text{grandDaughter}(G,P) \leftarrow \text{parent}(P,Q), \text{daughter}(G,Q)$

involved the invention of the `daughter` predicate (where `daughter(D,P)` holds iff person D is a daughter of person P), whose logic program hereafter is a by-product:

$\text{daughter}(D,P) \leftarrow \text{parent}(P,D), \text{female}(D)$

The invention of the `daughter` predicate was pragmatic since, although the `daughter` program could be unfolded into the program of the `grandDaughter` predicate, i.e. its invention was non-necessary, inventing it caused the `grandDaughter` program to become more compact, and since the `daughter` concept has now been defined and can be reused in the future.

Definition 1.6: Given a partially constructed logic program for the observational concepts in the evidence, predicate invention is *pragmatic* iff it is neither necessary nor useful.

The task of inductive inference amounts in the limit to finding a finite axiomatization for a given model. If the intended model cannot be finitely axiomatized within a language \mathcal{L} , inductive inference will never succeed. However, detecting this is undecidable. This follows from Rice's theorem (see [25]):

Theorem 1: Given a recursively enumerable set of ground atoms \mathcal{E} in a language \mathcal{L}_0 , it is undecidable whether \mathcal{E} is finitely axiomatizable in some language \mathcal{L} such that $\mathcal{L} \supseteq \mathcal{L}_0$.

Fortunately, introducing a new predicate allows finding a finite axiomatization, as proved by Kleene (see [25]):

Theorem 2: Any recursively enumerable set of formulas in a first-order language \mathcal{L} is finitely axiomatizable in the predicate calculus using additional predicate symbols not in \mathcal{L} .

In other words, Kleene’s theorem states that inductive inference will always succeed provided the system invents the appropriate new predicates. Thus, predicate invention is crucial in inductive inference.

1.1.6 Construction Modes and Admissibility

In this sub-section, we will introduce the concepts of construction modes and admissibility [10]. The informal definitions of these two concepts are as follows: a *construction mode* for a relation states which parameter(s) are used to “construct” the other parameters, also expressing whether such usage is mandatory or optional. Construction modes should not be confused with input/output modes, which state which parameters must be ground or may be variables at call/return-time. The concept of *admissibility* captures the notion of what it means for an *atom* to satisfy a construction mode for its relation. Now, let us give the formal definitions of these new concepts. In these definitions, when we want (or need) to group several terms into a single term, we represent this as a tuple, using angled brackets. For instance, $\langle f(X,Y), g(X,Y,Z) \rangle$ is a term representing the couple built of two terms $f(X,Y)$ and $g(X,Y,Z)$.

Definition 1.7: The *leaves* of a term t , denoted $leaves(t)$, are the set of the variables and constants occurring in t .

The *vertices* of a term t , denoted $vertices(t)$, are the multi-set of the variables and function symbols (including the constants symbols) occurring in t .

For instance, $leaves(1 \cdot B \cdot 1 \cdot nil) = \{1, B, nil\}$, and $leaves(a \cdot T) = \{a, T\}$, whereas $vertices(1 \cdot B \cdot 1 \cdot nil) = \{1, \cdot, B, \cdot, 1, \cdot, nil\}$, and $vertices(a \cdot T) = \{a, \cdot, T\}$.

Definition 1.8: Term s is *syntactically obtained* from term t iff $leaves(t) \subseteq leaves(s)$. We denote this by $t \subseteq s$.

Term s *syntactically contains* term t iff $vertices(t) \subseteq_m vertices(s)$, where \subseteq_m denotes multi-set inclusion. We denote this by $t \subseteq_m s$.

For instance, $\langle a, b \rangle$ is syntactically obtained from $\langle a, a \rangle$, because $leaves(\langle a, a \rangle) = \{a\} \subseteq \{a, b\} = leaves(\langle a, b \rangle)$. However, $\langle a, b \rangle$ does not syntactically contain $\langle a, a \rangle$, because $vertices(\langle a, a \rangle) = \{a, a\} \not\subseteq_m \{a, b\} = vertices(\langle a, b \rangle)$.

Definition 1.9: A construction mode m for a relation r of arity n is a total function from the set $\{1, 2, \dots, n\}$ into the set $\{may_1, \dots, may_n, may_{all}, must_1, \dots, res_1, \dots, res_n, not\}$, such that res_j is in the range of m iff may_j or $must_j$ also is in the range of m , and such that every res_j is at most once in the range of m . We also say $m(i)$ is the mode of the i^{th} parameter of r .

A construction mode m is often written in the more suggestive form $r(m(1), \dots, m(n))$. Do not confuse the position i of a parameter and the index j of its mode $m(i)$, say $must_j$. The intended semantics of a mode is as follows:

- mode $must_j$ means the parameter in the corresponding position is mandatory in syntactically constructing the parameter in the corresponding position of res_j ;
- mode may_j means the parameter in the corresponding position is optional for syntactically constructing the parameter in the corresponding position of res_j ;
- mode may_{all} means the parameter in the corresponding position is optional for syntactically constructing all other parameters;
- mode not means the parameter in the corresponding position is not used at all in syntactically constructing any of the parameter(s) in the corresponding position(s) of all res_j .

Let m be a mode for a relation r , and let $r(t_1, \dots, t_n)$ be the considered atom, where n is natural number. Let the indexes in m run from 1 to k inclusive, where k is a natural number. Let $Must_j = \langle t_i \mid m(i) = must_j \rangle$, and let $Must = \langle t_i \mid m(i) = must_j \text{ for some } j \rangle$. Similarly for May_j , May_{all} , May , Res_j , Res , and Not .

For instance, let the construction mode be $arelacion(may_{all}, must_1, must_2, res_1, res_2)$ and the atom be $arelacion(1, [b], [], [a, b], [a])$, then we have that $k = 2$, $Must_1 = \langle [b] \rangle$, $Must_2 = \langle [] \rangle$, $Must = \langle [b], [] \rangle$, $May_1 = May_2 = \langle \rangle$, $May = May_{all} = \langle 1 \rangle$, $Res_1 = \langle [a, b] \rangle$, $Res_2 = \langle [a] \rangle$, and $Res = \langle [a, b], [a] \rangle$.

Definition 1.10: A variable is linked in a clause if it occurs in the head or if it occurs in a literal L of the body and L contains a linked variable.

Definition 1.11: A clause that has no equality atoms and no recursive calls, no \top (see Definition 1.1) and no unlinked variables in the body:

$$r(X,Y,Z) \leftarrow C$$

is admissible with respect to a mode m for r iff

$$\forall 1 \leq j \leq k: Must_j \subseteq_m \langle Res_j, C' \rangle \quad (1)$$

where C' is a tuple built of the atoms (seen as terms) of conjunction C , and

$$leaves(Res) \setminus sharedLeaves(Res) \subseteq leaves(May, May_{all}, Must, C') \cup \{0, nil, \dots\} \quad (2)$$

where $sharedLeaves(t)$ denotes the set of leaves shared by all components of tuple t .

Now, we present the objective of the thesis based on the terminology and theoretical results given in Section 1.1.

1.2 The Objective of the Thesis

The learning of recursive logic programs (i.e. the class of logic programs where at least one clause is recursive, e.g. the `subset` program given in Section 1) from incomplete information, such as input/output examples, is a challenging subfield both of ILP (Inductive Logic Programming) and of the synthesis (in general) of logic programs from formal specifications. This is an extremely important class of logic programs, as the recent work on constructive induction [12] [25] shows that necessarily invented predicates (see Section 1.1.5) have recursive programs, and it even turns out that their induction is much harder than the one of non-recursive programs. We call this (inductive) program synthesis.

When it comes to programming applications, we believe the ideal technique is interactive (in the sense of DIALOGS [13]) and non-incremental, has a clausal evidence language plus type, mode, and multiplicity information (like SYNAPSE [11], DIALOGS), can handle semantic manipulation relations, actually uses (structured) background knowledge and a syntactic bias, which are both problem-independent and intensional (like in SYNAPSE), is guided by (and not just based on) at least the powerful divide-and-conquer schema of SYNAPSE and DIALOGS (using the implementation approach of METAINDUCE [18]), discovers additional base case and recursive case examples (like CILP [19]), can perform both necessary and useful predicate invention (like SYNAPSE, DIALOGS), even from sparse abduced evidence (like CILP), actually dis-

covers the recursive atoms, and makes a constructive usage of the negative evidence (through abduction, like the *Constructive Interpreter* [9] and SYNAPSE).

Our aim was thus to study this important class of logic programs, i.e. recursive logic programs, and to develop a system that induces logic programs of this class. The closest system to our considerations was DIALOGS (Dialogue-based Inductive and Abductive LOGic Program Synthesizer) [13]. Therefore, we improved this system into a new one called DIALOGS-II. Thus, our aim became to improve DIALOGS, whose ancestor was the SYNAPSE system [11] [14], which induces recursive logic programs from a set of positive examples, and a set of Horn clauses that are called *properties*. The drawbacks of SYNAPSE are that the specifier may not always provide properties that are needed to induce a logic program that is correct with respect to its specification, and that most positive examples are redundant with the properties.

DIALOGS-II is a schema-guided, interactive, and non-incremental synthesizer of recursive logic programs that takes the initiative and queries a (possibly naive) specifier for evidence in her/his conceptual language. DIALOGS-II needs no properties, and only asks for the minimal knowledge a specifier *must* have in order to want a (logic) program, and it can be used by any learner (including itself) that detects, or merely conjectures, the necessity of invention of a new predicate. Moreover, due to its powerful codification of “recursion-theory” into program schemata and schematic constraints, it needs very little evidence and is very fast.

DIALOGS-II is schema-guided. The reason why it is schema-guided is as follows: most (but not all) inductive/abductive synthesizers require large amounts of ground positive and negative examples of the intended concept. This is because ground examples are not an adequate way of communicating a concept to a computer and/or because the underlying “recursion theory” of the synthesizer is poor. In order to overcome this deficiency, some researchers used non-ground examples [20], or Horn clauses [11] [14] as evidence language instead of using only ground examples, and some experimented with schema-based synthesis [11] [14] to address the poor “recursion theory” problem [17]. We chose the schema-guided approach, because we think that it is the best approach to handle “recursion theory”. The schemata of DIALOGS-II are open programs and are available to the system together with their synthesis strategies. In other words, for a particular synthesis, a schema together with a synthesis strategy is chosen.

DIALOGS-II can be used to synthesize programs by making use of the available schemata and strategies that are already existing in the system. Moreover, the specifier can provide additional schemata using the declarative syntax of the schemas of the system to encode new schemata, and adding the code for strategies for those new schemata. In that way, the specifier can make syntheses of programs by executing the strategies that fit to the schemata added.

DIALOGS-II is interactive, because the specifier is assumed to be “lazy” in the sense that s/he is reluctant to take the initiative and type in evidence of the intended concept without knowing whether it will be “useful” to the synthesizer or not [13]. Therefore, DIALOGS-II takes the initiative and queries the specifier only for strictly necessary evidence. The query and answer languages are carefully designed so that even a computationally naive specifier can use the system. Moreover, it is guaranteed that the specifier can answer such queries, because otherwise the specifier would not need the synthesized program.

DIALOGS-II is a system that only induces recursive logic programs because we believe that inducing recursive logic programs is important [12], especially that they are strictly necessary (see Section 1.1.5).

DIALOGS-II is a recursive synthesizer, which means it recursively calls itself when a necessary predicate invention is conjectured during the synthesis. It is then a natural solution for the system to call itself recursively to make this new synthesis since the problem (of synthesizing a program for a necessary new predicate) has the same nature as the problem of synthesizing a program for the top-level predicate. That is, for both cases, the necessity of predicate invention is conjectured before starting a synthesis.

DIALOGS-II is non-incremental, because we believe that using an incremental approach is not practical for program synthesis [17]. Recursive programs are so fragile objects that they should be handled with utmost care. Therefore, we believe that using general-purpose induction techniques to synthesize programs by incrementally “debugging” the empty program (or an approximate program) according to incomplete evidence is not an appropriate way of synthesizing programs. Moreover, in incremental synthesis, the order of the evidence is important. That means the system can be forced into the synthesis of infinite, redundant, or dead code. We strongly believe that the only way to reliably and efficiently synthesize recursive programs from incomplete information is through guidance by a schema capturing a design methodology (e.g. a di-

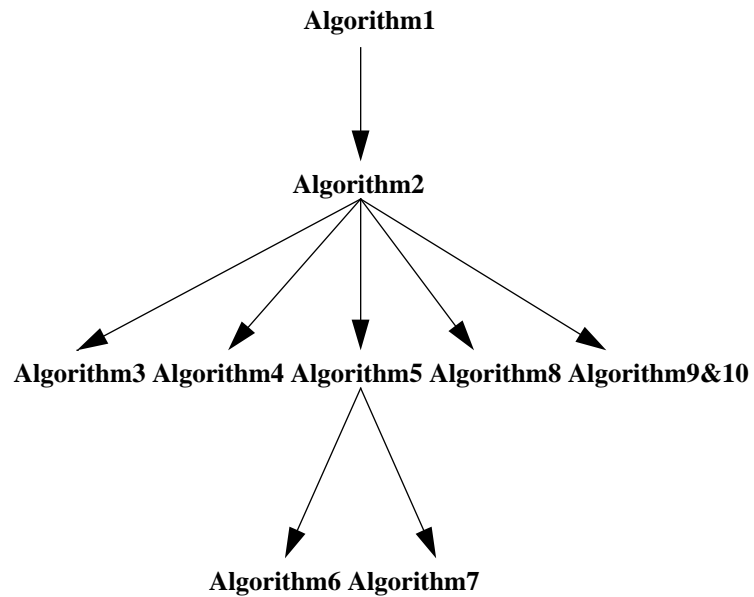
vide-and-conquer schema), as well as through non-incremental handling of the evidence.

In the remainder of this thesis, we will examine the DIALOGS-II technique closely in Chapter 2. This will be followed by a comparison of DIALOGS-II with current ILP systems in Chapter 3, and finally, we reach a conclusion in Chapter 4.

Chapter 2

The DIALOGS-II Technique

As mentioned earlier, DIALOGS-II is a schema-guided, interactive, recursive, and non-incremental recursion synthesizer that takes the initiative and queries a (possibly computationally naive) specifier for evidence in her/his conceptual language. In the following sub-sections, we will illustrate how the DIALOGS-II mechanism works by means of sample syntheses. First, we illustrate the synthesis of a program for the $\text{delOdds}(L,R)$ predicate, where $\text{delOdds}(L,R)$ holds iff R is L without its odd elements, where L, R are integer-lists. Next, we examine the synthesis of a program for the predicate $\text{reverse}(L,R)$, where $\text{reverse}(L,R)$ holds iff list R is the reverse of list L , to illustrate the recursive call of DIALOGS-II to itself. Before giving the sample syntheses, we give an algorithm call chart of the basic synthesis algorithm of how DIALOGS-II works and the basic synthesis algorithm itself:



Algorithm 1: schemaGuidedDialogs-II(Pgm)

Inputs: (none)

Outputs: Pgm

ask for the predicate declaration of the predicate for which a program is being synthesized

PredDecl := ask('Predicate Declaration')

ask for a schema and a strategy for the schema

selectSchemaStrategy(Schema,Strategy)

call Dialogs-II with Schema, Strategy and PredDecl to induce Pgm

dialogsII(Schema,Strategy,PredDecl,Pgm)

As shown in Algorithm 1, after executing the first two statements, the system executes the statement dialogsII(Schema,Strategy,PredDecl,Pgm) whose algorithm is given as follows:

Algorithm 2: dialogsII(Schema,Strategy,PredDecl,Pgm)

Inputs: Schema, Strategy, PredDecl

Outputs: Pgm

execute the strategy in order to obtain an open program from the schema, where the open program has open relations to be "closed" by the end of the next two statements (i.e. abduce and induce). ParamRoles denotes the information about

the names, types, and roles of the parameters (e.g. induction, result).

Strategy(PredDecl,Schema,OpenPgm,ParamRoles)

abduce the evidence necessary for “closing” the open relations p and q of the open program by means of querying the specifier, where the open relation of the non-recursive clause is p , whereas the open relation of the recursive clause is q .

