

FOUNDATIONS OF SATISFIABILITY MODULO THEORIES

SC² Summer School

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The University of Iowa

Many thanks to

- Clark Barrett
- Dejan Jovanovic
- 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 sample. Apologies to all authors whose work is not cited.

INTRODUCTION

Historically, automated reasoning \equiv uniform proof-search
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Limited success: is FOL the best compromise between expressivity and efficiency?

Most recently R&D has focused on:

- addressing mostly (expressive enough) decidable fragments of a certain logic
- incorporating domain-specific reasoning, e.g on:
 - arithmetic reasoning
 - equality
 - data structures (arrays, lists, stacks, ...)

Examples of this trend:

SAT: propositional formalization, Boolean reasoning

- + high degree of efficiency
- expressive (all NP-complete problems) but involved encodings

SMT: first-order formalization, Boolean + domain-specific reasoning

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These lectures: an overview of SMT and its formal foundations

Some problems are more naturally expressed in logics other than propositional logic, e.g:

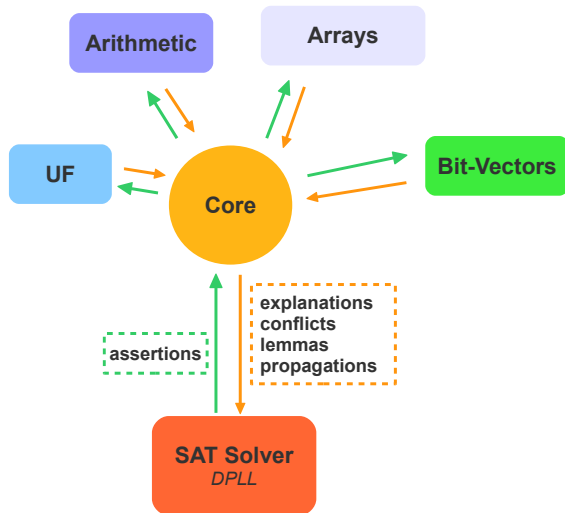
- Software verification needs reasoning about *equality*, *arithmetic*, *data structures*, ...

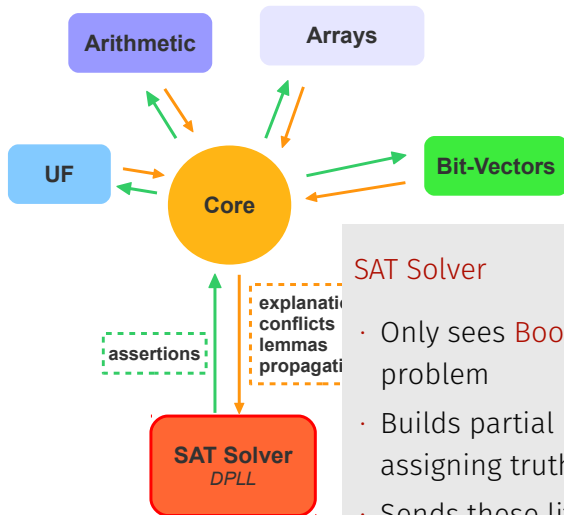
SMT is about deciding the satisfiability of a (usually quantifier-free) FOL formula with respect to some *background theory*

- Example (Equality with Uninterpreted Functions):

$$g(a) = c \quad \wedge \quad (f(g(a)) \neq f(c) \vee g(a) = d) \quad \wedge \quad c \neq d$$

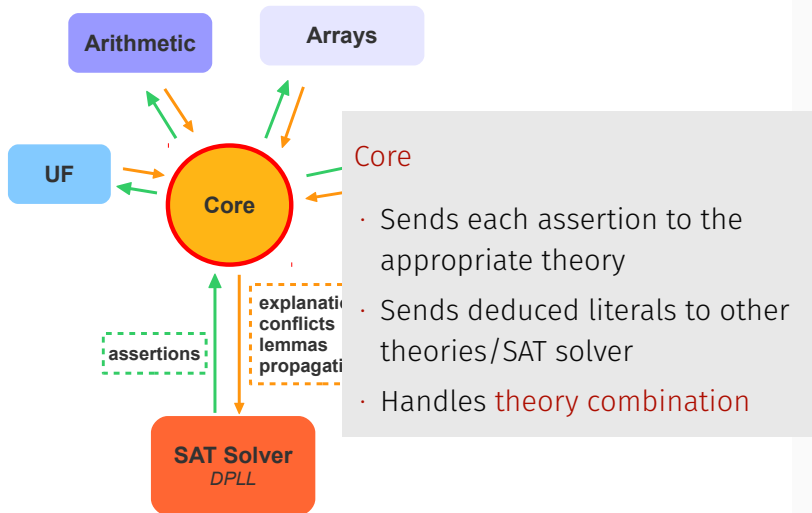
Wide range of *applications*: Extended Static Checking [FLL⁺02], Predicate abstraction [LNO06], Model checking [AMP06, HT08], Scheduling [BNO⁺08b], Test generation [TdH08], ...

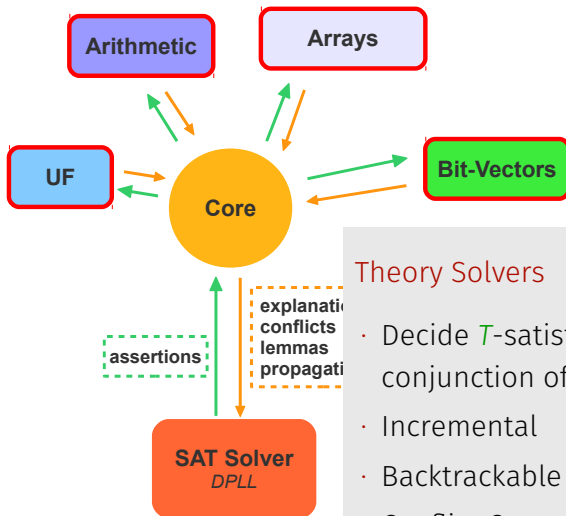




SAT Solver

- Only sees **Boolean skeleton** of problem
- Builds partial model by assigning truth values to literals
- Sends these literals to the core as **assertions**





Theory Solvers

- Decide T -satisfiability of a conjunction of theory literals
- Incremental
- Backtrackable
- Conflict Generation
- Theory Propagation

THEORIES

Equality ($=$) with Uninterpreted Functions [NO80, BD94, NO07a]

Typically used to abstract unsupported constructs, e.g.:

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- ALUs in circuits

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$$a * (|b| + c) = d \wedge b * (|a| + c) \neq d \wedge a = b$$

is **unsatisfiable**, but no arithmetic reasoning is needed

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is unsatisfiable, but no arithmetic reasoning is needed

If we abstract it to

$$\text{mul}(a, \text{add}(\text{abs}(b), c)) = d \wedge \text{mul}(b, \text{add}(\text{abs}(a), c)) \neq d \wedge a = b$$

it is still unsatisfiable

Very useful, for obvious reasons

Restricted fragments (over the reals or the integers) support more efficient methods:

- Bounds: $x \bowtie k$ with $\bowtie \in \{<, >, \leq, \geq, =\}$ [BBC⁺05a]
- Difference logic: $x - y \bowtie k$, 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]

Used in software verification and hardware verification (for memories) [SBDL01, BNO⁺08a, dMB09a]

Two interpreted function symbols $_[_]$ and **store**

Axiomatized by:

- $\forall a \forall i \forall v. \text{store}(a, i, v)[i] = v$
- $\forall a \forall i \forall j \forall v. i \neq j \Rightarrow \text{store}(a, i, v)[j] = a[j]$

Sometimes also with *extensionality*:

- $\forall a \forall b. (\forall i. a[i] = b[i] \Rightarrow a = b)$

Is the following set of literals satisfiable in this theory?

$$\text{store}(a, i, x) \neq b, b[i] = y, \text{store}(b, i, x)[j] = y, a = b, i = j$$

Useful both in hardware and software verification [BCF⁺07, BB09a, HBJ⁺14b]

Universe consists of (fixed-sized) vectors of bits

Different types of operations:

- *String-like*: concat, extract, ...
- *Logical*: bit-wise not, or, and, ...
- *Arithmetic*: add, subtract, multiply, ...
- *Comparison*: <, >, ...

Is this formula satisfiable over bit vectors of size 3?

$$a[1:0] \neq b[1:0] \wedge (a \mid b) = c \wedge c[0:0] = 0 \wedge a[1:0] + b[1:0] = 0$$

- Floating point arithmetic [BDG⁺14, ZWR14]
- Ordinary differential equations [GKC13]
- (Co)Algebraic data-types [BST07, RB16]
- Strings and regular expressions [LRT⁺14, KGG⁺09]
- Finite sets with cardinality [BRBT16]
- Finite relations [MRTB17]
- ...

THEORY SOLVERS

Given a theory T , a *Theory Solver* for T takes as input a set Φ of *literals* and determines whether Φ is T -satisfiable.

Φ is *T -satisfiable* iff there is some model M of T such that each formula in Φ holds in M .

