

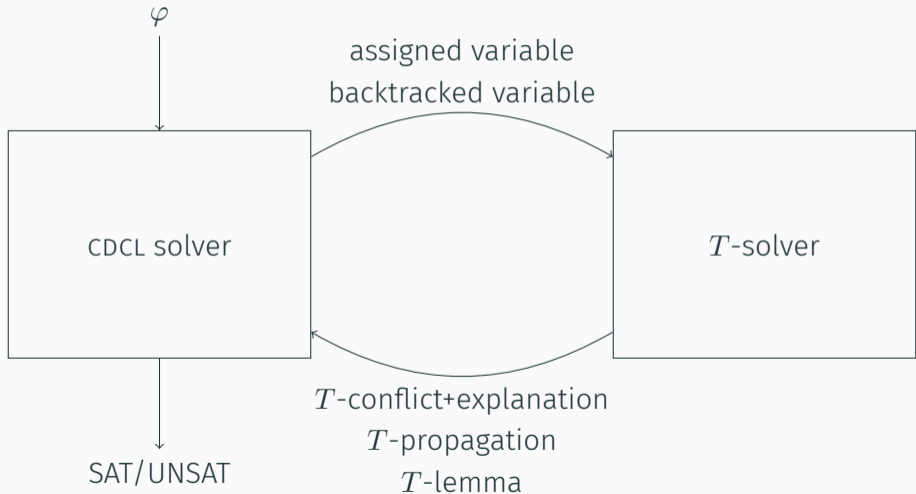
# Selected Algorithms for Theory Solvers

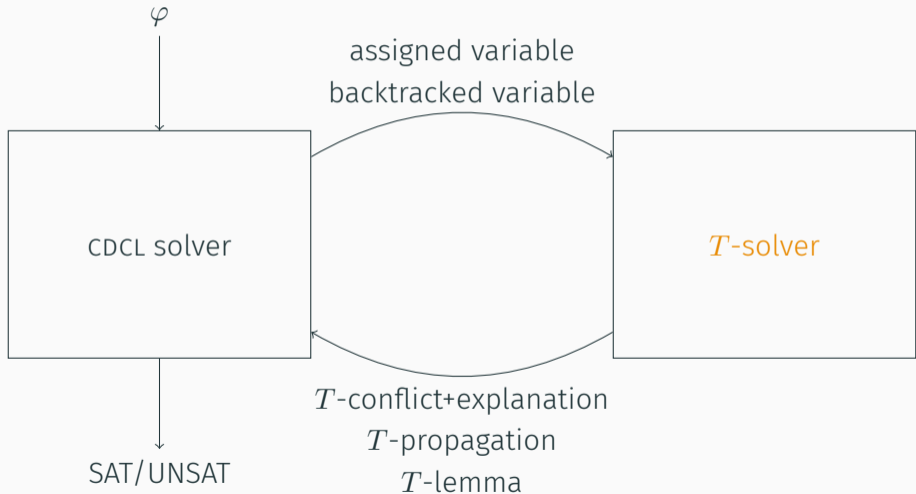
IA085: Satisfiability and Automated Reasoning

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FI MUNI, Spring 2024





### Possible interface of the $T$ -solver

- `void notifyAtom(lit)`
- `void assignLiteral(lit)`
- `void push()`
- `void pop()`
- `result checkSat()`
- `option<result> checkSat_approx()`
- `list<lit> getConflictReason()`
- `option<lit> getPropagatedLiteral()`
- `list<lit> getExplanation(lit)`

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## Equality and uninterpreted functions

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# Theory of equality and uninterpreted functions

## Signature

- a countable set of function symbols  $\Sigma^f = \{f, g, h, \dots\}$
- single predicate symbol  $\Sigma^p = \{=\}$

## Theory

- $T_{UF}$  is a set of **all  $\Sigma$ -structures**

## Idea

- we only know that  $=$  is **equivalence** and
- the symbols in  $\Sigma^f$  are interpreted as functions  $\rightarrow$  for equal arguments return equal results  $\rightarrow$  equality is **congruence**

$$f(x, y) = x \wedge f(f(x, y), y) = z \wedge g(x) \neq g(z)$$



# Applications of uninterpreted functions

- verification of software and hardware (represent unknown external components/functions)
- abstraction of complex parts of the system
- SMT-solving – overapproximation; cheaper unsatisfiability check
- ...

