

AN OVERVIEW OF SATISFIABILITY MODULO THEORIES AND ITS APPLICATIONS

ETAPS 2019

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Many thanks to **Clark Barrett**, **Alberto Griggio**, **Liana Hadarean**, **Dejan Jovanovic**, and **Albert Oliveras** for contributing some of the material used in these slides.

Disclaimer: The literature on SMT and its applications is vast. The bibliographic references provided here are just a small and highly incomplete sample. Apologies to all authors whose work is not cited.

Introduction

SMT Solver Functionality

Background Theories

Applications

- Model Checking

- Software Verification

- Synthesis

- Misc

References

INTRODUCTION

Historically:

Automated logical reasoning achieved through **uniform theorem-proving procedures** for **First Order Logic** (e.g., resolution, superposition, and tableaux calculi)

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Automated logical reasoning achieved through **uniform theorem-proving procedures** for **First Order Logic** (e.g., resolution, superposition, and tableaux calculi)

Limited success:

Uniform proof procedure for FOL are not always the best **compromise** between **expressiveness** and **efficiency**

Last 20 years: R&D has focused on

- expressive enough **decidable fragments** of various logics
- incorporating **domain-specific** reasoning, e.g., on:
 - temporal reasoning
 - arithmetic reasoning
 - equality reasoning
 - reasoning about certain data structures
(arrays, lists, finite sets, ...)
- combining specialized reasoners **modularly**

Two successful examples of this trend:

SAT: propositional formalization, Boolean reasoning

+ high degree of efficiency

– expressive (all NP-complete problems) but
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This tutorial: an overview of SMT and its applications

Determining the **satisfiability** of a logical formula **wrt** some combination T of **background theories**

Example

$$n > 3 * m + 1 \wedge (f(n) \leq \text{head}(l_1) \vee l_2 = f(n) :: l_1)$$

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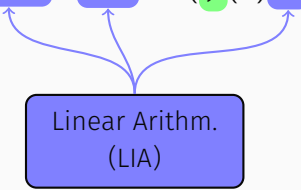
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(LIA)



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SMT formulas are formulas in
many-sorted FOL with **built-in symbols**

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Are **changing the way people solve problems** in Computer Science and beyond:

- instead of building a **special-purpose** tool
- **translate** problem into a logical formula
- use an SMT solver as **backend reasoner**

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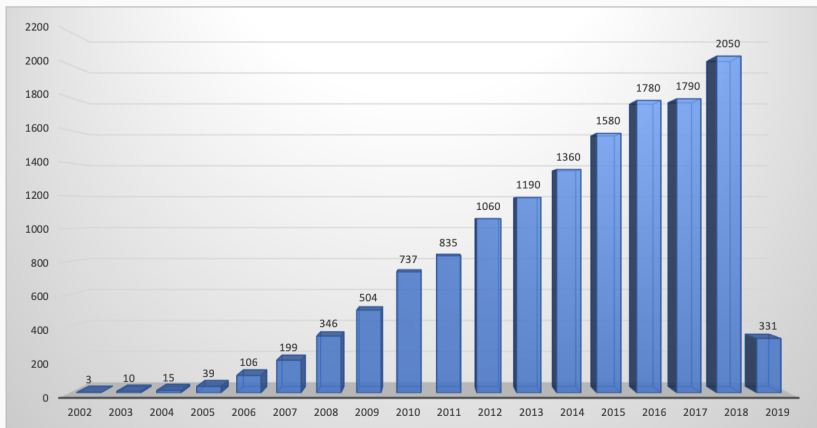
- instead of building a **special-purpose** tool
- **translate** problem into a logical formula
- use an SMT solver as **backend reasoner**

Not only easier, **often better**

THE EXPLOSION OF SMT



"Satisfiability Modulo Theories" OR "SMT Solver"



	Citations	Google Scholar Hits
Z3	5,068 ¹	7,870
CVC Lite, CVC 3, 4	1,560 ²	2,030
Yices 1, 2	972 ³	2,430
MathSat 3, 4, 5	628 ⁴	1,010

¹[DMB08], 2018 ETAPS Test of Time Award to Z3 developers

² [BT07, BT07, BCD⁺11]

³[DdM06b, Dut14]

⁴[BBC⁺05b, BCF⁺08, CGSS13]

Model Checking

- (in)finite-state systems
- hybrid systems
- abstraction refinement
- state invariant generation
- interpolation

Type Checking

- dependent types
- semantic subtyping
- type error localization

Program Analysis

- symbolic execution

- program verification
- verification in separation logic
- (non-)termination
- loop invariant generation
- procedure summaries
- race analysis
- concurrency errors detection

Software Synthesis

- syntax-guided function synthesis
- automated program repair
- synthesis of reactive systems
- synthesis of self-stabilizing systems
- network schedule synthesis

Security

- automated exploit generation
- protocol debugging
- protocol verification
- analysis of access control policies
- run-time monitoring

Compilers

- compilation validation
- optimization of arithmetic computations

Software Engineering

- system model consistency
- design analysis
- test case generation
- verification of ATL transformations
- semantic search for code reuse
- interactive (software) requirements prioritization
- generating instances of meta-models
- behavioral conformance of web services

