

Synthesis and Satisfiability Modulo Theory Solvers

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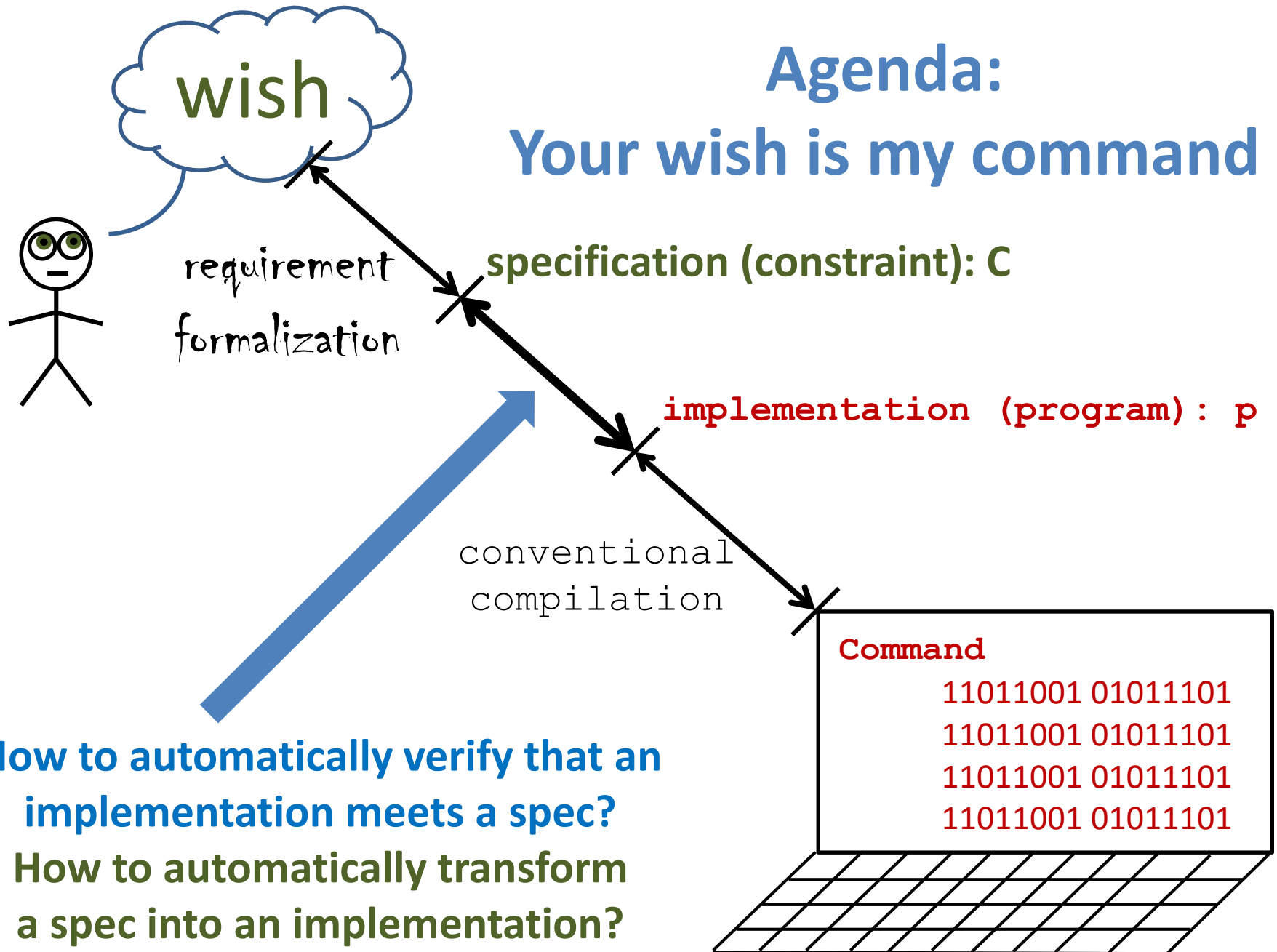
<http://lara.epfl.ch>

<http://leon.epfl.ch>

<http://cvc4.cs.nyu.edu/web/>

Agenda:

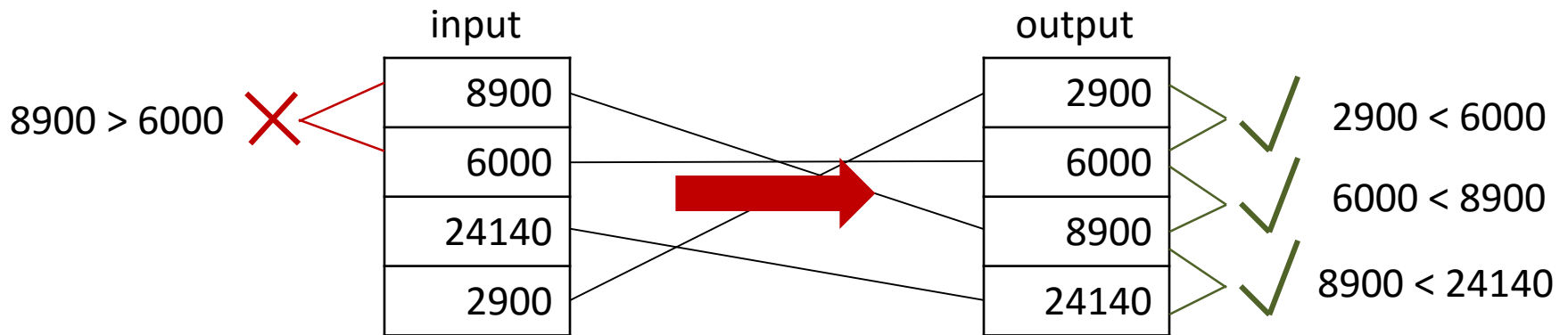
Your wish is my command



How to automatically verify that an implementation meets a spec?

How to automatically transform a spec into an implementation?