The atoms of these relations are supposed to be the last atoms of the non-recursive and recursive clauses of the open program respectively.

abduce(OpenPgm,ParamRoles,PredDecl,pEvidence,qEvidence)

induce the programs for the open relations by using the Program Closing Method based on the evidence abducted in the previous step according to the construction modes $pMode$ and $qMode$ of the relations p and q respectively.

induce(pEvidence,qEvidence,pMode,qMode,pClauses,qClauses)

evaluate the result of the Program Closing Method to conjecture if there is a need for inventing a new predicate

evaluate(Schema,Strategy,OpenPgm,pClauses,qClauses,PredDecl,ParamRoles,Pgm)

Now, we go through the statements of the basic synthesis algorithm (Algorithm 1) for the synthesis of a program for the $delOdds(L,R)$ predicate. We will first discuss the first two statements of this algorithm: asking for a predicate declaration, selecting a schema and a strategy in Section 2.1. Next we will go through the statements of Algorithm 2 by first discussing the execution of the strategy in Section 2.3 and abduction of evidence in Section 2.4, which is followed by the discussion of the induction of program clauses in Section 2.5, and finally by the evaluation of the program clauses to conjecture necessary predicate invention and sparseness problem in Section 2.5.2.

2.1 Asking For a Predicate Declaration, a Schema and a Strategy

DIALOGS-II first needs to know for which predicate it is synthesizing a program. Therefore, it asks the predicate declaration of the predicate. The specifier *must* be able to give such a declaration, because otherwise s/he would not have the need to have a program

for this predicate. Thus, the first step in the synthesis is prompting the specifier for a predicate declaration and obtaining it:

```
Predicate declaration? delOdds(L:list(int),R:list(int))
```

where the type of L and R is list(int). Other available types are in the set {atom, int, nat, list(_),...}.

As mentioned earlier, DIALOGS-II is a schema-guided synthesizer. Therefore, it needs a schema and a strategy for the schema in order to be able to start a synthesis. Thus, the next step in the synthesis is prompting the specifier for a schema and a strategy for this schema.

A basic algorithm for selecting a schema and a strategy for it is given below, where SchemaDefaults is a parameter representing the list of available schemata in the system, Schema is a schema in SchemaDefaults, and Strategy is a strategy for Schema.

Algorithm 3: selectSchemaStrategy(Schema,Strategy)

Inputs: none

Outputs: Schema, Strategy

ask the specifier to select Schema from SchemaDefaults in the system

Schema := ask('Schema', SchemaDefaults)

determine StrategyDefaults, the list of available strategies for Schema

StrategyDefaults := determineStrategyDefaults(Schema)

ask the specifier to select Strategy from StrategyDefaults

Strategy := ask('Strategy', StrategyDefaults)

Now, let us see how is this done during the synthesis of a program for delOdds(L,R). Note that the questions of this dialog are in the typewriter font, the specifier's answers are in helvetica font, and the default answers of the system are given inside curly braces, i.e. {}, and suppose that one of the schemata available in the system is a "divide-and-conquer" schema together with a strategy for it:

```
Schema? {divide-and-conquer1} divide-and-conquer1
```

```
Strategy? {divide-and-conquer-Strategy1} divide-and-conquer-Strategy1
```

Now, DIALOGS-II knows that it will use a divide-and-conquer schema with a particular strategy, i.e. *divide-and-conquer-Strategy1*.

2.2 Execution of the Strategy

The next step is to execute the strategy selected by the specifier. Before giving the algorithm of a particular strategy, let us see what the considered divide-and-conquer schema looks like. The considered schema is:

$$\begin{aligned} r(X, Y, \mathbf{Z}) &\leftarrow \text{solve_r}(X, Y, \mathbf{Z}) \\ r(X, Y, \mathbf{Z}) &\leftarrow \text{decompose_r}(X, \mathbf{HX}, \mathbf{TX}), r(\text{TX}_1, \text{TY}_1, \mathbf{Z}), \dots, r(\text{TX}_t, \text{TY}_t, \mathbf{Z}), \\ &\quad \text{compose_r}(\mathbf{HX}, \mathbf{TY}, Y, \mathbf{Z}) \end{aligned}$$

where $\mathbf{HX} = \text{HX}_1, \dots, \text{HX}_h$, $\mathbf{TX} = \text{TX}_1, \dots, \text{TX}_t$, $\mathbf{TY} = \text{TY}_1, \dots, \text{TY}_t$, and $\mathbf{Z} = \text{Z}_1, \dots, \text{Z}_z$.

A *divide-and-conquer* program for a predicate r over parameters X , Y , and \mathbf{Z} works as follows. Suppose that X is the induction parameter, Y is the result parameter, and \mathbf{Z} the (repetitive) passive parameter(s), where a *passive* parameter is a parameter that does not change through a recursive call. There are two possibilities of how Y can be computed: the first one is that Y is directly computed from X and \mathbf{Z} by means of $\text{solve_r}(X, Y, \mathbf{Z})$. There could be more than one way in which Y is directly computed from X and \mathbf{Z} (in other words, there could be more than one clause whose head is $\text{solve_r}(X, Y, \mathbf{Z})$ in the final synthesized program). In the second one, first X is decomposed into h heads and t tails by means of $\text{decompose_r}(X, \mathbf{HX}, \mathbf{TX})$. Next, t recursive calls are done, one for each TX_i . Last, the result parameter Y is constructed from \mathbf{HX} , \mathbf{TY} , and \mathbf{Z} by means of $\text{compose_r}(\mathbf{HX}, \mathbf{TY}, Y, \mathbf{Z})$. To be precise, the \mathbf{HX} are processed and composed with the \mathbf{TY} and \mathbf{Z} in order to yield Y . Again, there could be more than one way of computing Y from \mathbf{HX} , \mathbf{TY} , and \mathbf{Z} . The schema given above is a representation of this algorithm description.

So, in order to generate an open program from this schema according to the strategy *divide-and-conquer-Strategy1*, the system must determine and use the roles of the parameters, the number of passive parameter(s) (if any), i.e. $0 \leq z$, the program for the open relation decompose_r , and h and t .

Now, let us give the algorithm for executing the strategy for the divide-and-conquer schema given above:

Algorithm 4: divide-and-conquer-Strategy1(PredDecl,Schema,Pgm,
ParamRoles)

Input: PredDecl, Schema

Output: Pgm, ParamRoles

determine the induction parameter, which is of an inductively defined type, the result parameter (if any), and the passive parameter(s) (if any), and the number of result and passive parameters, i.e. y and z respectively, from the predicate declaration PredDecl

$\langle \text{ParamRoles}, y, z \rangle := \text{paramRoles}(\text{PredDecl})$

determine decompose using the system-defined decomposition operators, i.e.

DecomposeDefaults

$\langle \text{decompose}_r, h, t \rangle := \text{selectDecompose}(\text{DecomposeDefaults})$

$\text{Pgm} := \text{generateOpenPgm}(\text{Schema}, \text{decompose}_r, h, t, z)$

Now, we examine the execution of the strategy *divide-and-conquer-Strategy1* by means of the synthesis of a program for $\text{delOdds}(L, R)$.

First, we show the determination of the parameter roles using the predicate declaration $\text{delOdds}(L:\text{list}(\text{int}), R:\text{list}(\text{int}))$: DIALOGS-II creates a sequence of potential induction parameters, which are of inductively defined types, keeps the first one as the (first) default answer, and the remaining ones as default ones upon backtracking. Similarly for the result parameter, which is also likely to be of an inductively defined type: from the currently remaining parameters, DIALOGS-II can create a sequence of potential result parameters, keep the first one as the (first) default answer, and the remaining ones as default answers upon backtracking. Finally, DIALOGS-II can propose as the passive parameter(s) (if any) the remaining parameter(s) (if any). Providing default answers is good for naive specifiers, where naive specifiers are the ones who do not have the capability for answering every question of the system, since if s/he has no idea of determining the roles of the parameters, s/he can simply accept the default answers and go on with the synthesis without blocking at this step. Note that a passive parameter may accidentally be declared as a result parameter, without any influence on the existence of a correct program: it would be found to be always equal to its tail by post-synthesis transformations, where in that case the synthesis would be a bit slower, because unnecessary computations would need to be done for its construction using its tail, **HX**, and the actually declared passive parameters.

How the parameter roles of `delOdds(L,R)` are determined is shown by the dialogue below, supposing that the specifier accepts the default answers proposed by the system. First, the specifier is prompted for the induction parameter, where the system proposes the parameter `L` as the default answer:

```
Induction parameter? {L} L
```

Next, the specifier is prompted for the result parameter, where the system proposes the (remaining) parameter `R` as the result parameter since there is only one remaining parameter according to the predicate declaration and it has to be a result parameter since the result parameter is asked before passive parameters.

```
Result parameter? {R} R
```

Note that there is (are) no passive parameter(s).

The strategy selected by the specifier makes `DIALOGS-II` create a sequence of potential decomposition operators using available decomposition operators in the system, keep the first one as the (first) default answer, and the remaining ones as default ones upon backtracking. The specifier can select the default one or can write her/his own `decompose_delOdds` as an answer, where the predicates in the body must already be defined as procedures in the system; let us assume that the specifier selects the default one, which is a head-tail decomposition of the list:

```
Decomposition operator? {decompose_delOdds(L,HL,TL)←
L=[HL|TL]}
decompose_delOdds(L,HL,TL) ← L=[HL|TL]
```

The other pre-defined decomposition operators of the type `list(_)` are given below, where h denotes the number of heads and t denotes the number of tails:

```
decompose_r(L,H1,H2,T) ← L=[H1,H2|T]           h/2, t/1
...
decompose_r(L,H,T1,T2) ← L=[H|T], partition(T,H,T1,T2)   h/1, t/2
decompose_r(L,T1,T2) ← L=[_|_|], halves(L,T1,T2)       h/0, t/2
...
```

Similar sequences are also available for other inductively defined types, e.g. `nat`. Next, h and t are instantiated according to the selected decomposition operator: for head-tail decomposition, both h and t are 1. At this time of the synthesis, from a programming point of view, all creative decisions have been taken, but alternative decisions are ready

for any occurrence of backtracking (either because DIALOGS-II fails due to some decision at a later step of Algorithm 2, or because the specifier wants another program after successful completion of all the steps).

Knowing `decompose_delOdds`, and the values of h , t , z , the following open program for `delOdds(L,R)` is generated from the input schema:

```
delOdds(A,B) ← solve_delOdds(A,B)
delOdds(A,B) ← decompose_delOdds(A,C,D), delOdds(D,E),
                compose_delOdds(C,E,B)
decompose_delOdds(F,G,H) ← F=[G|H]
```

Note that the relations `solve_delOdds` and `compose_delOdds` are open: they will be “closed” after the execution of the second and the third statements (abduction of evidence and induction of clauses) of Algorithm 2. This open program is passed as an input to the second statement of Algorithm 2.

2.3 Abduction of Evidence for the Open Relations of the Open Program

Let the open relations of an open program be p and q , where p is the open relation of a non-recursive clause and q is the open relation of a recursive clause of the open program.

In DIALOGS-II, the process of finding programs for the open relations p and q is also interactive and is based on the notions of abduction through (naive) unfolding and querying, and induction through the Program Closing Method (computation of least general generalizations).

We will illustrate naive unfolding and querying by means of the open relations of an open program of the divide-and-conquer schema given previously. The basic principle of (naive) unfolding and querying is as follows. Based on an open program

```
r(A,B) ← solve_r(A,B)
r(A,B) ← decompose_r(A,C,D), r(D,E), compose_r(C,E,B)
decompose_r(F,G,H) ← F=[G|H]
```

whose induction parameter is A , result parameter is B , decomposition operator is a head-tail one, and open relations are `solve_r` and `compose_r` (where `solve_r` denotes

p and compose_r denotes q , respectively), the possible computation “traces” for various most general values of the induction parameter are:

$$\begin{aligned} r([], D_1) &\leftarrow \text{solve_r}([], D_1) \\ r([E_1], F_1) &\leftarrow \text{solve_r}([E_1], F_1) \\ r([E_1], F_1) &\leftarrow r([], F_2), \text{compose_r}(E_1, F_2, F_1) \\ r([G_1, G_2], H_1) &\leftarrow \text{solve_r}([G_1, G_2], H_1) \\ r([G_1, G_2], H_1) &\leftarrow r([G_2], H_2), \text{compose_r}(G_1, H_2, H_1) \\ &\dots \end{aligned}$$

The basic principle is to (i) query the specifier for an instance of the last atom of each trace, using previous answers to resolve recursive calls, (ii) induce a program for solve_r from some of the answers so that it is not an open relation afterwards, (iii) induce a program for compose_r from the other answers so that it is not an open relation after this induction. The criterion of how to make such a division of the answers follows from the construction modes (see Section 1.1.6) of the schema. Before giving the steps above in detail, we introduce a new concept.

Definition 2.1: (Most general form of a parameter)

The most general form of a parameter of a certain type t and of a certain size s is denoted by

$$\text{mostGenForm}(t,s)$$

and is found using type-specific programs. For instance, for type `list`, the program is as follows:

$$\begin{aligned} \text{list}(\text{nil}, 0) &\leftarrow \\ \text{list}(H.T, M) &\leftarrow \text{list}(T, N), \text{ M is N+1} \end{aligned}$$

The most general form X of a parameter of type `list` and of size 3 is computed by SLD resolution of the goal

$$\leftarrow \text{list}(X, 3)$$

with the program given above yielding the list `A.B.C.nil`. Similarly, for type `nat`, the program is

$$\begin{aligned} \text{nat}(0, 0) &\leftarrow \\ \text{nat}(s(N), M) &\leftarrow \text{nat}(N, T), \text{ M is N+1} \end{aligned}$$

The most general form X of a parameter of type `nat` and of size 2 is computed by SLD resolution of the goal

$\leftarrow \text{nat}(X,2)$

with the program given above yielding the natural number $s(s(0))$.

Step(i) is realized by means of a basic loop: for each most general form of the induction parameter a goal for the top-level predicate is generated. For each clause whose head unifies with that goal, the atom of an open relation in the body of the clause is found by resolving the body atoms (“executing” the body) using the primitives, specifier-introduced predicates (which are introduced while the specifier gives answers to the queries about the predicate for which a program is being synthesized), and the clauses abducted during the previous iterations of the loop. And for each such an “open” atom, a query is generated. From the specifier’s answer to the query, some evidence is abducted for the open relation. This basic loop is repeated until the user answers a query by *stop-it*.

Let us now give an algorithm for Step(i) (note that Step(ii) and Step(iii) will be discussed in the following sub-sections). The algorithm abduces evidence, i.e. p Evidence and q Evidence, for the open relations p and q , where Pgm is the open program, $ParamRoles$ is information about the parameters of Pgm , i.e. names, types, and the positions of the parameters in the heads of the clauses of Pgm , which is computed using the predicate declaration $PredDecl$ by Algorithm 4, and $TopPred$ is the name of the predicate for which a program is being induced.