## Flashbacks from Mathematical Foundations of Computer Science

- each equivalence  $\sim \subseteq X^2$  partitions  $X$  to a set of **equivalence** classes  $X/\sim$
- the equivalence class of  $x \in X$  is denoted  $[x]_\sim$

## Flashbacks from Algebra

- given a set of functions  $\Sigma^f$ , an equivalence  $\sim$  is **congruence** if for any  $f \in \Sigma^f$  of arity  $k$  and  $x_i, y_i \in X$

$$(x_1 \sim y_1) \wedge (x_2 \sim y_2) \wedge \dots \wedge (x_k \sim y_k) \implies f(x_1, x_2, \dots, x_k) \sim f(y_1, y_2, \dots, y_k)$$

## Herbrand universe

- a set of **all terms** over the signature  $\Sigma^f$
- denoted  $\mathcal{T}$
- can be turned to  $\Sigma^f$ -structure: all functions work syntactically  $f$  applied to  $t_1$  and  $t_2$  returns  $f(t_1, t_2)$

## Theorem

A set of equalities  $\mathcal{E}$  and disequalities  $\mathcal{D}$  is  $T_{\text{UF}}$  satisfiable if and only if there exists a congruence  $\sim$  on  $\mathcal{T}$  such that

- $t_l \sim t_r$  for all  $(t_l = t_r) \in \mathcal{E}$ , and
- $t_l \not\sim t_r$  for all  $(t_l \neq t_r) \in \mathcal{D}$ .

## Proof.

- “ $\Rightarrow$ ”: Set  $t_1 \sim t_2$  iff  $\llbracket t_1 \rrbracket = \llbracket t_2 \rrbracket$  in the model.
- “ $\Leftarrow$ ”: Construct model from the equality classes.



## Idea

- terms not occurring in the formula  $\varphi$  are not relevant to its satisfiability
- define a set  $\mathcal{T}_\varphi$  of all **subterms** of  $\varphi$
- **compute equivalence classes only of  $\mathcal{T}_\varphi$**

## Example

Given  $\varphi = f(x, y) = x \wedge f(f(x, y), y) = z \wedge g(x) \neq g(z)$

$$\mathcal{T}_\varphi = \{x, y, z, f(x, y), g(x), g(z), f(f(x, y), y)\}$$

## Congruence closure

Congruence closure of  $R \subseteq X^2$

- the smallest congruence  $R^*$  that contains  $R$
- alternatively:  $R^* = \bigcap_{S \text{ is congruence, } R \subseteq S} S$

# Congruence closure

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- alternatively:  $R^* = \bigcap_{S \text{ is congruence, } R \subseteq S} \{S\}$

## Computing congruence closure

Compute the least fixed point of  $R_i$  defined by

$$R_0 = R$$

$$R_{i+1} = R_i \cup \text{id}_X \cup$$

$$\{(x, y) \in X^2 \mid (y, x) \in R_i\} \cup$$

$$\{(x, z) \in X^2 \mid (x, y) \in R_i, (y, z) \in R_i\} \cup$$

$$\{(f(x_1, \dots, x_k), f(y_1, \dots, y_k)) \in X^2 \mid (x_j, y_j) \in R_i \text{ for all } 1 \leq j \leq \text{Ar}(f)\}$$

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Does it have to terminate?



**Theorem**  
*A formula*

$$\varphi = \bigwedge \mathcal{E} \wedge \bigwedge \mathcal{D},$$

where  $\mathcal{E}$  is a set of equalities and  $\mathcal{D}$  is a set of disequalities, is  $T_{\text{UF}}$  satisfiable if and only if the congruence closure of  $\{(t_l, t_r) \mid (t_l = t_r) \in \mathcal{E}\}$  over  $\mathcal{T}_\varphi$  does not contain any  $(t_l, t_r)$  such that  $(t_l \neq t_r) \in \mathcal{D}$ .