Planning

- motion planning
- nonlinear PDDL planning

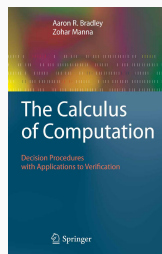
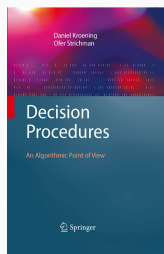
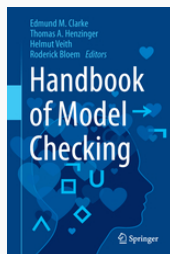
Machine Learning

- verification of deep NNs

Business

- verification of business rules
- spreadsheet debugging

Handbook chapters and books [BSST09, BT18, BM07, KS08]



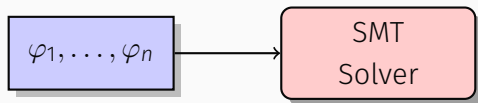
Online

- SMT-LIB at <http://smt-lib.org>
- SMT-COMP at <http://smt-comp.org>

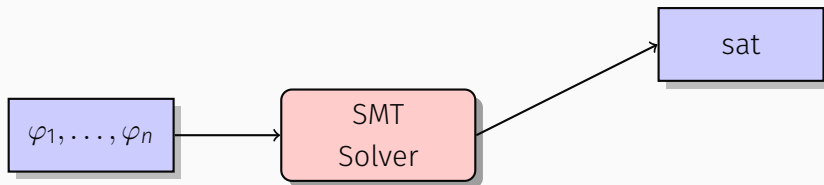
FUNCTIONALITY

- v value — i.e., distinguished variable-free term
- $\varphi[\mathbf{x}]$ formula with free vars from $\mathbf{x} = (x_1, \dots, x_n)$
- $\varphi[\mathbf{x} \mapsto \mathbf{v}]$ formula obtained by replacing free occurrences of variables from \mathbf{x} in φ with corresponding values from $\mathbf{v} = (v_1, \dots, v_n)$
- $\mathbf{x} = \mathbf{v}$ $x_1 = v_1 \wedge \dots \wedge x_n = v_n$
- $\mathbf{z} \subseteq \mathbf{x}$ every element of \mathbf{z} occurs in \mathbf{x}
- $M \models \varphi$ model M satisfies formula φ
- $\varphi \models_T \psi$ formula φ entails formula ψ in theory T

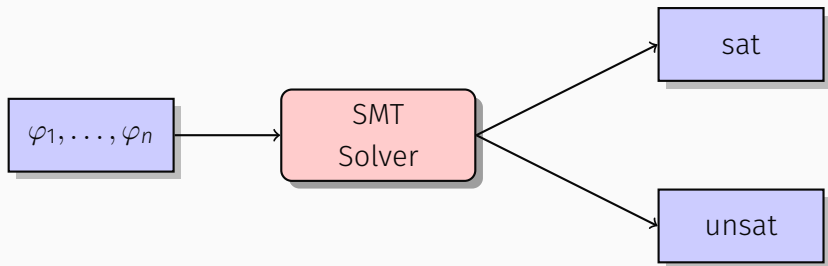
Background theory T



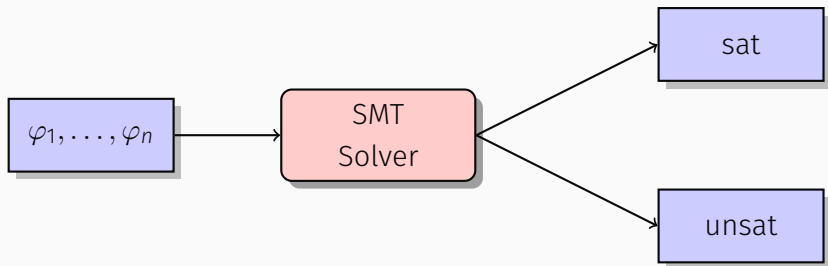
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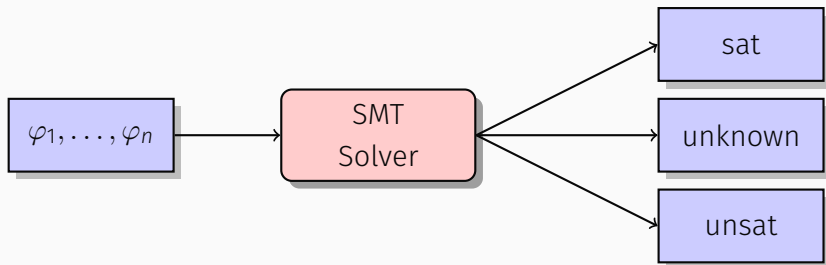
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sat/unsat: there is **a**/no model M of T such that

$$M \models \varphi_1 \wedge \dots \wedge \varphi_n$$

Background theory T

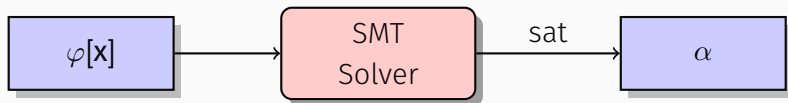


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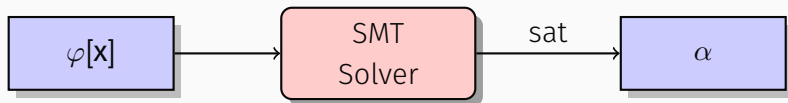
unknown: inconclusive – because of resource limits or incompleteness

