

Regular Language Type Inference with Term Rewriting – ICFP 2020

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joined work with Thomas Genet and Thomas Jensen

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Introduction

Fully Automatic Verification of Higher-Order Functional Programs processing Algebraic Datatypes

For instance, we want to **automatically** solve this kind of safety problems:

```
# let rec sort cmp = function
  | [] -> []
  | x::l -> insert cmp x (sort cmp l) ;;
val insert : ('a -> 'a -> bool) -> 'a -> 'a list -> 'a list
val sort : ('a -> 'a -> bool) -> 'a list -> 'a list
# let rec sorted cmp = function
  | _ -> true
  | x::y::l -> cmp x y && sorted (y::l) ;;
val sorted : ('a -> 'a -> bool) -> 'a list -> bool
# forall l. sorted cmp (sort cmp l) ;;
```

Our goal is to answer the last query.

We do not deal with polymorphism (yet)

In this example instead, we define a very simple type equipped with a comparison function:

```
# type ab = A | B ;;  
type ab = A | B  
  
# let rec cmp x y = match x y with  
  | B, A -> false  
  | _, _ -> true  
val cmp : ab -> ab -> bool
```

Our polymorphic signatures become:

```
val insert : (ab -> ab -> bool) -> ab -> ab list -> ab list  
val sort : (ab -> ab -> bool) -> ab list -> ab list  
val sorted : (ab -> ab -> bool) -> ab list -> bool
```

This is a type checking problem

Our safety problem becomes:

```
val insert : (ab -> ab -> bool) -> ab -> ab list -> ab list
val sort : (ab -> ab -> bool) -> ab list -> ab list
val sorted : (ab -> ab -> bool) -> ab list -> bool

# forall l. sorted cmp (sort cmp l) ;;
```

This is equivalent to the following refinement-type checking problem:

```
# function l -> ( sorted cmp (sort cmp l) : { true } ) ;;
```

Solving by refining the signature of the program

We need to infer precise types for `sorted` and `sort` to type the pattern:

```
( sorted cmp (sort cmp (l : ab list)) : { true } ) ;;
```

A possible solution is:

```
insert : { cmp } -> ab -> { sorted lists } -> { sorted lists }  
sort : { cmp } -> ab list -> { sorted list }  
sorted : { cmp } -> { sorted lists } -> { true }
```

This is a **regular abstraction** of the program:

- all the (refinement) types can be described with **regular tree languages**
- it captures all the **invariants** necessary to prove our property

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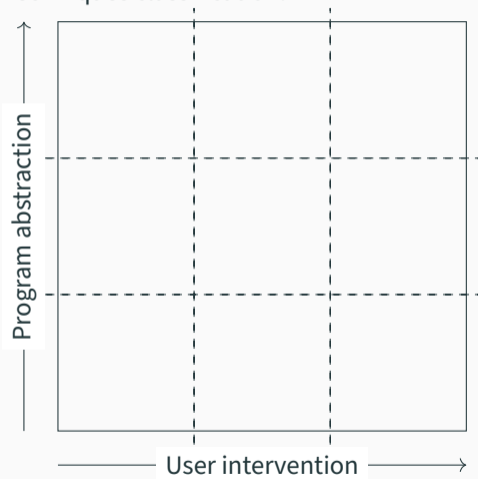
This is a **regular abstraction** of the program:

- all the (refinement) types can be described with **regular tree languages**
- it captures all the **invariants** necessary to prove our property

Our goal is to automatically infer such regular abstractions.

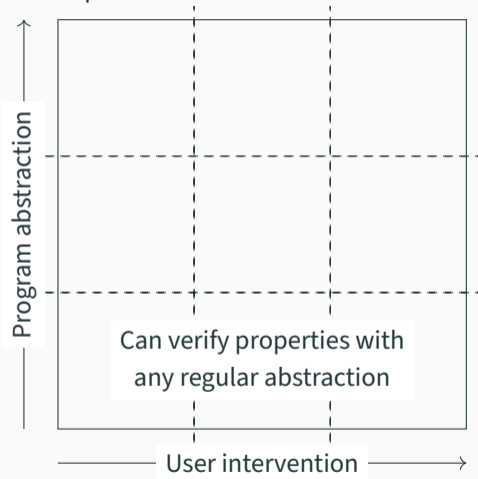
Current verification techniques vs. regular problems

Techniques classification:



