

Satisfiability Modulo Theories (SMT) solving

Amar Shah

Come to my Office Hours!!

(CIC 2206, Wednesdays 11am – noon)

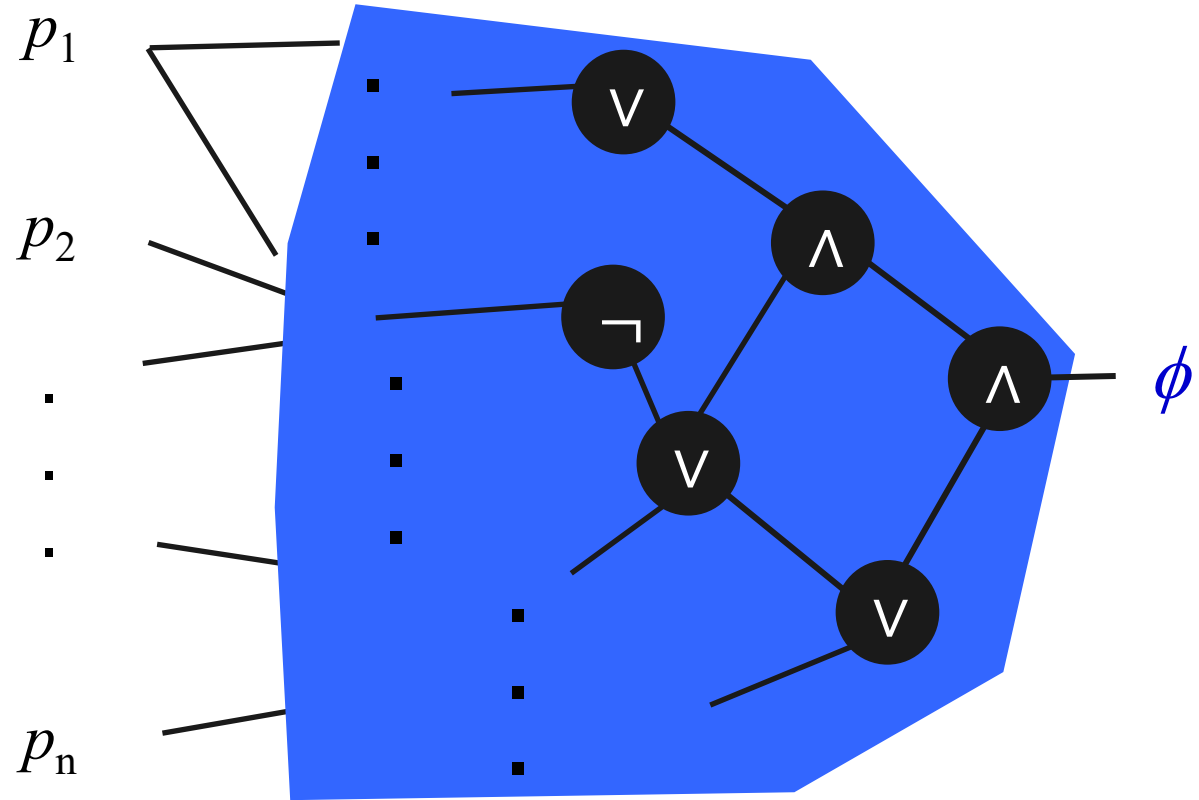
- Slides inspired by previous presentations by Sanjit Seshia, Clark Barrett, Andrew Reynolds, Leonardo de Moura, and Alberto Oliveras
- SMT solving is a pretty broad field, which is continuing to evolve. This presentation is only a taste

1. Introduction

- 2. Defining SMT
- 3. Eager Approach
- 4. Lazy Approach
- 5. Universal Quantifiers
- 6. Future of SMT solving

This class so far: Boolean SAT solving

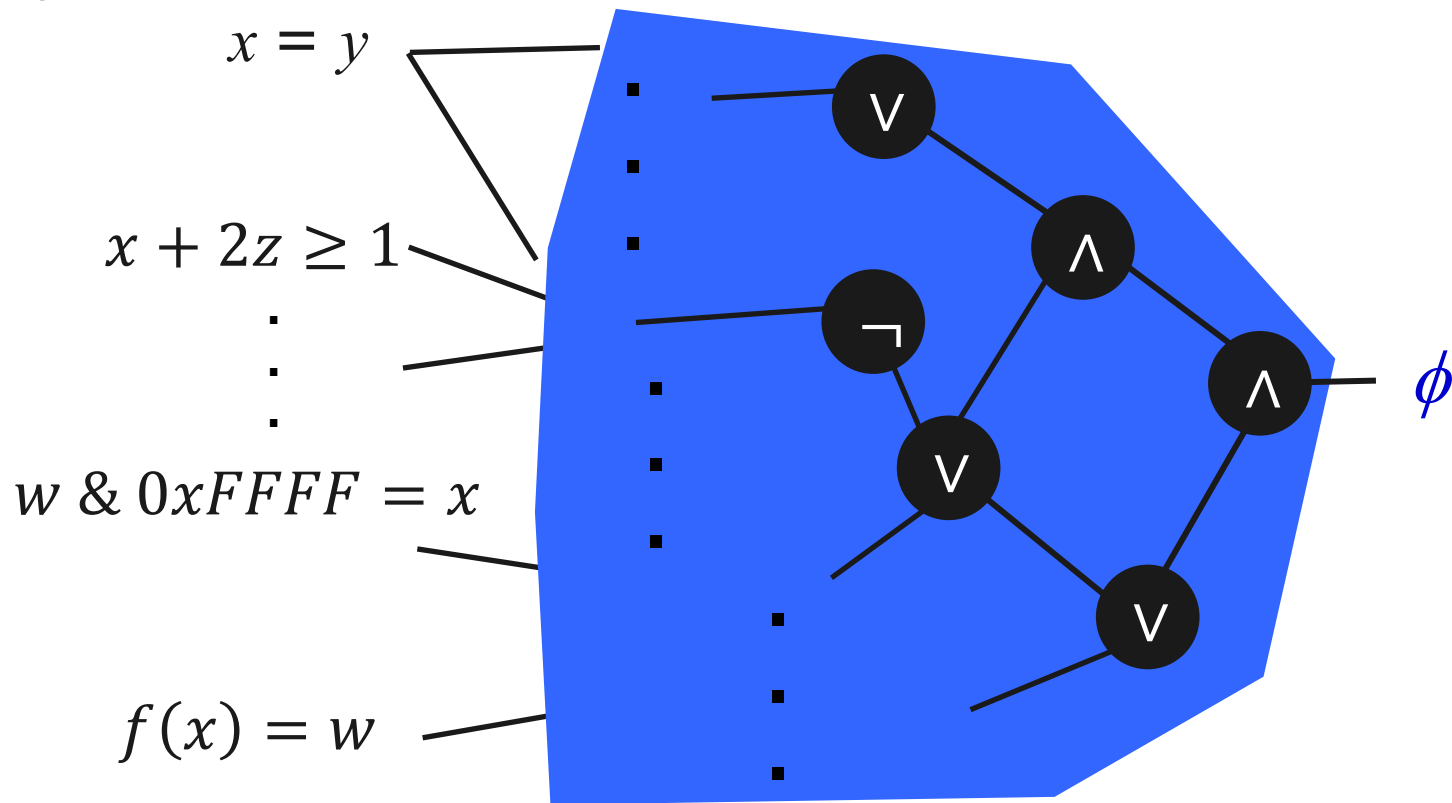
Is there an assignment to the $p_1, p_2, \dots, p_n$ variables such that ϕ evaluates to true?