Background theory T



α is a *satisfying assignment* for $\mathbf{x} = (x_1, \dots, x_n)$:

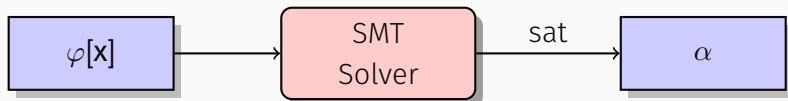
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α is a *satisfying assignment* for $\mathbf{x} = (x_1, \dots, x_n)$:

1. $\alpha = \{x_1 \mapsto v_1, \dots, x_n \mapsto v_n\}$ for some values $\mathbf{v} = (v_1, \dots, v_n)$
2. $M \models \varphi[\mathbf{x} \mapsto \mathbf{v}]$ for some model M of T

Background theory T



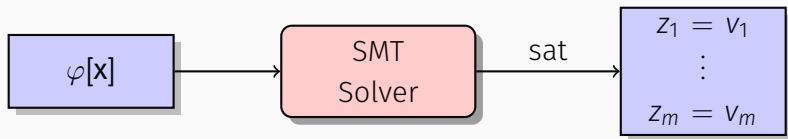
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Note.

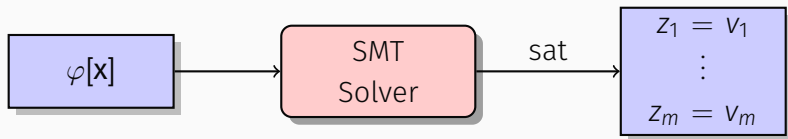
\mathbf{x} may consist of first- and second-order variables
(aka, *uninterpreted constants and function symbols*)

Background theory T



$z = v$ is a *backbone* for φ :

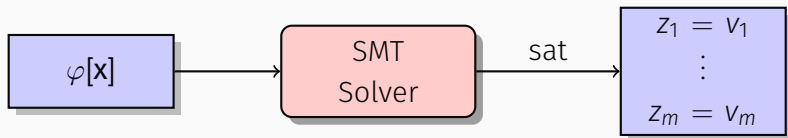
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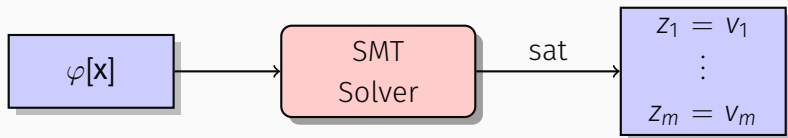
1. $\mathbf{z} \subseteq \mathbf{x}$
2. $\varphi \models_T \mathbf{z} = \mathbf{v}$
3. \mathbf{z} is maximal (or largish)

Background theory T



$z = v$ is a *sat core* for φ :

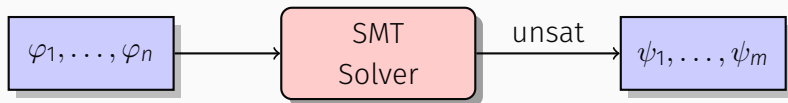
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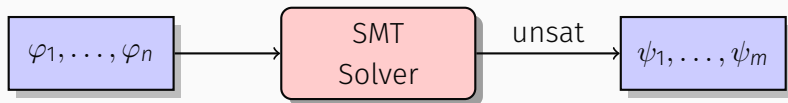
1. $z \subseteq x$
2. $y = x \setminus z$
3. $\forall y (\varphi \wedge z = v)$ is satisfiable in T
4. z is minimal (or smallish)

Background theory T



ψ_1, \dots, ψ_m is a *unsat core* of $\{\varphi_1, \dots, \varphi_n\}$:

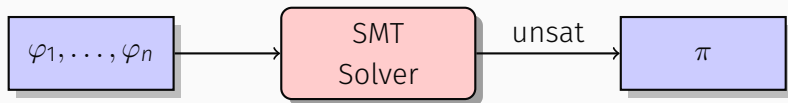
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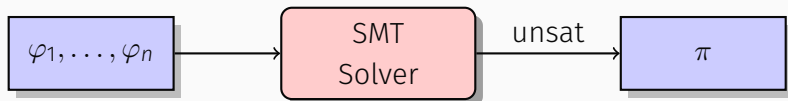
1. $\{\psi_1, \dots, \psi_m\} \subseteq \{\varphi_1, \dots, \varphi_n\}$
2. $\{\psi_1, \dots, \psi_m\}$ is unsat in T
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π is a checkable *proof object* for $\{\varphi_1, \dots, \varphi_n\}$:

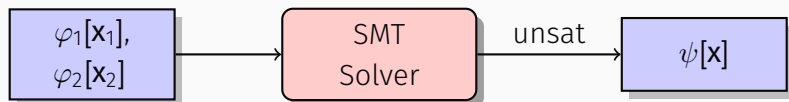
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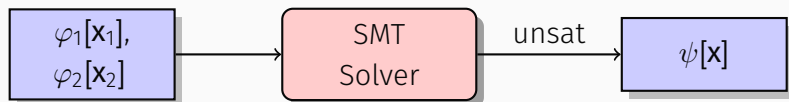
1. π is a proof term in some formal proof system
2. π expresses a *refutation* of $\{\varphi_1, \dots, \varphi_n\}$
3. π can be *efficiently checked* by an external proof checker