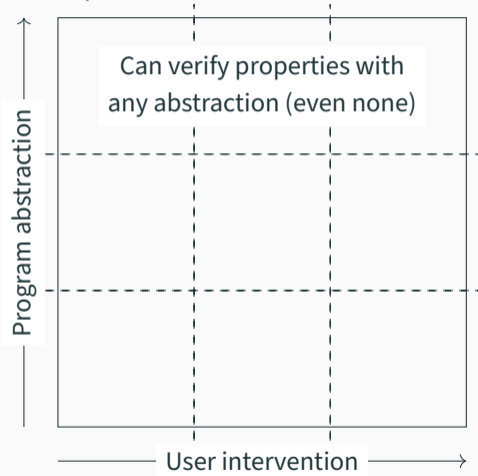
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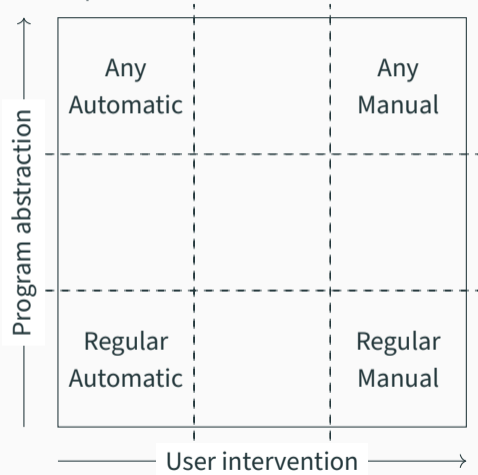
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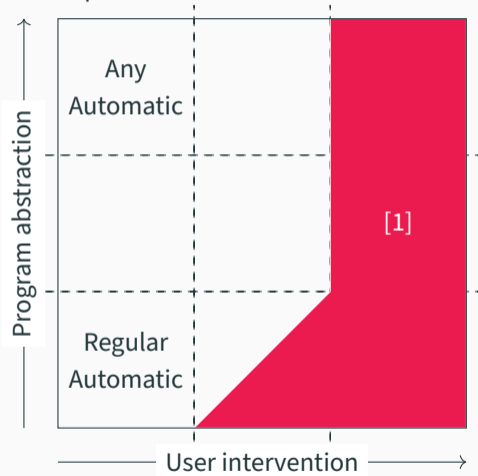
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Current verification techniques vs. regular problems

Techniques classification:



1. Proof assistants (Coq, Isabelle HOL, etc.):

- Large range of properties
- Requires a lot of manual work



Inria. "The coq proof assistant", 2016



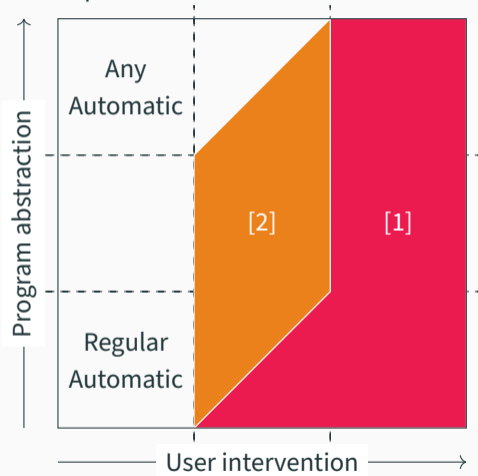
Tobias Nipkow, Lawrence C. Paulson, and Markus Wenzel. "Isabelle/HOL – A Proof Assistant for Higher-Order Logic", 2002.



...and more.

Current verification techniques vs. regular problems

Techniques classification:



1. Proof assistants (Coq, Isabelle HOL, etc.):
 - Large range of properties
 - Requires a lot of manual work
2. Refinement types (Liquid Types, F^*):
 - Only require some annotations
 - Reduced range of properties



Patrick Maxim Rondon, Ming Kawaguchi, and Ranjit Jhala. "Liquid types", 2018



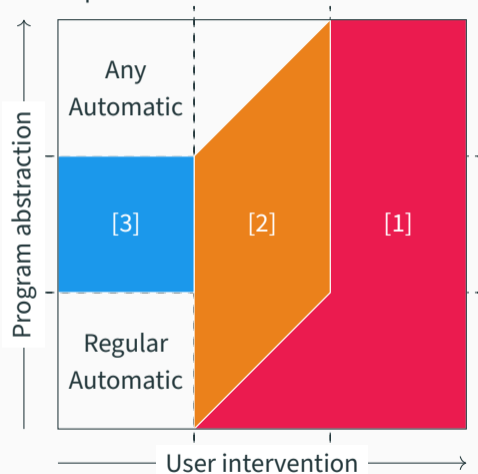
Microsoft Research and Inria. " F^* ", 2013.



...and more.

Current verification techniques vs. regular problems

Techniques classification:



3. HORS, Predicate abstraction, CHC:

- No annotations needed
- Focus on relational properties over numerical datatypes



C.-H. Luke Ong and Steven J. Ramsay. "Verifying higher-order functional programs with pattern-matching algebraic data types", 2006



Naoki Kobayashi. "Types and higher-order recursion schemes for verification of higher-order programs", 2009



Naoki Kobayashi, Ryosuke Sato, and Hiroshi Unno. "Predicate abstraction and CEGAR for higher-order model checking", 2010



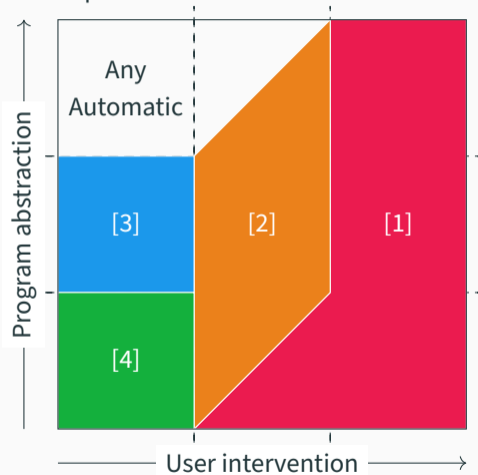
Adrien Champion, Naoki Kobayashi, Ryosuke Sato. "Holce: An ICE-Based Non-linear Horn Clause Solver", 2018