Algorithm5: $\text{abduce}(Pgm, ParamRoles, PredDecl, pEvidence, qEvidence)$

Inputs: Pgm , $ParamRoles$, $PredDecl$

Outputs: $pEvidence$, $qEvidence$

Shortcuts are abducted clauses for the open relations p , q and for $TopPred$

Shortcuts := {}

$pEvidence$:= {}

$qEvidence$:= {}

i := 0

repeat

let X_i be the most general form of the induction parameter of type t of size i

$X_i := \text{mostGenForm}(t,i)$

construct a goal using X_i and variable result and passive parameter(s)

TopPred := predName(PredDecl)

Goal := TopPred(X_j, Y, Z)

find a clause (in Pgm) whose head unifies with Goal and whose body unifies with Body (under the same substitution)

Body := pgmClause(Pgm, Goal)

prove Body in order to find an atom of open relation p or q

demo(Body, Pgm, TopPred, Shortcuts, Background, Assumptions, ResidueAtom)

query the specifier to abduce evidence for the open relation of Body

askQuery(Goal, ResidueAtom, Assumptions, Answer)

if Answer \neq "false" and Answer \neq "stop-it" then

abduce evidence for open relations p or q using the answer Answer to the query made in askQuery

if ResidueAtom is of predicate p then

$\langle pExs, Shortcut \rangle := abduceClauses(Answer, ResidueAtom)$

assert pExs and Shortcut

pEvidence := pEvidence \cup pExs

else

$\langle qExs, Shortcut \rangle := abduceClauses(Answer, ResidueAtom)$

assert qExs and Shortcut

qEvidence := qEvidence \cup qExs

Shortcuts := Shortcuts \cup {Shortcut}

else

abduce nothing

increment i

until Answer = "stop-it"

retract all Shortcuts to prevent them being used for further syntheses

Now let us give the algorithm for *demo*:

Algorithm 6: $\text{demo}(\text{Goal}, \text{Pgm}, \text{TopPred}, \text{Shortcuts}, \text{Background},$
 $\text{Assumptions}, \text{ResidueAtom})$

Input: Goal, Pgm, TopPred, Shortcuts, Background

Output: Assumptions, ResidueAtom

$(\text{Pgm} + \text{Shortcuts}) \cup \text{Background} \cup \text{Assumptions} \cup \text{ResidueAtom} \vdash_{\text{SLD}}$
Goal

Let us explain how *demo* works now: the proof of **Goal** is done by using **Shortcuts** and **Background**. **Shortcuts** are abduced clauses for the open relations **p**, **q** and for **TopPred**, where these clauses have precedence over the clauses of during SLD resolution (note that abduced clauses for the open relations **p**, **q** are also called evidence since they will be used as evidence for closing these open relations). That is, when the head of a shortcut clause unifies with an atom in **Goal**, these shortcut clauses are used instead of the clauses of **Pgm** (note that + is used instead of \cup to indicate this precedence in Algorithm 6). If there is neither a shortcut for an atom nor a clause in **Pgm** whose head unifies with that atom, then resolution is impossible and the resolution of **Goal** stops there, where this atom is **ResidueAtom**. The resolution of **Goal** also stops when **Goal** is proved to be *true*. **Background** is a set of programs for pre-defined primitives, such as “=”, “>”, etc. The atoms of specifier-introduced predicates (introduced by the answers that the specifier gives to the queries) encountered in **Goal** are called **Assumptions**, meaning that these atoms are assumed to be true during the SLD resolution since these atoms are introduced by the specifier and, thus, there is not any program for the specifier-introduced predicates that is known to the system, which implies that the resolution will be blocked by the atoms of the specifier-introduced predicates, if they are not assumed to be true. **Assumptions** are collected (through conjunction) in order to be passed to the query-asking during which the query is designed by considering **Assumptions** to be true (see *askQuery* below). Now, we give the algorithm for asking queries:

Algorithm 7: $\text{askQuery}(\text{Goal}, \text{ResidueAtom}, \text{Assumptions}, \text{Answer})$

Inputs: Goal, ResidueAtom, Assumptions

Output: Answer

if **ResidueAtom** is *true* then

do not query the specifier, because there is no atom for which any evidence should be abduced, thus Answer is an empty set

```

Answer := {}
else if ResidueAtom is an atom of the relation p or q then
  if Assumptions = [ ] then
    ask the query: "When does Goal hold?" and get Answer from the specifier
    Answer := ask('When does' Goal 'hold?')
  else
    ask the query: "When does Goal hold, assuming Assumptions?" and get
    Answer from the specifier
    Answer := ask('When does' Goal 'hold assuming' Assumptions?)

```

Now we know how the evidence is abduced for the open relations of an open program. Let us now examine how the abduction of evidence for the open relations `solve_delOdds` and `compose_delOdds` is done during the synthesis of a program for the `delOdds` predicate by considering the open program

```

delOdds(A,B) ← solve_delOdds(A,B)
delOdds(A,B) ← decompose_delOdds(A,C,D), delOdds(D,E),
                compose_delOdds(C,E,B)
decompose_delOdds(F,G,H) ← F=[G|H]

```

and considering that the relation `solve_delOdds` plays the role of the relation `p`, and the relation `compose_delOdds` plays the role of the relation `q`. This correspondence of the relations is due to the fact that `solve_delOdds` is the open relation of the non-recursive clause of the open program, and `compose_delOdds` is the open relation of the recursive clause of the open program.

First Iteration for Abducing Evidence. The specifier must know the value of the result parameter when the induction parameter is the empty list, otherwise s/he would not have the need for a program for `delOdds`. Thus, the first most general form of the induction parameter `A` is `[]`, where the query generation process proceeds by first resolving the goal `delOdds([],B)` with the head of the recursive clause of the open program and finding a goal for resolution. But, this attempt fails after resolving `decompose` with the recursive clause since the induction parameter has a value, i.e. `[]`, that cannot be decomposed. Therefore, the non-recursive clause is considered next. The recursive clause of the open program is tried first, because in that way the answers that the specifier gives to the queries are shorter (thus it is less boring for the specifier to answer the

queries) than in the case where the non-recursive clause is used first. This is because during the resolution of a goal that has been generated by resolving the goal with the head of the recursive clause, more assumptions are likely to be collected to be passed to the queries than in the case where the goal is resolved with the non-recursive clause. More assumptions during the querying causes the specifier to write less conditions in order to make the goal (the one that includes a most general form of the induction parameter) hold.

Thus, next the goal $\text{delOdds}([],B)$ is resolved with the non-recursive clause of the open program yielding the goal:

$\leftarrow \text{solve_delOdds}(A,B)$

Resolving this goal is impossible, so the unfolding process stops here, and DIALOGS-II extracts the following query to abduce evidence for solve_delOdds :

When does $\text{delOdds}([],B)$ hold?

from this goal (see Algorithm 7). Note that the specifier should be able to answer this query, since otherwise s/he would not need a program for the predicate delOdds , in that sense the specifier is guaranteed to answer the queries. The answer should be a formula $\mathcal{F}[B]$, where only B may be free, explaining how to compute B from $[]$ such that $\text{delOdds}([],B)$ holds. In other words, $\text{solve_delOdds}([],B)$ should be “equivalent” to $\mathcal{F}[B]$. The answer to the query is: $B=[]$. Using this answer, DIALOGS-II abduces the following evidence and shortcuts for solve_delOdds and delOdds (see Algorithm 5):

$\text{solve_delOdds}([],A) \leftarrow A=[]$
 $\text{delOdds}([],A) \leftarrow A=[]$ (s1)

Second Iteration for Abducing Evidence. The specifier must also know the result when the induction parameter is a one-element list. The query generation process starts by unifying the goal $\text{delOdds}([A],B)$ with the head of the recursive clause of the open program yielding the goal:

$\leftarrow \text{decompose_delOdds}(A,C,D), \text{delOdds}(D,E),$
 $\text{compose_delOdds}(C,E,B)$

Resolving $\text{decompose_delOdds}(A,C,D)$ and resolving the resulting equality atom gives

$\leftarrow \text{delOdds}([],E), \text{compose_delOdds}(C,E,B)$

Using the shortcut `s1` and resolving the resulting equality atom yields:

$$\leftarrow \text{compose_delOdds}(C,[],B)$$

Now the following query can be extracted from this goal since resolving this goal is impossible. The specifier answers the query as follows (note that the comma “,” stands for conjunction, and the semi-colon “;” stands for disjunction, where the comma has a higher precedence than the semi-colon):

$$\text{When does } \text{delOdds}([A],B) \text{ hold? } B=[], \text{ odd}(A); B=[A], \text{ even}(A).$$

Note that the predicates *odd* and *even* are introduced by the specifier, where the atoms *odd(X)* and *even(X)* are from now on assumed by the system to be true. Otherwise, resolving these atoms would be impossible and the resolution will be blocked because there are no programs for the predicates *odd* and *even*. Instead of blocking when such atoms are encountered, the system keeps these atoms to pass them to the queries (see the third iteration for abducing evidence given below). Using this answer to the query, DIALOGS-II abduces the following evidence and shortcuts (note the correspondence between the answers in the answer disjunct and the bodies of the shortcut and evidence clauses):

$$\text{compose_delOdds}(A,[],B) \leftarrow B=[], \text{ odd}(A) \quad (\text{s2})$$

$$\text{compose_delOdds}(A,[],B) \leftarrow B=[A], \text{ even}(A) \quad (\text{s3})$$

$$\text{delOdds}([A],B) \leftarrow B=[], \text{ odd}(A) \quad (\text{s4})$$

$$\text{delOdds}([A],B) \leftarrow B=[A], \text{ even}(A) \quad (\text{s5})$$

Now, upon backtracking, unifying the goal `delOdds([A],B)` with the head of the non-recursive clause of the open program yields the goal:

$$\leftarrow \text{solve_delOdds}([A],B)$$

where resolving this goal is impossible. In this case, DIALOGS-II directly collects evidence for `solve_delOdds` using the shortcuts `s4` and `s5` instead of generating a query that would be identical to the one made for the abduction of evidence for `compose_delOdds(C,[],B)`. Thus, the evidence collected for `solve_delOdds` is the following:

$$\text{solve_delOdds}([A],B) \leftarrow B=[], \text{ odd}(A)$$

$$\text{solve_delOdds}([A],B) \leftarrow B=[A], \text{ even}(A)$$

Third Iteration for Abducing Evidence. Next, the specifier is queried for the result when the induction parameter is a two-element list. Again, the specifier *must* know the answer. DIALOGS-II first creates the following clause by unifying the goal $\text{delOdds}([A,B],C)$ with the head of the recursive clause of the open program yielding the goal:

$$\leftarrow \text{decompose_delOdds}([A,B],HA,TA), \text{delOdds}(TA,TB), \\ \text{compose_delOdds}(HA,TB,C)$$

Resolving $\text{decompose_delOdds}([A,B],HA,TA)$ and the resulting equality atom, and using the shortcut **s4** reduces this goal to:

$$\leftarrow \text{odd}(B), \text{compose_delOdds}(A,[],C)$$

Note that the atom $\text{odd}(B)$ is an atom of the specifier-introduced predicate odd , and remember that during the SLD resolution of a goal, if such an atom is encountered, then this atom is assumed to be true since it was introduced by the specifier, and kept since it is passed to the next query. Thus, the goal becomes:

$$\leftarrow \text{compose_delOdds}(A,[],C)$$

Using **s2**, this becomes:

$$\leftarrow \text{odd}(A), C=[]$$

Again note that $\text{odd}(A)$ is assumed to be true since it is an atom of the specifier-introduced predicate, and it is kept for the next query. So, now Assumptions becomes equal to the set $\{\text{odd}(B), \text{odd}(A)\}$. Thus, the goal becomes:

$$\leftarrow C=[]$$

which is finally resolved to:

$$\leftarrow \text{true}$$

Since there is no atom of any open relation in that goal, no query can be generated from it (thus, in that case the assumptions collected are not used).

Next, upon backtracking, by the use of the other shortcut, i.e. **s5**, the following goal is obtained:

$$\leftarrow \text{even}(B), \text{compose_delOdds}(A,[B],C)$$

where the atom $\text{even}(B)$ is assumed to be true and collected as an assumption to be passed to the next query, again because it is an atom of a specifier-introduced predicate. Thus, the goal becomes:

$\leftarrow \text{compose_delOdds}(A,[B],C)$

where resolving this goal is impossible, so that the following query is generated (note the usage of the assumption $\text{even}(B)$ in the query):

When does $\text{delOdds}([A,B],C)$ hold, assuming $\text{even}(B)$?

$C=[B], \text{odd}(A); C=[A,B], \text{even}(A).$

The following shortcuts and evidence are abduced from the answer:

$\text{compose_delOdds}(A,[B],C) \leftarrow C=[B], \text{odd}(A)$

$\text{compose_delOdds}(A,[B],C) \leftarrow C=[A,B], \text{even}(A)$

$\text{delOdds}([A,B],C) \leftarrow C=[B], \text{odd}(A), \text{even}(B)$ (s6)

$\text{delOdds}([A,B],C) \leftarrow C=[A,B], \text{even}(A), \text{even}(B)$ (s7)

Unifying the goal $\text{delOdds}([A,B],C)$ with the head of the non-recursive clause of the open program would yield the goal

$\leftarrow \text{solve_delOdds}([A,B],C)$

Since the system now knows when $\text{delOdds}([A,B],C)$ holds (see shortcuts s6 and s7), the specifier is not queried, and by using the shortcuts s6 and s7, DIALOGS-II directly abduces the evidence:

$\text{solve_delOdds}([A,B],C) \leftarrow C=[B], \text{odd}(A), \text{even}(B)$

$\text{solve_delOdds}([A,B],C) \leftarrow C=[A,B], \text{even}(A), \text{even}(B)$

If first the goal $\text{delOdds}([A,B],C)$ had been unified with the non-recursive clause yielding the goal

$\leftarrow \text{solve_delOdds}([A,B],C)$

where resolving this goal is impossible, then the specifier would have been queried as follows:

When does $\text{delOdds}([A,B],C)$ hold?

where s/he should have answered this query as:

$C=[B], \text{odd}(A), \text{even}(B); C=[A,B], \text{even}(A), \text{even}(B); C=[], \text{odd}(A), \text{odd}(B);$
 $C=[A], \text{even}(A), \text{odd}(B)$

Note that the specifier would have to write a longer answer for this query than for the one that was asked for compose_delOdds . That is why the goal is unified first with the recursive clause rather than the non-recursive one as explained earlier.

Stopping the Query Session. Next, the specifier is queried for the result when the induction parameter is a three-element list. Suppose that the specifier is bored or believes having said sufficiently many useful things about `delOdds` and does not want to answer any queries anymore. In that case, the specifier answers the query by the keyword “stop-it”, so that the query session is ended:

```
When does delOdds([A,B,C],D) hold, assuming even(B),  
even(C)? stop-it.
```

Stopping the querying is thus fully manual (specifier-dependent). Actually, there are two other possibilities to stop querying: the first one is fully automatic, the second one is semi-automatic.

In the first one, a heuristic is used to conjecture whether the system has to stop querying or not. The heuristic is as follows: after abducing evidence for p and q after each query, all the abduced evidence for p and q is processed (by the Program Closing Method) and compared with the result of the same process done on the evidence collected for the previous query. If the results of these two processes are the same, then it is assumed that the potential next queries would also yield the same results, so it is conjectured that the system can stop querying and rely on the evidence that was collected until that time. This method is fully-automatic, because the system makes its decision without any interaction with the specifier. But, due to its being a heuristic, the system can be defeated.

The second method is a combination of the other two methods. The system processes all the evidence after each query, and if the last two successive results are the same, it asks the specifier to conjecture whether to continue querying or not, since there is a possibility that the abduced evidence is adequate for induction of a correct program. If the specifier thinks that this much evidence is sufficient to induce a correct program, then a program is induced from this evidence, otherwise s/he is further queried until s/he decides that the abduced evidence is adequate.

We think that the DIALOGS-II method is the most appropriate one. Its method is better than the fully-automatic one since it leaves the decision to the specifier, so that it is always possible to induce a correct program either by a first correct decision of the specifier on stopping querying, or by successive syntheses that would let the specifier synthesize a correct program in the end, by making the specifier learn that s/he should answer more queries each time the system is re-run. This method has a drawback be-

cause of its being a heuristic. It fails when a correct program can only be induced after some other queries. That is, abduction of some more new evidence could cause a change in the result of each process done after each query. In that case, the program induced could be incomplete/incorrect. The second method is mostly for expert specifiers since the decision whether the abduced evidence is adequate or not is not an easy decision for a naive specifier, where a specifier who has the knowledge and capability to make such a decision is considered an expert specifier, whereas a specifier who is not capable of making such a decision is considered a naive one. However, the naive specifier could decide to stop querying the first time the system asks to make a decision. In that case, this method boils down to a combination of the other two methods.

Now, let us see how the abduced evidence for `solve_delOdds` and `compose_delOdds` will be processed in order to find programs for these relations.

2.4 Induction of Clauses: The Program Closing Method

The Program Closing Method discussed in this section is based on the Program Closing Method discussed in [10]. There, the open program has only one relation that will be closed using evidence for that relation. According to our Program Closing Method, there are two open relations of the open program, i.e. p and q . Let us see now how it works.

The evidence abduced for the open relations p and q during the execution of the third statement of Algorithm 2 is divided into subsets such that the $lg\theta$ of each subset yields a clause for either p or q . In order to understand how this division into subsets and taking the $lg\theta$ of each subset is done, we have to first analyze the dataflow of the programs that have p and q as open relations. In other words, we have to look inside the open relations p and q .