- Literals are of the form $t_1 = t_2$ and $t_1 \neq t_2$
- Can be decided in $O(n \log(n))$ based on congruence closure
- Efficient theory propagation for equalities
- Can generate:
 - small explanations [DNS05]
 - minimal (i.e., non-redundant) explanations [NO07b]
 - smallest explanations (NP-hard) [FFHP]
- Typically the core of the SMT solver and used in other theories

Main idea: apply congruence axiom:

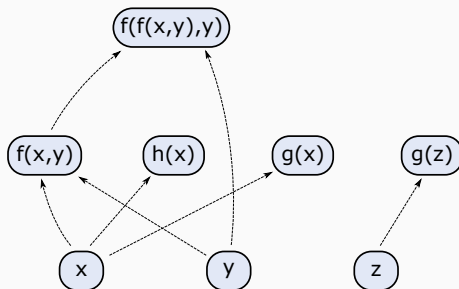
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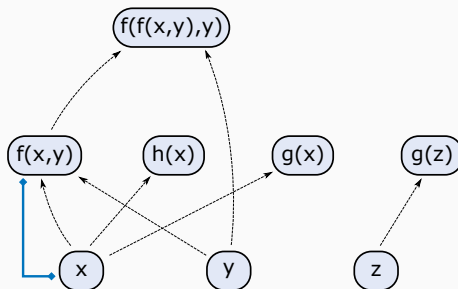


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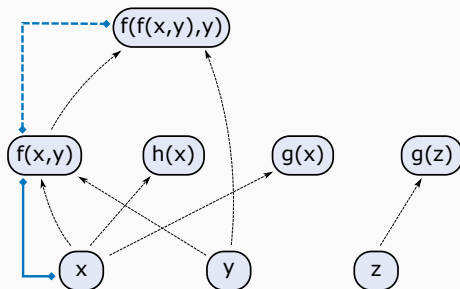


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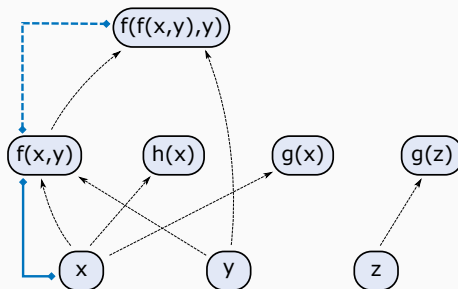


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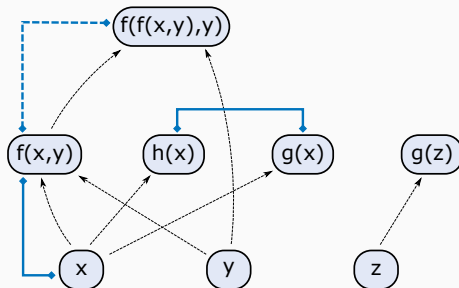


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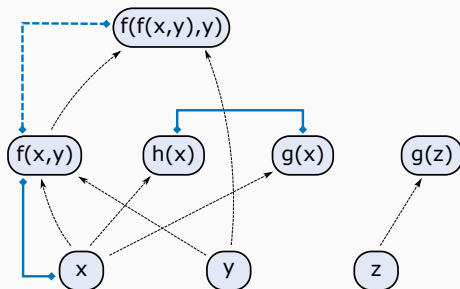


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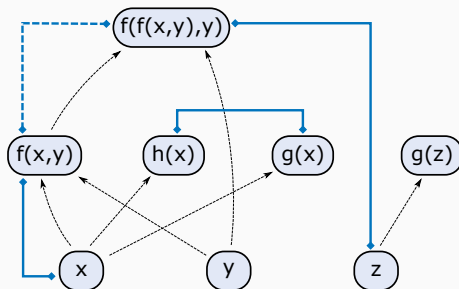


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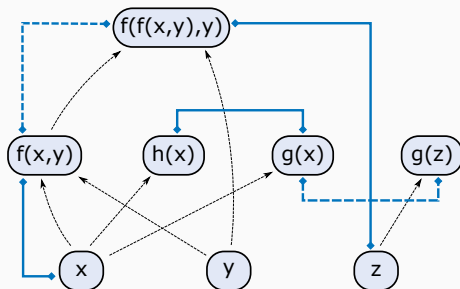


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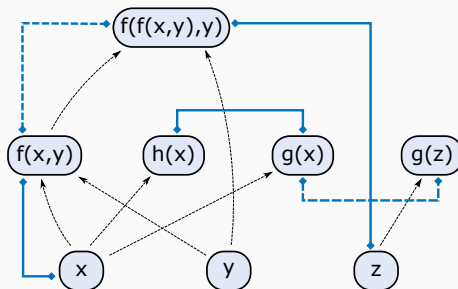


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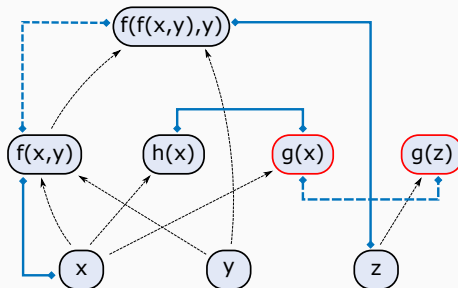
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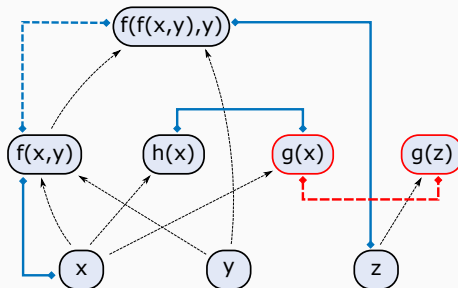
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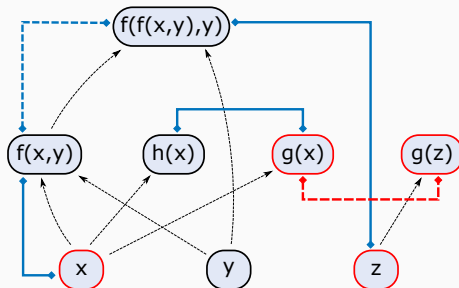
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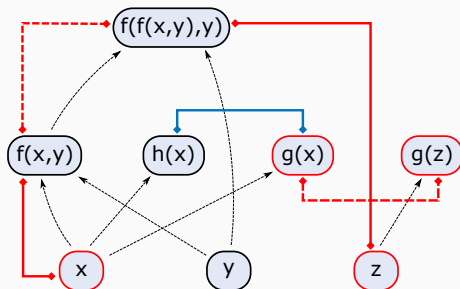
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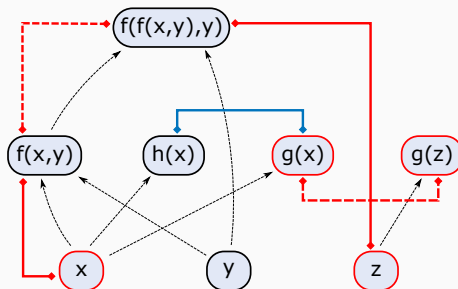
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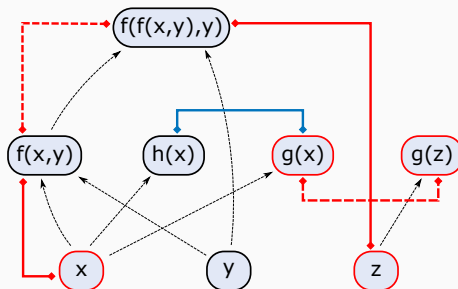
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Conflict Set:

1. $g(x) \neq g(z)$
2. $f(f(x, y), y) = z$
3. $f(x, y) = x$



$$\forall a, i, e : \text{store}(a, i, e)[i] = e$$

$$\forall a, i, j, e : i \neq j \Rightarrow \text{store}(a, i, e)[j] = a[j]$$

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

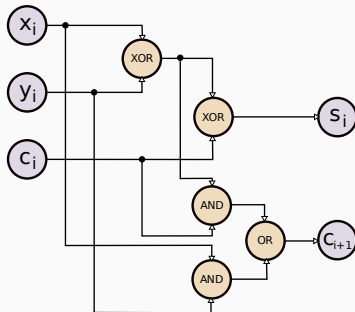
Common approach:

- UF + lemmas on demand [BB09b, DMB09b]
- Use EUF as if `store` and `_[_]` were uninterpreted
- If UNSAT in EUF, then UNSAT in arrays too
- If SAT and solution satisfies array axioms, then SAT (lucky case)
- If not, then refine by instantiating violated axioms

Common approach:

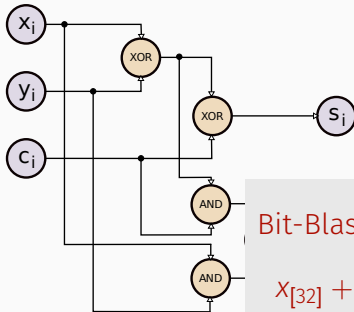
1. Simplify/preprocess (heavily)
2. Encode to SAT (aka, bit blasting)
3. Send to a SAT solver