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4. Regular language-based techniques:

- Uncomplete (Timbuk)
- Non-scalable (Matsumoto et al. 2015)



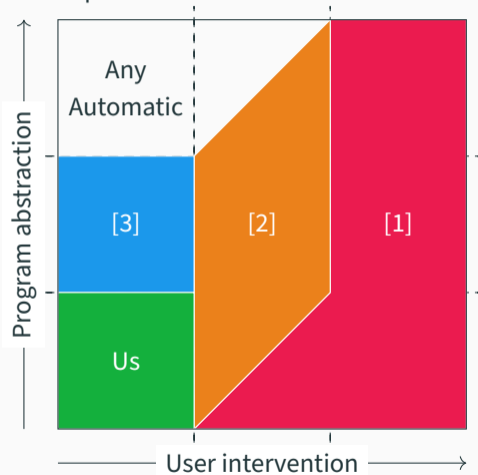
Yuma Matsumoto, Naoki Kobayashi, and Hiroshi Unno. "Automata-based abstraction for automated verification of higher-order tree-processing programs", 2015



Thomas Genet, Timothée Haudebourg, and Thomas Jensen. "Verifying higher-order functions with tree automata", 2018

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Preliminaries

Terms, Rewriting Systems & Regular Languages

We use **terms** to represent the states and values of a program.

Definition (Term)

Let Σ be **ranked-alphabet**. A term defined over Σ is written:

$$f \in \Sigma \quad \text{sub-term} \quad f(t_1, \dots, t_n) \quad \text{arity}(f) = n$$

We write $ar(f)$ the **arity** of the symbol f .

Definition (Pattern)

Let \mathcal{X} be a set of **variables**. We name **pattern** a term defined over $\Sigma \cup \mathcal{X}$.



Baader, F., Nipkow, T. "Term Rewriting and All That", 1998



Hubert Comon et al. "Tree Automata Techniques and Applications", 2007

From OCaml to terms

Here are some examples of terms encoding OCaml expressions:

OCaml expression	Term
<code>true</code>	<i>true</i>
<code>false</code>	<i>false</i>
<code>0</code>	<i>0</i>
<code>1</code>	<i>s(0)</i>
<code>2</code>	<i>s(s(0))</i>
<code>[]</code>	<i>nil</i>
<code>a :: []</code>	<i>cons(a, nil)</i>
<code>a :: b :: []</code>	<i>cons(a, cons(b, nil))</i>
<code>f a b</code>	<i>f(a, b)</i>
<code>if a then b else c</code>	<i>ite(a, b, c)</i>
<code>let g = function x -> x + 1 in body</code>	<i>body</i>

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<code>1</code>	<i>s(0)</i>
<code>2</code>	<i>s(s(0))</i>
<code>[]</code>	<i>nil</i>
<code>a :: []</code>	<i>cons(a, nil)</i>
<code>a :: b :: []</code>	<i>cons(a, cons(b, nil))</i>
<code>f a b</code>	<i>@(@(f, a), b)</i>
<code>if a then b else c</code>	<i>ite(a, b, c)</i>
<code>let g = function x -> x + 1 in body</code>	<i>body</i>



John Reynolds. "Automatic Computation of Data Set Definitions", 1969



Richard Kennaway, Jan Willem Klop, M. Ronan Sleep, and Fer-Jan de Vries. "Comparing Curried and Uncurried Rewriting", 1996

Term Rewriting Systems

We use **term rewriting systems** (TRSs) to represent the program itself.

Definition (Term Rewriting System)

A TRS \mathcal{R} is a set of rewriting rules of the form

$$l \rightarrow r$$

where

- l and r are **patterns**
- $Var(r) \subseteq Var(l)$
- r is **linear** (for the purpose of the presentation)



TeReSe. "Term Rewriting Systems", 2011



Franz Baader, Tobias Nipkow. "Term Rewriting and All That", 1999

We write " $s \rightarrow_{\mathcal{R}} t$ " when there exists a rule $l \rightarrow r$ in \mathcal{R} , a position p and a substitution σ such that $s|_p = l\sigma \wedge t = s[r\sigma]_p$.

We write $\rightarrow_{\mathcal{R}}^*$ the reflexive transitive closure of $\rightarrow_{\mathcal{R}}$.

From OCaml to TRSs

Here is an example of TRS encoding our previous sorted function:

```
let rec sorted cmp = function
  | _ -> true
  | x::y::l -> cmp x y && sorted (y::l)
```

This can be translated into the following TRS (variables are underlined):

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This can be translated into the following TRS (variables are underlined):

$$@(@(\underline{sorted}, \underline{cmp}), \underline{nil}) \rightarrow \underline{true}$$
$$@(@(\underline{sorted}, \underline{cmp}), \underline{cons}(\underline{x}, \underline{nil})) \rightarrow \underline{true}$$
$$@(@(\underline{sorted}, \underline{cmp}), \underline{cons}(\underline{x}, \underline{cons}(\underline{y}, \underline{l}))) \rightarrow \underline{and}(@(@(\underline{cmp}, \underline{x}), \underline{y}), @(@(\underline{sorted}, \underline{cmp}), \underline{cons}(\underline{y}, \underline{l})))$$

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This can be translated into the following TRS (variables are underlined):

$$\text{sorted } \underline{cmp} \text{ nil} \rightarrow \text{true}$$
$$\text{sorted } \underline{cmp} \text{ cons}(\underline{x}, \text{nil}) \rightarrow \text{true}$$
$$\text{sorted } \underline{cmp} \text{ cons}(\underline{x}, \text{cons}(\underline{y}, \underline{l})) \rightarrow \text{and}(\underline{cmp} \underline{x} \underline{y}, \text{sorted } \underline{cmp} \text{ cons}(\underline{y}, \underline{l}))$$

We write “ $f a b$ ” instead of “ $@(@ (f, a), b)$ ”...