Here, we analyze the data-flow of divide-and-conquer programs, which have `solve_r` and `compose_r` as open relations (see the divide-and-conquer schema on page 23).

Using general knowledge of the divide-and-conquer design methodology, it is possible to conjecture that, in general, the construction mode (see Section 1.1.6) of `compose_r(HX, TY, Y, Z)` is

`compose_r(may, must, res, may),`

where the first *may* denotes *may, ..., may* with h occurrences of *may*, the second *may* denotes *may, ..., may* with z occurrences of *may*, and *must* denotes *must, ..., must* with t occurrences of *must* (remember that h is the number of heads HX_i , that z is the number of passive parameters, and that t is the number of tails TY_i).

Indeed, the TY_i being obtained through recursion, they must all somehow be used to construct Y , because some of the recursive calls would otherwise have been useless. The HX_i need not always be used to construct Y , as it depends on the particular program. So there is no fixed mode for the head(s) of the induction parameter, and their most general mode thus is *may*. The passive parameter(s) Z also need not always be used to construct Y . So there also is no fixed mode for the passive parameter(s), and their most general mode thus also is *may*.

Similarly, one can argue that the mode of `solve_r(X, Y, Z)` is `solve_r(may, res, may)`, where *may* denotes *may, ..., may* with z occurrences of *may*. The inductive parameter X and the passive parameter(s) Z need not always be used to construct the result parameter Y . So there are no fixed modes for X and Z , and their most general mode thus is *may*.

The evidence abduced for the open relations p and q needs to be processed according to the Program Closing Method so that admissible clauses (see Section 1.1.6) for the open relations p and q are obtained. We give an algorithm below for the realization of this process (note that `solve_r` plays the role of p , and `compose_r` plays the role of q):

Algorithm 8: `induce(pEvidence, qEvidence, pMode, qMode, pClauses, qClauses)`

Inputs: `pEvidence, qEvidence, pMode, qMode`

Outputs: `pClauses, qClauses`

divide the (evidence) clause set for q , i.e. `qEvidence`, into a minimal number of subsets (called cliques) of which any two elements have an admissible $lg\theta$, i.e.

`qCliques` (see [10] for an efficient algorithm for this NP-complete problem)

`qCliques := division(qEvidence, qMode)`

analyze every such clique: if the $lg\theta$ of the counterpart subset of the clauses for p is also admissible, then delete the clique from the clauses for q ; otherwise

delete that counterpart subset from the clauses for p , and thus obtain

`NewqCliques` and `NewpEvidence`.

$\langle \text{NewqCliques}, \text{NewpEvidence} \rangle := \text{prune}(\text{qCliques}, \text{pMode}, \text{pEvidence})$

take the $\text{lg}\theta$ s of the remaining cliques, i.e. NewqCliques , as clauses of q , i.e.

$q\text{Clauses}$

$q\text{Clauses} := \{\text{lg}\theta(c) \mid c \in \text{NewqCliques}\}$

divide the remaining clause set for p , i.e. NewpEvidence , into a minimal number of cliques such that any two elements in each clique have an admissible $\text{lg}\theta$, i.e.

$p\text{Cliques}$

$p\text{Cliques} := \text{division}(\text{NewpEvidence}, \text{pMode})$

build admissible clauses, i.e. $p\text{Clauses}$, of the p from their $\text{lg}\theta$ s, i.e. $p\text{Cliques}$

$p\text{Clauses} := \{\text{lg}\theta(c) \mid c \in p\text{Cliques}\}$

Let us now turn back to the synthesis of a program for delOdds and see how the “closing” of open relations solve_delOdds and compose_delOdds is done according to Algorithm 8. The evidence collected for the open relations solve_delOdds and compose_delOdds is (see previous sub-section):

$\text{solve_delOdds}([A,B],[A,B]) \leftarrow \text{even}(A), \text{even}(B)$	(1)	$\text{compose_delOdds}(A,[B],[A,B]) \leftarrow \text{even}(A)$
$\text{solve_delOdds}([A,B],[B]) \leftarrow \text{odd}(A), \text{even}(B)$	(2)	$\text{compose_delOdds}(A,[B],[B]) \leftarrow \text{odd}(A)$
$\text{solve_delOdds}([A],[A]) \leftarrow \text{even}(A)$	(3)	$\text{compose_delOdds}(A,[],[A]) \leftarrow \text{even}(A)$
$\text{solve_delOdds}([A],[]) \leftarrow \text{odd}(A)$	(4)	$\text{compose_delOdds}(A,[],[]) \leftarrow \text{odd}(A)$
$\text{solve_delOdds}([],[]) \leftarrow$	(5)	(no counterpart)

Following the statements of Algorithm 8, DIALOGS-II first divides the compose_delOdds evidence into the following cliques:

$\text{compose_delOdds}(A,B,[A|B]) \leftarrow \text{even}(A)$ (1,3)

$\text{compose_delOdds}(A,B,B) \leftarrow \text{odd}(A)$ (2,4)

where the first clique is constructed by taking the $\text{lg}\theta$ of (1) and (3), and the second one by taking the $\text{lg}\theta$ of (2) and (4) of the compose_delOdds evidence. Next, it analyzes the counterpart sets for solve_delOdds . That is, it takes the $\text{lg}\theta$ of (1) and (3), as well as the $\text{lg}\theta$ of (2) and (4) of the solve_delOdds evidence, and thus obtains:

$\text{solve_delOdds}([A|B],[A|B]) \leftarrow \text{even}(A), \text{even}(C)$ (1,3)

$\text{solve_delOdds}([A|B],B) \leftarrow \text{odd}(A), T$ (2,4)

None of these two clauses is admissible since the first one contains a literal, i.e. $\text{even}(C)$, in its body, which has an unlinked variable, i.e. C . And, the second one is not admissible because the body contains T (see Section 1.1.6). Thus, the counterpart sets of solve_delOdds , i.e. $\{(1), (3)\}$ and $\{(2), (4)\}$ are eliminated from the

`solve_delOdds` evidence set and the cliques of `compose_delOdds` are kept. The $\text{lg}\theta$ s of these two cliques become thus clauses of `compose_delOdds`, namely the clauses that will be in the final program. The remaining set for `solve_delOdds` is

$$\text{solve_delOdds}([],[]) \leftarrow$$

and since this set is a clique and is admissible, its $\text{lg}\theta$ (i.e. itself) becomes a clause for `solve_delOdds`.

Now, the open relations `solve_delOdds` and `compose_delOdds` are “closed”, that is they have an interpretation, and the open program constructed from the initial schema is also “closed” since it has no open relations. The final step in the synthesis is adding the clauses of the open relations to the open program to close the open program. In that way, the final program becomes:

$$\begin{aligned} \text{delOdds}(A,B) &\leftarrow \text{solve_delOdds}(A,B) \\ \text{delOdds}(A,B) &\leftarrow \text{decompose_delOdds}(A,C,D), \text{delOdds}(D,E), \\ &\qquad\qquad\qquad \text{compose_delOdds}(C,E,B) \\ \text{decompose_delOdds}(F,G,H) &\leftarrow F=[G|H] \\ \text{solve_delOdds}([],[]) &\leftarrow \\ \text{compose_delOdds}(A,B,[A|B]) &\leftarrow \text{even}(A) \\ \text{compose_delOdds}(A,B,B) &\leftarrow \text{odd}(A) \end{aligned}$$

This program is correct with respect to its specification. Post-synthesis transformations that optimize the final programs are not our concern in this thesis. See [6] if you want to know more about them.

2.5 Evaluation of the Program Closing Method

Finding a program for the open relation of the recursive clause of an open program, i.e. the relation q , via the Program Clausuring Method assumes that there is a finite non-recursive program for that relation. However such is not always the case. That is, there might be a recursive one instead. In other words, the system might have to do a necessary predicate invention.

2.5.1 Necessary Predicate Invention

How can the system possibly decide that the result of the Program Closing Method is wrong, that is that the finite non-recursive program that was induced for the relation q via the Program Closing Method is incomplete, and that it has to invent a predicate with a recursive program after rejecting the result of the Program Closing Method? These questions imply that some heuristic needs to be used for detecting and handling necessary predicate invention [12] [25].

Since the Program Clausuring Method has been devised to always succeed (indeed, in the worst case, it divides a clause set into cliques of one element each), a heuristic is needed for rejecting the results of the Program Clausuring Method and conjecturing necessity of the predicate invention. For the time being, we do not have an acceptable heuristic that frequently correctly conjectures necessary predicate invention, whenever there is a need to synthesize a recursive program. Therefore, in DIALOGS-II, the decision of predicate invention is specifier-dependent. That is, the specifier is asked whether the system should reject the result of the Program Closing Method and synthesize a recursive program (do a necessary predicate invention), or whether it should use the result of the Program Closing Method. If the result of the Program Clausuring Method is rejected by the specifier, then DIALOGS-II re-invokes itself under the assumption that a recursive logic program exists for the open relation.

In general, DIALOGS-II is called with a *start program*: this is the empty set in the case of a new synthesis (for the *top-level predicate*), or a set of clauses for a (unique) top-level predicate and its (directly or indirectly) used predicates, in case DIALOGS-II is used (possibly by itself) for a necessary invention of a predicate that is (directly or indirectly) used by the top-level predicate. In case there is a predicate invention, the new program synthesized for the new predicate is added to the start program, otherwise the clauses induced by the Program Closing Method are added to the start program, yielding the final program.

We saw how query generation and answering take place when there is no predicate invention and how the result of the Program Closing Method is used for “closing” the open relations during the synthesis of a program for $delOdds$. Now let us see how this is done in case of necessary predicate invention: when a necessity of predicate invention is conjectured, query generation during the synthesis of the new predicate is always done for the top-level predicate, but resolution will eventually be blocked by an

open relation of the current predicate and thus the system will extract a question for it in terms of the top-level one. This is because the user does not always (see the next sub-section for an exceptional case) need to know the predicate being invented, but s/he has to know the top-level predicate since otherwise s/he would not even have the need for a program for the top-level predicate. Thus, DIALOGS-II generates queries for the new predicate in terms of the top level predicate, but resolution is eventually blocked by an open atom of the program of the new predicate, i.e. current predicate, and extract a question for it in terms of the top-level one.

Now, we introduce two new concepts: the concept of giving hints and the concept of calling DIALOGS-II in a certain mode: *aloud* or *mute*. Let us first discuss the concept of giving hints: hints about the roles of the parameters of a certain parameter declaration can be given to the system. In a recursive call of DIALOGS-II itself, it is possible to hint about the parameter roles of the new predicate (how this is done will be explained later). So, we can say that DIALOGS-II can be called with hints about the roles of the parameters (if there are any hints), where the initial call of DIALOGS-II for the top-level predicate is done with an empty hint list. DIALOGS-II has preference of hints over defaults. In other words, if there are any hints, then the system uses these hints instead of using the defaults.

Now let us introduce the concept of calling the system in *mute* or *aloud* mode: DIALOGS-II is said to be in *aloud* mode when it asks the specifier for a predicate declaration, a schema, a strategy, parameter roles and a decomposition operator, and gets the answer from the specifier whereas it is said to be in *mute* mode when the specifier is queried for nothing, where the system itself answers the questions by itself. By default, the system is in *aloud* mode when it starts synthesis, but it is called in *mute* mode when there is necessary predicate invention. Now, we give an algorithm that realizes all the observations and discussions explained so far. What this algorithm basically does is that it *evaluates* the result of the Program Closing Method based on the specifier's evaluation of the Program Closing Method and calls DIALOGS-II recursively, in *mute* mode, to synthesize a recursive program for the predicate q if predicate invention is necessary, otherwise it uses the result of the Program Closing Method to produce a non-recursive program for the relation q :

Algorithm 9: evaluate(Schema,Strategy,CurrOpenPgm,
pClauses,qClauses,PredDecl,ParamRoles,Pgm)

Inputs: Schema, Strategy, CurrOpenPgm, pClauses, qClauses,
PredDecl, ParamRoles

Outputs: Pgm

display the result of the Program Closing Method

display(pClauses,qClauses)

ask the specifier if predicate invention is necessary

Answer := ask('Please evaluate the Program Closing Method: need for
recursive synthesis? [yes/no]')

if Answer=yes then

*determine the predicate declaration for the new predicate for which a
recursive program is being synthesized using ParamRoles of TopPred,
where TopPred is the name of the predicate given in PredDecl*

NewPredDecl := predDeclL(ParamRoles)

TopPred := predName(PredDecl)

*add the clauses for the relation p, i.e. SelectedpClauses, which are from
pClauses and have no counterparts among the clauses of qClauses, to
CurrOpenPgm to obtain NewOpenPgm*

SelectedpClauses := select(pClauses,qClauses)

NewOpenPgm := CurrOpenPgm \cup SelectedpClauses

construct hints about the roles of the parameters

Hints := constructHints(NewPredDecl)

setMode(mute)

*call DIALOGS-II recursively with the new predicate declaration and hints to
induce a program for the new predicate*

dialogsII(Schema,Strategy,NewPredDecl,NewOpenPgm,Hints,
TopPred,Pgm)

else

add the clauses pClauses and qClauses to CurrOpenPgm to obtain Pgm

Pgm := CurrOpenPgm \cup pClauses \cup qClauses

Note that in Algorithm 9, DIALOGS-II is now called with NewPredDecl and Hints
about the parameter roles, where the final program for the new predicate will be added

to NewOpenPgm, which is an open program (whose relation q is still open) for Top-Pred.

Synthesis of a Program for reverse(L,R). Now we will illustrate how Algorithm 9 works by means of the synthesis of a program for reverse(L,R), where reverse(L,R) holds iff list R is the reverse of list L. Since we already discussed the first two statements of the basic synthesis algorithm, i.e. Algorithm 1, and first two statements of Algorithm 2, in terms of the synthesis of a program for delOdds, we will skip these statements in the illustration of the synthesis of a program for reverse(L,R), where we will only give the results of these statements.

By the execution of the first statement (execution of the strategy) of Algorithm 2, the following open program for reverse(L,R) has been generated:

```
reverse(A,B) ← solve_reverse(A,B)
reverse(A,B) ← decompose_reverse(A,C,D), reverse(D,E),
                                     compose_reverse(C,E,B)
decompose_reverse(F,G,H) ← F=[G|H]
```

where A is the induction parameter and B is the result parameter.

Remember that by executing the second statement of Algorithm 2, the evidence for the open relations, i.e. p and q , is abducted. So, at the end of the second statement, the evidence for solve_reverse and compose_reverse is as given below in the form of counterparts:

solve_reverse([A],[A]) ←	(1)	compose_reverse(A,[],[A]) ←
solve_reverse([A,B],[B,A]) ←	(2)	compose_reverse(A,[B],[B,A]) ←
solve_reverse([A,B,C],[C,B,A]) ←	(3)	compose_reverse(A,[B,C],[B,C,A]) ←
solve_reverse([A,B,C,D],[D,C,B,A]) ←	(4)	compose_reverse(A,[B,C,D],[B,C,D,A]) ←
solve_reverse([],[]) ←	(5)	(no counterpart)

where the Program Closing Method results in the following clauses for the open relation solve_reverse (note that there is no compose_reverse clause):

```
solve_reverse(A,A) ←
solve_reverse([A,B],[B,A]) ←
solve_reverse([A,B,C],[C,B,A]) ←
```

Now, it is time to query the specifier about the result of the Program Closing Method to conjecture whether predicate invention is necessary or not.

Please evaluate the Program Closing Method: need for recursive synthesis? [yes/no] **yes**

The specifier here answers the query by *yes* believing that there exists a recursive program for the predicate of the recursive clause of the open program, i.e. `compose_reverse`, rejecting the result of the Program Closing Method. Since the system now knows that it should synthesize a recursive program for the `compose_reverse` predicate, it needs to call itself recursively. But, before doing this it should first elaborate a predicate declaration for the predicate, and construct hints about the parameter roles, and compute the new start program for the new synthesis by adding the clauses for the relation `solve_reverse` that have no counterparts among the clauses of `compose_reverse`.