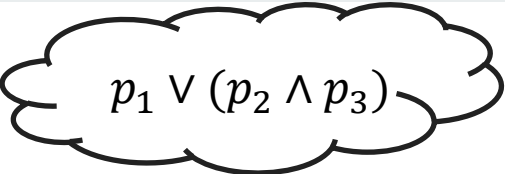
Satisfiability Modulo Theories

Is there an assignment to f, w, x, y, z such that ϕ evaluates to true?

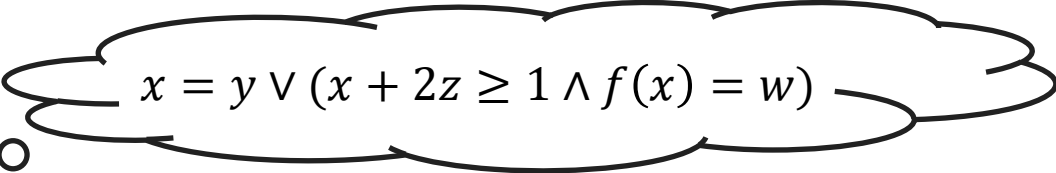
i.e. is the formula satisfiable?



What is SMT?


$$p_1 \vee (p_2 \wedge p_3)$$

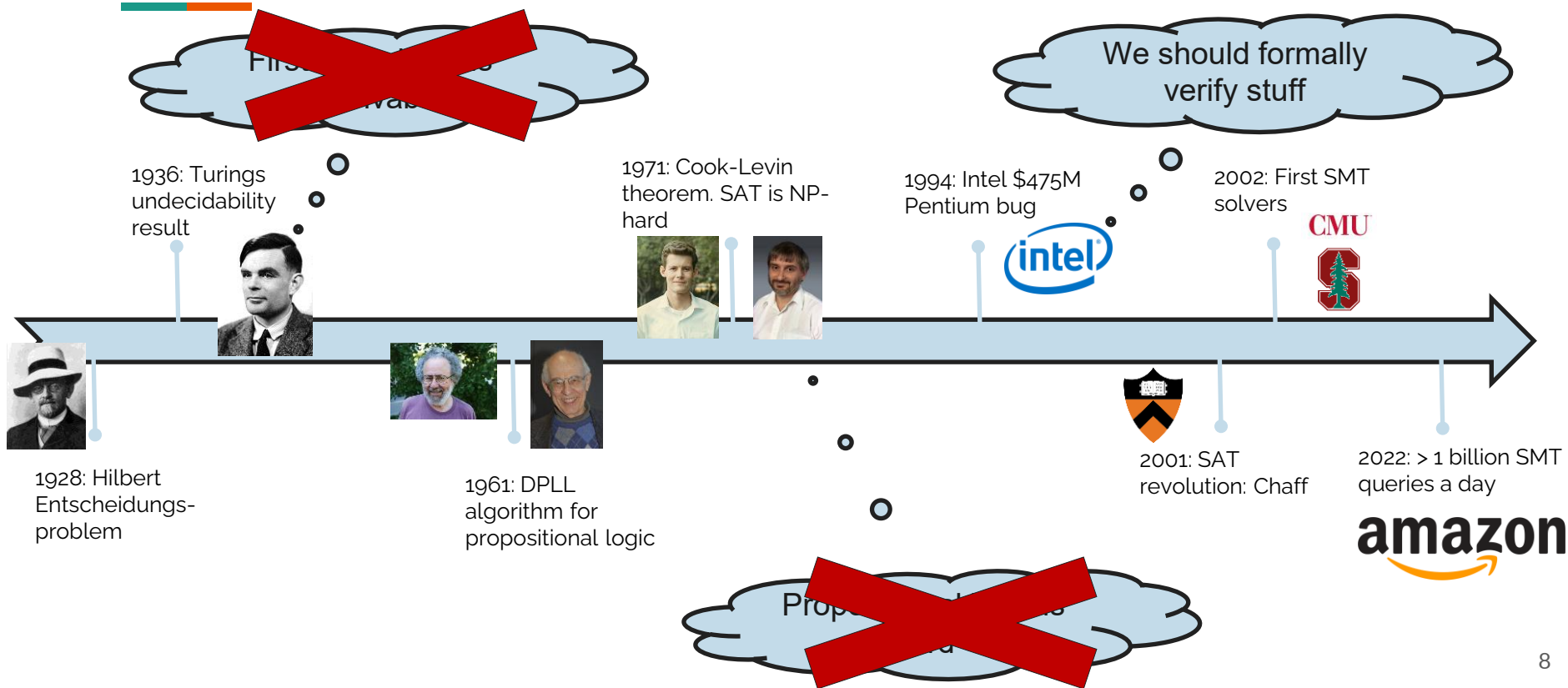
- SAT: use **propositional logic** as the formalization language
 - + high degree of efficiency
 - expressive (all NP-complete) but involved encodings


$$x = y \vee (x + 2z \geq 1 \wedge f(x) = w)$$

- SMT: propositional logic + **domain-specific reasoning (multi-sorted first-order logic)**
 - + improves the expressivity
 - certain (but acceptable) loss of efficiency

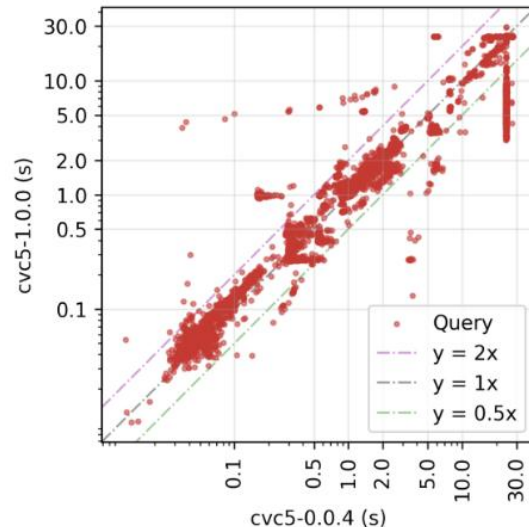
Goal: Introduce SMT solving and its main techniques