Regular tree languages

A set of term is called a **tree language**.

It is **regular** iff it can be recognized by a **tree automaton**.

Definition (Bottom-Up Tree Automaton)

A tree automaton \mathcal{A} is a quadruple $\langle \Sigma, Q, Q_f, \Delta \rangle$ where

- Σ is a ranked alphabet
- Q a set of **states**
- $Q_f \subseteq Q$ a set of **final states**
- Δ a rewriting system of **transitions** of the form

$$f(q_1, \dots, q_n) \rightarrow q \quad \text{or} \quad q' \rightarrow q \quad (\epsilon\text{-transition})$$

A term t is **recognized** by \mathcal{A} if there exists a final state $q \in Q_f$ such that $t \rightarrow_{\Delta}^* q$.

We write $\mathcal{L}(\mathcal{A})$ the language recognized by \mathcal{A} .



Hubert Comon et al. "Tree Automata Techniques and Applications", 2007

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Regular Language Types

Regular languages can be used to represent OCaml types (and more).

For instance, the following OCaml type definition:

```
type ab = A | B ;;
```

...can be translated into the tree automaton

$$\langle \Sigma = \{ A : 0, B : 0 \}, Q = \{ ab \}, Q_f = Q, \Delta \rangle$$

where Δ is defined by:

$$A \rightarrow ab$$

$$B \rightarrow ab$$

Regular Language Types

Regular languages can be used to represent OCaml types (and more).

For instance, the following OCaml type definition:

```
type nat = 0 | S of nat ;;
```

...can be translated into the tree automaton

$$\langle \Sigma = \{ 0 : 0, S : 1 \}, Q = \{ nat \}, Q_f = Q, \Delta \rangle$$

where Δ is defined by:

$$0 \rightarrow nat$$

$$S(nat) \rightarrow nat$$

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For instance, the following OCaml type definition:

```
type nlist = Nil | Cons of nat * nlist ;;
```

...can be translated into the tree automaton

$$\langle \Sigma = \{ 0 : 0, S : 1, Nil : 0, Cons : 2 \}, Q = \{ nat, nlist \}, Q_f = \{ nlist \}, \Delta \rangle$$

where Δ is defined by:

$$\begin{array}{ll} 0 \rightarrow nat & Nil \rightarrow nlist \\ S(nat) \rightarrow nat & Cons(nat, nlist) \rightarrow nlist \end{array}$$

Regular Language Types

Regular languages can be used to represent OCaml types (and more).

For instance, the following OCaml type definition:

```
???
```

...can be translated into the tree automaton

$$\langle \Sigma = \{ 0 : 0, S : 1 \}, Q = \{ \text{even}, \text{odd} \}, Q_f = \{ \text{even} \}, \Delta \rangle$$

where Δ is defined by:

$$0 \rightarrow \text{even}$$

$$S(\text{even}) \rightarrow \text{odd}$$

$$S(\text{odd}) \rightarrow \text{even}$$

Regular Language Types

Regular languages can be used to represent OCaml types (and more).

For instance, the following OCaml type definition:

```
???
```

...can be translated into the tree automaton

$$\langle \Sigma = \{ 0 : 0, S : 1 \}, Q = \{ even, odd \}, Q_f = \{ even \}, \Delta \rangle$$

where Δ is defined by:

$$0 \rightarrow even$$

$$S(even) \rightarrow odd$$

$$S(odd) \rightarrow even$$

This is a **regular language type**

Regular Verification

Regular verification problem (OCaml perspective)

Input problem

```
insert : (ab -> ab -> bool) -> ab -> ab list -> ab list  
sort   : (ab -> ab -> bool) -> ab list -> ab list  
sorted : (ab -> ab -> bool) -> ab list -> bool
```

```
sorted cmp (sort cmp (l : ab list)) : { true }
```

Output

```
insert : { cmp } -> ab -> { sorted lists } -> { sorted lists }  
sort   : { cmp } -> ab list -> { sorted list }  
sorted : { cmp } -> { sorted lists } -> { true }
```

...with the definitions of { cmp }, { true } and { sorted lists }.