Alternatives [HBJ⁺14a, ZWR16] not yet mature



Translation to CNF

- Each node a new variables
- XOR introduces 4 clauses
- AND introduces 3 clauses
- OR introduces 3 clauses
- **17 new clauses**
- **5 new variables**



Translation to CNF

- Each node a new variables
- XOR introduces 4 clauses
- AND introduces 3 clauses

Bit-Blasting Addition/Multiplication

$x_{[32]} + y_{[32]}$ 544 new clauses, 160 new variables

$x_{[32]} \times y_{[32]}$ 10016 new clauses, 3008 new variables

Language:

- Literals of the form

$$x - y \leq k$$

with x and y variables (integer or real) and k constant (integer or real)

- Reductions:

$$x - y = k \longrightarrow x - y \leq k \wedge y - x \leq k$$

$$x - y < k \longrightarrow x - y \leq k - 1 \quad (\text{integers})$$

$$x - y < k \longrightarrow x - y \leq k - \delta \quad (\text{reals})$$

- Any solution to a set of literals can be shifted:
 - if v is a satisfying assignment, so is $v' = \lambda x. v(x) + k$
- We can use this to also process simple bounds $x \leq k$:
 - introduce fresh variable z (for zero),
 - rewrite each $x \leq k$ to $x - z \leq k$,
 - given a solution v , shift it so that $v'(z) = 0$
- If we allow (dis)equalities as literals,
 - in reals, satisfiability is polynomial
 - in integers, satisfiability is NP-hard

Common approach: Cycle detection

1. Construct a graph from literals
2. Check if there is a negative path

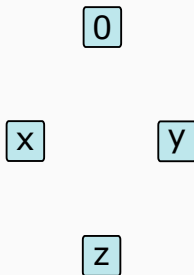
Theorem *Literals unsatisfiable $\Leftrightarrow \exists$ negative path*

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Theorem *Literals unsatisfiable* $\Leftrightarrow \exists$ negative path

Example

$$[x \leq 1, x - y \leq 2, y - z \leq 3, z - x \leq -6]$$

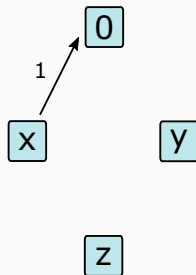


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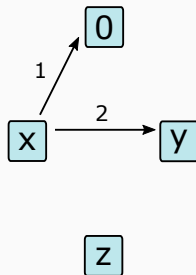


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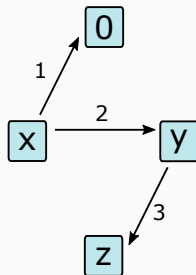


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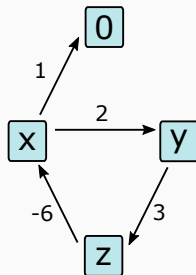


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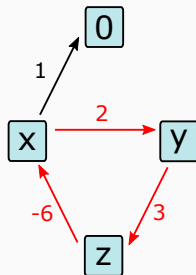


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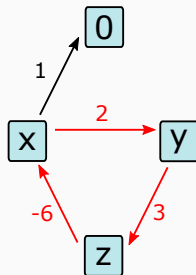
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Conflict Set:

$$x - y \leq 2, y - z \leq 3, z - x \leq -6$$



Language:

- Literals of the form $a_1x_1 + \dots + a_nx_n \bowtie b$
with a_1 positive and $\bowtie \in \{\leq, \geq\}$

$$t = b \longrightarrow t \leq b \wedge t \geq b$$

- Reductions: $t < b \longrightarrow t \leq b - 1$ (integer arith.)
 $t < b \longrightarrow t \leq b - \delta$ (real arith.)

Common approach: variant of Simplex designed for
SMT [DDM06b]

- Incremental
- Cheap backtracking
- Can do theory propagation
- Can generate minimal explanations
- Worst case exponential but fast in practice

Rewrite each $\sum a_i x_i \bowtie b$ as $s \bowtie b$ with $s = \sum a_i x_i$

We get tableau of equations + simple bounds on variables

- Tableau is fixed (modulo pivoting and substitutions)
- Bounds can be asserted and retracted

Tableau

$$s_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_j + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$s_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_j + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$s_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_j + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq s_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_i + \cdots + a_{1,n} \cdot x_n$$

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Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_i + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_i + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_i + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq s_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

- lhs variables are basic, rhs variables are non-basic

Tableau

$$s_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_i + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$s_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_i + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$s_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_i + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq s_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

- lhs variables are **basic**, rhs variables are non-basic

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_i + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_i + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_i + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq S_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

- lhs variables are basic, rhs variables are non-basic

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_j + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_j + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_j + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq S_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

- lhs variables are basic, rhs variables are non-basic
- Keep an assignment v of all variables:
 - v satisfies the tableau,
 - v satisfies bounds on the non-basic variables

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_j + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_j + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_j + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq S_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

- lhs variables are basic, rhs variables are non-basic
- Keep an assignment v of all variables:
 - v satisfies the tableau,
 - v satisfies bounds on the non-basic variables
- Initially $v(x) = 0$ and $-\infty \leq x \leq +\infty$

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_j + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_j + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_j + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq S_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

Case 1:

- v satisfies bound on the basic variables too
- **Satisfiable**, v is the model!

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_j + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_j + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_j + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq s_i \leq u_i$$

$$\vdots$$

$$l_j \leq x_j \leq u_j$$

$$\vdots$$

Case 2:

- v doesn't satisfy bound on some s_i and all x_j 's that s_i depends on are at their bounds (can't fix)
- **Unsatisfiable**, the row is the explanation

Tableau

$$S_1 = a_{1,1} \cdot x_1 + \cdots + a_{1,i} \cdot x_j + \cdots + a_{1,n} \cdot x_n$$

$$\vdots$$

$$S_i = a_{i,1} \cdot x_1 + \cdots + a_{i,i} \cdot x_j + \cdots + a_{i,n} \cdot x_n$$

$$\vdots$$

$$S_m = a_{m,1} \cdot x_1 + \cdots + a_{m,i} \cdot x_j + \cdots + a_{m,n} \cdot x_n$$

Bounds

$$\vdots$$

$$l_i \leq S_i \leq u_i$$

$$\vdots$$

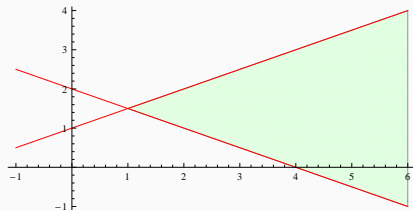
$$l_j \leq x_j \leq u_j$$

$$\vdots$$

Case 3:

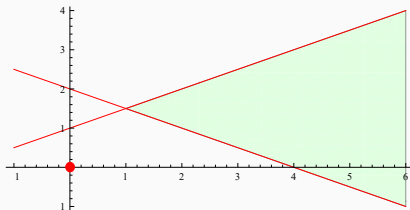
- v doesn't satisfy bound on some S_i , and some x_j 's that S_i depends on has some slack
- Pivot, substitute, and continue

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[2y - x - 2 \leq 0, -2y - x + 4 \leq 0]$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq +\infty$$

$$-\infty \leq s_2 \leq +\infty$$

Assignment

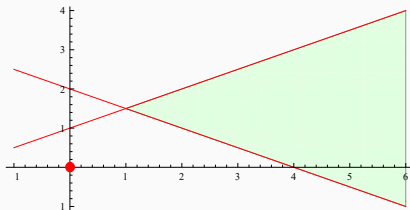
$$x \mapsto 0$$

$$y \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq +\infty$$

$$-\infty \leq s_2 \leq +\infty$$

Assignment

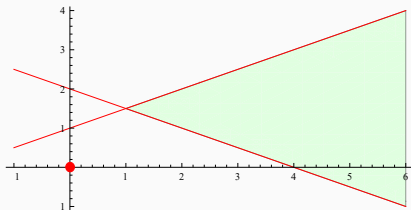
$$x \mapsto 0$$

$$y \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq +\infty$$

Assignment

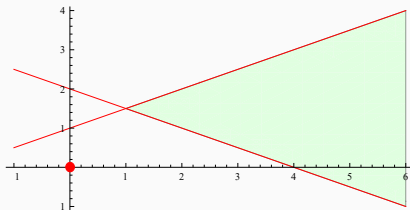
$$x \mapsto 0$$

$$y \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq +\infty$$

Assignment

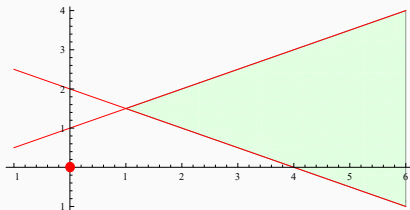
$$x \mapsto 0$$

$$y \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

$$x \mapsto 0$$

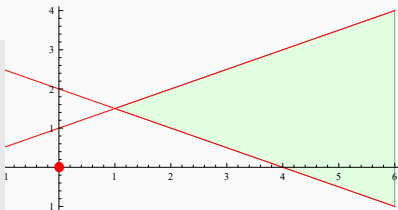
$$y \mapsto 0$$

$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

$$-\infty \leq x \leq +\infty$$

$$x \mapsto 0$$

$$-\infty \leq y \leq +\infty$$

$$y \mapsto 0$$

$$-\infty \leq s_1 \leq 2$$

$$s_1 \mapsto 0$$

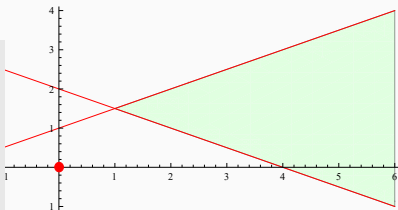
$$-\infty \leq s_2 \leq -4$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