Background theory T



ψ is a logical *interpolant* of φ_1 and φ_2 :

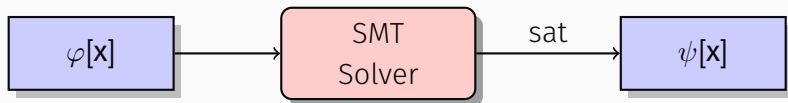
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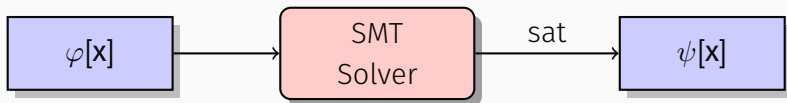
1. $\varphi_1 \models_T \psi$ and $\psi \models_T \neg\varphi_2$
2. $\mathbf{x} = \mathbf{x}_1 \cap \mathbf{x}_2$

Background theory T



ψ is a *prime implicate* of φ :

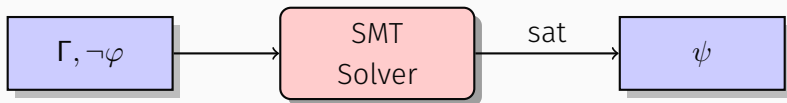
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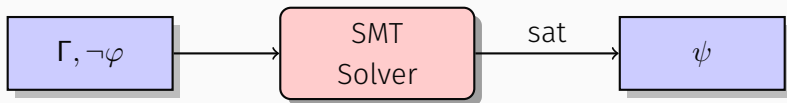
1. ψ is a disjunction of literals
2. $\varphi \models_T \psi$
3. there is no disjunction of literals $\psi' \notin \{\varphi, \psi\}$ s.t.
 $\varphi \models_T \psi'$ and $\psi' \models_T \psi$

Background theory T



ψ is an *abduction hypothesis* for φ wrt Γ :

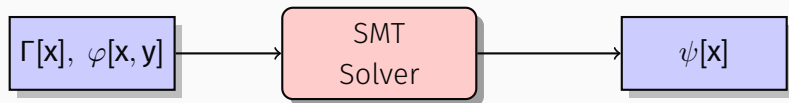
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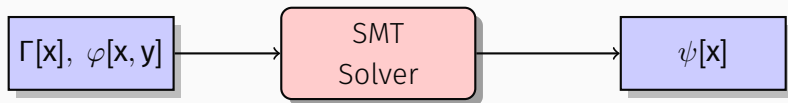
ψ is an *abduction hypothesis* for φ wrt Γ :

1. Γ, ψ is satisfiable in T
2. $\Gamma, \psi \models_T \varphi$
3. ψ is *maximal*, e.g., with respect to \models_T
(if ψ' satisfies 1 and 2 and $\psi \models_T \psi'$ then $\psi' \models_T \psi$)

Background theory T

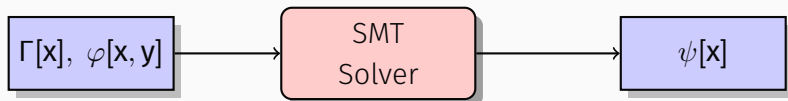


Background theory T



ψ is a *projection* of φ over \mathbf{y} with respect to Γ :

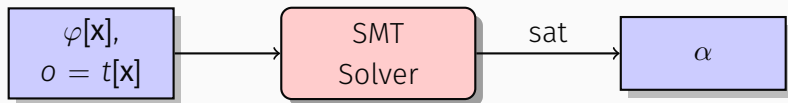
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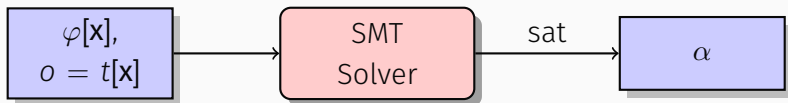
1. $\Gamma \models_T \psi \Leftrightarrow \exists \mathbf{y} \varphi$

Background theory T



α is a *an optimal assignment* for φ :

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α is a *an optimal assignment* for φ :

1. $\alpha = \{x_1 \mapsto v_1, \dots, x_n \mapsto v_n\}$ for some *values* v_1, \dots, v_n
2. $M \models \varphi[\mathbf{x} \mapsto \mathbf{v}]$ for some model M of T
3. α minimizes/maximizes *objective* o

BACKGROUND THEORIES

Uninterpreted Funs	$x = y \Rightarrow f(x) = f(y)$
Integer/Real Arithmetic	$2x + y = 0 \wedge 2x - y = 4 \rightarrow x = 1$
Floating Point Arithmetic	$x + 1 \neq NaN \wedge x < \infty \Rightarrow x + 1 > x$
Bit-vectors	$4 \cdot (x \gg 2) = x \& \sim 3$
Strings and RegExs	$x = y \cdot z \wedge z \in ab^* \Rightarrow x > y $
Arrays	$i = j \Rightarrow \text{store}(a, i, x)[j] = x$
Algebraic Data Types	$x \neq \text{Leaf} \Rightarrow \exists l, r : \text{Tree}(\alpha). \exists a : \alpha.$ $x = \text{Node}(l, a, r)$
Finite Sets	$e_1 \in x \wedge e_2 \in x \setminus e_1 \Rightarrow \exists y, z : \text{Set}(\alpha).$ $ y = z \wedge x = y \cup z \wedge y \neq \emptyset$
Finite Relations	$(x, y) \in r \wedge (y, z) \in r \Rightarrow (x, z) \in r \bowtie s$