Sorting Specification as a Program



Given a list of numbers, make **this list sorted**

wish

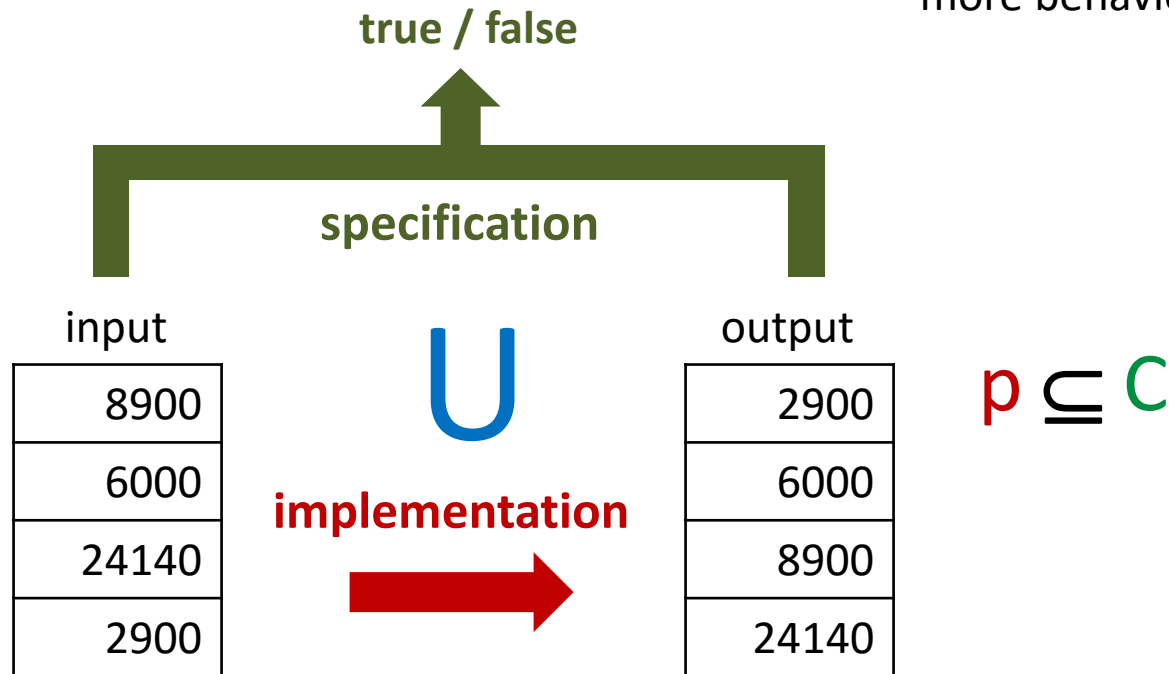
```
def sort_spec(input : List, output : List) : Boolean =  
  content(output) == content(input) && isSorted(output)
```

Specification (for us) is a *program* that checks, for a **given input**, whether the **given output** is acceptable

Specification vs Implementation

```
def C(i : List, o : List) : Boolean = constraint on the output  
content(o)==content(i) && isSorted(o)
```

more behaviors



fewer behaviors

```
def p(i : List) : List = function that computes the output  
sort i using a sorting algorithm (equality constraint on output)
```

<http://leon.epfl.ch> system for Scala

Scala: invented at EPFL by Prof. Martin Odersky <http://scala-lang.org/>

- hundreds of thousands of Scala programmers, used in:

[Twitter](#), [Foursquare](#), [Coursera](#), [The Guardian](#), [New York Times](#), [Huffington Post](#), [UBS](#), [LinkedIn](#), [Meetup](#), [Verizon](#), Intel, ...

Typesafe Inc. supports Scala commercially

EPFL: industrial advisory board, courses, open source development

Chisel: “...an open-source hardware construction language developed at UC Berkeley that supports advanced hardware design using highly parameterized generators and layered domain-specific hardware languages.” – other Scala DSLs: e.g. OptiML

Apache Spark: “an open-source cluster computing framework with in-memory processing to speed analytic applications up to 100 times faster compared to technologies on the market today. Developed in the AMPLab at UC Berkeley, Apache Spark can help reduce data interaction complexity, increase processing speed and enhance mission-critical applications with deep intelligence.”

“...IBM is making a major commitment to the future of [Apache Spark](#), with a series of initiatives announced today. IBM will offer Apache Spark as a service on [Bluemix](#); commit 3,500 researchers to work on Spark-related projects; donate IBM SystemML to the Spark ecosystem; and offer courses to train 1 million data scientists and engineers to use Spark.”

DEMO

<http://leon.epfl.ch>

a) **Check assertion** while program **p** runs: **$C(i, p(i))$**

b) **Verify** whether program always meets the spec:

$\forall i. C(i, p(i))$

c) **Constraint programming**: once **i** is known, find **o** to satisfy a given constraint: **find o such that $C(i, o)$**

repair

d) **Synthesis**: solve **C** symbolically to obtain program **p** that is correct by construction, for all inputs: **find p such that $\forall i. C(i, p(i))$** i.e. **$p \subseteq C$**

Approaches and Their Guarantees

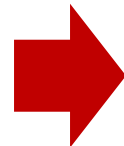
both specification **C** and program **p** are given:

a) **Check assertion** while program **p** runs: **$C(i, p(i))$**

b) **Verify** that program always meets spec:
 $\forall i. C(i, p(i))$

only specification **C** is given:

c) **Constraint programming**: once **i** is known, find **o** to satisfy a given constraint: **find o such that $C(i, o)$**
run-time



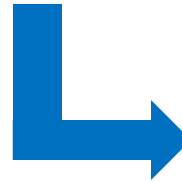
d) **Synthesis**: solve **C** symbolically to obtain program **p** that is correct by construction, for all inputs: **find p such that $\forall i. C(i, p(i))$** i.e. **$p \subseteq C$**

compile-time

Synthesizing Sort in Leon System

```
def insertSorted(lst : List, v: Int): List = {  
  require(isSorted(lst))  
  choose { (r: List) =>  
    isSorted(r) && content(r) == content(lst) ++ Set(v) } }  
  
def sort(lst : List): List = choose { (r: List) =>  
  isSorted(r) && content(r) == content(lst) }
```

<http://leon.epfl.ch>



```
def insertSorted(lst: List, v: Int): List = {  
  require(isSorted(lst))  
  lst match {  
    case Nil => Cons(v, Nil)  
    case Cons(h, tail) =>  
      val r = insertSorted(t, v)  
      if (v > h) Cons(h, r)  
      else if (h == v) r  
      else Cons(v, Cons(h, t)) }  
  
def sort(lst : List): List = lst match {  
  case Nil => Nil  
  case Cons(h, t) => insertSorted(sort(t), h) }
```