$$-\infty \leq x \leq +\infty$$

$$x \mapsto 0$$

$$-\infty \leq y \leq +\infty$$

$$y \mapsto 0$$

$$-\infty \leq s_1 \leq 2$$

$$s_1 \mapsto 0$$

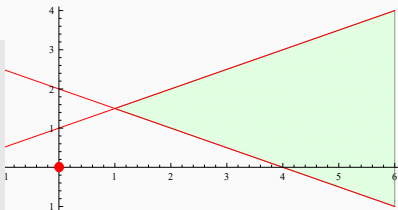
$$-\infty \leq s_2 \leq -4$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y
- Pivot s_2 and y



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = 2y - x$$

$$s_2 = -2y - x$$

$$-\infty \leq x \leq +\infty$$

$$x \mapsto 0$$

$$-\infty \leq y \leq +\infty$$

$$y \mapsto 0$$

$$-\infty \leq s_1 \leq 2$$

$$s_1 \mapsto 0$$

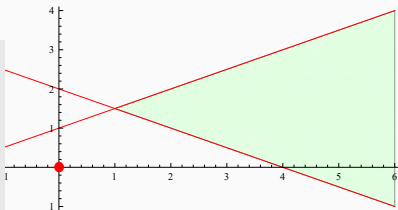
$$-\infty \leq s_2 \leq -4$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y
- Pivot s_2 and y



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 0$$

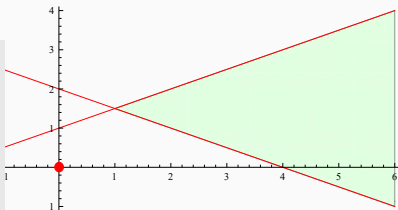
$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y
- Pivot s_2 and y
- Update s_2 value



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 0$$

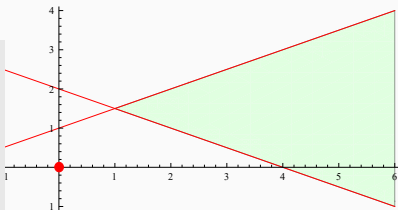
$$s_1 \mapsto 0$$

$$s_2 \mapsto 0$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y
- Pivot s_2 and y
- Update s_2 value



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 0$$

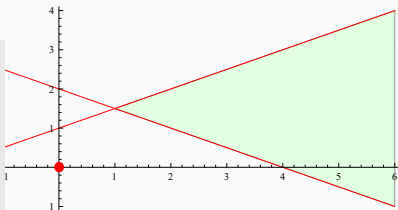
$$s_1 \mapsto 0$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y
- Pivot s_2 and y
- Update s_2 value
- Update basic vars



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

$$-\infty \leq x \leq +\infty$$

$$x \mapsto 0$$

$$-\infty \leq y \leq +\infty$$

$$y \mapsto 0$$

$$-\infty \leq s_1 \leq 2$$

$$s_1 \mapsto 0$$

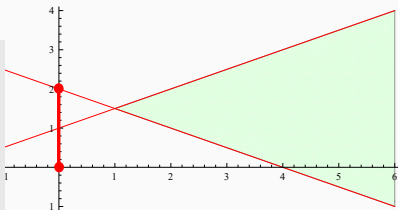
$$-\infty \leq s_2 \leq -4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_2 's bound violated

- There is slack in y
- Pivot s_2 and y
- Update s_2 value
- Update basic vars



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

$$-\infty \leq x \leq +\infty$$

$$x \mapsto 0$$

$$-\infty \leq y \leq +\infty$$

$$y \mapsto 2$$

$$-\infty \leq s_1 \leq 2$$

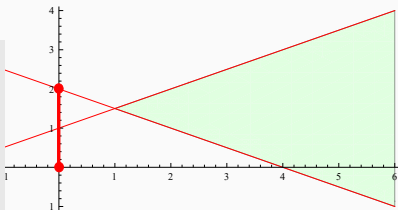
$$s_1 \mapsto 4$$

$$-\infty \leq s_2 \leq -4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

$$x \mapsto 0$$

$$y \mapsto 2$$

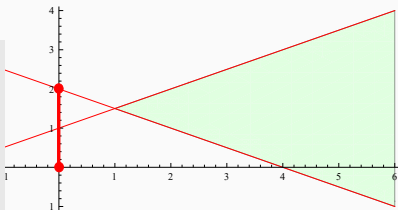
$$s_1 \mapsto 4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 2$$

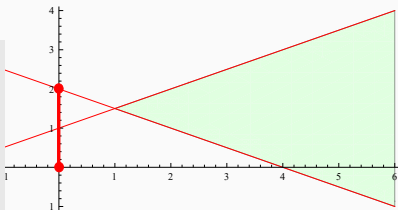
$$s_1 \mapsto 4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x
- Pivot s_1 and x



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$s_1 = -s_2 - 2x$$

$$y = -\frac{1}{2}s_2 - \frac{1}{2}x$$

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

$$x \mapsto 0$$

$$y \mapsto 2$$

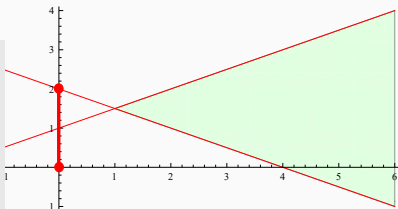
$$s_1 \mapsto 4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x
- Pivot s_1 and x



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$x = -\frac{1}{2}s_1 - \frac{1}{2}s_2$$

$$y = \frac{1}{4}s_1 - \frac{1}{4}s_2$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 2$$

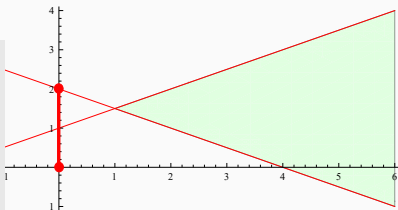
$$s_1 \mapsto 4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x
- Pivot s_1 and x
- Update s_1 value



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$x = -\frac{1}{2}s_1 - \frac{1}{2}s_2$$

$$y = \frac{1}{4}s_1 - \frac{1}{4}s_2$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 2$$

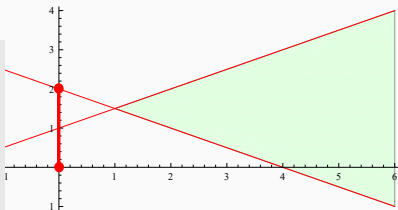
$$s_1 \mapsto 4$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x
- Pivot s_1 and x
- Update s_1 value



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$x = -\frac{1}{2}s_1 - \frac{1}{2}s_2$$

$$y = \frac{1}{4}s_1 - \frac{1}{4}s_2$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 0$$

$$y \mapsto 2$$

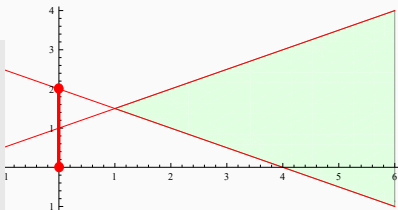
$$s_1 \mapsto 2$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x
- Pivot s_1 and x
- Update s_1 value
- Update basic vars



$$[s_1 \leq 2, s_2 \leq -4]$$

Bounds

Assignment

Tableau

$$x = -\frac{1}{2}s_1 - \frac{1}{2}s_2$$

$$y = \frac{1}{4}s_1 - \frac{1}{4}s_2$$

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

$$x \mapsto 0$$

$$y \mapsto 2$$

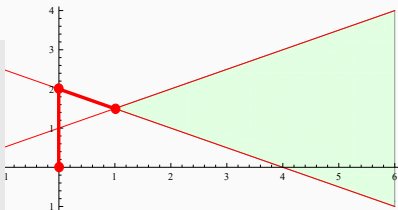
$$s_1 \mapsto 2$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE

s_1 's bound violated

- There is slack in x
- Pivot s_1 and x
- Update s_1 value
- Update basic vars



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$x = -\frac{1}{2}s_1 - \frac{1}{2}s_2$$

$$y = \frac{1}{4}s_1 - \frac{1}{4}s_2$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