History Lesson: trying to solve first-order logic



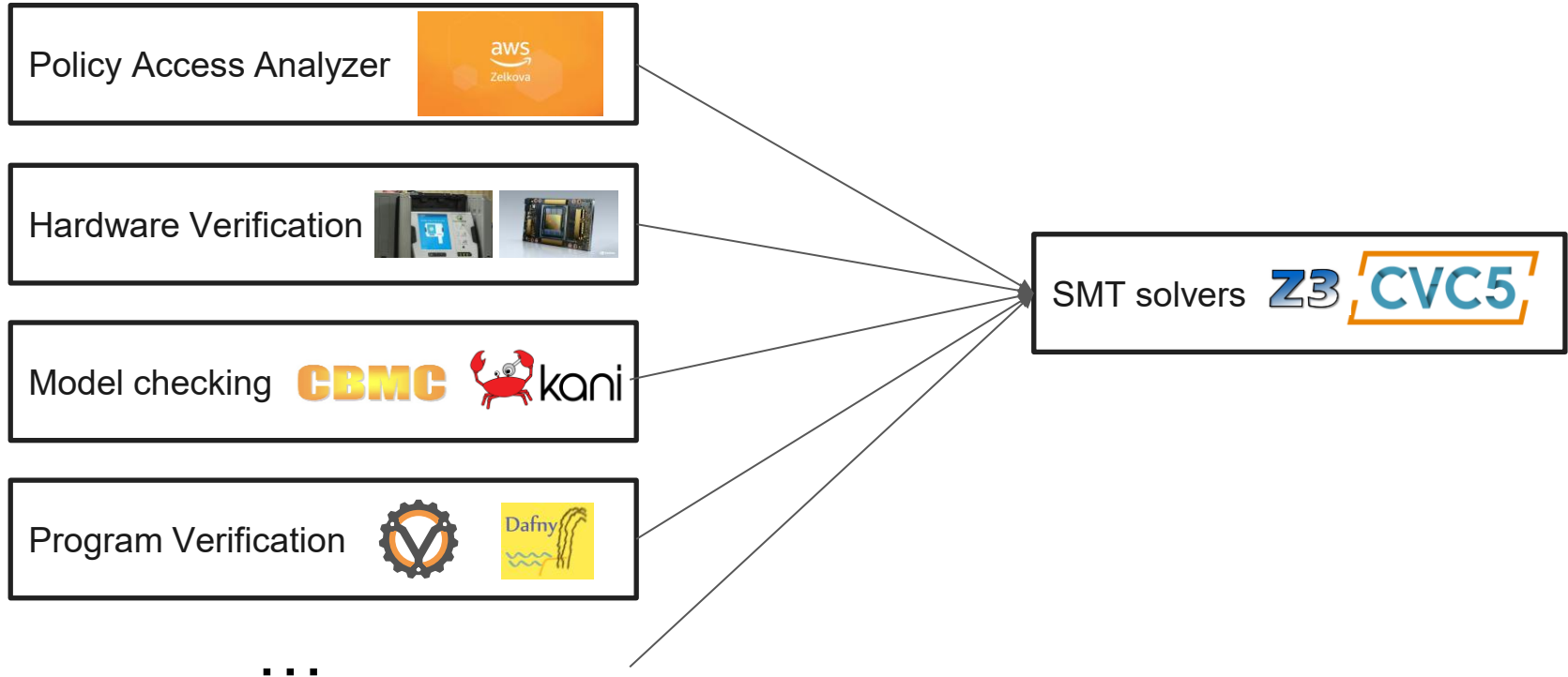
A Billion SMT Queries a Day (Invited Paper)

(roughly the same number as ChatGPT)



- How did we get here?
 - “Just because a problem is undecidable, it doesn’t stop being important”
 - Modelling solvers after the problems we want to solve
 - Combination of multiple approaches: eager and lazy
- Where do we go next?
 - Find the right problems
 - A trillion SMT queries a day?

Many Applications



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Formally: multi-sorted first-order logic

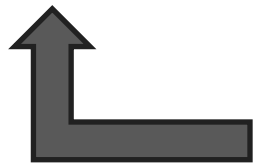


- Can define a set of sorts S for example $S = \{Bool, Int, Real, String\}$
 - Can have user-defined (uninterpreted) sorts
- A signature Σ with:
 - Function symbols $\Sigma_F = \{f : \sigma_1 \times \dots \times \sigma_l \rightarrow \sigma, \dots\}$
 - Predicate Symbols $\Sigma_R = \{P : \sigma_1 \times \dots \times \sigma_k\}$
 - $\sigma_i \in S$
- Function symbols with arity 0 are called constants
- A set of variables V
- From these build up a set of terms, atomic formulas, and formulas
- A theory is a set of formulas closed under logical deduction
- *Bool* and $=$ are part of all theories

Example: Uninterpreted Functions (UF)



```
(declare-sort T 0)
(declare-fun x () T)
(declare-fun f (T) T)
(assert (= (f (f (f x))) x))
(assert (= (f (f (f (f (f x))))) x))
(assert (not (= (f x) x)))
```



This is SMT syntax. Sorry!

- Also called the “free theory”
- Because function symbols can take any meaning
- Only property required is *congruence*: that functions map identical arguments to identical values i.e., $x = y \Rightarrow f(x) = f(y)$
- $S = \{Bool, T, \dots\}$ where T is a user defined sorts
- $\Sigma_F = \{f, g, \dots\}$ where f, g are user defined functions

Example: Linear Integer Arithmetic



```
(declare-fun x () Int)
(declare-fun y () Int)
(declare-fun z () Int)
(assert (or (<= (+ x y) 10)
            (= x z)))
(assert (<= (- (+ x x) y) 5))
```

- $S = \{Bool, Int\}$
- $\Sigma_F = \{0, 1, +, -\}$
- $\Sigma_P = \{\leq, =\}$
- Can express larger integers and $\geq, <, >$
- Boolean combination of linear constraints of the form $(a_1 x_1 + a_2 x_2 + \dots + a_n x_n \sim b)$
- There are also nonlinear and real arithmetic theories

Example: (Real) Integer Difference Logic



```
(declare-fun x () Int)
(declare-fun y () Int)
(declare-fun z () Int)
(assert (or (>= (- x y) 10)
            (= x 2)))
(assert (<= (- x z) 5))
```

- $S = \{Bool, Int\}$
- $\Sigma_F = \{0, 1, -\}$
- $\Sigma_P = \{\leq, =\}$
- Boolean combination of linear constraints of the form
$$x_i - x_j \sim c_{ij} \quad \text{or} \quad x_i \sim c_i$$
- Applications
 - Processor datapath verification
 - Job shop scheduling / real-time systems
 - Timing verification for circuits

Example: Bitvectors



```
(declare-fun x () (_ BitVec 8))
```

```
(declare-fun y () (_ BitVec 8))
```

```
(declare-fun z () (_ BitVec 8))
```

```
(assert (= x #b10101010))
```

```
(assert (= z (bvand x y)))
```

```
(assert (= (bvadd x y) #xff))
```