Simplest first-order theory with equality, applications of uninterpreted functions, and variables of uninterpreted sorts

For all sorts σ, σ' and function symbols $f: \sigma \rightarrow \sigma'$

Reflexivity: $\forall x: \sigma. x = x$

Symmetry: $\forall x, y: \sigma. x = y \Rightarrow y = x$

Transitivity: $\forall x, y, z: \sigma. x = y \wedge y = z \Rightarrow x = z$

Congruence: $\forall x, y: \sigma. x = y \Rightarrow f(x) = f(y)$

Example

$$f(f(f(a))) = b \quad g(f(a), b) = a \quad f(a) \neq a$$

Operates over sorts $\text{Array}(\sigma_i, \sigma_e)$, σ_i , σ_e and function symbols

$$_[-] : \text{Array}(\sigma_i, \sigma_e) \times \sigma_i \rightarrow \sigma_e$$

$$\text{store} : \text{Array}(\sigma_i, \sigma_e) \times \sigma_i \times \sigma \rightarrow \text{Array}(\sigma_i, \sigma_e)$$

For any index sort σ_i and element sort σ_e

Read-Over-Write-1: $\forall a, i, e. \text{store}(a, i, e)[i] = e$

Read-Over-Write-2: $\forall a, i, j, e. i \neq j \Rightarrow \text{store}(a, i, e)[j] = a[j]$

Extensionality: $\forall a, b, i. a \neq b \Rightarrow \exists i. a[i] \neq b[i]$

Example

$$\text{store}(\text{store}(a, i, a[j]), j, a[i]) = \text{store}(\text{store}(a, j, a[i]), i, a[j])$$

Restricted fragments, over the reals or the integers, support efficient methods:

- Bounds: $x \bowtie k$ with $\bowtie \in \{<, >, \leq, \geq, =\}$ [BBC⁺05a]
- Difference constraints: $x - y \bowtie k$, with $\bowtie \in \{<, >, \leq, \geq, =\}$ [NO05, WIGG05, CM06]
- UTVPI: $\pm x \pm y \bowtie k$, with $\bowtie \in \{<, >, \leq, \geq, =\}$ [LM05]
- Linear arithmetic, e.g: $2x - 3y + 4z \leq 5$ [DdM06a]
- Non-linear arithmetic, e.g:
 $2xy + 4xz^2 - 5y \leq 10$ [BLNM⁺09, ZM10, JdM12]

Family of user-definable theories

Example

$$\text{Color} := \text{red} \mid \text{green} \mid \text{blue}$$

$$\text{List}(\alpha) := \text{nil} \mid (\text{head} : \alpha) :: (\text{tail} : \text{List}(\alpha))$$

Distinctiveness: $\forall h, t. \text{nil} \neq h :: t$

Exhaustiveness: $\forall l. l = \text{nil} \vee \exists h, t. h :: t$

Injectivity: $\forall h_1, h_2, t_1, t_2.$

$$h_1 :: t_1 = h_2 :: t_2 \Rightarrow h_1 = h_2 \wedge t_1 = t_2$$

Selectors: $\forall h, t. \text{head}(h :: t) = h \wedge \text{tail}(h :: t) = t$

(Non-circularity: $\forall l, x_1, \dots, x_n. l \neq x_1 :: \dots :: x_n :: l$)

- Strings and regular expressions [KGG⁺09, LRT⁺14]
- Floating point arithmetic [BDG⁺14, ZWR14]
- Finite sets with cardinality [BRBT16]
- Finite relations [MRTB17]
- Transcendental Functions [GKC13]
- Ordinary differential equations [GKC13]
- ...

APPLICATIONS

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Software Verification

Synthesis

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References

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3. Encode system M as T -formulas $(I[\mathbf{x}], R[\mathbf{x}, \mathbf{x}'])$ where
 - I encodes M 's initial state condition and
 - R encodes M 's transition relation

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 - I encodes M 's initial state condition and
 - R encodes M 's transition relation
4. Encode S as a T -formula $B[\mathbf{x}]$

To check the **reachability** of a class S of *bad states* for a *system model* M :

1. Choose a theory T decided by an SMT solver (e.g., quantifier-free linear arithmetic and EUF)
2. Represent system states as values for a tuple \mathbf{x} of *state vars*
3. Encode system M as T -formulas $(I[\mathbf{x}], R[\mathbf{x}, \mathbf{x}'])$ where
 - I encodes M 's initial state condition and
 - R encodes M 's transition relation
4. Encode S as a T -formula $B[\mathbf{x}]$
5. Find a k such that $I[\mathbf{x}_0] \wedge R[\mathbf{x}_0, \mathbf{x}_1] \wedge \cdots \wedge R[\mathbf{x}_{k-1}, \mathbf{x}_k] \wedge B[\mathbf{x}_k]$ is satisfiable in T