Regular verification problem (example)

Input problem

TRS \mathcal{R} (the program):

$insert\ \underline{cmp}\ x\ \underline{l} \rightarrow \dots$

$sort\ \underline{cmp}\ \underline{l} \rightarrow \dots$

$sorted\ \underline{cmp}\ \underline{l} \rightarrow \dots$

Property:

$\forall \underline{l}. sorted\ cmp\ (sort\ cmp\ \underline{l}) \rightarrow_{\mathcal{R}}^* true$

Output

An TRS abstraction $\mathcal{R}^\#$ of \mathcal{R} :

$insert\ cmp^\#\ ab^\#\ sorted_lists^\# \rightarrow sorted_lists^\#$

$sort\ cmp^\#\ ab_list^\# \rightarrow sorted_lists^\#$

$sorted\ cmp^\#\ sorted_lists^\# \rightarrow true^\#$

...with the definition of $cmp^\#, true^\#, sorted_lists^\#$ with an automaton $\Delta^\#$ such that:

$sorted\ cmp^\#\ (sort\ cmp^\#\ ab_list^\#) \rightarrow_{\mathcal{R}^\# \cup \Delta^\#}^* true^\#$

Regular verification problem (general)

Input problem

- A TRS \mathcal{R} defined over Σ
- A pattern p with π mapping each variable of p to a language
- A target output language O

$$\forall \sigma, \sigma(x) \in \pi(x). \quad p\sigma \rightarrow_{\mathcal{R}}^* v \Rightarrow v \in O \quad \text{with } v \in IRR(\mathcal{R})$$

Output

- An abstraction $\mathcal{R}^\#$ of \mathcal{R}
- An abstract domain (tree automaton) $\Lambda = \langle \Sigma^\#, \Delta^\# \rangle$ such that
 - $O^\# \in \Sigma^\#$ with $O = \mathcal{L}(\Lambda, O^\#)$
 - for all $x \in \text{Var}(p)$, $\sigma(x)^\# \in \Sigma^\#$ with $\sigma(x) = \mathcal{L}(\Lambda, O^\#)$
 - $p\sigma^\# \rightarrow_{\mathcal{R}^\# \cup \Delta^\#}^* O^\#$ where $\sigma^\#(x) = \sigma(x)^\#$.

Abstraction

Definition (TRS Abstraction)

An abstraction $\mathcal{R}^\#$ of \mathcal{R} in $\Lambda = \langle \Sigma^\#, \Delta^\# \rangle$:

- rules of the form “ $f(a_1^\#, \dots, a_n^\#) \rightarrow a^\#$ ” with $a_i^\#, a^\# \in \Sigma^\#$
- defines a simulation:

$$\begin{array}{ccc} s & \longrightarrow & t \\ \left. \vphantom{s} \right\} & & \left. \vphantom{t} \right\} \\ s^\# & \dashrightarrow & t^\# \end{array}$$

$\mathcal{R}^\#$ can be seen as a set of **type signatures** for the program's functions.

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- rules of the form “ $f(a_1^\#, \dots, a_n^\#) \rightarrow a^\#$ ” with $a_i^\#, a^\# \in \Sigma^\#$
- defines a simulation:

$$\begin{array}{ccc} s & \xrightarrow{\mathcal{R}} & t \\ \mathcal{R}^\# \cup \Delta^\# \downarrow * & & * \downarrow \mathcal{R}^\# \cup \Delta^\# \\ s^\# & \equiv & t^\# \end{array}$$

Note: $\mathcal{R}^\# \cup \Delta^\#$ is not necessarily deterministic (multiple abstractions for one term).

$\mathcal{R}^\#$ can be seen as a set of **type signatures** for the program's functions.

Abstraction inference using a backward typing algorithm

Algorithm

1. **Extract a template signature for the topmost symbol**
 - The output type is always known
 - Constant inputs are known
2. **Analyse function**
 - Using this algorithm recursively or;
 - Fixpoint learning procedure.
3. **Check**
4. **Repeat with the sub-terms**

$\forall \underline{l}. \text{sorted cmp} (\text{sort cmp } \underline{l}) \rightarrow_{\mathcal{R}}^* \text{true}$

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sorted cmp (sort (l : ablist[#])) : true[#]

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$sorted\ cmp\ (sort\ (\underline{l} : ablist^\#) : sorted^\#) : true^\#$

$sorted\ cmp^\# \mathbf{sorted}^\# \rightarrow true^\# \in \mathcal{R}^\#$

$sort\ (\underline{l} : ablist^\#) : sorted^\#$

$sort\ ablist^\# \rightarrow sorted^\# \in \mathcal{R}^\#$

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Example

- $ite(true, \underline{x}, \underline{y}) \rightarrow \underline{x}$
- $ite(false, \underline{x}, \underline{y}) \rightarrow \underline{y}$
- ite
- $\{A\}$

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Example

Non-recursive case: computed using the **previous backward typing algorithm**.

Function analysis

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Output

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Example

$$ite(true, \underline{x}, \underline{y}) \rightarrow \underline{x} : A^\#$$

Non-recursive case: computed using the **previous backward typing algorithm**.

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Example

- $ite(true, \underline{x}, \underline{y}) \rightarrow \underline{x}$
- $ite(false, \underline{x}, \underline{y}) \rightarrow \underline{y}$
- ite
- $\{A\}$

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Example

$$ite(true, \underline{x} : A^\#, \underline{y}) \rightarrow \underline{x} : A^\#$$

Non-recursive case: computed using the **previous backward typing algorithm**.

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Example

- $ite(true, \underline{x}, \underline{y}) \rightarrow \underline{x}$
- $ite(false, \underline{x}, \underline{y}) \rightarrow \underline{y}$
- ite
- $\{A\}$

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Example

$$ite(true, \underline{x} : A^\#, \underline{y} : \top^\#) \rightarrow \underline{x} : A^\#$$

Non-recursive case: computed using the **previous backward typing algorithm**.