- $S = \{Bool, BV1, BV2, BV3 \dots\}$
- Fixed width data words
 - Can model int, short, long, .
- Arithmetic operations
 - add/subtract/multiply/divide & comparisons
 - Two's complement and unsigned operations
- Bit-wise logical operations
 - E.g., and/or/xor, shift/extract and equality
- Boolean connectives

Example: Arrays



```
(declare-fun arr () (Array Int Int))  
(declare-fun i () Int)  
(declare-fun j () Int)  
(assert (= (select arr i) 42))  
(assert (= (select (store arr j 10) i) 10))  
(assert (not (= i j)))
```

- Two interpreted functions: select and store
 - select(A,i) -> Read from A at index i
 - store(A,i,d) -> Write d to A at index i
- Two main axioms:
 - for $i \neq j$
- One other axiom:
 $(\forall i. \text{select}(A, i) = \text{select}(B, i)) \rightarrow A = B$

Others



- Strings
- Floating Points
- Algebraic Datatypes
- Higher Order Functions
- Finite Fields

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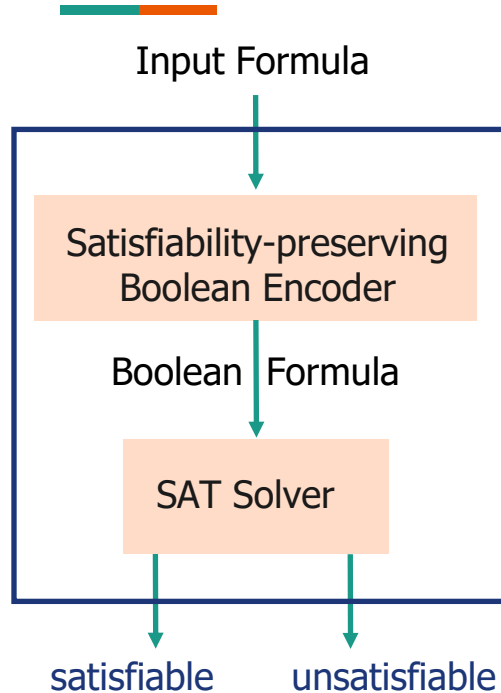
Have to combine theory reasoning with SAT solving



Two main approaches:

1. “Eager”
 - translate into an equisatisfiable propositional formula
 - feed it to any SAT solver
2. “Lazy”
 - abstract the input formula to a propositional one
 - feed it to a (DPLL-based) SAT solver
 - use a theory decision procedure to refine the formula and guide the SAT solver

Eager Approach to SMT



EAGER ENCODING

SAT Solver involved in
Theory Reasoning

Key Ideas:

Small-domain encoding

Constrain model search

Rewrite rules

Abstraction-based methods (eager + lazy)

Example Solvers:

NFA2SAT, Bitwuzla, Algoroba, UCLID, STP

...

Ackerman's Encoding: UF to SAT

$f(x) \longrightarrow fx$

$f(y) \longrightarrow fy$

For each pair of function applications add:

$$x = y \Rightarrow fx = fy$$

```
(declare-sort T 0)
(declare-fun x () T)
(declare-fun f (T) T)
(assert (= (f (f (f x))) x))
(assert (= (f (f (f (f (f x))))) x))
(assert (not (= (f x) x)))
```

What is the Ackerman encoding of this?

Another example: bitvectors

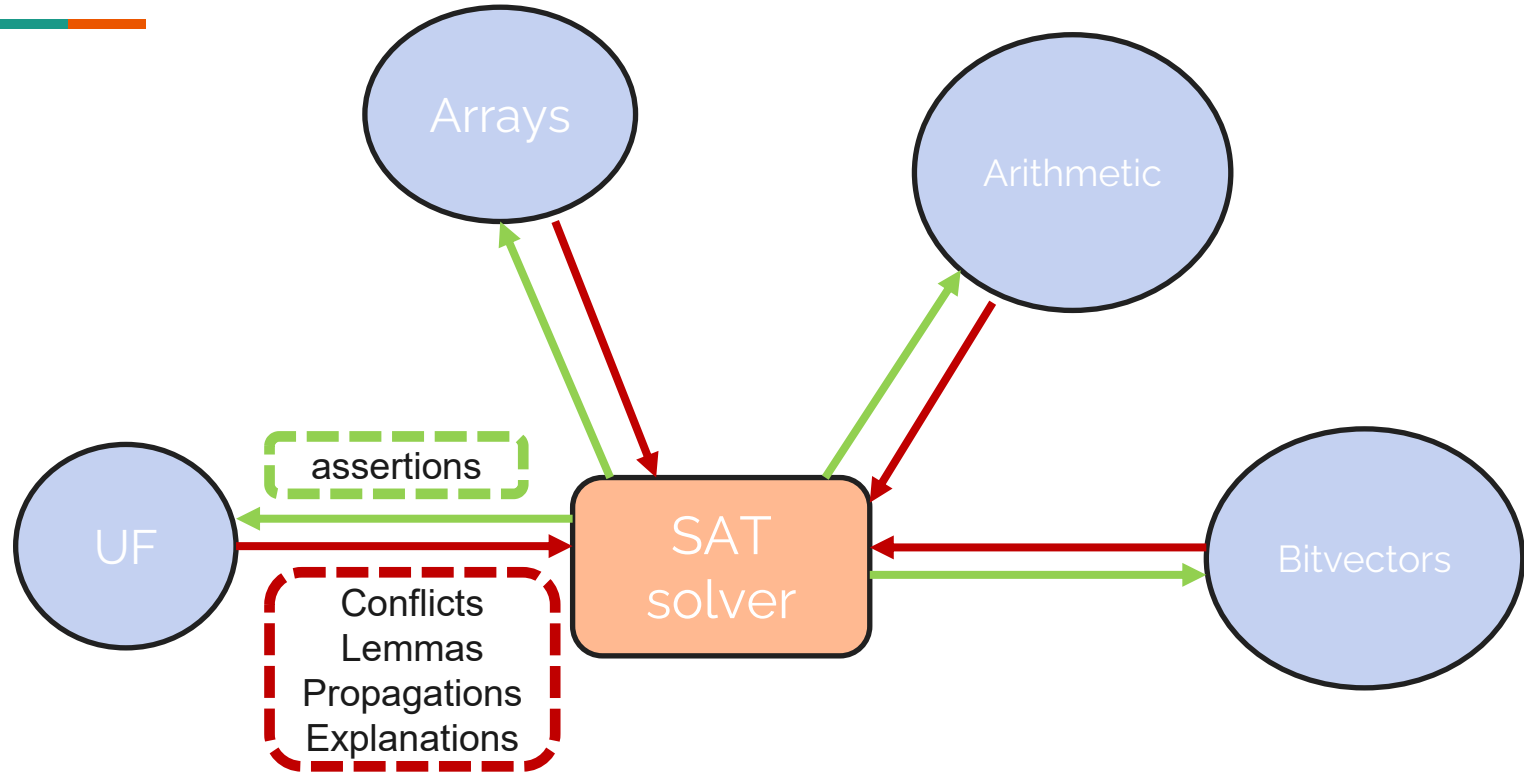


```
(declare-fun x () (_ BitVec 8))  
(declare-fun y () (_ BitVec 8))  
(declare-fun z () (_ BitVec 8))  
(assert (= x #b10101010))  
(assert (= z (bvand x y)))
```