OOPSLA 2013:
Synthesis Modulo Recursive Functions



Etienne Kneuss



Ivan Kuraj



Philippe Suter

Recursion Schemas + STE in Action

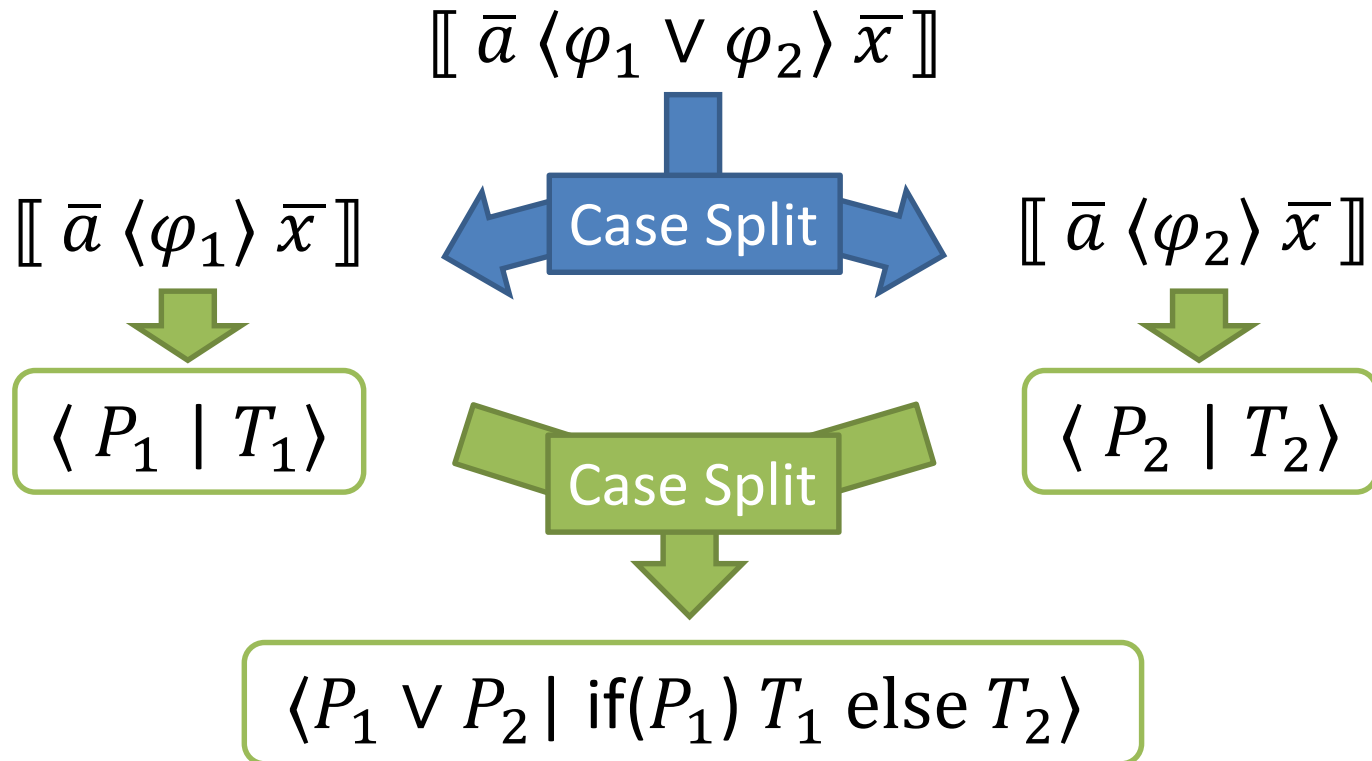
```
def delete(in1: List, v: Int) = choose {  
  (out: List) => content(out) == content(in1) -- Set(v)  
}
```

```
def delete(in1: List, v: Int) = {  
  def rec(in: List, v: Int): List = in match {  
    case Cons(h,t) =>  
      val r = rec(t,v)  
      if (h == v) {  
        CEGIS r  
      } else {  
        CEGIS Cons(h, r)  
      }  
    case Nil =>  
      CEGIS Nil  
  } ensuring { content(_) == content(in1) -- Set(v) }  
  rec(in1, v)  
}
```

ADT Induction

EQ Split

Decomposition Example: Case Split



Symbolic Term Exploration (STE)

Symbolic search over many expressions of bounded size

$$T(\bar{b}) = \text{if } (b_0) a_0 \\ \text{elseif } (b_1) \text{ Nil} \\ \text{elseif } (b_2) \text{ Cons}(0, \text{Nil}) \\ \text{elseif } (b_3) \dots$$

SMT solver searches exponentially many expressions given by polynomially many Boolean variables

Concrete execution prunes search space

Leon's verifier validates candidate expressions

- using again SMT solvers

Building Block: Complete Synthesis

```
def secondsToTime(totalSeconds: Int) : (Int, Int, Int) =
```

```
  choose((h: Int, m: Int, s: Int) => (  
    h * 3600 + m * 60 + s == totalSeconds  
    && h ≥ 0  
    && m ≥ 0 && m < 60  
    && s ≥ 0 && s < 60  ))
```

← spec in integer
linear arithmetic

```
def secondsToTime(totalSeconds: Int) : (Int, Int, Int) =
```

```
  val t1 = totalSeconds div 3600  
  val t2 = totalSeconds - 3600 * t1  
  val t3 = t2 div 60  
  val t4 = totalSeconds - 3600 * t1 - 60 * t3  
  (t1, t3, t4)
```

synthesis
procedure

Implementation

- V1) Quantifier elimination procedure for Presburger arith.
 - optimization for some important cases
 - inefficient in general
- V2) Automata-based procedure for int. arithmetic with bitwise operations (Hamza, Jobstmann, K., FMCAD'10)
 - handles larger subset
 - bad at e.g. multiplication by large constants - we do not know good general techniques for sequential circuits
- V3) Inside an SMT solver: **Andrew Reynolds**, Morgan Deters, Cesare Tinelli, Clark Barrett, K. – CAV'15
 - ➔ focus today

Synthesis Problem for an SMT Solver

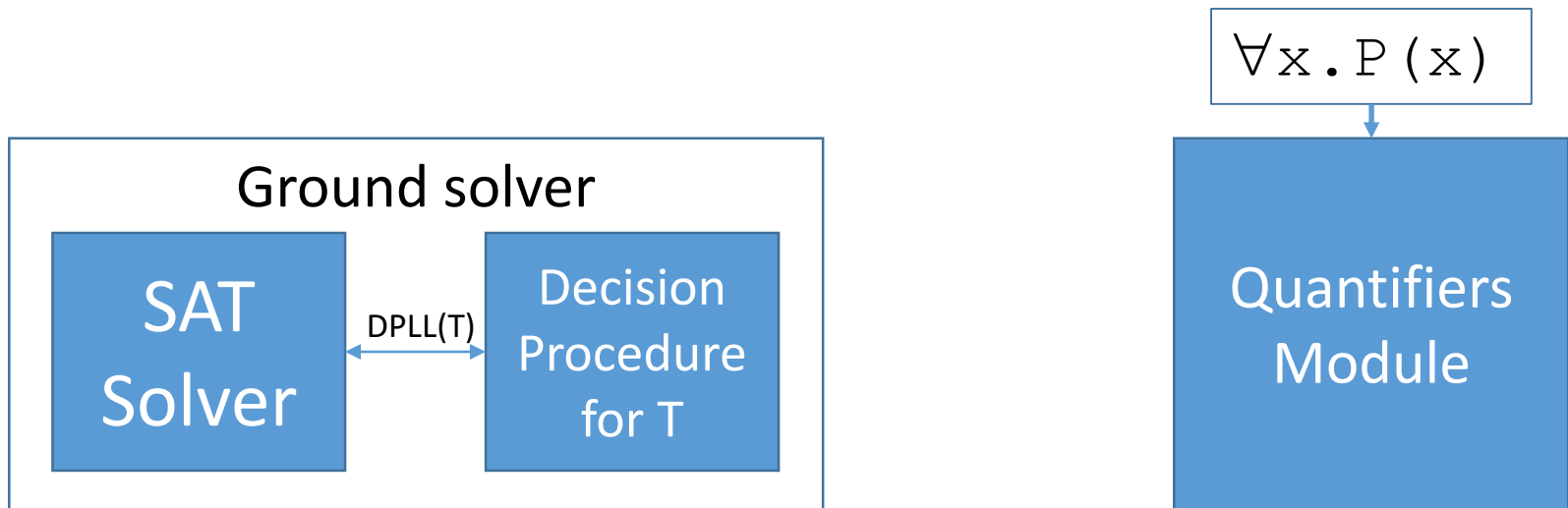
- Synthesis Problem : $\exists f . \forall x . P (f , x)$