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Example

- $ite(true, \underline{x}, \underline{y}) \rightarrow \underline{x}$
- $ite(false, \underline{x}, \underline{y}) \rightarrow \underline{y}$
- ite
- $\{A\}$

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Example

$$ite(true, A^\#, T^\#) \rightarrow A^\#$$

Non-recursive case: computed using the **previous backward typing algorithm**.

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Example

- $ite(true, \underline{x}, \underline{y}) \rightarrow \underline{x}$
- $ite(false, \underline{x}, \underline{y}) \rightarrow \underline{y}$
- ite
- $\{A\}$

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Example

$$ite(true, A^\#, \top^\#) \rightarrow A^\#$$
$$ite(false^\#, \top^\#, A^\#) \rightarrow A^\#$$

Non-recursive case: computed using the **previous backward typing algorithm**.

Function analysis

Input

- The TRS \mathcal{R}
- The function's symbol $f \in \Sigma$
- The target output language O

Example

- $even(0) \rightarrow true$ $even(s(0)) \rightarrow false$
 $even(s(s(\underline{x}))) \rightarrow even(\underline{x})$
- $even$
- $\{true\}$

Output

- A set of rules of the form:

$$f(l_1^\#, \dots, l_n^\#) \rightarrow O^\#$$

...a **complete** abstraction of the function.

Example

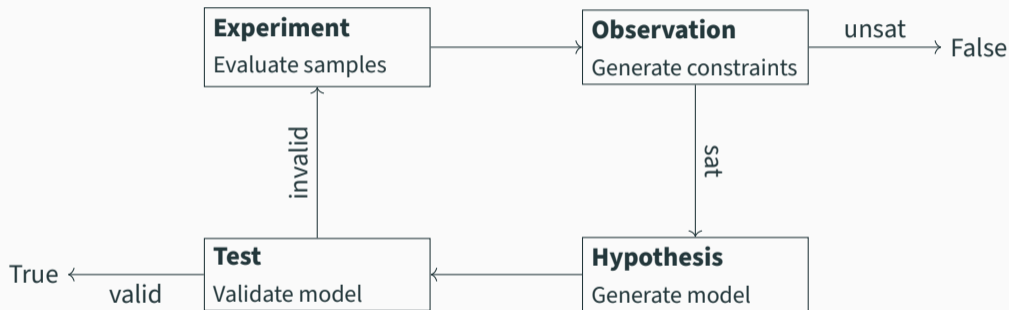
$$even(Even^\#) \rightarrow true^\#$$

Recursive-case: we need a **regular invariant learning procedure**.

Regular Invariant Learning

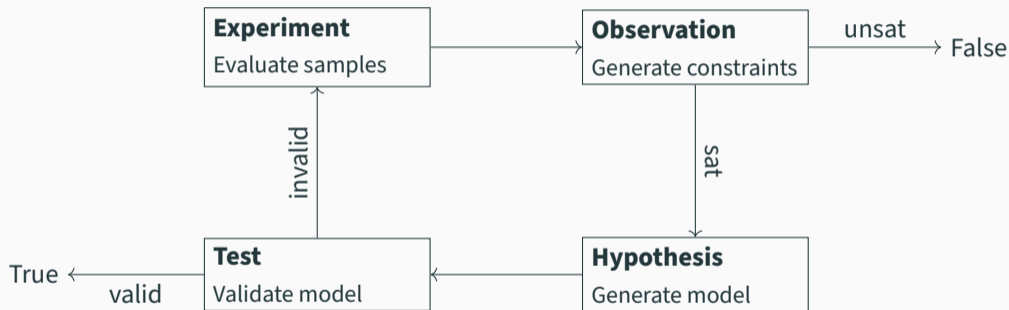
Regular Invariant Learning Procedure

A 4 steps SMT-based CEGAR-like procedure: the scientific method!



Regular Invariant Learning Procedure

A 4 steps SMT-based CEGAR-like procedure: the scientific method!



Note: In our setting a counter example is a term $f(t_1, \dots, t_n)$ that rewrites into $O^\#$ **and** $\bar{O}^\#$.

Step 1: Experiment

Definition (Experimental automaton \mathcal{A}_i)

- Finite, deterministic, reduced
- Recognizes a fragment of $\{ f(t_1, \dots, t_n) \}$
- \mathcal{R} -closed using the **Tree Automaton Completion algorithm**

\mathcal{A}_i contains all our **knowledge** about the function f at iteration i .