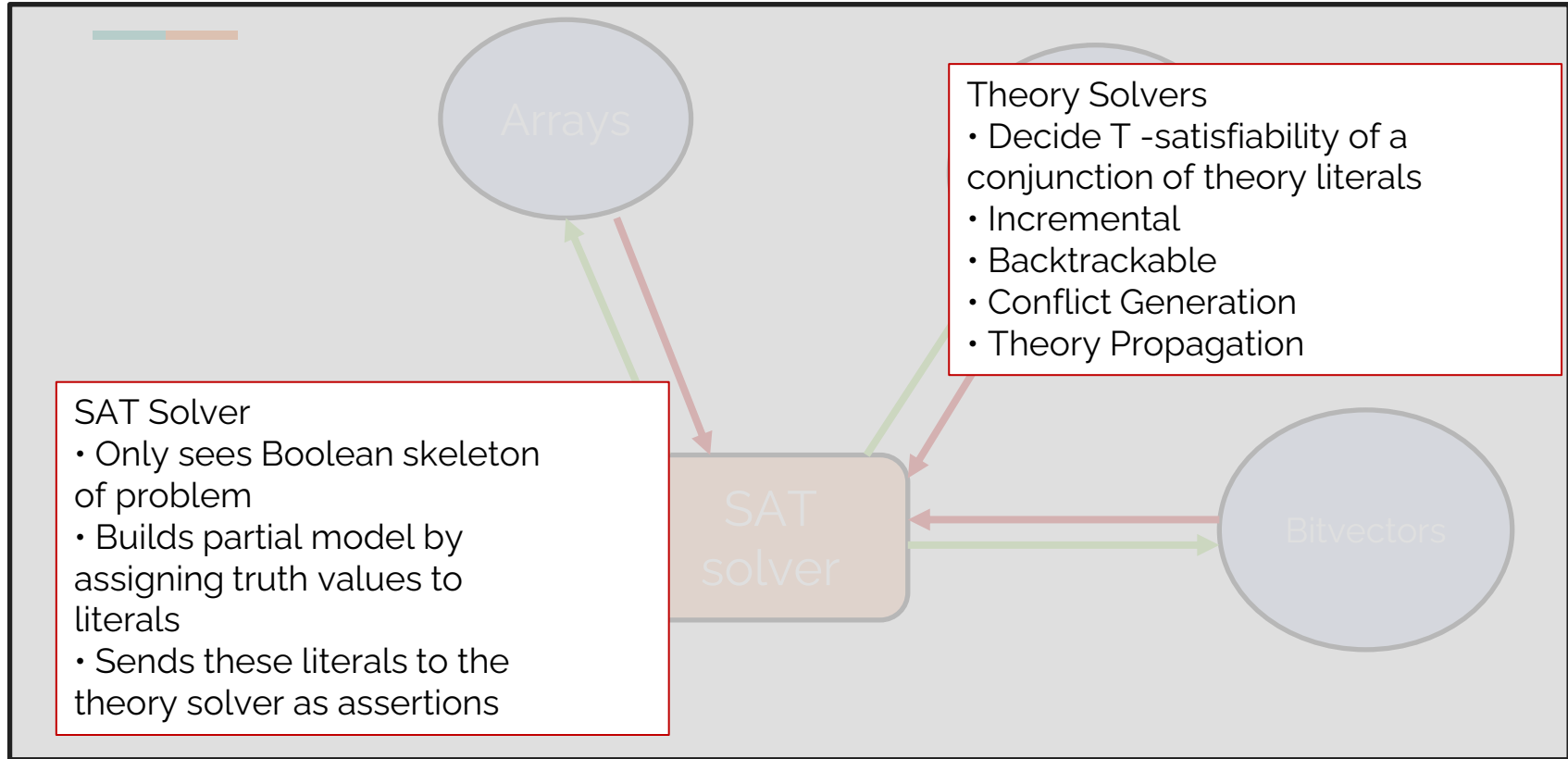
Can we bit-blast this to SAT?

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Lazy SMT solving: often called CDCL(T)

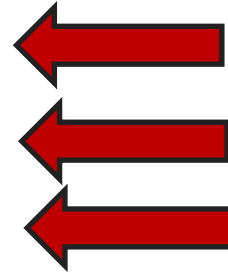


Lazy SMT solving: often called CDCL(T)

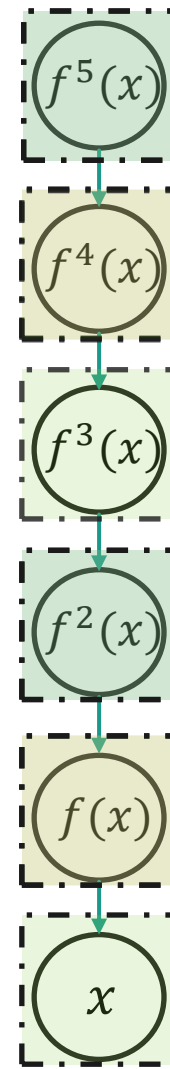


Congruence Closure Algorithm (for UF)

$$\begin{aligned} &x = f(f(f(x))) \wedge \\ &x = f(f(f(f(f(x))))) \wedge \\ &x \neq f(x) \end{aligned}$$



We learn a conflict clause and backtrack!