There exists a function f such that for all x , $P (f , x)$

- Most existing approaches for synthesis
 - Rely on specialized solver that makes **subcalls** to an SMT Solver
- Goal: approach implemented *entirely inside SMT solver*

SMT Solver + Quantified Formulas



SMT solver consists of:

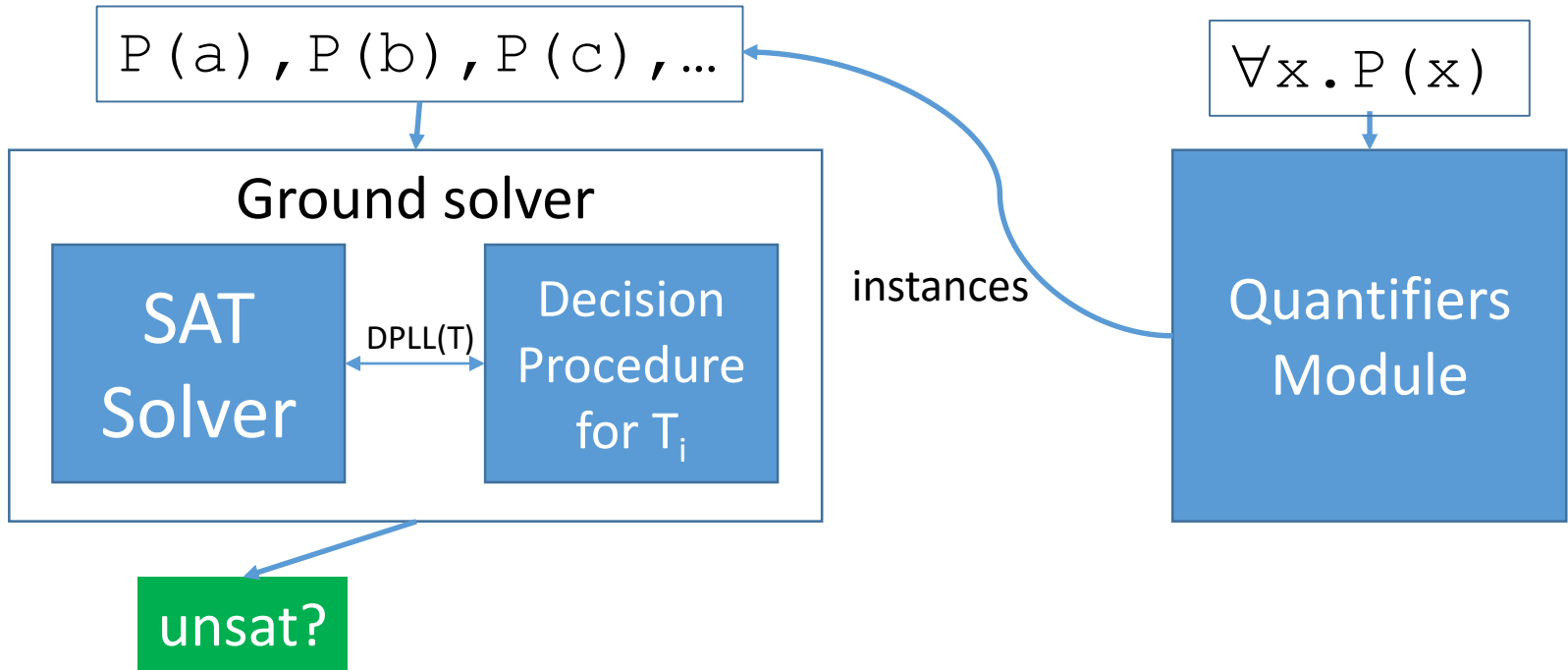
- **Ground solver** maintains a set of ground (variable-free) constraints
- **Quantifiers Module** maintains a set of quantified formulas: $\forall x . P(x)$

Using SMT solvers: game-changer in automated software verification

Increasingly relevant in industry:

- symbolic execution of systems code, microcode, word-level reasoning
- Anders Franzén, Alessandro Cimatti, Alexander Nadel, Roberto Sebastiani, Jonathan Shalev: **Applying SMT in symbolic execution of microcode**. [FMCAD 2010](#): 121-128 – **Best Paper Award**

SMT Solver + Quantified Formulas



- Goal : add **instances** of axioms until ground solver can answer “unsat”

Running Example : Max of Two Integers

$$\exists f . \forall x y . (f (x , y) \geq x \wedge f (x , y) \geq y \wedge (f (x , y) = x \vee f (x , y) = y))$$

- Specifies that f computes the maximum of integers x and y
- A solution:

$$f := \lambda x y . \text{ite} (x \geq y , x , y)$$

Approach: Refutation-Based Synthesis

$$\neg \exists f . \forall \mathbf{x} . P(f, \mathbf{x})$$

- What if we **negate** the synthesis conjecture?
- If we are in a *satisfaction-complete* theory T (e.g. LIA, BV):
 - F is T -satisfiable if and only if $\neg F$ is T -unsatisfiable

⇒ Will suffice for us to show the above formula is **unsat**

Challenge: Second-Order Quantification

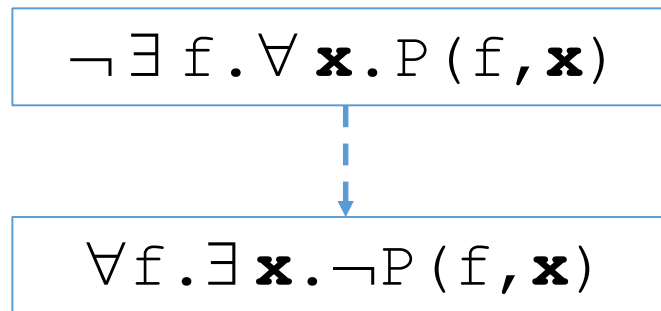
$$\neg \exists f . \forall \mathbf{x} . P(f, \mathbf{x})$$



$$\forall f . \exists \mathbf{x} . \neg P(f, \mathbf{x})$$

- Challenge: negation introduces universal \forall **over function f**
 - No SMT solvers directly support second-order quantification