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5. Prove that $P[\mathbf{x}]$ holds in all reachable states of $(I[\mathbf{x}], R[\mathbf{x}, \mathbf{x}'])$

Example (Parametric Resettable Counter)

System

Vars

input pos int, n_0
input bool r
int c, n

Initialization

c := 1
n := n_0

Transitions

$n' := n$
 $c' :=$ if (r' or $c = n$)
 then 1
 else $c + 1$

Property

$c \leq n$

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input pos int, n_0
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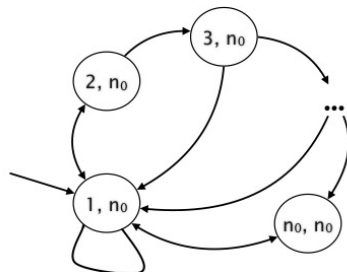
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The transition relation contains infinitely many instances of the schema above, one for each $n_0 > 0$

Example (Parametric Resettable Counter)

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Property

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Encoding in $T = \text{LIA}$

$x := (c, n, r, n_0)$

$I[x] := c = 1$
 $\wedge n = n_0$

$R[x, x'] := n' = n$
 $\wedge (\neg r' \wedge c \neq n \vee c' = 1)$
 $\wedge (r' \vee c = n \vee c' = c + 1)$

$P[x] := c \leq n$

$$M = (I[x], R[x, x'])$$

$$M = (I[x], R[x, x'])$$

To prove $P[x]$ invariant for M it suffices to show that it is *inductive* for M ,
i.e.,

$$(1) \quad I[x] \models_T P[x] \quad (\text{base case})$$

and

$$(2) \quad P[x] \wedge R[x, x'] \models_T P[x'] \quad (\text{inductive step})$$

$M = ($

Problem: Not all invariants are inductive

For the parametric resettable counter,

$P := c \leq n + 1$ is invariant but (2) is falsifiable

e.g., by $(c, n, r) = (4, 3, false)$ and $(c, n, r)' = (5, 3, false)$

To prove
to show
i.e.,

(1) $I[x] \models_T P[x]$ (base case)

and

(2) $P[x] \wedge R[x, x'] \models_T P[x']$ (inductive step)

$$(1) I[x] \models_T P[x]$$

$$(2) P[x] \wedge R[x, x'] \models_T P[x']$$

Various approaches:

$$(1) I[x] \models_T P[x]$$

$$(2) P[x] \wedge R[x, x'] \models_T P[x']$$

Various approaches:

Strengthen P : find a property Q such that $Q[x] \models_T P[x]$ and prove Q inductive

(ex., interpolation-based MC, IC3, CHC)

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Strengthen R : find an auxiliary invariant $Q[x]$ and use $Q[x] \wedge R[x, x'] \wedge Q[x']$ instead of $R[x, x']$

(ex., Houdini, invariant sifting)

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Strengthen R : find an auxiliary invariant $Q[x]$ and use $Q[x] \wedge R[x, x'] \wedge Q[x']$ instead of $R[x, x']$

(ex., Houdini, invariant sifting)

Lengthen R : Consider increasingly longer R -paths

$$R[x_0, x_1] \wedge \cdots \wedge R[x_{k-1}, x_k] \wedge R[x_k, x_{k+1}]$$

(ex., k -induction)

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Example

```
void swap(int* a, int* b) {  
    *a = *a + *b;  
    *b = *a - *b;  
    *a = *a - *b;  
}
```

Check if the swap is correct:

- Heap: $Array(BV_{32}) \mapsto BV_{32}$
- Update heap line by line
- Check that $a^* = \text{old}(b^*)$ and $b^* = \text{old}(a^*)$

Example

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$$h_1 = \text{store}(h_0, a, h_0[a] +_{32} h_0[b])$$

$$h_2 = \text{store}(h_1, b, h_1[a] -_{32} h_1[b])$$

$$h_3 = \text{store}(h_2, a, h_2[a] -_{32} h_2[b])$$

$$\neg(h_3[a] = h_0[b] \wedge h_3[b] = h_0[a])$$

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```
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- Heap: $Array(BV_{32}) \mapsto BV_{32}$
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- Check that
 $a^* = \text{old}(b^*)$ and $b^* = \text{old}(a^*)$
- **Incorrect:** aliasing

SMT solver solution

$a \mapsto 0, \quad b \mapsto 0$
 $h_0[0] \mapsto 1, \quad h_1[0] \mapsto 2$
 $h_2[0] \mapsto 0, \quad h_3[0] \mapsto 0$

$\neg (h_3[a] = h_0[b] \wedge h_3[b] = h_0[a])$

Example (Binary Search)

```

//@assume 0 <= n <= |a| &&
//          foreach i in [0..n-2]. a[i] <= a[i+1]
//@ensure (0 <= res ==> a[res] = k) &&
//          (res < 0 ==> foreach i in [0..n-1]. a[i] != k)
int BinarySearch(int[] a, int n, int k) {
    int l = 0; int h = n;
    while (l < h) { // Find middle value
        //@invariant 0 <= low < high <= len <= |a| &&
        //          foreach i in [0..low-1]. a[i]<k &&
        //          foreach i in [high..len-1]. a[i] > k
        int m = l + (h - l) / 2; int v = a[m];
        if (k < v) { l = m + 1; } else if (v < k) { h = m; }
        else { return m; }
    }
    return -1;
}