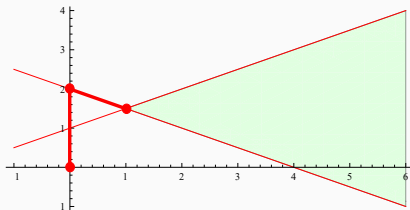
$$x \mapsto 1$$

$$y \mapsto \frac{3}{2}$$

$$s_1 \mapsto 2$$

$$s_2 \mapsto -4$$

LINEAR RATIONAL ARITHMETIC: EXAMPLE



$$[s_1 \leq 2, s_2 \leq -4]$$

Tableau

$$x = -\frac{1}{2}s_1 - \frac{1}{2}s_2$$

$$y = \frac{1}{4}s_1 - \frac{1}{4}s_2$$

Bounds

$$-\infty \leq x \leq +\infty$$

$$-\infty \leq y \leq +\infty$$

$$-\infty \leq s_1 \leq 2$$

$$-\infty \leq s_2 \leq -4$$

Assignment

$$x \mapsto 1$$

$$y \mapsto \frac{3}{2}$$

$$s_1 \mapsto 2$$

$$s_2 \mapsto -4$$

Classic NP-complete problem [Pap81]

Admits quantifier elimination [Coo72]

Common approach:

- Simplex + Branch-And-Bound [DDM06b, Gri12, Kin14]
- Use Simplex to solve real relaxation (treat variables as real)
- If UNSAT over reals, then UNSAT over integers too
- If SAT and solution v is integral, then SAT (lucky case)
- Otherwise, refine:
 - Add branch-and-bound lemmas: $x \leq \lfloor v(x) \rfloor \vee x \geq \lceil v(x) \rceil$
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Alternatives [JdM13, BSW15] not yet mature

$$f(y, x) = a_m \cdot x^{d_m} + a_{m-1} \cdot x^{d_{m-1}} + \dots + a_1 \cdot x^{d_1} + a_0$$

f is in $\mathbb{Z}[y, x]$, a_i are in $\mathbb{Z}[y]$

Examples

$$\begin{aligned} f(x, y) &= (x^2 - 1)y^2 + (x + 1)y - 1 \in \mathbb{Z}[x, y] \\ g(x) &= 16x^3 - 8x^2 + x + 16 \in \mathbb{Z}[x] \end{aligned}$$

Polynomial Constraints

$$f(x, y) > 0 \wedge g(x) < 0$$

$$p_1 > 0 \vee (p_2 = 0 \wedge p_3 < 0) \quad p_1, p_2, p_3 \in \mathbb{Z}[x_1, \dots, x_n]$$

Projection (Saturation)

Project polynomials using a projection P

$$\{p_1, p_2, p_3\} \mapsto \{p_1, p_2, p_3, p_4, \dots, p_n\}$$

Lifting (Model construction)

For each variable x_k

1. Isolate roots of $p_i(\alpha, x_k)$
2. Choose a cell C and assign $x_k \mapsto \alpha_k \in C$, continue
3. If no more cells, backtrack

Model Construction

Build partial model by assigning variables to values

$$[\dots, C_1, C_2, \dots, x \mapsto \sqrt{2}/2, \dots]$$

Unit Reasoning

Reason about unit constraints

$$C_1 \equiv (x^2 + y^2 < 1) \quad C_2 \equiv (xy > 1)$$

Explain Conflicts

Explain conflicts using valid clausal reasons

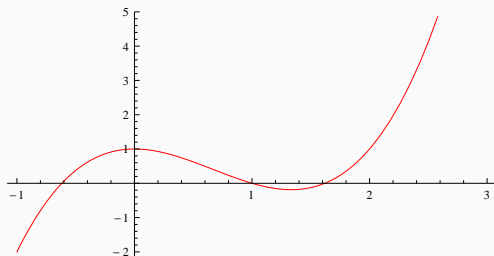
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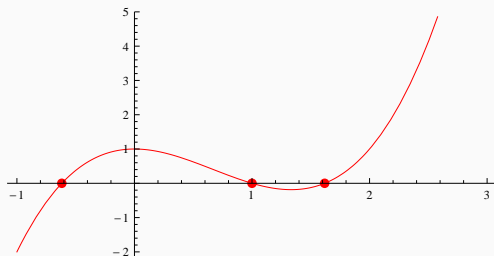
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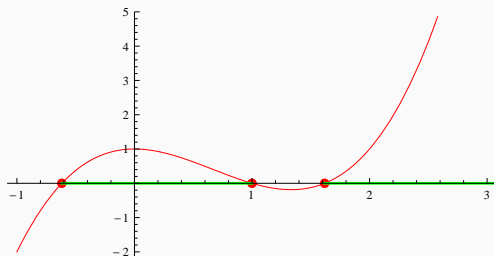
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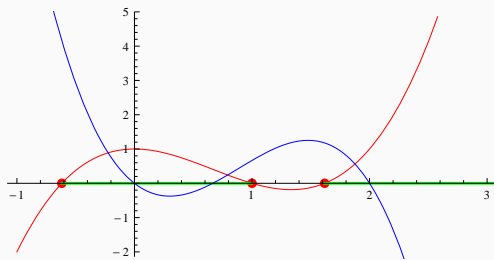
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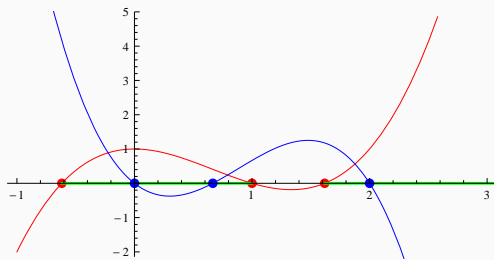


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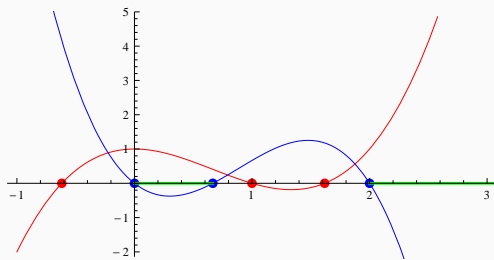
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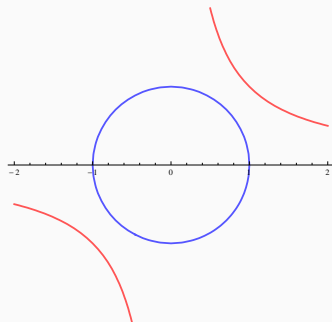
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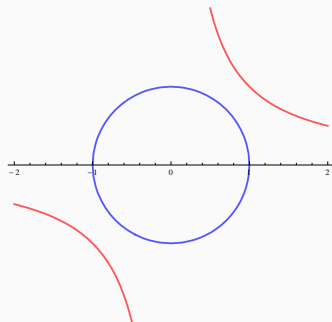
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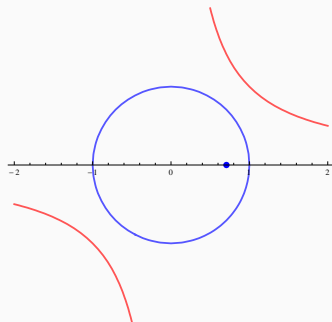
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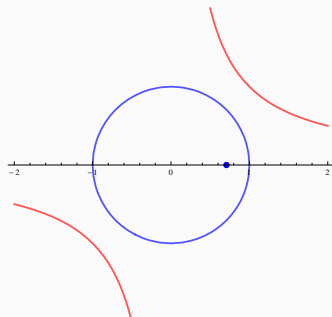
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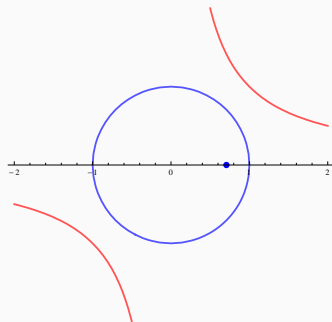
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Unit Constraint Reasoning

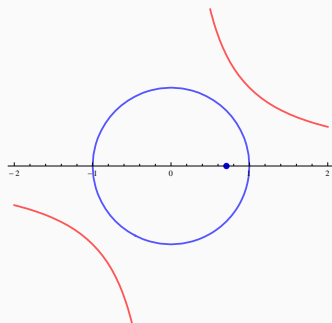
$$x^2 + y^2 < 1 \Rightarrow -\sqrt{3/2} < y < \sqrt{3/2}$$
$$-2y - x + 4 < 0 \Rightarrow y > \sqrt{2}$$