Challenge: Second-Order Quantification



- Challenge: negation introduces universal \forall over function f
 - No SMT solvers directly support second-order quantification
- However, we can avoid this quantification using two approaches:
 1. When property P is **single invocation** for f ← focus now
 2. When f is given **syntactic restrictions**

Single Invocation Properties

$$\forall f. \exists x y. (f(x, y) < x \vee f(x, y) < y \vee (f(x, y) \neq x \wedge f(x, y) \neq y))$$

Single Invocation Properties

$$\forall f. \exists x y. (f(x, y) < x \vee f(x, y) < y \vee (f(x, y) \neq x \wedge f(x, y) \neq y))$$

- *Single invocation* properties
 - Are properties such that:
 - All occurrences of f are of a particular form, e.g. $f(x, y)$ above
 - Are a common class of properties useful for:
 - Software Synthesis (post-conditions describing the result of a function)
 - Given solution, it can be checked without replicating solution
- NOT single invocation: “ f is commutative”

Single Invocation Properties

$$\forall f . \exists x y . (f (x , y) < x \vee f (x , y) < y \vee (f (x , y) \neq x \wedge f (x , y) \neq y))$$



Push \forall quantification inwards

$$\exists x y . \forall g . (g < x \vee g < y \vee (g \neq x \wedge g \neq y))$$

- Occurrences of $f (x , y)$ are replaced with integer variable g
- Resulting formula is equisatisfiable, and **first-order**

Single Invocation Properties

$$\forall f. \exists x y. (f(x, y) < x \vee f(x, y) < y \vee (f(x, y) \neq x \wedge f(x, y) \neq y))$$

$$\exists x y. \forall g. (g < x \vee g < y \vee (g \neq x \wedge g \neq y))$$

Skolemize, for fresh **a** and **b**

$$\forall g. (g < \mathbf{a} \vee g < \mathbf{b} \vee (g \neq \mathbf{a} \wedge g \neq \mathbf{b}))$$

Solving Max Example

Ground
solver

$\forall g. (g < a \vee g < b \vee (g \neq a \wedge g \neq b))$



Quantifiers
Module

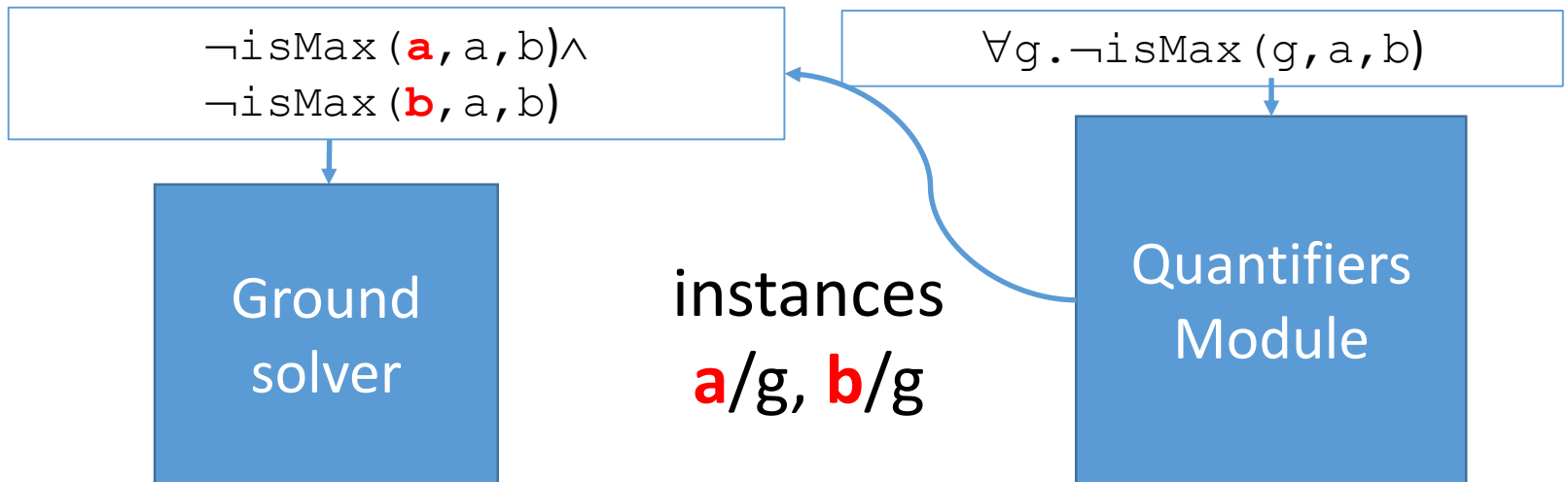
Solving Max Example

Ground
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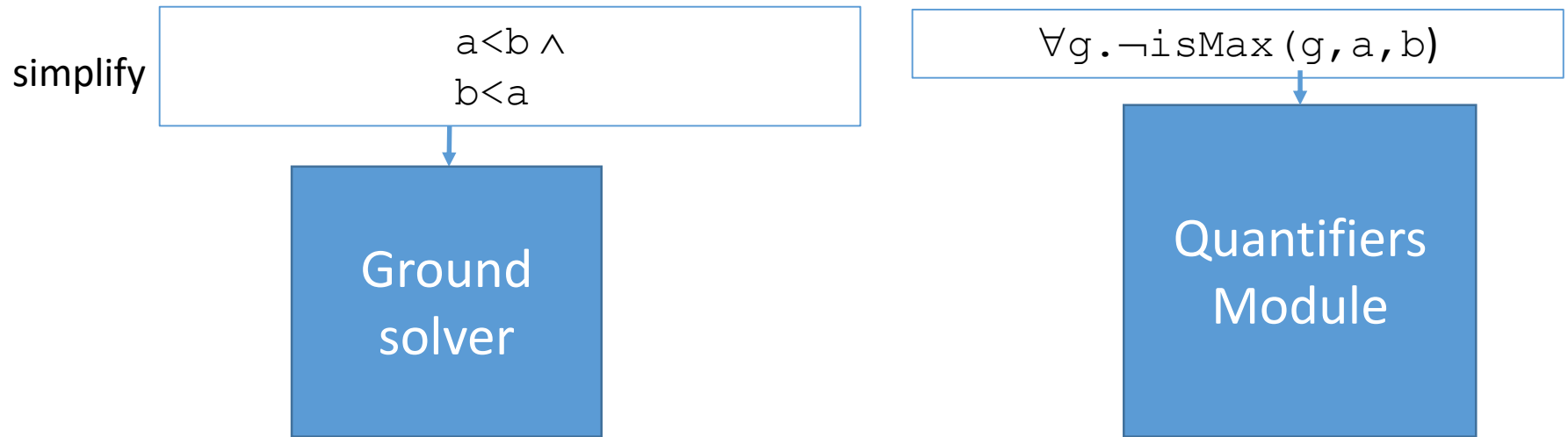
$\forall g. \neg \text{isMax}(g, a, b)$