Notice $x = f^2(x)$

Hence $f(x) = f^3(x)$

Conflict

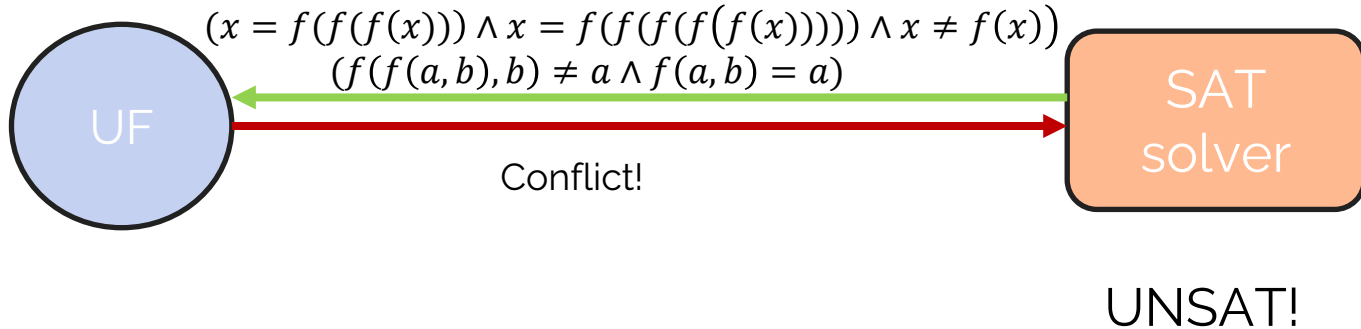
More examples



1. $f(f(a, b), b) \neq a \wedge f(a, b) = a$
2. $f(a, a) = b \wedge g(c, a) = c \wedge g(c, f(a, a)) =$
 $f(g(c, a), g(c, a)) \wedge f(c, c) \neq g(c, b)$
3. $f(a, a) = b \wedge g(c, a) = c \wedge g(c, f(a, a)) =$
 $f(g(c, a), g(c, a)) \wedge f(h(d), d) =$
 $f(g(a, c), g(c, a)) \wedge h(a) = h(c)$

Putting it all together: lazy SMT solving

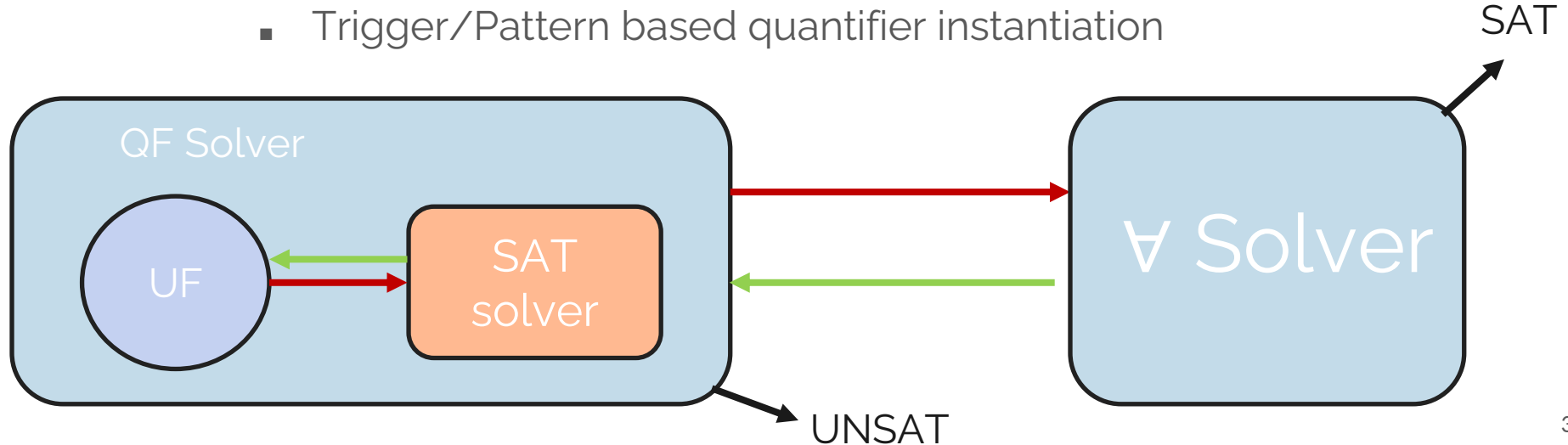
$$\begin{aligned} & [(x = f(f(f(x))) \wedge x = f(f(f(f(f(x)))))) \wedge x \neq f(x)) \\ & \vee (f(f(a, b), b) \neq a \wedge f(a, b) = a)] \\ & \wedge [(x \neq f(f(f(x))) \vee x \neq f(f(f(f(f(x)))))) \vee x = f(x)] \\ & \wedge [f(f(a, b), b) = a \vee f(a, b) \neq a] \end{aligned}$$



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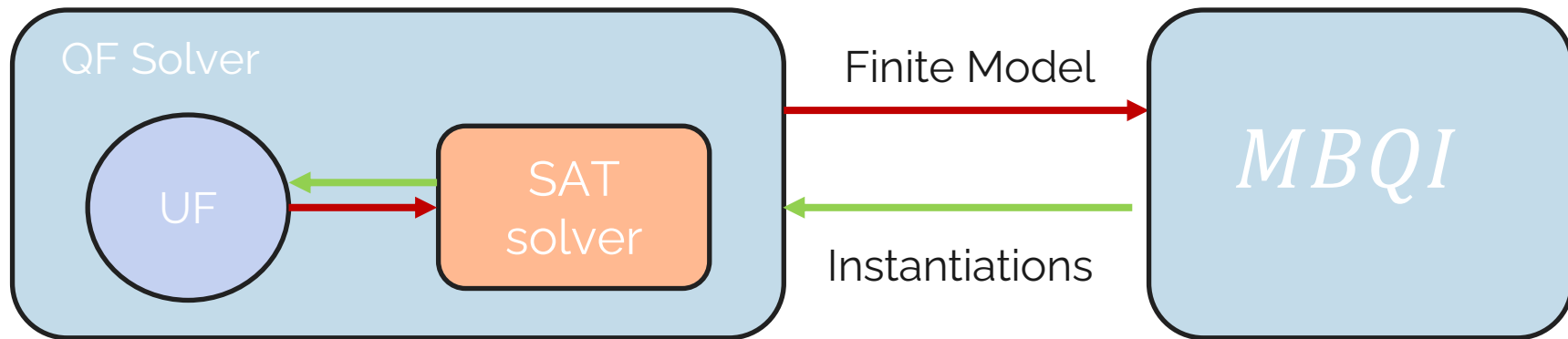
FOL is undecidable: because of quantifiers

- How do we reason about "forall" statements?
- Could enumerate all cases?
- Two smarter approaches?
 - Model-based quantifier instantiation
 - Trigger/Pattern based quantifier instantiation