Determination of a Predicate Declaration for the New Predicate. Now, let us go through the steps of determination of a predicate declaration for the new predicate one by one to see how they are realized. First, we discuss how the new predicate declaration is elaborated. A predicate declaration has two components: the name of the predicate and the list of parameters together with their types. The name of the new predicate is already known, which is `compose_reverse`. The list of parameters together with their types is elaborated as follows: it is known that the new predicate has three parameters. The type of the first parameter is found to be `int`, since in the open program given above the first parameter of `compose_reverse`, i.e. `C`, is the head (namely, an element) of the list `A`, where the type of the parameter `A` is `list(int)`. The type of the second parameter is found to be `list(int)`, since in the open program the second parameter of `compose_reverse`, i.e. `E`, is the result parameter of the recursive call, i.e. `reverse(D,E)`, where the result parameter of the `reverse` predicate is of type `list(int)`. Finally, the third parameter is found to be of type `list(int)`, since it also is the result parameter of the open program, where its type is `list(int)`. Thus, using the information about the name of the new predicate and the parameters together with their types, the predicate declaration for the new predicate is constructed as shown below:

```
compose_reverse(HL:int,TR:list(int),R:list(int))
```

Now, the system has a predicate declaration of the new predicate for which it will call itself to induce a program.

Construction of Hints. Next, it has to construct hints about the parameter roles, i.e. which parameter is the induction parameter, which one is (are) the result parameter(s) (if any), and which one is (are) the passive parameter(s) (if any), in order to call itself in *mute* mode with these hints (remember that DIALOGS-II has a preference of hints over defaults in *mute* mode). It is reasonable that R (see the predicate declaration above) is hinted as the result parameter since the corresponding parameter B in the open program (see the open program on page 44) is the result parameter of the program, and it is also reasonable to hint TR as the induction parameter since it is of an inductively defined type, and finally to hint the remaining parameter HL as the passive parameter. In general, the result parameter of the open relation q in the open program can be hinted as a result parameter for the new predicate, a parameter which is the result parameter of the recursive call in the open program can be hinted as an induction parameter if it is of an inductively defined type, and the remaining parameters as the passive parameters. Here we described the determination of hints about the parameter roles for a divide-and-conquer schema, since we are illustrating the synthesis of a program that fits a divide-and-conquer schema. The construction of hints would be different if the schema were another one, e.g. descending-generalization, since the parameter roles of the schema would be different.

Construction of a Start Program for the New Synthesis. What DIALOGS-II does after elaboration of the new predicate declaration and construction of hints is that it constructs a start program for the new synthesis by using the evidence clauses abduced during execution of the second statement of Algorithm 2. How this is done is as follows: the system adds the abduced clauses for the relation p that have no counterparts among the abduced clauses for the relation q to the open program to obtain the start program for the new synthesis. The clause

$$\text{solve_reverse}([],[]) \leftarrow$$

has no counterparts among the abduced clauses for the relation `compose_reverse` (see page 45). Thus the start program for the new synthesis is:

$$\begin{aligned} \text{reverse}(A,B) &\leftarrow \text{solve_reverse}(A,B) \\ \text{reverse}(A,B) &\leftarrow \text{decompose_reverse}(A,C,D), \text{reverse}(D,E), \\ &\qquad\qquad\qquad \text{compose_reverse}(C,E,B) \end{aligned}$$

```

decompose_reverse(F,G,H) ← F=[G|H]
solve_reverse([],[]) ←

```

Now, it is time for the system to re-invoke itself on this start program using the new predicate declaration and the hints.

Calling DIALOGS-II Recursively. Before calling the system recursively, the synthesis mode is converted into *mute* mode. DIALOGS-II first determines the roles of the parameters that are given inside the predicate declaration using the hints and the decomposition operator using the defaults. Next, an open program is generated for the new predicate, and this open program is added to the start program to obtain the new open program that will be used for the abduction of the new evidence for the open relations of the open program of the new predicate. The second and third statements (abduction of evidence and induction of clauses) of Algorithm 2 are then executed using this new open program to “close” the open relations of the open program of the new predicate. Let us now see how all this is done during the synthesis of a program for the `reverse(L,R)` predicate.

DIALOGS-II first determines the roles of the parameters of the predicate declaration:

```

compose_reverse(HL:int,TR:list(int),R:list(int))

```

using the hints determined previously. That is, the induction parameter is `TR`, the result parameter is `R`, and the passive parameter is `HL` (note that DIALOGS-II does not query the user for that since it uses the hints).

```

Induction parameter? {TR} TR
Result parameter? {R} R
Passive parameter(s)? {[HL]} [HL]

```

Next, it determines `decompose_compose_reverse` by using the default one.

```

Decomposition operator?
{decompose_compose_reverse(L,H,T) ← L=[H|T]}
decompose_compose_reverse(L,H,T) ← L=[H|T]

```

Next, it generates the following open program using `decompose_compose_reverse`:

```

reverse(A,B) ← solve_reverse(A,B)
reverse(A,B) ← decompose_reverse(A,C,D), reverse(D,E),

```

```

compose_reverse(C,E,B)
decompose_reverse(F,G,H) ← F=[G|H]
solve_reverse([],[]) ←
compose_reverse(G,M,N) ← solve_compose_reverse(G,M,N)
compose_reverse(G,H,I) ← decompose_compose_reverse(H,J,K),
    compose_reverse(G,K,L), compose_compose_reverse(J,L,I,G)
decompose_compose_reverse(F,G,H) ← F=[G|H]

```

Now, it is time to abduce evidence for the open relations, i.e. `compose_compose_reverse` and `solve_compose_reverse`, where during the abduction of the evidence for these relations, the system does not query the specifier, but uses the shortcuts for the top-level predicate `reverse`, except in the case where there is no shortcut left after using the available shortcuts:

```

reverse([],[]) ← (s1)
reverse([A],[A]) ← (s2)
reverse([A,B],[B,A]) ← (s3)
reverse([A,B,C],[C,B,A]) ← (s4)
reverse([A,B,C,D],[D,C,B,A]) ← (s5)

```

After that point, the SLD resolution of a goal for the top-level predicate is blocked by an open atom, and the system extracts a query for this open atom, where the answer to that query is found using the shortcuts of the top-level predicate. Let us now go through the steps of “closing” the open relations of the open program given above.

The most general form of the goal when the size of the induction parameter, i.e. `A`, is 0 is the following: `← reverse([],X)`. This goal is first tried to be resolved with the recursive clause of the `reverse` predicate, where this attempt fails since resolving `decompose_reverse` when the induction parameter `A` is the empty list, i.e. `[]`, is impossible. The system next resolves the goal with the non-recursive clause of the open program. As a result of this resolution, the goal

```
← true
```

is reached because the predicate `solve_reverse` is already closed (there are clauses for the `solve_reverse` predicate), and thus there is no need to abduce evidence for it.

Next, the goal $\leftarrow \text{reverse}([X],Y)$, where the induction parameter is a one-element list, is resolved with the clauses of the `reverse` predicate, first with the recursive clause yielding the goal:

$$\begin{aligned} \leftarrow & \text{decompose_reverse}([X],C,D), \text{reverse}(D,E), \\ & \text{compose_reverse}(C,E,Y) \end{aligned}$$

Resolving `decompose_reverse`($[X],C,D$) and the resulting equality atom, and using the shortcut `s1` gives:

$$\leftarrow \text{compose_reverse}(X,[],Y)$$

Since there is no shortcut for `compose_reverse`($X,[],Y$) (shortcuts obtained before starting the new synthesis are not kept, to prevent them from being accidentally used by the new synthesis as shortcuts, see Algorithm 5), the goal $\leftarrow \text{compose_reverse}(X,[],Y)$ is resolved with the non-recursive clause of `compose_reverse` (note that the recursive clause cannot be resolved since the induction parameter, i.e. $[]$, cannot be decomposed) yielding the goal:

$$\leftarrow \text{solve_compose_reverse}(X,[],Y)$$

There is neither a shortcut nor a program for the predicate `solve_compose_reverse`, so resolving this goal is impossible. Therefore, it is time to make a query out of this goal. Since there is a shortcut, i.e. `s2`, the system uses the shortcut `s2` to abduce the following evidence and shortcuts for `solve_compose_reverse` and `compose_reverse` without any need for a query:

$$\begin{aligned} \text{solve_compose_reverse}(X,[],[X]) & \leftarrow \\ \text{compose_reverse}(X,[],[X]) & \leftarrow \end{aligned} \tag{s6}$$

The system resolves the goal $\leftarrow \text{reverse}([X],Y)$ with the non-recursive clause of the `reverse` predicate. As a result of this resolution, the goal

$$\leftarrow \text{true}$$

is reached because the predicate `solve_reverse` is already closed (there are clauses for the `solve_reverse` predicate), and thus there is no need to abduce evidence for it.

Next, the goal $\leftarrow \text{reverse}([X,Y],W)$ is resolved with the recursive clause of the `reverse` predicate, yielding the goal:

$$\begin{aligned} \leftarrow & \text{decompose_reverse}([X,Y],C,D), \text{reverse}(D,E), \\ & \text{compose_reverse}(C,E,W) \end{aligned}$$

Resolving `decompose_reverse([X,Y],C,D)` and the resulting equality atom, and using the shortcut (s2), the goal becomes

$\leftarrow \text{compose_reverse}(X,[Y],W)$

Since there is no shortcut for `compose_reverse(X,[Y],W)`, it is resolved with the recursive clause of the `compose_reverse` predicate yielding the goal

$\leftarrow \text{decompose_compose_reverse}([Y],J,K), \text{compose_reverse}(X,K,L),$
 $\text{compose_compose_reverse}(J,L,W,X)$

Resolving `decompose_compose_reverse([Y],J,K)` and the resulting equality atom, and using (s6) gives

$\leftarrow \text{compose_compose_reverse}(Y,L,W,X)$

Resolving this goal is impossible since there is neither a shortcut nor a clause for `compose_compose_reverse`. So, the following evidence and shortcuts are abduced using the shortcut s3:

$\text{compose_compose_reverse}(Y,[X],[Y,X],X) \leftarrow$
 $\text{compose_reverse}(X,[Y],[Y,X]) \leftarrow \tag{s7}$

Upon backtracking, the goal

$\leftarrow \text{compose_reverse}(X,[Y],W)$

is resolved with the non-recursive clause of the `compose_reverse` yielding the goal:

$\leftarrow \text{solve_compose_reverse}(X,[Y],W)$

Using the shortcut s7 the following evidence is abduced:

$\text{solve_compose_reverse}(X,[Y],[Y,X]) \leftarrow$

Next, this resolution process is also done for the most general values of the induction parameter *A* of the `reverse` predicate (see the open program for `reverse`) when the size of the induction parameter is three and then four, i.e. `reverse([X,Y,W],V)`, and `reverse([X,Y,W,V],Z)`. And, as a result of this process the following evidence and shortcuts are abduced:

$\text{solve_compose_reverse}(X,[W,Y],[W,Y,X]) \leftarrow \tag{2}$

$\text{solve_compose_reverse}(X,[V,W,Y],[V,W,Y,X]) \leftarrow \tag{1}$

$\text{compose_compose_reverse}(X,[W,Y],[X,W,Y],Y) \leftarrow \tag{2}$

$\text{compose_compose_reverse}(X,[W,Y,V],[X,W,Y,V],V) \leftarrow \tag{1}$

$\text{compose_reverse}(X,[W,Y],[W,Y,X]) \leftarrow$
 $\text{compose_reverse}(X,[V,W,Y],[V,W,Y,X]) \leftarrow$

The evidence given above for `solve_compose_reverse` and `compose_compose_reverse` together with the following evidence

$\text{solve_compose_reverse}(X,[],[X]) \leftarrow$ (no counterpart)
 $\text{solve_compose_reverse}(X,[Y],[Y,X]) \leftarrow$ (3)
 $\text{compose_compose_reverse}(Y,[X],[Y,X],X) \leftarrow$ (3)
 $\text{compose_reverse}(X,[],Y) \leftarrow$
 $\text{compose_reverse}(X,[Y],[Y,X]) \leftarrow$

abduced previously is input to the Program Closing Method in order to find programs for these open relations. Following Algorithm 8, DIALOGS-II first divides the `compose_compose_reverse` evidence into a clique and computes its $\text{lg}\theta$:

$\text{compose_compose_reverse}(L,[M|N],[L,M|N],P) \leftarrow$ (1,2,3)

where the clique is constructed by taking the $\text{lg}\theta$ of (1), (2) and (3) of `compose_compose_reverse` evidence (see Algorithm 8). Next, it analyzes the counterpart set for `solve_compose_reverse`. That is, it takes the $\text{lg}\theta$ of (1), (2), and (3) of the `solve_compose_reverse` evidence, and thus obtains:

$\text{solve_compose_reverse}(A,[B|C],[B,D|E]) \leftarrow$ (1,2,3)

Since this clause is not admissible (see Section 1.1.6), it is not kept. The remaining clause

$\text{solve_compose_reverse}(X,[],[X]) \leftarrow$

that has no counterpart is kept in the final program. Thus, as a result of the Program Closing Method the following two clauses are induced

$\text{solve_compose_reverse}(K,[],[K]) \leftarrow$
 $\text{compose_compose_reverse}(L,[M|N],[L,M|N],P) \leftarrow$

Adding these clauses to the open program gives the following program for the reverse predicate, which is correct with respect to its specification:

$\text{reverse}(A,B) \leftarrow \text{solve_reverse}(A,B)$
 $\text{reverse}(A,B) \leftarrow \text{decompose_reverse}(A,C,D), \text{reverse}(D,E),$
 $\text{compose_reverse}(C,E,B)$
 $\text{decompose_reverse}(F,G,H) \leftarrow F=[G|H]$

```

solve_reverse([],[]) ←
compose_reverse(P,V,W) ← solve_compose_reverse(P,V,W)
compose_reverse(P,Q,R) ← decompose_compose_reverse(Q,S,T),
    compose_reverse(P,T,U), compose_compose_reverse(S,U,R,P)
decompose_compose_reverse(F,G,H) ← F=[G|H]
solve_compose_reverse(K,[],[K]) ←
compose_compose_reverse(L,[M|N],[L,M|N],P) ←

```

2.5.2 Handling the Sparseness Problem

DIALOGS-II faces the *sparseness* problem [19] when not every value of the induction parameter of the new predicate, i.e. q , is “reachable” by the values of the induction parameter of the top-level predicate. That is, queries about the new predicate cannot always be asked in terms of the top-level one. To show how we solve this problem, we will examine the synthesis of a program for the **factorial** predicate, where **factorial**(N,F) holds iff natural number F is the factorial of natural number N . What happens during the synthesis of a **factorial** program, in short, is that the synthesis requires the invention of the multiplication predicate, where **multiplication**(A,B,C) holds iff natural number C is the product of natural numbers A and B , but actually only uses a sparse subset of the multiplication relation. That is, it uses the following subset of the multiplication relation.

```

multiplication(s(0),s(0),s(0))
multiplication(s(0),s2(0),s2(0))
multiplication(s2(0),s3(0),s6(0))
multiplication(s6(0),s4(0),s24(0))
...

```

So, the evidence abduced for the open relations of the open program of the **multiplication** relation is a sparse set of evidence from which it is not possible to induce a correct and complete **multiplication** program, nor in turn a correct and complete **factorial** program with respect to its specification. Here, we introduce a new solution to the sparseness problem. Before explaining this new approach, let us first give a new conjecture.

The “Yılmaz Conjecture”. We conjecture that if there is a relation such that during the synthesis of a program for that relation the sparseness problem occurs, then the

specifier should be able to answer the queries related to the relations that are intrinsic to the relation being induced (this is the exception that was mentioned in Section 2.5).

For instance, during the synthesis of a program for **factorial**, if the specifier is able to answer the query

When does `factorial(s3(0),L)` hold?

then s/he should also be able to answer the following query about **multiplication**, after having seen some evidence of the **multiplication** relation that was abduced and is different from the one given below:

When does `multiplication(s2(0),s3(0),M)` hold?

since, what s/he is actually doing while finding an answer to the query of the **factorial** relation is that s/he is using the **multiplication** relation, because otherwise s/he would not be able to answer the query about the **factorial** relation. In other words, **multiplication** is “intrinsic” to **factorial**.