↓
Quantifiers
Module

Solving Max Example



Solving Max Example



Solving Max Example

$a < b \wedge$
 $b < a$

Ground
solver

unsat

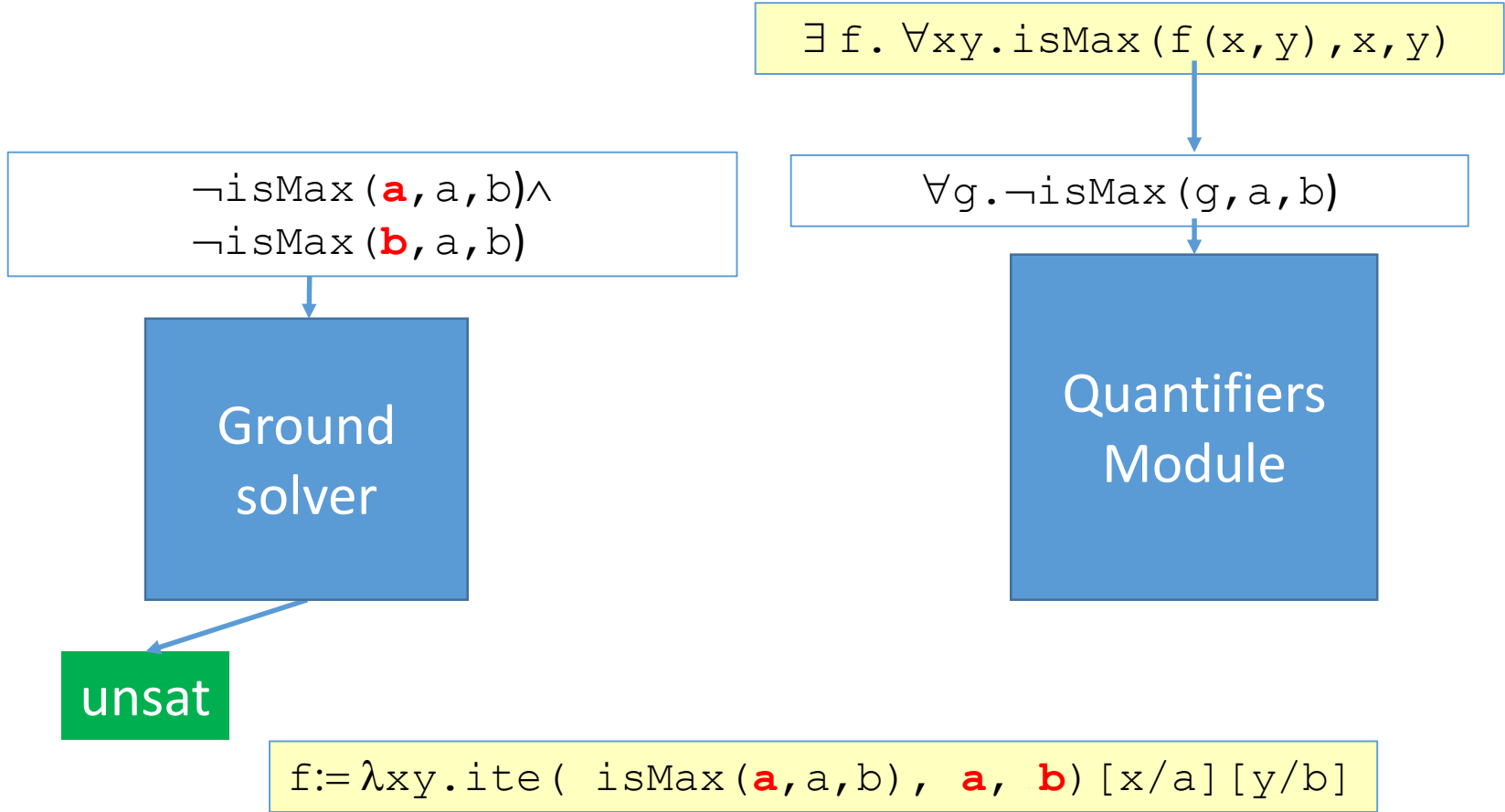
$\Rightarrow \forall g. \neg \text{isMax}(g, a, b)$ is **unsatisfiable**
by instances $a/g, b/g,$