Model-based quantifier instantiation

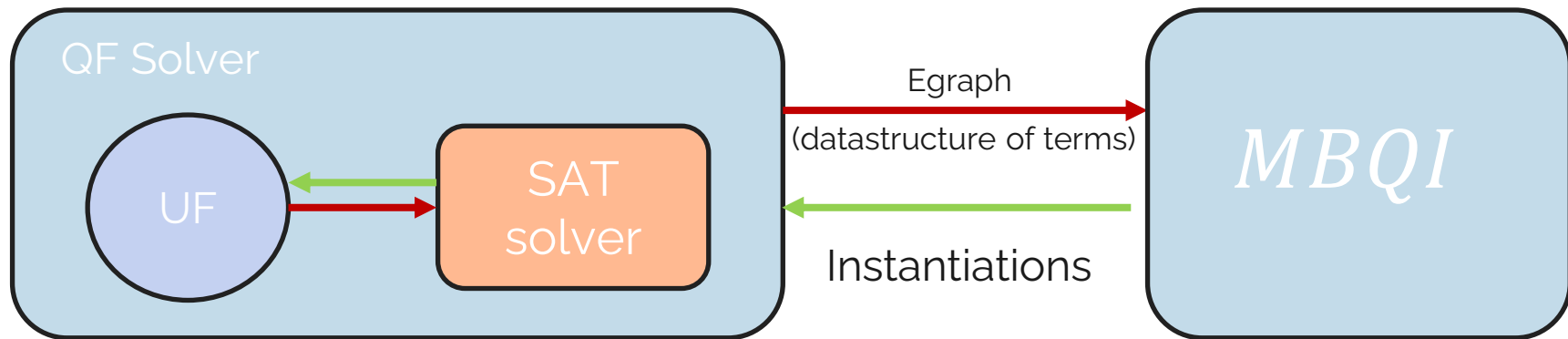
- Idea: Instantiate quantifiers based on (complete) QF models
- Complete for certain fragments, e.g. EPR, essentially uninterpreted
- Can be useful for answering “sat”



Example: $f(t) \wedge g(s) \wedge [\forall x. \neg f(x) \vee \neg g(x)]$

Pattern/Trigger based quantifier instantiation

- Idea: Instantiations found by pattern matching
- Implemented in early SMT solvers (e.g. simplify) as well as z3, cvc5
- Key applications: Software verification (dafny, Verus)



Example: $f(t) \wedge g(s) \wedge [\forall x. f(g(x)) : pattern\ g(x)] \longrightarrow$ Instantiation: $f(g(s))$

Pattern-based Instantiations Example (from Verus)

We have the assertion

```
(assert
  (not (=> (and (is_member x (set_union s1 s2))
                (not (is_member x s1)))
            (is_member x s2))))
```

Trying to instantiate quantifier

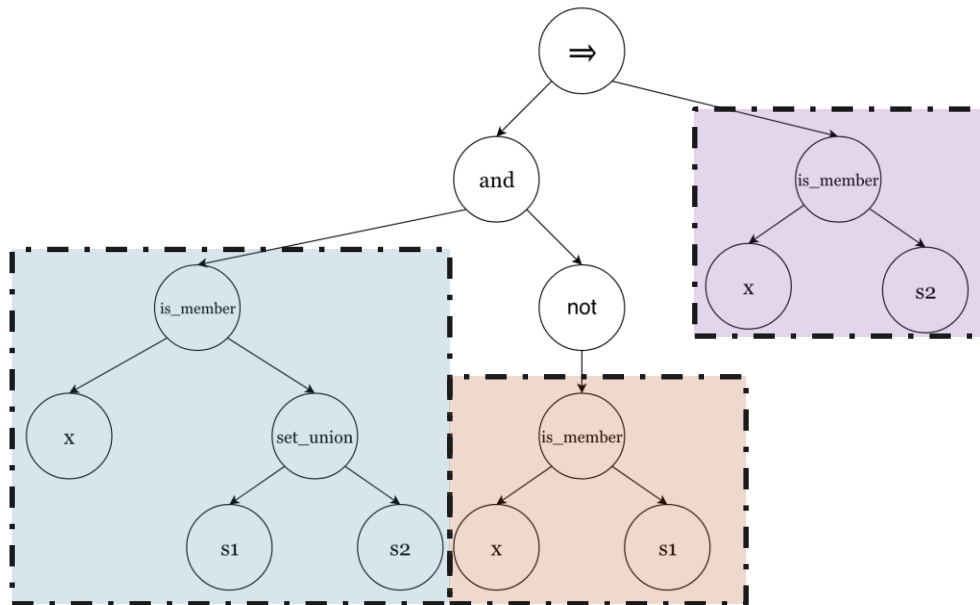
```
(forall ((y Poly) (set Poly)) (!
  (= (is_member y set) (Set.contains set y))
  :pattern ((is_member y set))))
```

Adding the instantiations:

```
(= (is_member x (set_union s1 s2))
   (Set.contains (set_union s1 s2) x))
```

```
(= (is_member x s1) (Set.contains s1 x))
```

```
(= (is_member x s2) (Set.contains s2 x))
```



Other quantifier instantiation techniques

- 
- Enumerative instantiation
 - Counter-example guided instantiation
 - Syntax Guided instantiation

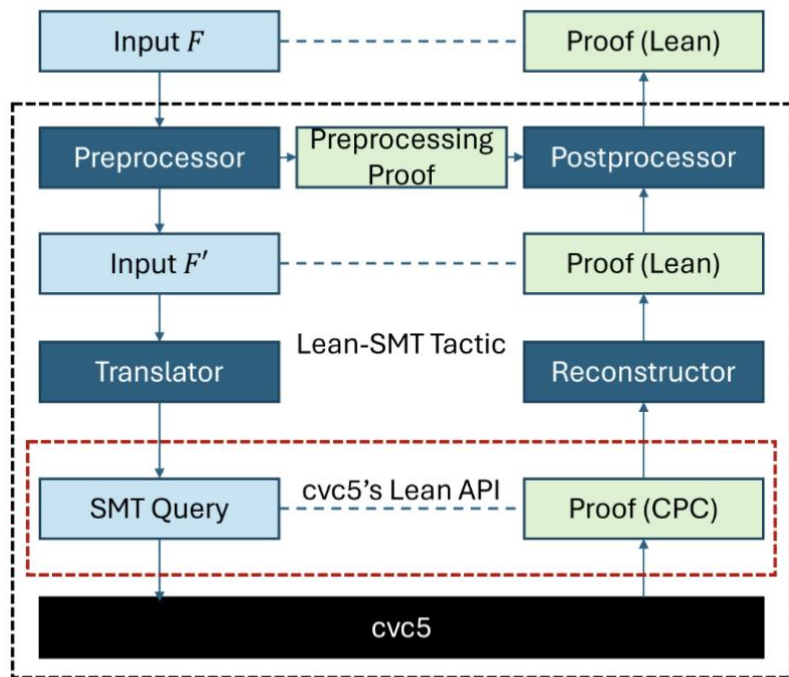
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Where else are SMT solvers useful?