In our approach to handling the sparseness problem, we use the idea given by the conjecture above. Before explaining how we use this idea, let us first investigate how the system conjectures that there is a sparseness problem. In **DIALOGS-II**, this detection is done by means of a heuristic. How this heuristic works is as follows: if the abduced evidence for the open relations in the open program for the new predicate (the evidence for the `solve_compose_q` and `compose_compose_q`) is *unbalanced*, that is, if there are at least three more `solve_compose_q` clauses than `compose_compose_q` clauses, then the system conjectures that there is a sparseness problem. The number three has been determined empirically (e.g. based on the results obtained during the synthesis of a program for the **factorial** predicate). When the system conjectures that there is a sparseness problem, the evidence abduced for `solve_compose_q` and `compose_compose_q` is discarded, and a new synthesis, in *aloud* mode, is started for the `q` predicate, after letting the specifier know that there will be a new synthesis for the new predicate, and s/he would need to answer the queries of that new synthesis. Let us now refine the algorithm *evaluate* (Algorithm 9) such that it conjectures the sparseness problem:

Algorithm 10: evaluate(Schema,Strategy,CurrOpenPgm,
pClauses,qClauses,PredDecl,ParamRoles,Pgm)

Inputs: CurrOpenPgm, pClauses,qClauses,TopPred,ParamRoles

Outputs: Pgm

display the result of the Program Closing Method

display(pClauses,qClauses)

ask the specifier if predicate invention is necessary

Answer := ask('Please evaluate the Program Closing Method: need for
recursive synthesis? [yes/no]')

if Answer = yes then

*determine the predicate declaration for the new predicate for which a
recursive program is being synthesized using ParamRoles of TopPred,
where TopPred is the name of the predicate given in PredDecl*

NewPredDecl := predDeclL(ParamRoles)

TopPred := predName(PredDecl)

*add the clauses for the relation p, i.e. SelectedpClauses, which are from
pClauses and have no counterparts among the clauses of qClauses, to
CurrOpenPgm to obtain NewOpenPgm*

SelectedpClauses := select(pClauses,qClauses)

NewOpenPgm := CurrOpenPgm \cup SelectedpClauses

*check if there is sparseness problem by calling DIALOGS-II recursively with
the new predicate declaration in mute mode using the heuristic*

qAndpEvidence := collectAssertedEvidence(q,p)

SynthesisMode := getMode()

setMode(mute)

dialogsII(Schema,Strategy,NewPredDecl,NewOpenPgm,[],

TopPred, Pgm)

setMode(SynthesisMode)

Sparseness := sparsenessHeuristic(qAndpEvidence)

if Sparseness = no then

Hints := constructHints(NewPredDecl)

setMode(mute)

call DIALOGS-II recursively with the new predicate declaration and

hints to induce a program for the new predicate such that final Pgm is obtained

dialogsII(Schema,Strategy,NewPredDecl,NewOpenPgm,Hints,
TopPred,Pgm)

else

let the specifier know that a new synthesis for new predicate is being started and display the abduced clauses for the new predicate

NewPred := predName(NewPredDecl)

Clauses := collectAssertedEvidence(NewPred)

display(Clauses)

setMode(aloud)

Hints := constructHints(NewPredDecl)

call DIALOGS-II recursively with the new predicate declaration and an empty hint list to induce a program for the new predicate in aloud mode

dialogsII(Schema,Strategy,NewPredDecl,NewOpenPgm,Hints,
NewPred,Pgm)

else

add the clauses pClauses and qClauses to the CurrOpenPgm to obtain

Pgm

Pgm := CurrOpenPgm \cup pClauses \cup qClauses

Note that if there is a sparseness problem, then the system will call DIALOGS-II recursively to induce a new program for the new predicate.

Now, let us examine the synthesis of a program for the factorial predicate. Suppose that the following open program for the factorial predicate is generated at the end of the execution of the first statement of Algorithm 2:

factorial(A,B) \leftarrow solve_factorial(A,B)

factorial(A,B) \leftarrow decompose_factorial(A,C,D), factorial(C,E),

compose_factorial(D,E,B)

decompose_factorial(F,G,H) \leftarrow F=s(G), H=F

And, also suppose that the Program Closing Method yields the following clauses for solve_factorial (note that no clause for compose_factorial has been induced):

```

solve_factorial(s3(0),s6(0)) ←
solve_factorial(s(A),s(A)) ←
solve_factorial(0,s(0)) ←

```

Now, suppose that the specifier is asked to evaluate the result of the Program Closing Method, and s/he rejects it (s/he thinks that predicate invention is necessary), and thus the open program given above becomes (see Algorithm 10):

```

factorial(A,B) ← solve_factorial(A,B)
factorial(A,B) ← decompose_factorial(A,C,D), factorial(C,E),
compose_factorial(D,E,B)
decompose_factorial(F,G,H) ← F=s(G), H=F
solve_factorial(0,s(0)) ←

```

Next, the predicate declaration for `compose_factorial` is determined, as it was done for the `compose_reverse` predicate, which is:

```

compose_factorial(A:nat,B:nat,C:nat)

```

Now, it is time for the system to detect if there is a sparseness problem. The sparseness problem is detected by calling `DIALOGS-II` in *mute* mode using the new predicate declaration (the shortcuts abduced for the `factorial` predicate previously are used for this new synthesis). Thus, the system abduces the following evidence for `solve_compose_factorial` (note that no clause for `compose_compose_factorial` is induced) at the end of the Program Closing Method of this new synthesis:

```

solve_compose_factorial(s(0),s(0),s(0)) ←
solve_compose_factorial(s2(0),s(0),s2(0)) ←
solve_compose_factorial(s3(0),s2(0),s6(0)) ←

```

The system now uses the heuristic to see if there is any sparseness problem: the number of clauses for `solve_compose_factorial` is three (at least three) more than the number of clauses for `compose_compose_factorial`. So, a correct program for the relations `solve_compose_factorial` and `compose_compose_factorial`, and thus for `compose_factorial` in turn, cannot be induced from this evidence, and therefore the evidence is eliminated. Thus, a new synthesis for a program for `compose_factorial` is started in *aloud* mode by letting the specifier know about this:

```

You must know the relation compose_factorial since it
is intrinsic to the factorial relation. The clauses of

```

this relation obtained during the synthesis are given below. The system is starting a new synthesis for that relation, so please answer the queries about it:

```
compose_factorial(s(0),s(0),s(0)) ←
compose_factorial(s2(0),s(0),s2(0)) ←
compose_factorial(s3(0),s2(0),s6(0)) ←
```

Note that the relation `compose_factorial` is actually the multiplication relation. And, with the new predicate declaration, the system is called recursively with an empty hint list yielding the resulting open program for `compose_factorial` at the end of the execution of these statements:

```
compose_factorial(T,U,V) ← solve_compose_factorial(T,U,V)
compose_factorial(T,U,V) ← decompose_compose_factorial(T,W,X),
    compose_factorial(W,U,Y), compose_compose_factorial(X,Y,V,U)
decompose_compose_factorial(F,G,H) ← F=s(G), H=F
```

Next, the query session for the synthesis of programs for `solve_compose_factorial` and `compose_compose_factorial` takes place to abduce evidence for these relations, where the specifier answers the queries:

```
When does compose_factorial(0,A,B) hold? B=0.
When does compose_factorial(s(0),A,B) hold? B=A.
When does compose_factorial(s(s(0)),A,B) hold? B=A+A.
When does compose_factorial(s(s(s(0))),A,B) hold?
B=A+A+A.
```

The abduced evidence from this query is:

```
solve_compose_factorial(0,A,0) ←
solve_compose_factorial(s(0),A,A) ←
solve_compose_factorial(s(s(0)),A,A+A) ←
solve_compose_factorial(s(s(s(0))),A,A+A+A) ←
compose_compose_factorial(s(0),0,A,A) ←
compose_compose_factorial(s(s(0)),A,A+A,A) ←
compose_compose_factorial(s(s(s(0))),A+A,A+A+A,A) ←
```

From this evidence, using the Program Closing Method, the following program is induced for `compose_factorial`:


```

compose_factorial(T,U,V) ← solve_compose_factorial(T,U,V)
compose_factorial(T,U,V) ← decompose_compose_factorial(T,W,X),
    compose_factorial(W,U,Y), compose_compose_factorial(X,Y,V,U)
decompose_compose_factorial(F,G,H) ← F=s(G), H=F
solve_compose_factorial(0,S,0) ←
solve_compose_factorial(s(0),R,R) ←
compose_compose_factorial(s(s(O)),P,P+Q,Q) ←

```

Finally, this new program for `compose_factorial` is added to the open program for `factorial` yielding the following program for the `factorial` predicate:

```

factorial(A,B) ← solve_factorial(A,B)
factorial(A,B) ← decompose_factorial(A,C,D), factorial(C,E),
    compose_factorial(D,E,B)
decompose_factorial(F,G,H) ← F=s(G), H=F
solve_factorial(0,s(0)) ←
compose_factorial(T,U,V) ← solve_compose_factorial(T,U,V)
compose_factorial(T,U,V) ← decompose_compose_factorial(T,W,X),
    compose_factorial(W,U,Y), compose_compose_factorial(X,Y,V,U)
decompose_compose_factorial(F,G,H) ← F=s(G), H=F
solve_compose_factorial(0,A,0) ←
solve_compose_factorial(s(0),B,B) ←
compose_compose_factorial(s(s(C)),D,D+E,E) ←

```

where this `factorial` program is correct with respect to its specification. If we partially evaluate this program, then we obtain the following program that is more “readable”:

```

factorial(0,s(0)) ←
factorial(A,B) ← A=s(C), factorial(C,E), compose_factorial(A,E,B)
compose_factorial(0,A,0) ←
compose_factorial(s(0),B,B) ←
compose_factorial(T,U,V) ← T=s(W), compose_factorial(W,U,Y),
    compose_compose_factorial(T,Y,V,U)
compose_compose_factorial(s(s(C)),D,D+E,E) ←

```

Chapter 3

Comparison of DIALOGS-II with other ILP Systems

We compare DIALOGS-II with other ILP systems in terms of the evidence given as input to the system, and in terms of the power of their schemata. We first discuss (in Section 3.1) the evidence given in the form of examples and given in the form of syntactic bias (see Section 1.1.3), and then (in Section 3.2) we compare other ILP systems with DIALOGS-II in terms of the schemata available to these systems.

3.1 Comparison in Terms of the Evidence

FOIL [24] is a general purpose system that induces recursive and non-recursive logic programs. In order to learn a recursive program for $\text{length}(A,L)$, where $\text{length}(A,L)$ holds iff natural number L is the length of the list A , it needs in the order of thousands of positive and negative examples. On the other hand, DIALOGS-II can synthesize a recursive logic program for $\text{length}(A,L)$ from as few as three positive examples. The reason for FOIL to consume that many examples for the synthesis of such a simple recursive program is that it is a general purpose synthesizer that does not differentiate between the synthesis of non-recursive programs and the synthesis of recursive ones. This leads to poor “recursion” handling, and, as a result, the necessity of thousands of

examples for “encoding” the recursion. As advocated by Biermann [4], we believe that it is more efficient to try a suite of fast and reliable class-specific synthesizers (and, if necessary, to fall back onto a general purpose synthesizer) than to simply run such a slow, if not unreliable, general-purpose synthesizer.

The TRACY system [3] gets a description of the hypothesis space in the form of a syntactic bias and induces recursive logic programs using that bias. Suppose that for the `append` predicate (where `append(A,B,C)` holds iff list `C` is the concatenation of list `A` in front of list `B`), the following bias, positive and negative examples, and mode declaration are given as inputs, where the program and mode declaration of the `=` predicate are considered given as background knowledge:

```
append(A,B,C) ← {B=C, A=[]}
append(A,B,C) ← {A=[H|T], B=[E|F], append(T,{E,B,A},{D,F}), C=[H|D]}
+append([a],[b],[a,b])
-append([a],[b],[a])
-append([a],[b],[b])
append_inout(in,in,out)
```

The curly braces used for writing the body atoms and the parameters denote one element of the powerset of the elements inside the braces. After generating all possible clauses in the hypothesis space encoded by the bias above, the set of clauses used in the derivation of the positive example such that these clauses do not cover any of the two negative examples yields the final program:

```
append(A,B,C) ← B=C, A=[]
append(A,B,C) ← A=[H|T], append(T,B,D), C=[H|D]
```

Note that the recursive call is already encoded in the bias: the technique itself cannot discover recursion. In that sense, the source already knows how to write a possible program for `append`. If the same synthesis would be done with DIALOGS-II, the source would not need to know how to write a program for `append`. In fact, this is the ideal scenario since the very aim of a synthesizer is to synthesize a program that is unknown (or not completely known) to the source; it is not to extract a possible program from the evidence that encodes this program.

In summary, DIALOGS-II synthesizes recursive logic programs from little evidence, and the source can use DIALOGS-II to synthesize a recursive logic program that is unknown to it.

3.2 Comparison in Terms of Schemata

METAINDUCE [18] is almost exactly a subset of DIALOGS-II. Its schema is a particular case of the divide-and-conquer schemata of DIALOGS-II, namely for ternary relations, induction parameter of type list, exactly one base clause (when the list is empty), exactly one recursive clause (when the list is non-empty), and head-tail decomposition of the list (i.e. exactly one recursive call). In other words, the divide-and-conquer schemata that can be used by DIALOGS-II is more powerful: the induction parameter is not necessarily of type list, as it can be of any type that is inductively defined, multiple base clauses and multiple recursive clauses are possible, and the decomposition is not necessarily a head-tail one.

CRUSTACEAN [1] [2] synthesizes recursive logic programs of the following schema:

$$\begin{aligned} p(A_1, \dots, A_n) &\leftarrow \\ p(A_1, \dots, A_n) &\leftarrow p(B_1, \dots, B_n) \end{aligned}$$

where the A_i and B_i are terms. This is a very restricted schema compared to the possible divide-and-conquer schemata of DIALOGS-II. It has only one base clause and one recursive clause. Moreover, because of the schema, there is no possibility of any kind of predicate invention.

The schema of the CILP system [19] is superior to that of CRUSTACEAN:

$$\begin{aligned} p(\dots) &\leftarrow \\ p(\dots) &\leftarrow p(\dots) \end{aligned}$$

or, in the case of necessary predicate invention, it is:

$$\begin{aligned} q(\dots) &\leftarrow \\ q(\dots) &\leftarrow q(\dots), \text{newp}(\dots) \\ \text{newp}(\dots) &\leftarrow \\ \text{newp}(\dots) &\leftarrow \text{newp}(\dots) \end{aligned}$$

The CILP schema is superior to the schema of CRUSTACEAN. When there is no predicate invention, the schema of CILP is the same as that of CRUSTACEAN; when there is predicate invention, the schema has one base clause and one recursive clause, which has an invented predicate whose program has only one base clause and one recursive clause. When there is predicate invention, DIALOGS-II invents predicates whose programs are

also be of the divide-and-conquer schemata of DIALOGS-II, which implies DIALOGS-II can make use of divide-and-conquer schemata that are more general than that of CILP.

The hypothesis language of the FORCE2 system [8] is two-clause linear and closed recursive *ij*-determinate logic programs. A clause is linear and closed recursive if the body of the clause has a single recursive atom that is closed, i.e. has no output variables. Thus, the schema is:

$$\begin{aligned} p(\dots) &\leftarrow q_1(\dots), \dots, q_m(\dots) \\ p(\dots) &\leftarrow r_1(\dots), \dots, r_n(\dots), p(\dots) \end{aligned}$$

where each q_k and r_k is an *ij*-determinate literal that is defined in the background knowledge, and the recursive atom $p(\dots)$ has no output variables. This schema is restricted with respect to the possible divide-and-conquer schemata of DIALOGS-II since it has only one base clause and only one recursive clause, where the recursive clause has only one recursive call. Moreover, the schema above is further restricted by *ij*-determinacy, where the divide-and-conquer schemata of DIALOGS-II have no such constraint.

In summary, there exist divide-and-conquer schemata that can be used by DIALOGS-II, which are superior to those of all other ILP systems known to us.

3.3 Comparison of DIALOGS-II with DIALOGS

DIALOGS-II enables the specifier to select a certain schema together with a strategy, whereas DIALOGS does not have such a concept of selection of a schema and a strategy, i.e. the concept of schema-guidedness; however, DIALOGS is schema-based (has a hard-wired divide-and-conquer schema together with a strategy), and it was thus the first step towards the schema-guidedness of DIALOGS-II, and, to the best of our knowledge, DIALOGS-II is the first in schema-guided synthesis in the field of ILP.