implies original synthesis conjecture has a solution

$\forall g. \neg \text{isMax}(g, a, b)$

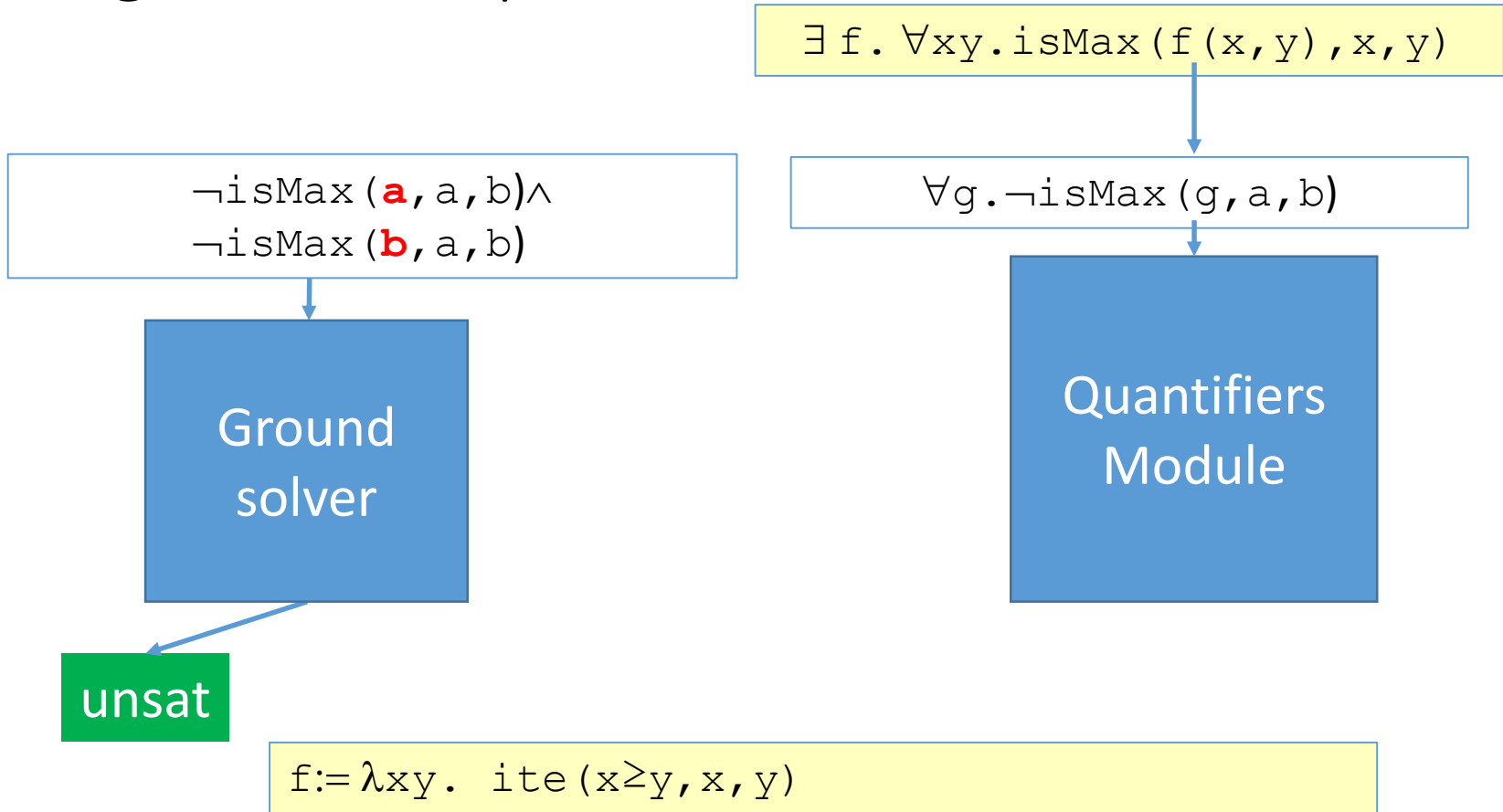
Quantifiers
Module

Solving Max Example



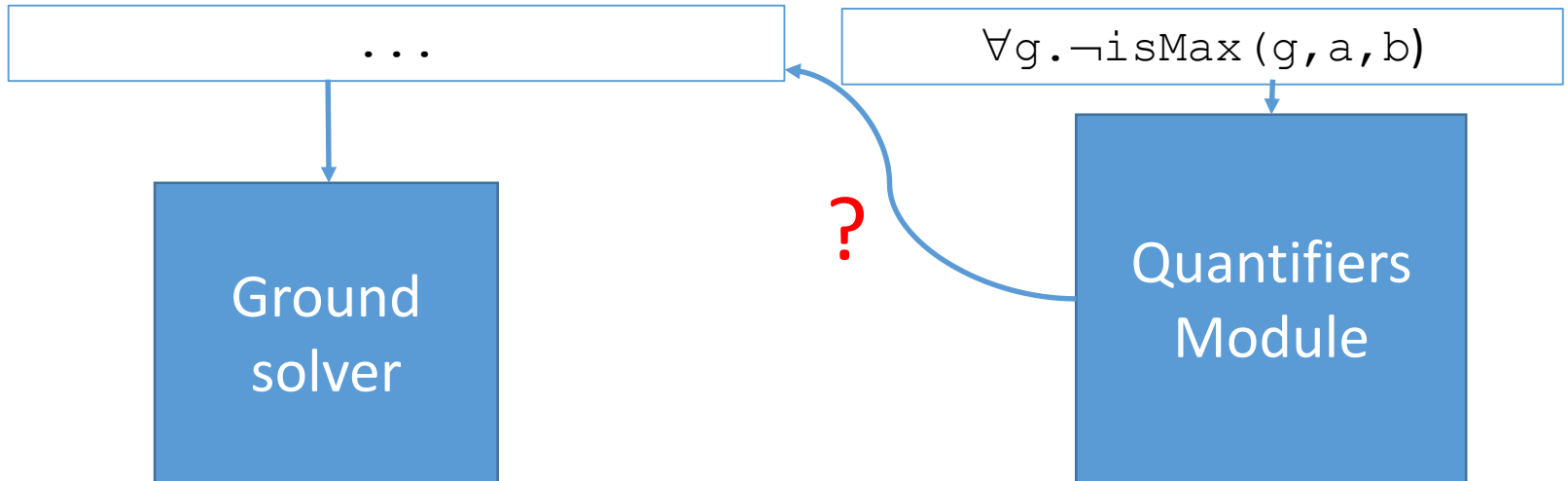
⇒ Extract solution from **unsatisfiable core of instantiations a/g, b/g**

Solving Max Example

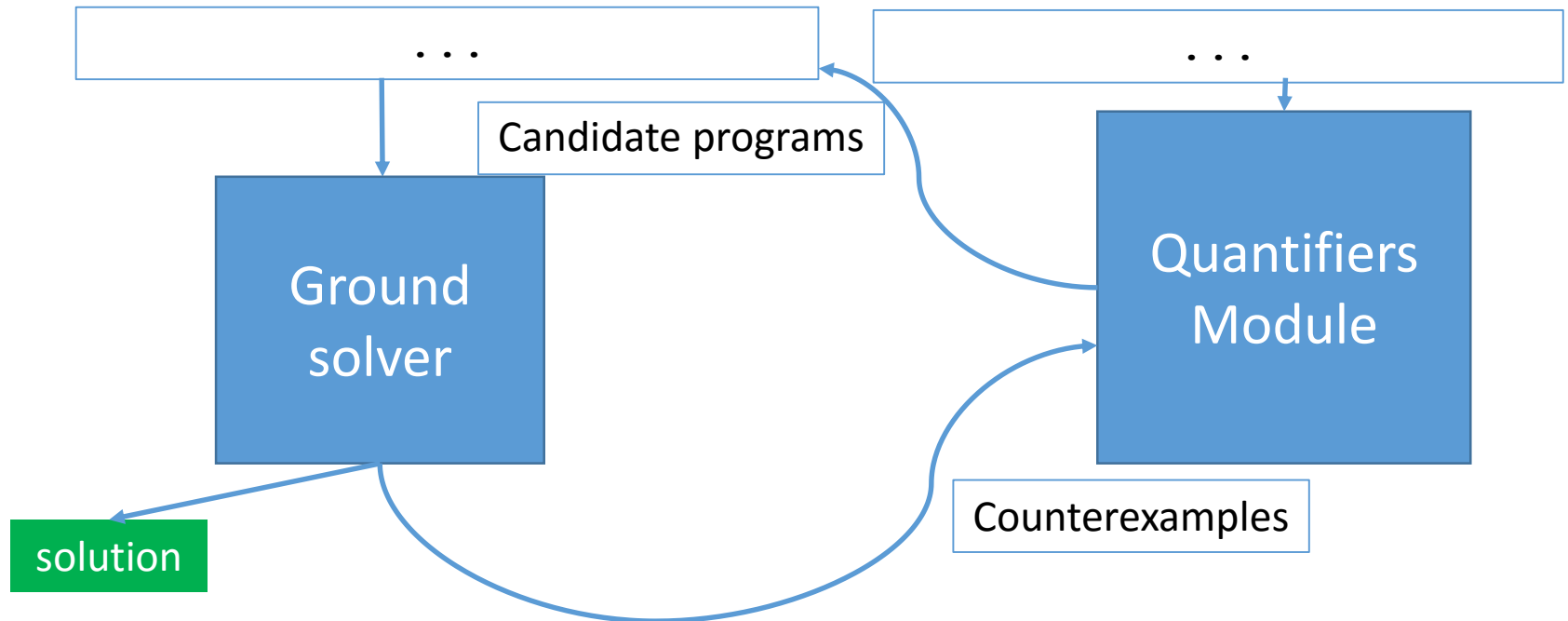


⇒ Desired function, after simplification

How do we Choose Relevant Instances?