# Congruence closure algorithm: theoretical description

## Algorithm

1. start with singleton equivalence classes  $\{t\}$  for each  $t \in \mathcal{T}_\varphi$
2. for each  $(t_l = t_r) \in \mathcal{E}$ , merge the equivalence classes  $[t_l]$  and  $[t_r]$  and all classes that need to be merged due to congruence
3. if at any point  $[t_l] = [t_r]$  for some  $(t_l \neq t_r) \in \mathcal{D}$ , return unsat
4. otherwise return sat

## Congruence closure: example

$$f(x, y) = x \wedge f(f(x, y), y) = z \wedge g(x) \neq g(z)$$

## Questions

1. how to represent the equivalence classes?
2. how to merge the equivalence classes?
3. how to decide if two terms are in the same equivalence class?
4. we have  $\mathcal{O}(|\varphi|)$  subterms, each of size  $\mathcal{O}(|\varphi|)$ ; do we really need  $\mathcal{O}(|\varphi|^2)$  memory to store the set  $\mathcal{T}_\varphi$ ?

## Union-Find

- a data structure to store **disjoint sets**
- allows creating a singleton sets, merging two sets into one, and computing a representative of a given set
- internally represented by a forest:
  - set = tree in the forest
  - representative = root of a tree
- each element stores its **parent** and **rank**

## Reminder: Union-Find

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```
1 make_singleton_set(value) {  
2     return { value: value; parent = value; rank: 1}  
3 }
```

---

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```
1 find(item) {  
2     repr ← item  
3     while (repr ≠ repr.parent) {  
4         repr = repr.parent  
5     }  
6     return repr  
7 }
```

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## Reminder: Union-Find

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```
1 union(item1, item2) {
2     repr1 ← find(item1)
3     repr2 ← find(item2)
4     if (repr1 = repr2) return
5
6     if (repr1.rank > repr2.rank) {
7         repr2.parent = repr1
8     } else if (repr1.rank < repr2.rank) {
9         repr1.parent = repr2
10    } else {
11        repr1.parent = repr2
12        repr2.rank++
13    }
14 }
```

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Efficient storage of terms with shared subterms.

## Nodes

- constant/variable with 0 children
- function symbol  $f$  of arity  $k$  with  $k$  children

## Example

Consider  $f(x, y) = x \wedge f(f(x, y), y) = z \wedge g(x) \neq g(z)$ .



Efficient storage of terms with shared subterms.

## Nodes

- constant/variable with 0 children
- function symbol  $f$  of arity  $k$  with  $k$  children

## Example

Consider  $f(x, y) = x \wedge f(f(x, y), y) = z \wedge g(x) \neq g(z)$ .

We extend each node with parent pointer and rank to store equivalence classes of terms à la union-find.

## Asserting new literals