DIALOGS-II uses the open program approach (a first-order approach) in representing schemata, whereas DIALOGS uses a second-order approach in representing its divide-and-conquer schema. Using the open program approach simplifies the representation and manipulation of the schemata of the system.

DIALOGS-II handles the sparseness problem, thus enabling the system to induce programs, e.g. *factorial*, that were not inducible by DIALOGS.

Another difference between DIALOGS and DIALOGS-II is that the DIALOGS implementation did not make a difference between the semantics of the answers *false* and *stop-it* to the queries. Actually, *false* means that there does not exist any condition such that the goal in the query might hold, whereas *stop-it* means that the specifier wants to stop the query session. In the DIALOGS-II implementation, *false* and *stop-it* have their intended meanings.

DIALOGS-II Uses Clause $lg\theta$. DIALOGS uses term $lg\theta$ s in its MSG Method, whereas DIALOGS-II uses clause $lg\theta$ s in its Program Closing Method, since clause $lg\theta$ is a more powerful way of handling generality among clauses.

If we had used term $lg\theta$ instead of clause $lg\theta$ in the Program Closing Method, then the order of the atoms inside a clause would matter. For instance, if the two clauses whose $lg\theta$ is to be computed were

$$\text{sort}([A,B,C],[B,A,C]) \leftarrow C \geq A, A \geq B \quad (c1)$$

$$\text{sort}([D,E,F],[E,D,F]) \leftarrow F \geq D, D \geq E \quad (c2)$$

then the clause $lg\theta$ of these two clauses would be:

$$\text{sort}([A,B,C],[B,A,C]) \leftarrow C \geq A, D \geq E, F \geq G, A \geq B$$

After reducing (see Definition 2.1) this clause, we would obtain the resulting clause

$$\text{sort}([A,B,C],[B,A,C]) \leftarrow C \geq A, A \geq B$$

If we write these two clauses in the form of two terms, i.e.

$$\begin{aligned} &\text{if}(\text{sort}([A,B,C],[B,A,C]),\text{and}(C \geq A, A \geq B)) \\ &\text{if}(\text{sort}([D,E,F],[E,D,F]),\text{and}(F \geq D, D \geq E)) \end{aligned} \quad (t2)$$

and then take their term $lg\theta$, the resulting term would be

$$\text{if}(\text{sort}([A,B,C],[B,A,C]),\text{and}(C \geq A, A \geq B))$$

where this $lg\theta$ corresponds to the clause obtained after taking the (reduced) clause $lg\theta$ of the two clauses $c1$ and $c2$.

Now, suppose that we change the order of the literals in the body of the clause $c2$, e.g.

$$\text{sort}([A,B,C],[B,A,C]) \leftarrow C \geq A, A \geq B$$

$$\text{sort}([D,E,F],[E,D,F]) \leftarrow D \geq E, F \geq D$$

and compute their clause $lg\theta$, i.e.

$$\text{sort}([A,B,C],[B,A,C]) \leftarrow D \geq E, C \geq A, A \geq B, F \geq G$$

After reducing this clause, we obtain the same clause that was computed above when the order of the literals was not changed:

$$\text{sort}([A,B,C],[B,A,C]) \leftarrow C \geq A, A \geq B \quad (\text{c3})$$

However, if we make this order change for term t_2 , and then take the term $\text{lg}\theta$ of the resulting terms, then we obtain the following term

$$\text{if}(\text{sort}([A,B,C],[B,A,C]),\text{and}(D \geq E, F \geq G))$$

where this $\text{lg}\theta$ does not correspond to the $\text{lg}\theta$ for clauses, i.e. c_3 . As we can see, this term is different from the one where the order has not been changed. So, changing the order of the terms matters when term $\text{lg}\theta$ is used, though it should not matter. Because of that reason, DIALOGS-II uses clause $\text{lg}\theta$ instead of term $\text{lg}\theta$; in that way it also guarantees that there are no second-order $\text{lg}\theta$ s. For instance, suppose that the two clauses, i.e.

$$\begin{aligned} \text{delOdds}([A,B],[A,B]) &\leftarrow \text{even}(A), \text{even}(B) \\ \text{delOdds}([A],[A]) &\leftarrow \text{even}(A) \end{aligned}$$

are given and their clause $\text{lg}\theta$ is computed as

$$\text{delOdds}([A|B],[A|B]) \leftarrow \text{even}(A), \text{even}(C)$$

Note that there is no second order variable in the clause $\text{lg}\theta$ of these two clauses. However, if we write these two clauses in the form of two terms, i.e.

$$\begin{aligned} \text{if}(\text{delOdds}([A,B],[A,B]),\text{and}(\text{even}(A),\text{even}(B))) \\ \text{if}(\text{delOdds}([A],[A]),\text{even}(A)) \end{aligned}$$

and then take their term $\text{lg}\theta$, the resulting term would be

$$\text{if}(\text{delOdds}([A|B],[A|B]),V)$$

where the variable V is a second-order variable.

DIALOGS-II Eliminates Redundant Answers. Another new concept related to the queries of DIALOGS-II is “elimination of redundant answers”. Before discussing this concept, we introduce some terminology. We assume that conjunctions of literals can also be viewn as sets of literals.

Definition 3.1: A conjunction of literals C_1 θ -subsumes a conjunction of literals C_2 (denoted $C_1 \geq C_2$) iff there exists a substitution σ such that $C_2\sigma \subseteq C_1$.

For instance, let C_1 be $B=[C], C=A$ and C_2 be $B=[A]$. The conjunction C_1 θ -subsumes C_2 since there exists a substitution σ , which is $\{A/C\}$, such that $C_2\{A/C\} \subseteq C_1$.

Theorem 3: $(C_1 \geq C_2) \Rightarrow (C_1 \Rightarrow C_2)$

Proof 3: From C_1 , we can build a clause, namely $\neg C_1$. From C_2 , we can build a clause, namely $\neg C_2$. Now, note that $C_1 \Rightarrow C_2$ is equivalent to $\neg C_2 \Rightarrow \neg C_1$. So, to check for $C_1 \Rightarrow C_2$, one may approximate this (correctly but incompletely) by checking for $\neg C_2 \geq \neg C_1$ (according to Plotkin's definition, i.e. Definition 2.1, for clauses) (since $\neg C_1$ and $\neg C_2$ are clauses), i.e. by finding a substitution σ such that $\neg C_2\sigma \subseteq \neg C_1$, which is obviously equivalent to $C_2\sigma \subseteq C_1$.

When, to a query (i.e. atom) Q , the specifier gives a DNF answer $C_1 \vee C_2 \vee \dots \vee C_n$ ($n \geq 0$), then the system must eliminate those C_i for which there exists j such that $C_i \geq C_j$ ($i \neq j$) (i.e. eliminate those that are more general than some other one), and then only build the clauses $Q \leftarrow C_k$, where k is in the set of remaining indices.

What happens when the system does not eliminate redundant answers? We illustrate this point by means of a case that occurs during the synthesis of a program for `efface(E,L,R)`, where `efface(E,L,R)` holds iff list R is list L without the first (existing) occurrence of term E in L. Let the query and its answer be:

When does `efface(A,[B,A],C)` hold? $C=[A], B=A; C=[B]$.

The system would abduce the following shortcuts and evidence from this answer:

`compose_efface(B,[],C,A) ← C=[A], B=A`

`compose_efface(B,[],C,A) ← C=[B]`

`solve_efface(A,[B,A],C) ← C=[A], B=A`

`solve_efface(A,[B,A],C) ← C=[B]`

`efface(A,[B,A],C) ← C=[A], B=A` (s1)

`efface(A,[B,A],C) ← C=[B]` (s2)

Next, the system generates the query

When does `efface(A,[B,A,A],C)` hold?

where the answer to the query is:

$C=[A,A], A=B; C=[B,A], B \neq A$.

Using the answer, the system would abduce the following shortcuts and evidence:

$$\text{compose_efface}(B,[A],C,A) \leftarrow C=[A,A], A=B \quad (\text{c4})$$

$$\text{compose_efface}(B,[A],C,A) \leftarrow C=[B,A], B \neq A \quad (\text{c5})$$

$$\text{efface}(A,[B,A,A],C) \leftarrow C=[A,A], A=B$$

$$\text{efface}(A,[B,A,A],C) \leftarrow C=[B,A], A \neq B$$

Upon backtracking to shortcut `s2`, the system would also abduce the following evidence:

$$\text{compose_efface}(B,[A],C,A) \leftarrow C=[A,A], A=B \quad (\text{c6})$$

$$\text{compose_efface}(B,[A],C,A) \leftarrow C=[B,A], B \neq A \quad (\text{c7})$$

Upon backtracking, the following evidence for `solve_efface` would be abduced using the answer to the query:

$$\text{solve_efface}(A,[B,A,A],C) \leftarrow C=[A,A], A=B$$

$$\text{solve_efface}(A,[B,A,A],C) \leftarrow C=[B,A], A \neq B$$

Now, note that `c4` and `c6` are identical, as well as `c5` and `c7`. This redundancy in the evidence clauses is due to the redundancy in the answer to the query asked for `efface(A,[B,A],C)`. There are now two more `compose_efface` clauses than `solve_efface` clauses. This means that in the resulting set of evidence clauses that is passed to the Program Closing Method, there will be more `compose_efface` clauses than `solve_efface`, which makes the Program Closing Method fail, because the division algorithm of the Program Closing Method (see Algorithm 8) works under the assumption that there are less `compose_efface` clauses than `solve_efface` clauses. This is a correct assumption since there should always be more number of `solve_r` clauses than the number of `compose_r` clauses, if the evidence is correctly abduced. This is due to the existence of a decomposition operator in the recursive clause, which does not resolve for some values of the induction parameter, e.g. `[]` for lists, `0` for natural numbers, which in turn causes less evidence to be abduced for the open relation of the recursive clause than the one of the non-recursive clause.

Thus, the system must eliminate the answer $C=[A], B=A$ (which is more general than $C=[B]$) from $C=[A], B=A; C=[B]$. So, $C=[A], B=A$ is redundant and is eliminated from the answer, leaving only $C=[B]$ as the answer to the query, where this elimination prevents the redundancy in the evidence clauses, which in turn makes the system to abduce a usable set of evidence clauses.

Chapter 4

Conclusion

The inductive synthesis of recursive (logic) programs is a challenging and important sub-field of ILP. Challenging because recursive programs are particularly delicate mathematical objects that must be designed with utmost care. Important because recursive programs (for certain predicates) are sometimes the only way to complete the induction of a finite hypothesis (involving these predicates).

When it comes to programming applications, we believe that the ideal technique is interactive (in the sense of DIALOGS [13]) and non-incremental, has a clausal evidence language plus type, mode, and multiplicity information (like SYNAPSE [11], DIALOGS), can handle semantic manipulation relations, actually uses (structured) background knowledge and a syntactic bias, which are both problem-independent and intensional (like in SYNAPSE), is guided by (and not just based on) at least the powerful divide-and-conquer schema of SYNAPSE and DIALOGS (using the implementation approach of METAINDUCE [18]), discovers additional base case and recursive case examples (like CILP [19]), can perform both necessary and useful predicate invention (like SYNAPSE, DIALOGS), even from sparse abduced evidence (like CILP), actually discovers the recursive atoms, and makes a constructive usage of the negative evidence (through abduction, like the *Constructive Interpreter* [9] and SYNAPSE).

Thus, we aimed to design and implement a synthesizer that induces recursive logic programs, which is non-incremental, schema-guided, and interactive, and finally developed DIALOGS-II, which is based on the system DIALOGS [13].

DIALOGS-II is a schema-guided, interactive, and non-incremental synthesizer of recursive logic programs that takes the initiative and queries a (possibly naive) specifier for evidence in her/his conceptual language. DIALOGS-II only asks for the minimal knowledge a specifier *must* have in order to want a (logic) program, and it can be used by any learner (including itself) that detects, or merely conjectures, the necessity of invention of a new predicate. Moreover, due to its powerful codification of “recursion-theory” into schemata and schematic constraints, it needs very little evidence and is very fast.

The main difference between DIALOGS-II and its ancestor DIALOGS is as follows: DIALOGS-II enables the specifier to select a certain schema together with a strategy, whereas DIALOGS does not have such a concept of selection of a schema and its strategy, i.e. the concept of schema-guidedness; indeed, DIALOGS is schema-based (has a hard-wired divide-and-conquer schema together with a strategy). To the best of our knowledge, DIALOGS-II is the first schema-guided synthesizer.

Other differences are that DIALOGS-II uses the open program approach (a first-order approach) to representing schemas, whereas DIALOGS uses a second-order approach to representing its divide-and-conquer schema. Using the open program approach simplifies the representation and manipulation of the schemas of the system.

DIALOGS-II handles the sparseness problem, thus enabling the system to induce programs that were not inducible by DIALOGS, e.g. for **factorial**.

DIALOGS uses term $lg\theta s$ in its MSG Method, whereas DIALOGS-II uses clause $lg\theta s$ in its Program Closing Method, since clause $lg\theta$ is a more powerful way of handling generality among clauses.

DIALOGS-II can induce correct recursive logic programs from less evidence than other ILP systems, e.g. FOIL [24] and TRACY [3]. Moreover, the divide-and-conquer schemata that can be used by the system may be more general than the ones of some other important ILP systems, e.g. CILP [19], CRUSTACEAN [1] [2], and METAINDUCE [18].

DIALOGS-II can be further improved in several ways: a heuristic for the necessary predicate invention would conjecture when to do predicate invention, and finding more powerful admissibility criteria for the evidence of the open relations of the divide-and-conquer schema would increase the probability of synthesizing a correct program.

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Appendix A: README file for DIALOGS-II

After loading the file `dialogsII.pl`, start a new synthesis by typing “d2.” (without the quotes). The system is composed of the following programs:

- `phase0.pl`: asks for predicate declaration, schema and strategy, and executes the strategy in order to obtain an open program to be passed to the next phase.
- `phase1and2.pl`: abduces evidence, induces program clauses by the Program Closing Method and evaluates the result of the Program Closing Method to conjecture necessary predicate invention.
- `schemas.pl`: contains the currently available schemata and the strategies of the system.
- `cliques.pl`: finds (admissible) cliques of clauses.
- `clausmsg.pl`: computes the $\lg\theta$ of two clauses.
- `primitives.pl`: contains primitives used by the system.
- `grammar.pl`: contains Definite Clause Grammar for parsing predicate declarations.
- `utilities.pl`: contains procedures frequently used by the system
- `dedotify.pl`: dedotifies initial schemata of the system to convert them to open programs

Variable names start with an uppercase letter; predicate names, functor and constants start with a lowercase letter. Conjunction is expressed by a comma (`,`), disjunction by a semi-colon (`;`), negation by wrapping the atom with a prefix `neg/1` functor, truth by “true”, and falsity by “false” (without the quotes). The available primitives are: `=/2`, `\==/2`, `length/2`, `append/3`, `member/2`, `nat/1`, `list/1`, `add/3`, `mult/3`, `lt/2`, `gt/2`, `le/2`,

`ge/2`, `partition/4`, and `halves/3` (see file `primitives.pl`). Natural numbers should be typed in as Peano numbers, using `0` for zero and prefix functor `s/1` for successor.

Please note that during the determination of the predicate declaration, parameter roles and decomposition operator, answers should not be terminated by a full-stop (`.`). The default answer (always between curly braces) can be selected by simply hitting the RETURN/ENTER key. You can force backtracking to a previous question using the answer “back” (without the quotes). Note that parameters that can be any number of (e.g. passive parameters) are indicated as lists, using the Prolog notation; that means the absence of such parameters is indicated using the empty list (`[]`). For the schema language please refer to [5]. A new schema can be added to the system using that schema language. You also need to make sure that the parameter roles of the parameters of the programs that fit to the schema, modes of the open relations, and the positions of the parameters inside the atoms of the open relations are defined (see file `schemas.pl`). Available types are `atom`, `term`, `nat`, `int`, `list(atom)`, `list(term)`, `list(nat)`, and `list(int)`. The type language can be inferred by looking at file `grammar.pl` (see `non-terminal type/1`). Similarly for the predicate declaration language. You can express your boredom with the questions (or unwillingness or inability to answer them) by answering “stop_it” (without the quotes). You will find some sample syntheses in the remainder of the thesis.