```

Example (Binary Search)

Main approach

1. Compile source and annotations to program in Dijkstra's core language:

$$S, T ::= x = t \mid \mathbf{havoc} \ x \mid \mathbf{assert} \ \varphi \mid \mathbf{assume} \ \varphi \mid S; T \mid S \parallel T$$

2. Convert core program to SMT using the weakest liberal precondition transformer wp :

$$\begin{aligned} wp(x = t, \varphi) &= \varphi\{x \mapsto t\} & wp(\mathbf{assert} \ \psi, \varphi) &= \psi \wedge \varphi \\ wp(\mathbf{assume} \ \psi, \varphi) &= \psi \Rightarrow \varphi & wp(\mathbf{havoc} \ x, \varphi) &= \forall x \ \varphi \\ wp(S; T, \varphi) &= wp(S, wp(T, \varphi)) \\ wp(S \parallel T, \varphi) &= wp(S, \varphi) \wedge wp(T, \varphi) \end{aligned}$$

Example (Binary Search)
$$pre = 0 \leq n \leq |a| \wedge \forall i : \text{Int } 0 \leq i \wedge i \leq n - 2 \Rightarrow a[i] \leq a[i + 1]$$
$$post = (0 \leq res \Rightarrow a[res] = k) \wedge$$
$$(res < 0 \Rightarrow \forall i : \text{Int } 0 \leq i \wedge i \leq n - 1 \Rightarrow a[i] \neq k)$$
$$inv = 0 \leq l \wedge l \leq h \wedge h \leq n \wedge n \leq |a| \wedge$$
$$\forall i : \text{Int } 0 \leq i \wedge i \leq l - 1 \Rightarrow a[i] < k \wedge$$
$$\forall i : \text{Int } h \leq i \wedge i \leq n - 1 \Rightarrow a[i] > k$$

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$$pre \wedge \neg \text{let } l = 0, h = n \text{ in } inv \wedge \forall : \text{Int } l, h. inv \Rightarrow$$

$$(\neg(l < h) \Rightarrow post\{res \mapsto -1\}) \wedge$$

$$(l < h \Rightarrow \text{let } m = l + (h - l)/2, v = a[m] \text{ in}$$

$$(k < v \Rightarrow inv\{l \mapsto m + 1\}) \wedge$$

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$$inv = 0 \leq l \wedge l \leq h \wedge h \leq n \wedge n \leq |a| \wedge$$

$$\forall i : \text{Int } 0 \leq i < l$$

$$\forall i : \text{Int } h < i < n$$

SMT solver answer: **Unsatisfiable**

$$pre \wedge \neg \text{let } l = 0, h = n \text{ in } inv \wedge \forall : \text{Int } l, h. inv \Rightarrow$$

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Synthesis

- Synthesize a function that satisfies a given high-level specification
- Already used extensively for hardware systems
- Particularly challenging for software

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Recent interest

- Major new efforts by several research groups
- New **syntax-guided** synthesis (SyGuS) format
- SyGuS competition started in 2014
- New technique: **Refutation-Based Synthesis in SMT** [RDK+15]

Formalization in **second**-order logic

- Let $P[f, \mathbf{x}]$ be a property (specification) for a function f over some variables $\mathbf{x} = (x_1, x_2)$
- The synthesis problem is to determine the satisfiability of

$$\exists f. \forall \mathbf{x}. P[f, \mathbf{x}]$$

Formalization in **second-order** logic

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Example

Maximum of 2 values

$$P[f, \mathbf{x}] = f(\mathbf{x}) \geq x_1 \wedge f(\mathbf{x}) \geq x_2 \wedge (f(\mathbf{x}) = x_1 \vee f(\mathbf{x}) = x_2)$$

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Problem: SMT only understands first-order logic

Single-invocation properties

- Every occurrence of f is of the form $f(\mathbf{x})$
 - Previous example is single-invocation
 - Not single-invocation: $\forall \mathbf{x}. f(x_1, x_2) = f(x_2, x_1)$
- When the synthesis property is single-invocation, it can be written as $\exists f. \forall \mathbf{x}. P[f(\mathbf{x}), \mathbf{x}]$

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- When the synthesis property is single-invocation, it can be written as $\exists f. \forall \mathbf{x}. P[f(\mathbf{x}), \mathbf{x}]$

Note that:

$$\exists f. \forall \mathbf{x}. P[f(\mathbf{x}), \mathbf{x}] \tag{1}$$

is equivalent to

$$\forall \mathbf{x}. \exists y P[y, \mathbf{x}] \tag{2}$$

because (1) is the Skolemization of (2) which is first-order!

Proving the validity of

$$\forall x. \exists y P[y, x]$$

Proving the validity of

$$\forall x. \exists y P[y, x]$$

is equivalent to proving the unsatisfiability of

$$\exists x. \forall y \neg P[y, x]$$

or the unsatisfiability of

$$\forall y \neg P[y, c]$$

for some fresh constants c

How does an SMT solver show that

$\forall y \neg P[y, c]$ is unsatisfiable?