```
1 def assert( $t = s$ ):
2     todo  $\leftarrow [(t, s)]$ 
3     while todo not empty:
4         ( $u, v$ )  $\leftarrow$  todo.pop()
5         if find( $u$ ) = find( $v$ ): continue
6         union( $u, v$ )
7         foreach  $f(u_1, \dots, u_k)$  and  $f(v_1, \dots, v_k)$  such that
8              $u_i = u, v_i = v$  for some  $i$  and
9             find( $u_j$ ) = find( $v_j$ ) for all  $j$ 
10            find( $f(u_1, \dots, u_k)$ )  $\neq$  find( $f(v_1, \dots, v_k)$ ):
11                todo.append( $(f(u_1, \dots, u_k), f(v_1, \dots, v_k))$ )
12    foreach  $(v \neq w) \in$  inequalities:
13        if find( $v$ ) = find( $w$ ): return false
14    return true
15
16 def assert( $t \neq s$ ):
17     if find( $t$ ) = find( $s$ ): return false
18     inequalities.append( $t \neq s$ )
19     return true
```

# Computing explanations

## Idea

- add explanations to each parent pointer (= edge of the E-graph)
- if  $\mathbf{find}(u) = \mathbf{find}(v)$ , the explanation is union of
  - sequence of explanations between  $u$  and the root  $\mathbf{find}(u)$  and
  - sequence of explanations between  $v$  and the root  $\mathbf{find}(v)$ .

## Each union

- of  $t$  and  $s$  due to  $\mathbf{assert}(t = s)$ 
  - explanation = the equality
- of  $f(u_1, \dots, u_n)$  and  $f(v_1, \dots, v_n)$  due to  $\mathbf{find}(u_i) = \mathbf{find}(v_i)$  for all  $1 \leq i \leq n$ 
  - explanation = the union of explanations of all  $\mathbf{find}(u_i) = \mathbf{find}(v_i)$

# Theory propagation

## Notation

- $t \in [s]$  =  $t$  is in the same subtree as  $s$

## After each union `merge`( $t, s$ )

- propagate  $t' = s'$ , where  $t' \in [t]$  and  $s' \in [s]$
- propagate  $t' \neq r$ , where  $t' \in [t]$ ,  $r' \in [r]$  and there exists  $s' \in [s]$  with  $s' \neq r' \in \text{inequalities}$

## After assertion of $t \neq s$

- propagate  $t' \neq s'$ , where  $t' \in [t]$  and  $s' \in [s]$

Implemented in most of the existing SMT solvers.

For efficient implementation and description of backtracking, see

- R. Nieuwenhuis, A. Oliveras: **Fast congruence closure and extensions**, 2007

## Difference logic

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## Difference logic

- all atoms of form  $(x - y) \bowtie k$  for  $\bowtie \in \{\leq, <, \geq, >, =, \neq\}$  and a number  $k$
- can be over any numeric theory:
  - DL( $\mathbb{Q}$ )
  - DL( $\mathbb{Z}$ )

# Applications of difference logic

- planning
- scheduling
- verification of timed automata
- ...

$$a_{end} - a_{start} \geq 10 \wedge$$

$$b_{start} - a_{end} \geq 0 \wedge$$

$$b_{end} - b_{start} \geq 5 \wedge$$

$$b_{end} - a_{start} \leq 13$$



# Normalization

Atoms can be normalized to  $x - y \leq k$

- $x - y \geq k \rightsquigarrow y - x \leq -k$
- $x - y < k \rightsquigarrow x - y \leq k'$  with  $k'$  a smaller number than  $k$  (theory-dependent)
- $x - y > k \rightsquigarrow x - y \geq k'$  with  $k'$  a bigger number than  $k$  (theory-dependent)
- $x - y = k \rightsquigarrow (x - y \leq k) \wedge (x - y \geq k)$
- $x - y \neq k \rightsquigarrow (x - y < k) \vee (x - y > k)$  (needs to be done in **the original formula**)

Need theory solver only for  $\varphi = \bigwedge (x_i - y_j \leq k_j)$

## Example

$$(x - y \leq 3) \wedge (y - z \leq -11) \wedge (x - z \leq -1) \wedge \\ (v - y \leq 15) \wedge (z - v \leq 5) \wedge (v - x \leq 2)$$

## Example

$$(x - y \leq 3) \wedge (y - z \leq -11) \wedge (x - z \leq -1) \wedge \\ (v - y \leq 15) \wedge (z - v \leq 5) \wedge (v - x \leq 2)$$