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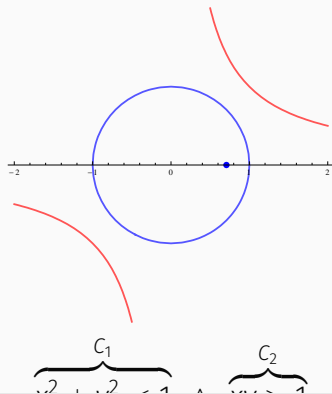
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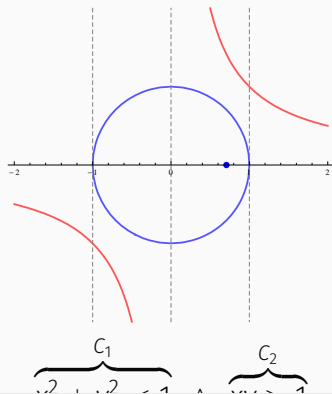
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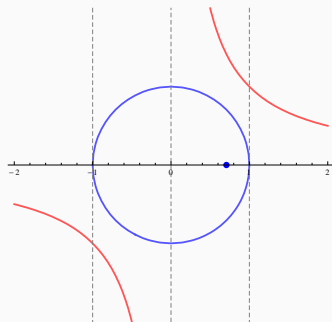
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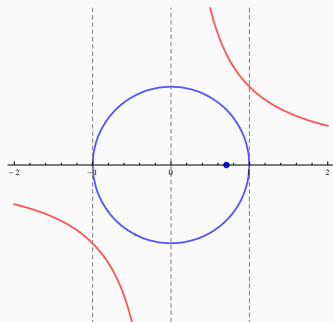
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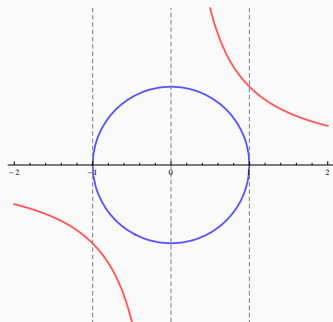
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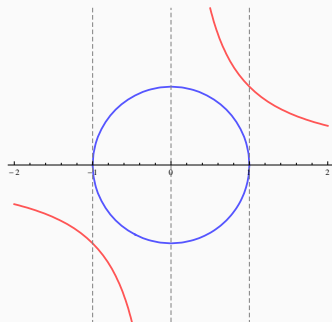
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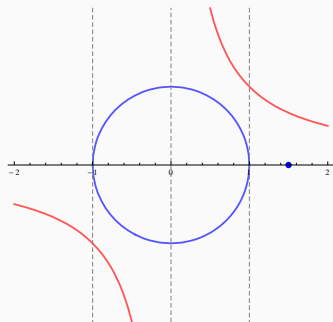
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EXTENDING THEORY SOLVERS TO QFFS

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Solution: Exploit propositional satisfiability technology

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1. “Eager” [PRSS99, SSB02, SLB03, BGV01, BV02]
 - translate into an equisatisfiable propositional formula
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Theory T: Equality with Uninterpreted Functions

Simplest setting:

- Off-line SAT solver
- Non-incremental *theory solver* for conjunctions of equalities and disequalities
- Theory atoms (e.g., $g(a) = c$) abstracted to propositional atoms (e.g., 1)

(VERY) LAZY APPROACH FOR SMT – EXAMPLE

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

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Done: the original formula is unsatisfiable in UF.

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- The theory solver works **only** with **conjunctions of literals**
- Modular approach:
 - SAT and theory solvers **communicate** via a **simple API** [GHN⁺04]
 - SMT for a **new theory** only requires **new theory solver**
 - An **off-the-shelf SAT solver** can be **embedded** in a lazy SMT system with few new lines of code (tens)

Several variants and enhancements of lazy SMT solvers exist

They can be modeled abstractly and declaratively as *transition systems*

A transition system is a binary relation over states, induced by a set of conditional transition rules

The framework can be first developed for SAT and then extended to lazy SMT [NOT06, KG07]

An abstract framework helps one:

- skip over implementation details and unimportant control aspects
- reason formally about solvers for SAT and SMT
- model advanced features such as non-chronological backtracking, lemma learning, theory propagation, ...
- describe different strategies and prove their correctness
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The one described next is a re-elaboration of those in

[NOT06, KG07]

- Modern SAT solvers are based on the DPLL procedure [DP60, DLL62]
- DPLL tries to build incrementally a satisfying truth assignment M for a CNF formula F
- M is grown by
 - deducing the truth value of a literal from M and F , or
 - guessing a truth value
- If a wrong guess for a literal leads to an inconsistency, the procedure backtracks and tries the opposite value

States:

fail or $\langle M, F \rangle$

where

- M is a *sequence of literals and decision points* • denoting a partial truth *assignment*
- F is a *set of clauses* denoting a CNF *formula*

Def. If $M = M_0 \bullet M_1 \bullet \dots \bullet M_n$ where each M_i contains no decision points

- M_i is *decision level* i of M
- $M[i] \stackrel{\text{def}}{=} M_0 \bullet \dots \bullet M_i$

States:

fail or $\langle M, F \rangle$

Initial state:

- $\langle (), F_0 \rangle$, where F_0 is to be checked for satisfiability

Expected final states:

- fail if F_0 is unsatisfiable
- $\langle M, G \rangle$ otherwise, where
 - G is equivalent to F_0 and
 - M satisfies G

States treated like records:

- M denotes the truth assignment component of current state
- F denotes the formula component of current state

Transition rules in *guarded assignment form* [KG07]

$$\frac{p_1 \quad \cdots \quad p_n}{[M := e_1] \quad [F := e_2]}$$

updating M , F or both when premises p_1, \dots, p_n all hold

Extending the assignment

$$\text{Propagate} \frac{l_1 \vee \dots \vee l_n \vee l \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad l, \bar{l} \notin M}{M := M \ l}$$

Note: When convenient, treat M as a set

Note: Clauses are treated modulo ACI of \vee

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Note: Clauses are treated modulo ACI of \vee

$$\text{Decide} \frac{l \in \text{Lit}(F) \quad l, \bar{l} \notin M}{M := M \cup l}$$

Note: $\text{Lit}(F) \stackrel{\text{def}}{=} \{l \mid l \text{ literal of } F\} \cup \{\bar{l} \mid l \text{ literal of } F\}$

Repairing the assignment

$$\text{Fail} \frac{l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad \bullet \notin M}{\text{fail}}$$

Repairing the assignment

$$\text{Fail} \frac{l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad \bullet \notin M}{\text{fail}}$$

Backtrack

$$\frac{l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M \quad M = M \bullet l \quad \bullet \notin N}{M := M \bar{l}}$$

Note: Last premise of **Backtrack** enforces **chronological** backtracking

To model conflict-driven backjumping and learning, add to states a third component **C** whose value is either **no** or a *conflict clause*

To model conflict-driven backjumping and learning, add to states a third component C whose value is either **no** or a *conflict clause*

States: **fail** or $\langle M, F, C \rangle$

Initial state:

- $\langle (), F_0, \text{no} \rangle$, where F_0 is to be checked for satisfiability

Expected final states:

- **fail** if F_0 is unsatisfiable
- $\langle M, G, \text{no} \rangle$ otherwise, where
 - G is equivalent to F_0 and
 - M satisfies G

Replace **Backtrack** with

Replace **Backtrack** with

$$\text{Conflict} \quad \frac{C = \text{no} \quad l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M}{C := l_1 \vee \dots \vee l_n}$$

$$\text{Explain} \quad \frac{C = l \vee D \quad l_1 \vee \dots \vee l_n \vee \bar{l} \in F \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

$$\text{Backjump} \quad \frac{C = l_1 \vee \dots \vee l_n \vee l \quad \text{lev } \bar{l}_1, \dots, \text{lev } \bar{l}_n \leq i < \text{lev } \bar{l}}{C := \text{no} \quad M := M^{[i]} l}$$

Note: $l \prec_M l'$ if l occurs before l' in M
 $\text{lev } l = i$ iff l occurs in decision level i of M

Replace **Backtrack** with

$$\text{Conflict} \quad \frac{C = \text{no} \quad l_1 \vee \dots \vee l_n \in F \quad \bar{l}_1, \dots, \bar{l}_n \in M}{C := l_1 \vee \dots \vee l_n}$$

$$\text{Explain} \quad \frac{C = l \vee D \quad l_1 \vee \dots \vee l_n \vee \bar{l} \in F \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

$$\text{Backjump} \quad \frac{C = l_1 \vee \dots \vee l_n \vee l \quad \text{lev } \bar{l}_1, \dots, \text{lev } \bar{l}_n \leq i < \text{lev } \bar{l}}{C := \text{no} \quad M := M^{[i]} l}$$

Maintain **invariant**: $F \models_P C$ and $M \models_P \neg C$ when $C \neq \text{no}$

Note: \models_P denotes propositional entailment

Modify **Fail** to

Modify **Fail** to

$$\text{Fail} \frac{C \neq \text{no} \quad \bullet \notin M}{\text{fail}}$$

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

M	F	C	rule
	F	no	

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

M	F	C	rule
	F	no	
1	F	no	by Propagate

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

M	F	C	rule
	F	no	
1	F	no	by Propagate
1 2	F	no	by Propagate

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

M	F	C	rule
	F	no	
1	F	no	by Propagate
1 2	F	no	by Propagate
1 2 • 3	F	no	by Decide

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
	1	F	no	by Propagate
	1 2	F	no	by Propagate
	1 2 • 3	F	no	by Decide
	1 2 • 3 4	F	no	by Propagate