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Example

Consider \mathcal{A}_0 recognizing the terms $even(0)$ and $even(s(s(0)))$:

$$\begin{array}{lll} 0 \rightarrow q_0 & s(q_0) \rightarrow q_1 & s(q_1) \rightarrow q_2 \\ even(q_0) \rightarrow q_{f_0} & even(q_1) \rightarrow q_{f_1} & even(q_2) \rightarrow q_{f_2} \end{array}$$

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Example

Consider \mathcal{A}_0 recognizing the terms $even(0)$ and $even(s(s(0)))$:

$$\begin{array}{lll} true \rightarrow q_{true} & false \rightarrow q_{false} & \\ 0 \rightarrow q_0 & s(q_0) \rightarrow q_1 & s(q_1) \rightarrow q_2 \\ even(q_0) \rightarrow q_{f0} & even(q_1) \rightarrow q_{f1} & even(q_2) \rightarrow q_{f2} \\ q_{true} \rightarrow q_{f0} & q_{false} \rightarrow q_{f1} & q_{true} \rightarrow q_{f2} \end{array}$$

Step 2: Observation (1/2)

Definition (SMT-constraints)

For all abstract value $v \in \mathcal{L}(O^\#)$ and $v' \in \mathcal{L}(\bar{O}^\#)$ such that

$$\begin{array}{ccc} f(t_1, \dots, t_n) \xrightarrow[\mathcal{R}]{*} v & & f(t'_1, \dots, t'_n) \xrightarrow[\mathcal{R}]{*} v' \\ \mathcal{A}_i \downarrow * & & \mathcal{A}_i \downarrow * \\ q & & q' \end{array}$$

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Example

$$\begin{array}{ccc} \text{even}(0) \xrightarrow[\mathcal{R}]{*} \text{true} & & \text{even}(s(0)) \xrightarrow[\mathcal{R}]{*} \text{false} \\ \mathcal{A}_i \downarrow * & & \mathcal{A}_i \downarrow * \\ q_{f0} & \neq & q_{f1} \end{array}$$

Step 2: Observation (2/2)

Definition (Determinism constraints)

For all abstract value $v \in \mathcal{L}(O^\#)$ and $v' \in \mathcal{L}(\bar{O}^\#)$ such that

$$f(q_1, \dots, q_n) \rightarrow q$$

$$f(q'_1, \dots, q'_n) \rightarrow q'$$

We generate the SMT-constraint

$$q \neq q' \Rightarrow q_1 \neq q'_1 \vee \dots \vee q_n \neq q'_n$$

Step 2: Observation (2/2)

Definition (Determinism constraints)

For all abstract value $v \in \mathcal{L}(O^\#)$ and $v' \in \mathcal{L}(\bar{O}^\#)$ such that

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We generate the SMT-constraint

$$q \neq q' \Rightarrow q_1 \neq q'_1 \vee \dots \vee q_n \neq q'_n$$

Example

$$\text{even}(q_0) \rightarrow q_{f0}$$

$$\text{even}(q_1) \rightarrow q_{f1}$$

We generate the SMT-constraint

$$q_{f0} \neq q_{f1} \Rightarrow q_0 \neq q_1$$

Step 2: Observation (2/2)

Definition (Determinism constraints)

For all abstract value $v \in \mathcal{L}(O^\#)$ and $v' \in \mathcal{L}(\bar{O}^\#)$ such that

$$f(q_1, \dots, q_n) \rightarrow q$$

$$f(q'_1, \dots, q'_n) \rightarrow q'$$

We generate the SMT-constraint

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Example

$$\text{even}(q_0) \rightarrow q_{f0}$$

$$\text{even}(q_1) \rightarrow q_{f1}$$

We generate the SMT-constraint

$$q_0 \neq q_1$$

Step 3: Hypothesis

Definition (Candidate abstraction generation)

We generate $\mathcal{A}_i^\# = \phi(\mathcal{A}_i)$ where ϕ is a solution to our SMT-constraints.

$$\phi : \mathcal{Q} \mapsto \Sigma^\#$$

ϕ maps each concrete state to an abstract value.

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Example

$$q_{true} \neq q_{false}$$

$$q_{f0} \neq q_{f1}$$

$$q_{f1} \neq q_{f2}$$

$$q_0 \neq q_1$$

$$q_1 \neq q_2$$

$$\Rightarrow$$

$$q_{true} \mapsto a^\#$$

$$q_{f0} \mapsto a^\#$$

$$q_{f2} \mapsto a^\#$$

$$q_0 \mapsto a^\#$$

$$q_2 \mapsto a^\#$$

$$q_{false} \mapsto b^\#$$

$$q_{f1} \mapsto b^\#$$

$$q_1 \mapsto b^\#$$

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$$q_1 \neq q_2$$

$$\Rightarrow$$

$$q_{true} \mapsto true^\#$$

$$q_{f0} \mapsto true^\#$$

$$q_{f2} \mapsto true^\#$$

$$q_0 \mapsto Even^\#$$

$$q_2 \mapsto Even^\#$$

$$q_{false} \mapsto false^\#$$

$$q_{f1} \mapsto false^\#$$

$$q_1 \mapsto Odd^\#$$

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Example

$$true \rightarrow q_{true}$$

$$false \rightarrow q_{false}$$

$$0 \rightarrow q_0$$

$$s(q_0) \rightarrow q_1$$

$$s(q_1) \rightarrow q_2$$

$$even(q_0) \rightarrow q_{f0}$$

$$even(q_1) \rightarrow q_{f1}$$

$$even(q_2) \rightarrow q_{f2}$$

$$q_{true} \rightarrow q_{f0}$$

$$q_{false} \rightarrow q_{f1}$$

$$q_{true} \rightarrow q_{f2}$$

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Example

$true \rightarrow true^\#$	$false \rightarrow false^\#$	
$0 \rightarrow Even^\#$	$s(Even^\#) \rightarrow Odd^\#$	$s(Odd^\#) \rightarrow Even^\#$
$even(Even^\#) \rightarrow true^\#$	$even(Odd^\#) \rightarrow false^\#$	$even(Even^\#) \rightarrow Odd^\#$
$true^\# \rightarrow true^\#$	$false^\# \rightarrow false^\#$	$true^\# \rightarrow true^\#$

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ϕ maps each concrete state to an abstract value.