## Example

$$(x - y \leq 3) \wedge (y - z \leq -7) \wedge (x - z \leq -1) \wedge \\ (v - y \leq 15) \wedge (z - v \leq 5) \wedge (v - x \leq 2)$$

# Constraint graph

Given a formula  $\varphi = \bigwedge(x_i - y_j \leq k_j)$ , we can construct a **constraint graph**  $G_\varphi$ .

## Nodes

- variables of  $\varphi$

## Edges

- edge between  $x$  and  $y$  of weight  $k$  for each conjunct  $(x - y \leq k)$  of  $\varphi$

### Theorem

*The formula  $\varphi = \bigwedge (x_i - y_j \leq k_j)$  is DL-satisfiable if and only if  $G_\varphi$  does not contain negative cycle.*

## Theorem

The formula  $\varphi = \bigwedge (x_i - y_j \leq k_j)$  is DL-satisfiable if and only if  $G_\varphi$  does not contain negative cycle.

## Proof.

- “ $\Rightarrow$ ”: Show by induction that if there is a path between  $x$  and  $y$  in  $G_\varphi$  of weight  $k$ , then  $\varphi \models_{\text{DL}} (x - y) \leq k$
- “ $\Leftarrow$ ”: Construct a model from shortest paths.



## Algorithm

1. Construct the graph  $G_\varphi$ .
2. Add a new node  $s$  with edges of weight 0 to all nodes of  $G_\varphi$
3. Run Bellman-Ford algorithm from  $s$ .
4. If the algorithm finds negative cycle, return unsat; otherwise return sat.

## Idea

- edges of  $G_\varphi$  = conjuncts  $\varphi$
- unsatisfiability reason = cycle of negative weight
- unsatisfiability explanation = conjuncts on the cycle



## Idea

1. Compute (or maintain) shortest paths between all pairs of vertices  $x$  and  $y$
2. If  $\text{dist}(x, y) = d$ , propagate all  $x - y \leq k$  with  $k \geq d$

## Efficient implementation

Implemented in most of the existing SMT solvers that deal with arithmetic.

For efficient implementation and description of backtracking and theory propagation, see

- A. Armando, C. Castellini, E. Giunchiglia, M. Maratea: **A SAT-Based Decision Procedure for the Boolean Combination of Difference Constraints**, SAT 2004
- S. Cotton, O. Maler: **Fast and Flexible Difference Constraint Propagation for DPLL(T)**, SAT 2006

## Other theories (quick overview)

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## Normalization

- atoms of form  $a_1x_1 + a_2x_2 + \dots + a_kx_k \leq b$

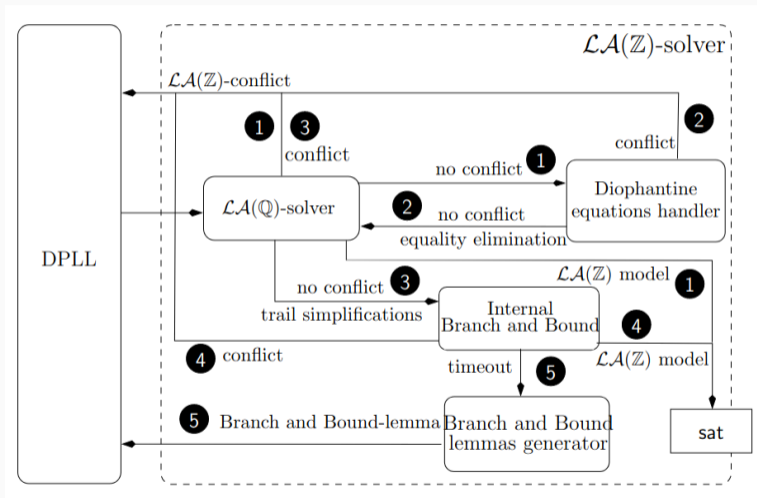
## Theory solver

- decide satisfiability **conjunctions** of atoms of form  $a_1x_1 + a_2x_2 + \dots + a_kx_k \leq b$  and their negations
- **simplex algorithm**
- needs changes to be incremental and backtrackable, see
  - B. Dutetre, L. de Moura: **A Fast Linear-Arithmetic Solver for DPLL(T)**, CAV 2006

Much more complicated. Combination of:

- simplex on the **LRA** relaxation of the formula
- branch and bound
- cutting planes
- diophantine equation solving
- ...

# Linear Integer Arithmetic



[from A.Griggio: A Practical Approach to Satisfiability Modulo Linear Integer Arithmetic]

Combinations of

1. heavy preprocessing
2. converting all operations to Boolean circuits that compute them (usually **and-inverter graphs**)
3. more preprocessing
4. computing **propositional** formula that encodes the circuit
5. often done **eagerly**

The conversion of bit-vector formula to the equisatisfiable propositional formula is called **bit-blasting**.

## Lazy approach

1. treat `read` and `write` as **uninterpreted functions**
2. check UF-satisfiability
3. if unsat, return unsat
4. if sat, check whether the model satisfies array axioms
  - if it does, return sat
  - if not, add the violated axioms and check UF-satisfiability again



## Next time

- combination of theories
- Nelson-Oppen algorithm