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
1		F	no	by Propagate
1 2		F	no	by Propagate
1 2 • 3		F	no	by Decide
1 2 • 3 4		F	no	by Propagate
1 2 • 3 4 • 5		F	no	by Decide

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
1		F	no	by Propagate
1 2		F	no	by Propagate
1 2 • 3		F	no	by Decide
1 2 • 3 4		F	no	by Propagate
1 2 • 3 4 • 5		F	no	by Decide
1 2 • 3 4 • 5 $\bar{6}$		F	no	by Propagate

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
1		F	no	by Propagate
1 2		F	no	by Propagate
1 2 • 3		F	no	by Decide
1 2 • 3 4		F	no	by Propagate
1 2 • 3 4 • 5		F	no	by Decide
1 2 • 3 4 • 5 $\bar{6}$		F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7		F	no	by Propagate

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
1		F	no	by Propagate
1 2		F	no	by Propagate
1 2 • 3		F	no	by Decide
1 2 • 3 4		F	no	by Propagate
1 2 • 3 4 • 5		F	no	by Decide
1 2 • 3 4 • 5 $\bar{6}$		F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7		F	no	by Propagate
1 2 • 3 4 • 5 $\bar{6}$ 7		F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
	1	F	no	by Propagate
	1 2	F	no	by Propagate
	1 2 • 3	F	no	by Decide
	1 2 • 3 4	F	no	by Propagate
	1 2 • 3 4 • 5	F	no	by Decide
	1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5} \vee 6$	by Explain with $\bar{1} \vee \bar{5} \vee 7$

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
	1	F	no	by Propagate
	1 2	F	no	by Propagate
	1 2 • 3	F	no	by Decide
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	1 2 • 3 4 • 5	F	no	by Decide
	1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5} \vee 6$	by Explain with $\bar{1} \vee \bar{5} \vee 7$
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5}$	by Explain with $\bar{5} \vee \bar{6}$

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
	1	F	no	by Propagate
	1 2	F	no	by Propagate
	1 2 • 3	F	no	by Decide
	1 2 • 3 4	F	no	by Propagate
	1 2 • 3 4 • 5	F	no	by Decide
	1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5} \vee 6$	by Explain with $\bar{1} \vee \bar{5} \vee 7$
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5}$	by Explain with $\bar{5} \vee \bar{6}$
	1 2 $\bar{5}$	F	no	by Backjump

EXECUTION EXAMPLE

$$F := \{1, \bar{1} \vee 2, \bar{3} \vee 4, \bar{5} \vee \bar{6}, \bar{1} \vee \bar{5} \vee 7, \bar{2} \vee \bar{5} \vee 6 \vee \bar{7}\}$$

	M	F	C	rule
		F	no	
	1	F	no	by Propagate
	1 2	F	no	by Propagate
	1 2 • 3	F	no	by Decide
	1 2 • 3 4	F	no	by Propagate
	1 2 • 3 4 • 5	F	no	by Decide
	1 2 • 3 4 • 5 $\bar{6}$	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	no	by Propagate
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$\bar{2} \vee \bar{5} \vee 6 \vee \bar{7}$	by Conflict
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5} \vee 6$	by Explain with $\bar{1} \vee \bar{5} \vee 7$
	1 2 • 3 4 • 5 $\bar{6}$ 7	F	$1 \vee \bar{2} \vee \bar{5}$	by Explain with $\bar{5} \vee \bar{6}$
	1 2 $\bar{5}$	F	no	by Backjump
	1 2 $\bar{5}$ • 3	F	no	by Decide
	...			

Also add

$$\text{Learn} \frac{F \models_p C \quad C \notin F}{F := F \cup \{C\}}$$

$$\text{Forget} \frac{C = \text{no} \quad F = G \cup \{C\} \quad G \models_p C}{F := G}$$

$$\text{Restart} \frac{}{M := M^{[0]} \quad C := \text{no}}$$

Note: Learn can be applied to [any](#) clause stored in C when $C \neq \text{no}$

At the core, current CDCL SAT solvers are implementations of the transition system with rules

Propagate, Decide,

Conflict, Explain, Backjump,

Learn, Forget, Restart

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Basic DPLL $\stackrel{\text{def}}{=}$

{ Propagate, Decide, Conflict, Explain, Backjump }

DPLL $\stackrel{\text{def}}{=}$ Basic DPLL + { Learn, Forget, Restart }

Some terminology:

Irreducible state: state for which no Basic DPLL rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

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Exhausted execution: execution ending in an irreducible state

Proposition (Strong Termination) Every execution in Basic DPLL is finite.

Note: This is not so immediate, because of Backjump.

Some terminology:

Irreducible state: state for which no Basic DPLL rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

Proposition (Strong Termination) Every execution in Basic DPLL is finite.

Lemma Every exhausted execution ends with either $C = \text{no}$ or fail.

Some terminology:

Irreducible state: state for which no Basic DPLL rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

Proposition (Soundness) For every exhausted execution starting with $F = F_0$ and ending with fail , the clause set F_0 is unsatisfiable.

Proposition (Completeness) For every exhausted execution starting with $F = F_0$ and ending with $C = \text{no}$, the clause set F_0 is satisfied by M .

- Applying
 - one Basic DPLL rule between each two **Learn** applications **and**
 - **Restart** less and less often
- ensures termination

- Applying
 - one Basic DPLL rule between each two **Learn** applications **and**
 - **Restart** less and less often**ensures termination**
- A **common basic strategy** applies the rules with the following priorities:
 1. If $n > 0$ conflicts have been found so far,
increase n and apply **Restart**

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- A common basic strategy applies the rules with the following priorities:
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 3. Keep applying **Explain** until **Backjump** is applicable

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 4. Apply **Learn**

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 4. Apply **Learn**
 5. Apply **Backjump**

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 - one Basic DPLL rule between each two **Learn** applications and **Restart** less and less oftenensures termination
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 1. If $n > 0$ conflicts have been found so far, increase n and apply **Restart**
 2. If a clause is falsified by **M**, apply **Conflict**
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 4. Apply **Learn**
 5. Apply **Backjump**
 6. Apply **Propagate** to completion

- Applying
 - one Basic DPLL rule between each two **Learn** applications and
 - **Restart** less and less oftenensures termination
- A common basic strategy applies the rules with the following priorities:
 1. If $n > 0$ conflicts have been found so far, increase n and apply **Restart**
 2. If a clause is falsified by **M**, apply **Conflict**
 3. Keep applying **Explain** until **Backjump** is applicable
 4. Apply **Learn**
 5. Apply **Backjump**
 6. Apply **Propagate** to completion
 7. Apply **Decide**

Same states and transitions but

- F contains quantifier-free clauses in some theory T
- M is a sequence of theory literals and decision points
- the DPLL system is augmented with rules

T -Conflict, T -Propagate, T -Explain

- maintains invariant: $F \models_T C$ and $M \models_P \neg C$ when $C \neq \text{no}$

Def. $F \models_T G$ iff every model of T that satisfies F satisfies G as well

Fix a theory T

$$\text{\textit{T-Conflict}} \quad \frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_T \perp}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

Fix a theory T

$$T\text{-Conflict} \quad \frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_T \perp}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

$$T\text{-Propagate} \quad \frac{l \in \text{Lit}(F) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \, l}$$

Fix a theory T

$$T\text{-Conflict} \quad \frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_T \perp}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

$$T\text{-Propagate} \quad \frac{l \in \text{Lit}(F) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \, l}$$

$$T\text{-Explain} \quad \frac{C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_T \bar{l} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

Note: \perp = empty clause

Note: \models_T decided by theory solver

T-Conflict is enough to model the naive integration of SAT solvers and theory solvers seen in the earlier UF example

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	1, $\bar{2} \vee 3, \bar{4}$	no	

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	1, $\bar{2} \vee 3, \bar{4}$	no	
1 $\bar{4}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺

MODELING THE VERY LAZY THEORY APPROACH

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
$1 \bar{4} \bullet \bar{2}$	$, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
$1 \bar{4} \bullet \bar{2}$	$, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
$1 \bar{4}$	$, \bar{1} \vee 2 \vee 4$	no	by Restart

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
$1 \bar{4} \bullet \bar{2}$	$, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
$1 \bar{4}$	$, \bar{1} \vee 2 \vee 4$	no	by Restart
$1 \bar{4} 2 3$	$, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
$1 \bar{4} \bullet \bar{2}$	$, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
$1 \bar{4}$	$, \bar{1} \vee 2 \vee 4$	no	by Restart
$1 \bar{4} 2 3$	$, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
$1 \bar{4} 2 3$	$, \bar{1} \vee 2 \vee 4, \bar{1} \vee \bar{3} \vee 4$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict, Learn