Example

$$\begin{array}{lll} \text{true} \rightarrow \text{true}^\# & \text{false} \rightarrow \text{false}^\# & \\ 0 \rightarrow \text{Even}^\# & s(\text{Even}^\#) \rightarrow \text{Odd}^\# & s(\text{Odd}^\#) \rightarrow \text{Even}^\# \\ \text{even}(\text{Even}^\#) \rightarrow \text{true}^\# & \text{even}(\text{Odd}^\#) \rightarrow \text{false}^\# & \end{array}$$

Step 4: Test

Termination

$\mathcal{A}_i^\#$ contains a valid abstraction $\mathcal{R}^\#$ and $\Lambda = \langle \Sigma^\#, \Delta^\# \rangle$ when:

- it is **\mathcal{R} -closed**
- it is **complete**

Otherwise we go back to Step 1: Experiment (with new samples).

Step 4: Test

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- it is **\mathcal{R} -closed**
- it is **complete**

Otherwise we go back to Step 1: Experiment (with new samples).

Example

Our previous candidate model is a valid abstraction:

$true \rightarrow true^\#$	$false \rightarrow false^\#$		
$0 \rightarrow Even^\#$	$s(Even^\#) \rightarrow Odd^\#$	$s(Odd^\#) \rightarrow Even^\#$	this is $\Delta^\#$
$even(Even^\#) \rightarrow true^\#$	$even(Odd^\#) \rightarrow false^\#$		this is $\mathcal{R}^\#$

Step 4: Test

\mathcal{R}

$even(0) \rightarrow true$
 $even(s(s(\underline{x}))) \rightarrow \underline{x}$

$even(s(0)) \rightarrow false$

Example

Our previous candidate model is a valid abstraction:

$true \rightarrow true^\#$	$false \rightarrow false^\#$		
$0 \rightarrow Even^\#$	$s(Even^\#) \rightarrow Odd^\#$	$s(Odd^\#) \rightarrow Even^\#$	this is $\Delta^\#$
$even(Even^\#) \rightarrow true^\#$	$even(Odd^\#) \rightarrow false^\#$		this is $\mathcal{R}^\#$

Properties of the Invariant Learning Procedure

Regular Completeness

If there exists a **regular** abstraction $\mathcal{R}^\#, \Delta^\#$ that separates $O^\#$ and $\bar{O}^\#$, we will eventually find it.

Completeness in Refutation

If there exists a counter example (a term t that rewrites into both $O^\#$ and $\bar{O}^\#$), then we will eventually find it.

Experiments

Experiments

Implementation in Timbuk 4, tested on 80+ problems, compared with Timbuk 3.

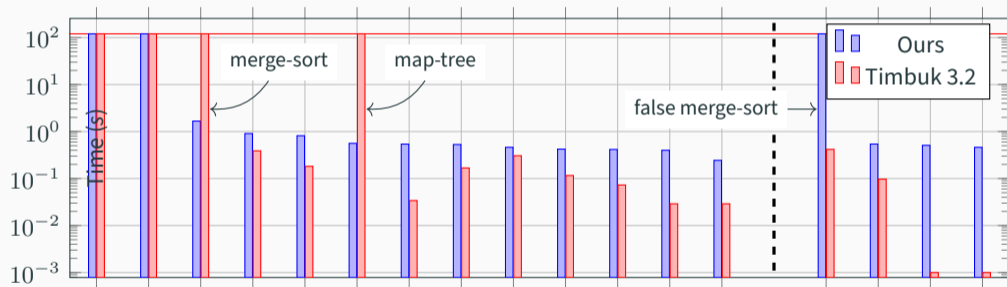
First-order programs



Experiments

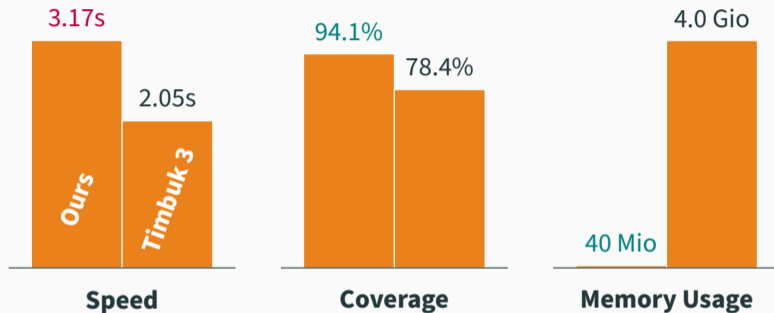
Implementation in Timbuk 4, tested on 80+ problems, compared with Timbuk 3.

Higher-order programs



Conclusion

Implementation in Timbuk 4, tested on 80+ problems, compared with Timbuk 3.



- “Regularly” complete, complete in refutation.
- Modular thanks to the type system approach.

<https://gitlab.inria.fr/regular-pv/timbuk/timbuk>