How does an SMT solver determine that

$\forall y \neg P[y, \mathbf{c}]$ is unsatisfiable?

SMT solvers use **heuristic instantiation** [GBT07, GdM09, RTGK13] to produce a set of unsatisfiable **quantifier-free** formulas:

$\{\neg P[t_1[\mathbf{c}], \mathbf{c}], \neg P[t_2[\mathbf{c}], \mathbf{c}], \dots, \neg P[t_n[\mathbf{c}], \mathbf{c}]\}$

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This also gives a **constructive** solution to the original synthesis problem:

$f = \lambda \mathbf{x}. \text{ite}(P[t_1[\mathbf{x}], \mathbf{x}], t_1[\mathbf{x}], (\dots \text{ite}(P[t_{n-1}[\mathbf{x}], \mathbf{x}], t_{n-1}[\mathbf{x}], t_n[\mathbf{x}]) \dots))$

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Example

Schedule n jobs, each composed of m consecutive tasks, on m machines.

Schedule in 8 time slots

$d_{i,j}$	Mach. 1	Mach. 2
Job 1	2	1
Job 2	3	1
Job 3	2	3

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Schedule in 8 time slots

$d_{i,j}$	Mach. 1	Mach. 2
Job 1	2	1
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Job 3	2	3

$$(t_{1,1} \geq 0) \wedge (t_{1,2} \geq t_{1,1} + 2) \wedge (t_{1,2} + 1 \leq 8)$$

$$(t_{2,1} \geq 0) \wedge (t_{2,2} \geq t_{2,1} + 3) \wedge (t_{2,2} + 1 \leq 8)$$

$$(t_{3,1} \geq 0) \wedge (t_{3,2} \geq t_{3,1} + 2) \wedge (t_{3,2} + 3 \leq 8)$$

$$((t_{1,1} \geq t_{2,1} + 3) \vee (t_{2,1} \geq t_{1,1} + 2))$$

$$((t_{1,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{1,1} + 2))$$

$$((t_{2,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{2,1} + 3))$$

$$((t_{1,2} \geq t_{2,2} + 1) \vee (t_{2,2} \geq t_{1,2} + 1))$$

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SMT solver solution

$$t_{1,1} \mapsto 5, \quad t_{1,2} \mapsto 7$$

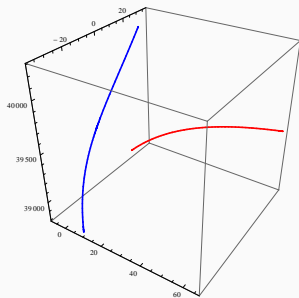
$$t_{2,1} \mapsto 2, \quad t_{2,2} \mapsto 6$$

$$t_{3,1} \mapsto 0, \quad t_{3,2} \mapsto 3$$

$$((t_{1,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{1,2} + 1))$$

$$((t_{2,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{2,2} + 1))$$

AIRCRAFT TRAJECTORY CONFLICT DETECTION



$$H = 5 \text{ nm} \quad V = 1000 \text{ ft} \quad 0 \leq t \leq \frac{1}{20} h$$

$$|T_1^z(t) - T_2^z(t)| \leq V$$

$$(T_1^x(t) - T_2^x(t))^2 + (T_1^y(t) - T_2^y(t))^2 \leq H^2$$

$$T_1^x(t) = 3.2484 + 270.7t + 433.12t^2 - 324.83999t^3$$

$$T_1^y(t) = 15.1592 + 108.28t + 121.2736t^2 - 649.67999t^3$$

$$T_1^z(t) = 38980.8 + 5414t - 21656t^2 + 32484t^3$$

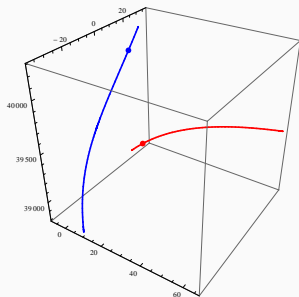
$$T_2^x(t) = 1.0828 - 135.35t + 234.9676t^2 + 3248.4t^3$$

$$T_2^y(t) = 18.40759 - 230.6364t - 121.2736t^2 - 649.67999t^3$$

$$T_2^z(t) = 40280.15999 - 10828t + 24061.9816t^2 - 32484t^3$$

Example from [NM12]

AIRCRAFT TRAJECTORY CONFLICT DETECTION



$$H = 5 \text{ nm} \quad V = 1000 \text{ ft} \quad 0 \leq t \leq \frac{1}{20} \text{ h}$$

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$$(T_1^x(t) - T_2^x(t))^2 + (T_1^y(t) - T_2^y(t))^2 \leq H^2$$

$$\begin{aligned} T_1^x(t) &= 1.0828 - 135.35t + 234.9676t^2 + 3248.4t^3 \\ T_1^y(t) &= 18.40759 - 230.6364t - 121.2736t^2 - 649.67999t^3 \\ T_1^z(t) &= 40280.15999 - 10828t + 24061.9816t^2 - 32484t^3 \end{aligned}$$

SMT solver solution

$$t \mapsto \frac{319}{16384} \approx 0.019470215$$

$$T_2^x(t) = 1.0828 - 135.35t + 234.9676t^2 + 3248.4t^3$$

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