Counterexample-Guided Quantifier Instantiation



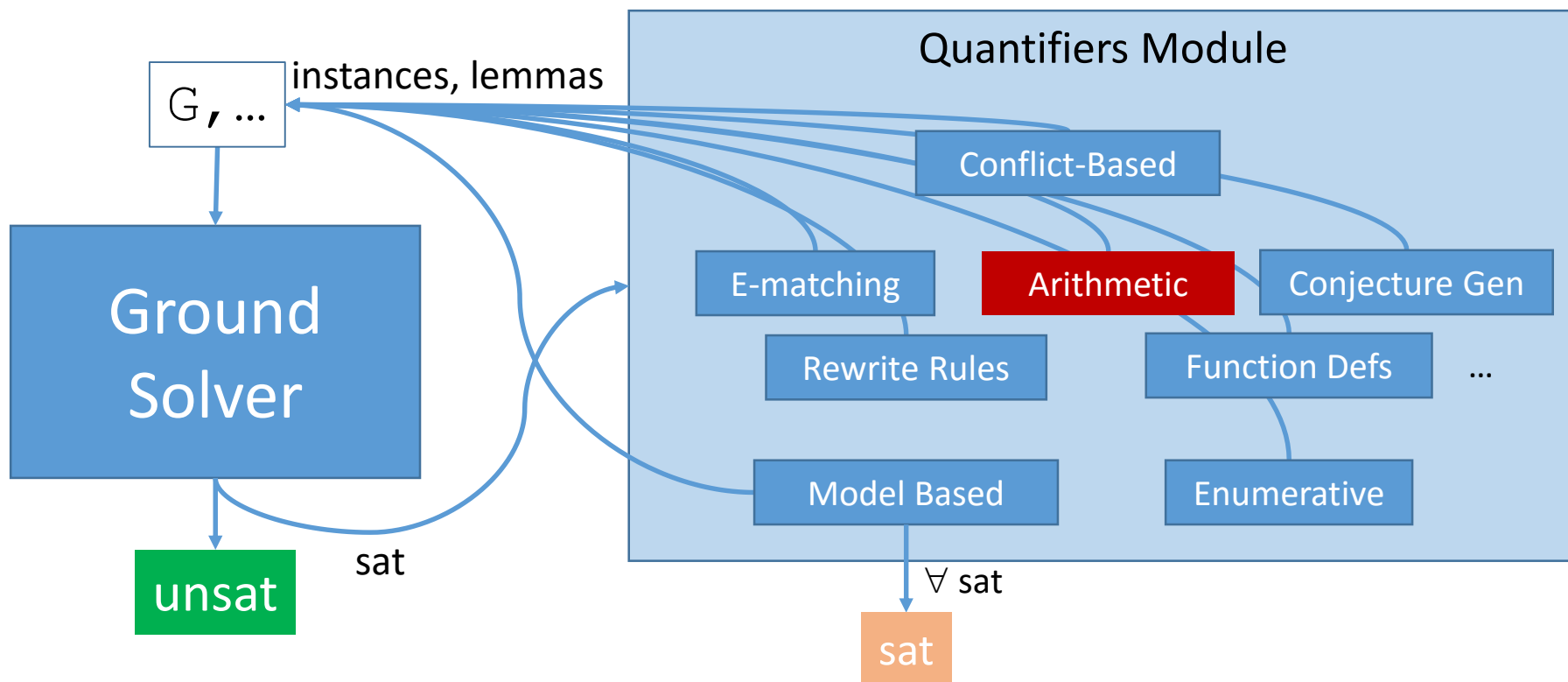
- Instances chosen using *counterexample-guided quantifier instantiation*

⇒ Follows counterexample-guided inductive synthesis (CEGIS) approach

In work under submission, we provide framework where such CEGIS choices are **complete** for linear arithmetic

Much better scalability than quantifier elimination approaches

Quantifiers Module of CVC4



Specialized technique for quantified arithmetic

[arXiv:1510.02642](https://arxiv.org/abs/1510.02642)

An Instantiation-Based Approach for Solving Quantified Linear Arithmetic

[Andrew Reynolds](#), [Tim King](#), [Viktor Kuncak](#)

CVC4 in Sygus Competition 2015

NSF Expedition on Computer Augmented Programm Engineering (ExCAPE)
led by Rajeev Alur, involves UPenn, MIT, Berkeley, Rice, Illinois, Maryland, UCLA, Michigan
organizes a competition of software synthesis tools

<https://excape.cis.upenn.edu/>

Our technique **Won** General and LIA tracks (= 2/3 tracks) in competition

LIA Track			
Solver	#solved	total-expr-size	average-expr-size
CVC4-1.5-syguscomp2015-v4	70	43726	624.66
AlchemistCSDT	47	6658	141.66
Alchemist CS	33	866	26.24

In LIA track, solved 70/73 benchmarks, 60 of these in <1 second

- Nearest competitor **AlchemistCSDT** solved 47/73 in a timeout of 1 hour

Max example : Sygus Comp 2015

n	2	3	4	5	6	7	8	9	10	11	12	13	14	15
cvc4+si	0.0	0.0	0.0	0.0	0.1	0.1	0.2	0.3	0.6	1.0	1.9	3.2	5.3	6.5
AlchemistCSDT	0.2	0.6	1.5	6.4	20.8	132.8	877.9	–	–	–	–	–	–	–
AlchemistCS	0.0	3.7	–	–	–	–	–	–	–	–	–	–	–	–

- Outperforms existing approaches by an order of magnitude or more

⇒ Our approach is efficient for synthesizing non-recursive functions that are defined by cases

Implementation available in the main branch of CVC4 SMT solver:

<http://cvc4.cs.nyu.edu/web/>

Conclusions: Synthesis and SMT Solvers

Leon system for verifying and synthesizing Scala programs

<http://leon.epfl.ch>

SMT solvers are essential for synthesis and verification

Given support for quantifiers, SMT solver can perform synthesis on its own!

Key challenge: efficient techniques to instantiate quantifiers

- CVC4 solution is state of the art for linear arithmetic: complete and fast

[arXiv:1510.02642](https://arxiv.org/abs/1510.02642)

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