- Disclaimer: SMT landscape is vast and I only gave you a brief taste of it
- Recent Applications in cryptography, programming languages, systems, and machine learning

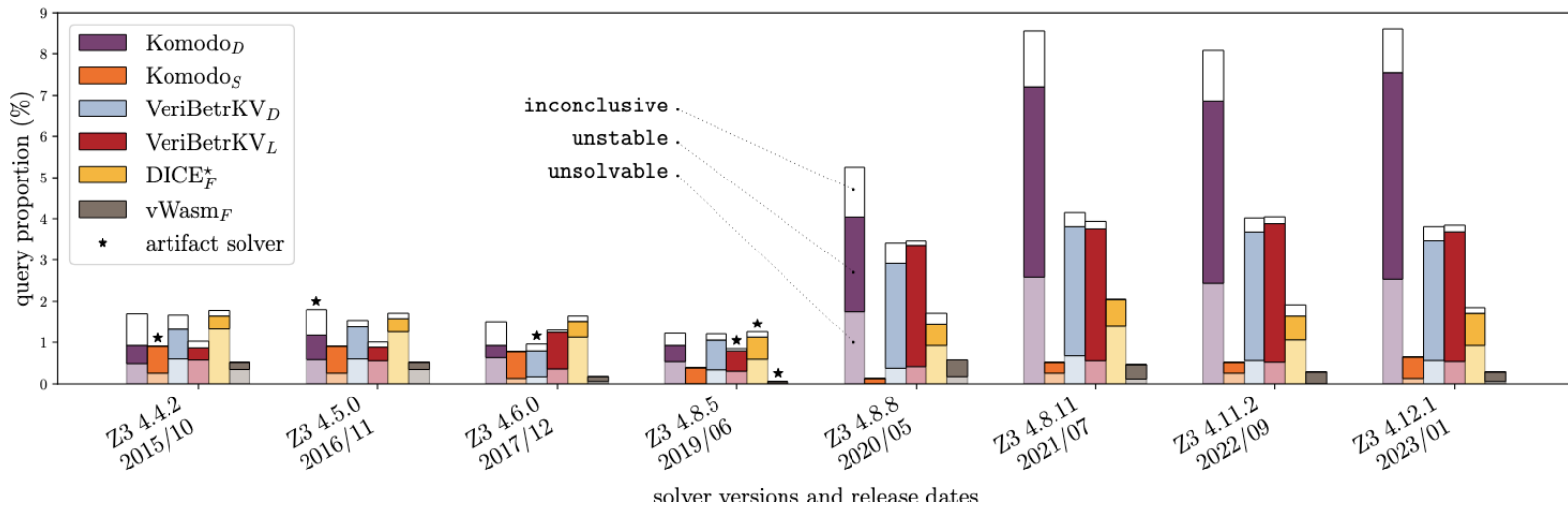
Programming Languages: A Hammer for Lean



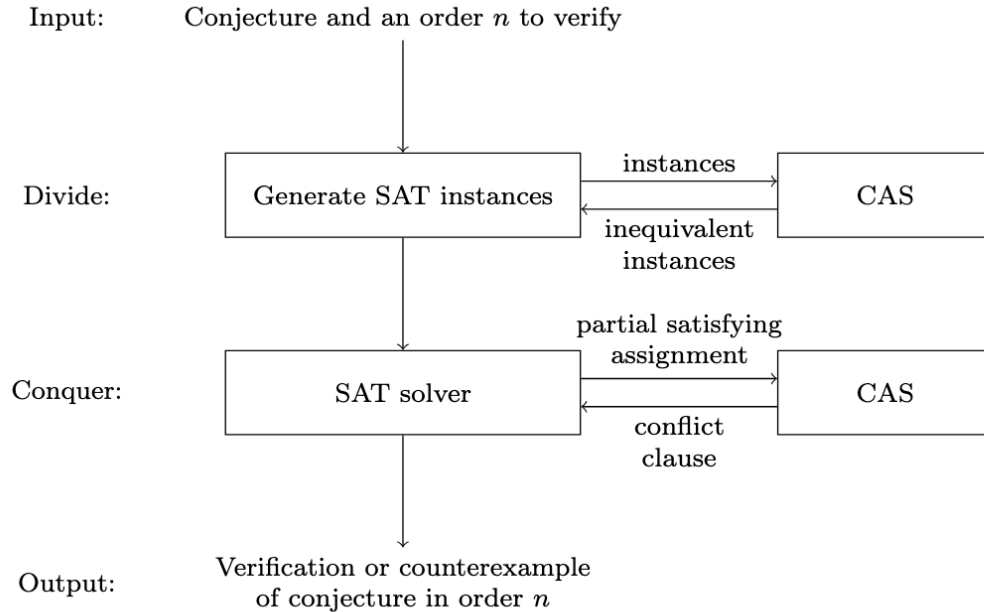
- Lean is a programming language/interactive theorem prover
- Calls tactics
- Two new tactics, `lean-auto` and `lean-smt` call SMT solvers to prove Lean goals

Systems: Reliability/Instability of SMT solvers

- SMT solvers used by systems with O(1 billion) queries
- They need to be robust/reliable



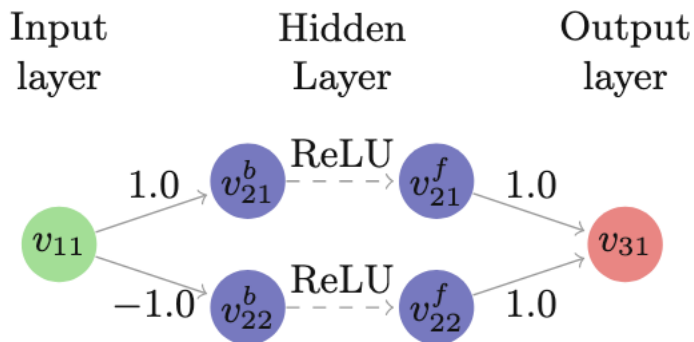
Mathematics: Enumerating matrices



- Combining a SAT solver with a computer algebra system
- Very good with enumerating matrices of certain properties

Machine Learning: Verifying Neural Networks

- Verifying Neural Networks—can mostly be modeled as a linear function
- Special decision procedure to deal with ReLU



Cryptography: Satisfiability modulo Finite Fields

$$x' + y' = \sum_{i=1}^{b+1} 2^{i-1} z'_i \quad \wedge \quad z' = \sum_{i=1}^b 2^{i-1} z'_i \quad \wedge \quad \bigwedge_{i=1}^{b+1} z'_i (z'_i - 1) = 0$$

Used to prove the correctness of a Zero Knowledge Proof compiler!

Definition 1 (Correctness). A ZKP compiler $\text{Compile}(\phi) \rightarrow (\phi', \text{Ext}_x, \text{Ext}_w)$ is **correct** if it is demonstrably complete and demonstrably sound.

- *demonstrable completeness*: For all $x \in \text{dom}(x), w \in \text{dom}(w)$ such that $\hat{\phi}(x, w) = \top$,

$$\hat{\phi}'(\text{Ext}_x(x), \text{Ext}_w(x, w)) = \top$$

- *demonstrable soundness*: There exists an efficient algorithm $\text{Inv}(x', w') \rightarrow w$ such that for all $x \in \text{dom}(x), w' \in \text{dom}(w')$ such that $\hat{\phi}'(\text{Ext}_x(x), w') = \top$,

$$\hat{\phi}(x, \text{Inv}(\text{Ext}_x(x), w')) = \top$$

How can SMT solvers be applied to work that you do?

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