Appendix B: Sample Syntheses

- `len(L,N)` iff natural number `N` is the length of the list `L`.

```
Predicate declaration? len(L:list(term),N:nat)
Schema? [dc,dg] dc
Strategy? [divide_and_conquer_strategy1]
divide_and_conquer _strategy1
Induction parameter? {L} L
Result parameter? {N} N
Decomposition Operator? {decompose(L,HL,TL)<--L=[HL|TL]}
decompose(L,HL,TL)<--L=[HL|TL]
```

When does `len([],A)` hold?

`A=0`.

When does `len([A],B)` hold?

`B=s(0)`.

When does `len([A,B],C)` hold?

`C=s(s(0))`.

When does `len([A,B,C],D)` hold?

`D=s(s(s(0)))`.

When does `len([A,B,C,D],E)` hold?

`stop_it`.

Result of the Program Closing Method:

Clauses for `compose_len`:

```
compose_len(A,B,s(B)) <--
```

Clauses for `solve_len`:

```
solve_len([],0) <--
```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] **no**

A possible program is:

```

len(A,B) <-- solve_len(A,B)
len(A,C) <-- decompose_len(A,D,E),len(E,F),compose_len(D,F,C)
decompose_len(G,H,I) <-- G=[H|I]
solve_len([],0) <--
compose_len(J,K,s(K)) <--
Do you want another logic program? {yes} yes

```

```

Decomposition Operator? {decompose(L,HL1,HL2,TL)<--L=[HL1,HL2|TL]}
decompose(L,HL1,HL2,TL)<--L=[HL1,HL2|TL]

```

Using shortcut(s) instead of querying...

When does len([A,B,C,D],E) hold?
E=s(s(s(0))).

When does len([A,B,C,D,E],F) hold?
stop_it.

Result of the Program Closing Method:

```

Clauses for compose_len:
compose_len(A,B,C,s(s(C))) <--

```

```

Clauses for solve_len:
solve_len([A],s(0)) <--
solve_len([],0) <--

```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] **no**

A possible program is:

```

len(A,B) <-- solve_len(A,B)
len(A,C) <-- decompose_len(A,D,E,F),len(F,G),compose_len(D,E,G,C)
decompose_len(H,I,J,K) <-- H=[I,J|K]
solve_len([],0) <--
solve_len([L],s(0)) <--
compose_len(M,N,P,s(s(P))) <--

```

Do you want another logic program? {yes} **no**

No (more) programs.

- **count(A,B,C)** iff natural number **C** is the number of elements that unify with the term **A** in list **B**.

Predicate declaration? `count(A:term,B:list(term),C:nat)`
 Schema? `[dc,dg] dc`
 Strategy? `[divide_and_conquer_strategy1]`
`divide_and_conquer _strategy1`
 Induction parameter? `{B} B`
 Result parameter? `{C} C`
 Passive parameter(s)? `{[A]} [A]`
 Decomposition Operator? `{decompose(B,HB,TB) <-- B=[HB|TB]}`
`decompose(B,HB,TB) <-- B=[HB|TB]`

When does `count(A,[],B)` hold?

`B=0.`

When does `count(A,[B],C)` hold?

`C=0,A==B;C=s(0),A=B.`

When does `count(A,[B,A],C)` hold?

`C=s(0),A==B;C=s(s(0)),A=B.`

When does `count(A,[B,A,A],C)` hold?

`C=s(s(0)),A==B;C=s(s(s(0))),A=B.`

When does `count(A,[B,A,A,A],C)` hold?

`stop_it.`

Result of the Program Closing Method:

Clauses for `compose_count`:

`compose_count(A,B,s(B),A) <--`
`compose_count(C,D,D,E) <-- E\==C`

Clauses for `solve_count`:

`solve_count(A,[],0) <--`

Please evaluate the Program Closing Method results: need for recursive synthesis? `[yes/no] no`

A possible program is:

`count(A,B,C) <-- solve_count(A,B,C)`
`count(A,D,E) <-- decompose_count(D,F,G),`
`count(A,G,H),compose_count(F,H,E,A)`
`decompose_count(I,J,K) <-- I=[J|K]`
`solve_count(L,[],0) <--`

```

compose_count(M,N,N,P) <-- P\==M
compose_count(Q,R,s(R),Q) <--
Do you want another logic program? {yes} no.
No (more) programs.

```

- **addlast(A,B,C)** iff list **C** is list **B** with the term **A** added in the end.

```

Predicate declaration? addlast(A:term,B:list(term),C:list(term))
Schema? [dc,dg] dc
Strategy? [divide_and_conquer_strategy1]
divide_and_conquer _strategy1
Induction parameter? {B} B
Result parameter? {C} C
Passive parameter(s)? {[A]} [A]
Decomposition Operator? {decompose(B,HB,TB)<--B=[HB|TB]}
decompose(B,HB,TB)<--B=[HB|TB]

```

When does `addlast(A,[],B)` hold?

`B=[A]`.

When does `addlast(A,[B],C)` hold?

`C=[B,A]`.

When does `addlast(A,[B,C],D)` hold?

`D=[B,C,A]`.

When does `addlast(A,[B,C,D],E)` hold?

`E=[B,C,D,A]`.

When does `addlast(A,[B,C,D,E],F)` hold?

`stop_it`.

Result of the Program Closing Method:

Clauses for `compose_addlast`:

```

compose_addlast(A,[B|C],[A,B|C],D) <--

```

Clauses for `solve_addlast`:

```

solve_addlast(A,[B],[B|A]) <--

```

```

solve_addlast(C,[],[C]) <--

```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] **no**

A possible program is:

```
addlast(A,B,C) <-- solve_addlast(A,B,C)
addlast(A,D,E) <-- decompose_addlast(D,F,G),
addlast(A,G,H),compose_addlast(F,H,E,A)
decompose_addlast(I,J,K) <-- I=[J|K]
solve_addlast(L,[],[L]) <--
solve_addlast(M,[N],[N|M]) <--
compose_addlast(P,[Q|R],[P,Q|R],S) <--
```

Do you want another logic program? {yes} no.

No (more) programs.

Do you want another synthesis with a different strategy? {yes} yes

There is no other strategy for schema dc!

- multiply(A,B,C) iff natural number C is the product of natural numbers A and B.

Predicate declaration? multiply(A:nat,B:nat,C:nat)

Schema? [dc,dg] dc

Strategy? [divide_and_conquer_strategy1]

divide_and_conquer _strategy1

Induction parameter? {A} A

Result parameter? {B} B

Passive parameter(s)? {[C]} [C]

Decomposition Operator? {decompose(A,HA,TA) <-- A=s(TA), HA=A}

decompose(A,HA,TA) <-- A=s(TA), HA=A

When does multiply(0,A,B) hold?

A=0.

When does multiply(s(0),A,B) hold?

A=B+0.

When does multiply(s(s(0)),A,B) hold?

A=B+(B+0).

When does multiply(s(s(s(0))),A,B) hold?

A=B+(B+(B+0)).

When does multiply(s(s(s(s(0))))),A,B) hold?

stop_it.

Result of the Program Closing Method:

Clauses for compose_multiply:

```
compose_multiply(s(A),B,C+B,C) <--
```

Clauses for solve_multiply:

```
solve_multiply(0,0,A) <--
```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] **no**

A possible program is:

```
multiply(A,B,C) <-- solve_multiply(A,B,C)
```

```
multiply(A,D,E) <-- decompose_multiply(A,F,G),
```

```
multiply(G,H,E),compose_multiply(F,H,D,E)
```

```
decompose_multiply(I,J,K) <-- I=s(K),J=I
```

```
solve_multiply(0,0,L) <--
```

```
compose_multiply(s(M),N,P+N,P)<--
```

Do you want another logic program? {yes} **no**.

No (more) programs.

- **compress(L,R)** iff list R is the compressed form of list L.

e.g. `compress([a,a,b,c,c,c,d],[a,s(s(0)),b,s(0),c,s(s(s(0))),d,s(0)])`

Predicate declaration? `compress(L:list(atom),R:list(atom))`

Schema? [dc,dg] dc

Strategy? [divide_and_conquer_strategy1]

`divide_and_conquer_strategy1`

Induction parameter? {L} L

Result parameter? {R} R

Decomposition Operator? {`decompose(L,HL,TL)<--L=[HL|TL]`}

`decompose(L,HL,TL)<--L=[HL|TL]`

When does `compress([],A)` hold?

`A=[]`.

When does `compress([A],B)` hold?

`B=[A,s(0)]`.

When does `compress([A,B],C)` hold?

`C=[A,s(s(0))],eq(A,B);C=[A,s(0),B,s(0)],diff(A,B)`.

When does `compress([A,B,C],D)` hold, assuming `eq(B,C)`?

`D=[A,s(s(s(0)))]],eq(A,B);D=[A,s(0),B,s(s(0))],diff(A,B)`.

When does `compress([A,B,C],D)` hold, assuming `diff(B,C)`?

`D=[A,s(s(0)),C,s(0)],eq(A,B);D=[A,s(0),B,s(0),C,s(0)],diff(A,B)`.

When does `compress([A,B,C,D],E)` hold, assuming `eq(B,C),eq(C,D)`?
`stop_it.`

Result of the Program Closing Method:

Clauses for `compose_compress`:

```
compose_compress(A,[B,s(C)|D],[A,s(0),B,s(C)|D]) <-- diff(A,B)
compose_compress(E,[F,s(G)|H],[E,s(s(G))|H]) <-- eq(E,F)
```

Clauses for `solve_compress`:

```
solve_compress([A],[A,s(0)]) <--
solve_compress([],[]) <--
```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] `no`

A possible program is:

```
compress(A,B) <-- solve_compress(A,B)
compress(A,C) <-- decompose_compress(A,D,E),
compress(E,F),compose_compress(D,F,C)
decompose_compress(G,H,I) <-- G=[H|I]
solve_compress([],[]) <--
solve_compress([J],[J,s(0)]) <--
compose_compress(K,[L,s(M)|N],[K,s(s(M))|N]) <-- eq(K,L)
compose_compress(P,[R,s(Q)|S],[P,s(0),R,s(Q)|S]) <-- diff(P,R)
```

- `s(L,S)` iff list `S` is (ascendingly) sorted version of list `L`.

Predicate declaration? `s(L:list(int),S:list(int))`

Strategy? `[divide_and_conquer_strategy1]`

`divide_and_conquer_strategy1`

Induction parameter? `{L} L`

Result parameter? `{S} S`

Decomposition Operator? `{decompose(L,HL,TL)<--L=[HL|TL]}`

`decompose(L,HL,TL)<--L=[HL|TL]`

When does `s([],A)` hold?

`A=[].`

When does `s([A],B)` hold?

B=[A].

When does $s([A,B],C)$ hold?

C=[A,B],le(A,B);C=[B,A],gt(A,B).

When does $s([A,B,C],D)$ hold, assuming $le(B,C)$?

D=[A,B,C],le(A,B);D=[B,A,C],gt(A,B),le(A,C);D=[B,C,A],gt(A,B),gt(A,C).

When does $s([A,B,C,D],E)$ hold, assuming $le(B,C),le(C,D)$?

stop_it.

Result of the Program Closing Method:

Clauses for compose_s:

```
compose_s(A,[B|C],[A,B|C]) <-- le(A,B)
```

Clauses for solve_s:

```
solve_s(A,A) <--
```

```
solve_s([B,C,D],[C,D,B]) <-- gt(B,C),gt(B,D),le(C,D)
```

```
solve_s([E,F],[F,E]) <-- gt(E,F)
```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] **yes**

Need for recursive synthesis detected!

Calling DIALOGS-II with the predicate declaration

```
compose_s(HL:int,TS:list(int),S:list(int))
```

Induction parameter? {[TS]} [TS]

Result parameter? {S} S

Passive parameter(s)? {[HL]} [HL]

Decomposition Operator? {decompose(TS,HTS,TTS)<--TS=[HTS|TTS]}

```
decompose(TS,HTS,TTS)<--TS=[HTS|TTS]
```

Current program:

```
s(A,B) <-- solve_s(A,B)
```

```
s(A,C) <-- decompose_s(A,D,E),s(E,F),compose_s(D,F,C)
```

```
decompose_s(G,H,I) <-- G=[H|I]
```

```
solve_s([],[]) <--
```

```
compose_s(J,K,L) <-- solve_compose_s(J,K,L)
```

```
compose_s(J,M,N) <-- decompose_compose_s(M,P,Q),
```

```
compose_s(J,Q,R),compose_compose_s(P,R,N,J)
```

```
decompose_compose_s(S,T,U) <-- S=[T|U]
```

When does $s([A,B,C],D)$ hold, assuming $le(B,C),le(A,C)$?

stop_it.

Result of the Program Closing Method:

compose_compose_s clauses:

(none)

solve_compose_s clauses:

solve_compose_s(A,[B],[B,A]) <-- gt(A,B)

solve_compose_s(C,[D],[C,D]) <-- le(C,D)

solve_compose_s(E,[],[E]) <--

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] **no**

A possible program is:

s(A,B) <-- solve_s(A,B)

s(A,C) <-- decompose_s(A,D,E),s(E,F),compose_s(D,F,C)

decompose_s(G,H,I) <-- G=[H|I]

solve_s([],[]) <--

compose_s(J,K,L) <-- solve_compose_s(J,K,L)

compose_s(J,M,N) <-- decompose_compose_s(M,P,Q),

compose_s(J,Q,R),compose_compose_s(P,R,N,J)

decompose_compose_s(S,T,U) <-- S=[T|U]

solve_compose_s(V,[],[V]) <--

solve_compose_s(W,[X],[W,X]) <-- le(W,X)

solve_compose_s(Y,[Z],[Z,Y]) <-- gt(Y,Z)

Do you want another logic program? {yes} **no**

- **reverse(A,B,C)** iff list **B** is the concatenation of reverse of list **A** and the list **C** itself.

Predicate declaration? **reverse(A:list(term),R:list(term),L:list(term))**

Schema? [dc,dg] **dg**

Strategy? [descend_gen_strategy1]

descend_gen_strategy1

Induction parameter? {A} A

Result parameter? {R} R

Passive parameter(s)? {[L]} []

Accumulation parameter(s)? {[L]} [L]

Decomposition Operator? {decompose(A,HA,TA) <--A=[HA|TA]}
decompose(A,HA,TA) <--A=[HA|TA]

When does reverse([],A,B) hold?

A=B.

When does reverse([A],B,C) hold?

B=[A|C].

When does reverse([A,B],[B|C],D) hold?

[B|C]=[B,A|D].

When does reverse([A,B,C],[C,B|D],E) hold?

[C,B|D]=[C,B,A|E].

When does reverse([A,B,C,D],[D,C,B|E],F) hold?

[D,C,B|E]=[D,C,B,A|F].

When does reverse([A,B,C,D,E],[E,D,C,B|F],G) hold?

stop_it.

Entering the Program Closing Method with the following evidence

solveAccu_reverse evidence:

```
solveAccu_reverse([],A,A) <--  
solveAccu_reverse([B],[B|C],C) <--  
solveAccu_reverse([D,E],[E,D|F],F) <--  
solveAccu_reverse([G,H,I],[I,H,G|J],J) <--  
solveAccu_reverse([K,L,M,N],[N,M,L,K|P],P) <--
```

extendAccu_reverse evidence:

```
extendAccu_reverse(A,B,[A|B]) <--  
extendAccu_reverse(C,D,[C|D]) <--  
extendAccu_reverse(E,F,[E|F]) <--  
extendAccu_reverse(G,H,[G|H]) <--
```

Result of the Program Closing Method:

Clauses for extendAccu_reverse:

```
extendAccu_reverse(A,B,[A|B]) <--
```

Clauses for solveAccu_reverse:

```
solveAccu_reverse([],A,A) <--
```

Please evaluate the Program Closing Method results: need for recursive synthesis? [yes/no] no

A possible program is:

```
reverse(A,B,C) <-- solveAccu_reverse(A,B,C)
reverse(A,D,E) <-- decompose_reverse(A,F,G),
                reverse(G,D,H),extendAccu_reverse(F,E,H)
```

```
decompose_reverse(I,J,K) <-- I=[J|K]
```

```
solveAccu_reverse([],L,L)<--
```

```
extendAccu_reverse(M,N,[M|N])<--
```

Do you want another logic program? {yes}

no