MODELING THE VERY LAZY THEORY APPROACH

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2 \vee 4$	by T-Conflict
$1 \bar{4} \bullet \bar{2}$	$, \bar{1} \vee 2 \vee 4$	$\bar{1} \vee 2 \vee 4$	by Learn
$1 \bar{4}$	$, \bar{1} \vee 2 \vee 4$	no	by Restart
$1 \bar{4} 2 3$	$, \bar{1} \vee 2 \vee 4$	no	by Propagate ⁺
$1 \bar{4} 2 3$	$, \bar{1} \vee 2 \vee 4, \bar{1} \vee \bar{3} \vee 4$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict, Learn
fail			by Fail

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- An *on-line* SAT engine,
which can accept new input clauses on the fly

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The very lazy approach can be improved considerably with

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 2. identify a small T -unsatisfiable subset of M once M becomes T -unsatisfiable

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M	F	C	rule
	1, $\bar{2} \vee 3$, $\bar{4}$	no	

A BETTER LAZY APPROACH

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺

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	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide

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$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Decide
$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2$	by T-Conflict

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$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2$	by T-Conflict
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by Backjump

A BETTER LAZY APPROACH

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	1, $\bar{2} \vee 3$, $\bar{4}$	no	
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1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3$, $\bar{4}$	no	by Decide
1 $\bar{4} \bullet \bar{2}$	1, $\bar{2} \vee 3$, $\bar{4}$	$\bar{1} \vee 2$	by T-Conflict
1 $\bar{4} 2$	1, $\bar{2} \vee 3$, $\bar{4}$	no	by Backjump
1 $\bar{4} 2 3$	1, $\bar{2} \vee 3$, $\bar{4}$	no	by Propagate

A BETTER LAZY APPROACH

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$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by Backjump
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict

A BETTER LAZY APPROACH

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$1 \bar{4} \bullet \bar{2}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee 2$	by T-Conflict
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by Backjump
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate
$1 \bar{4} 2 3$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{1} \vee \bar{3} \vee 4$	by T-Conflict
fail			by Fail

Ignoring **Restart** (for simplicity), a **common strategy** is to apply the rules using the following priorities:

1. If a clause is falsified by the current assignment **M**, apply **Conflict**
2. If **M** is **T**-unsatisfiable, apply **T-Conflict**
3. Apply **Fail** or **Explain+Learn+Backjump** as appropriate
4. Apply **Propagate**
5. Apply **Decide**

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3. Apply **Fail** or **Explain+Learn+Backjump** as appropriate
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5. Apply **Decide**

Note: Depending on the cost of checking the **T**-satisfiability of **M**, Step (2) can be applied with lower frequency or priority

With **T-Conflict** as the **only theory rule**, the theory solver is used just to **validate** the choices of the SAT engine

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With **T-Propagate** and **T-Explain**, it can also be used to **guide** the engine's search [Tin02]

$$\text{T-Propagate} \frac{l \in \text{Lit}(F) \quad M \models_T l \quad l, \bar{l} \notin M}{M := M \cup l}$$

$$\text{T-Explain} \frac{C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_T \bar{l} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

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M	F	C	rule
	1, $\bar{2} \vee 3, \bar{4}$	no	

THEORY PROPAGATION EXAMPLE

$$\underbrace{g(a) = c}_1 \wedge \underbrace{f(g(a)) \neq f(c)}_{\bar{2}} \vee \underbrace{g(a) = d}_3 \wedge \underbrace{c \neq d}_{\bar{4}}$$

M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺

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$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1 \models_T 2$)

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$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1 \models_T 2$)
$1 \bar{4} 2 \bar{3}$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1, \bar{4} \models_T \bar{3}$)

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	1, $\bar{2} \vee 3, \bar{4}$	no	
1 $\bar{4}$	1, $\bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
1 $\bar{4}$ 2	1, $\bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1 \models_T 2$)
1 $\bar{4}$ 2 $\bar{3}$	1, $\bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1, \bar{4} \models_T \bar{3}$)
1 $\bar{4}$ 2 $\bar{3}$	1, $\bar{2} \vee 3, \bar{4}$	$\bar{2} \vee 3$	by Conflict

THEORY PROPAGATION EXAMPLE

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M	F	C	rule
	$1, \bar{2} \vee 3, \bar{4}$	no	
$1 \bar{4}$	$1, \bar{2} \vee 3, \bar{4}$	no	by Propagate ⁺
$1 \bar{4} 2$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1 \models_T 2$)
$1 \bar{4} 2 \bar{3}$	$1, \bar{2} \vee 3, \bar{4}$	no	by T-Propagate ($1, \bar{4} \models_T \bar{3}$)
$1 \bar{4} 2 \bar{3}$	$1, \bar{2} \vee 3, \bar{4}$	$\bar{2} \vee 3$	by Conflict
fail			by Fail

Note: *T*-propagation eliminates search altogether in this case
no applications of **Decide** are needed

At the core, current lazy SMT solvers are implementations of the transition system with rules

- (1) Propagate, Decide, Conflict, Explain, Backjump, Fail
- (2) *T*-Conflict, *T*-Propagate, *T*-Explain
- (3) Learn, Forget, Restart

At the core, current lazy SMT solvers are implementations of the transition system with rules

- (1) **Propagate, Decide, Conflict, Explain, Backjump, Fail**
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Basic DPLL Modulo Theories $\stackrel{\text{def}}{=} (1) + (2)$

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Updated terminology:

Irreducible state: state to which no Basic DPLL MT rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

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Proposition (Termination) Every execution in which

(a) **Learn/Forget** are applied only *finitely many times* and

(b) **Restart** is applied with *increased periodicity*

is finite.

Lemma Every exhausted execution ends with either $C = \text{no}$ or *fail*.

Updated terminology:

Irreducible state: state to which no Basic DPLL MT rules apply

Execution: sequence of transitions allowed by the rules and starting with $M = \emptyset$ and $C = \text{no}$

Exhausted execution: execution ending in an irreducible state

Proposition (Soundness) For every exhausted execution starting with $F = F_0$ and ending with fail , the clause set F_0 is T -unsatisfiable.

Proposition (Completeness) For every exhausted execution starting with $F = F_0$ and ending with $C = \text{no}$, F_0 is T -satisfiable; specifically, M is T -satisfiable and $M \models_p F_0$.

The approach formalized so far can be implemented with a simple architecture named $\text{DPLL}(T)$ [GHN⁺04, NOT06]

$$\text{DPLL}(T) = \text{DPLL}(X) \text{ engine} + T\text{-solver}$$

The approach formalized so far can be implemented with a simple architecture named **DPLL(T)** [GHN⁺04, NOT06]

$$\text{DPLL}(T) = \text{DPLL}(X) \text{ engine} + T\text{-solver}$$

DPLL(X):

- Very **similar to a SAT solver**, enumerates Boolean models
- **Not allowed**: pure literal, blocked literal detection, ...
- **Required**: incremental addition of clauses
- **Desirable**: partial model detection

The approach formalized so far can be implemented with a simple architecture named $\text{DPLL}(T)$ [GHN⁺04, NOT06]

$$\text{DPLL}(T) = \text{DPLL}(X) \text{ engine} + T\text{-solver}$$

T -solver:

- Checks the T -satisfiability of conjunctions of literals
- Computes theory propagations
- Produces explanations of T -unsatisfiability/propagation
- Must be incremental and backtrackable

For certain theories, determining that a set M is T -unsatisfiable requires reasoning by cases.

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$$M = \{ \underbrace{r(w(a, i, x), j) \neq x}_1, \underbrace{r(w(a, i, x), j) \neq r(a, j)}_2 \}$$

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$i = j$) Then, $r(w(a, i, x), j) = x$. Contradiction with 1.

$i \neq j$) Then, $r(w(a, i, x), j) = r(a, j)$. Contradiction with 2.

Conclusion: M is T -unsatisfiable

A *complete* T -solver reasons by cases via (internal) case splitting and backtracking mechanisms

A *complete* *T*-solver reasons by cases via (internal) case splitting and backtracking mechanisms

An alternative is to lift case splitting and backtracking from the *T*-solver to the SAT engine

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Basic idea: encode case splits as sets of clauses and send them as needed to the SAT engine for it to split on them [BNOT06]

A *complete* *T*-solver reasons by cases via (internal) case splitting and backtracking mechanisms

An alternative is to *lift case splitting and backtracking* from the *T*-solver *to the SAT engine*

Basic idea: encode case splits as sets of clauses and send them as needed to the SAT engine for *it* to split on them [BNOT06]

Possible benefits:

- All case-splitting is coordinated by the SAT engine
- Only have to implement case-splitting infrastructure in one place
- Can learn a wider class of lemmas

Basic idea: encode case splits as a set of clauses and send them as needed to the SAT engine for it to split on them

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Basic Scenario:

$$M = \{ \dots, s = \underbrace{r(w(a, i, t), j)}_{s'}, \dots \}$$

- Main SMT module: “Is M T -unsatisfiable?”

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Basic Scenario:

$$M = \{ \dots, s = \underbrace{r(w(a, i, t), j)}_{s'}, \dots \}$$

- **Main SMT module:** “Is M T -unsatisfiable?”
- **T -solver:** “I do not know yet, but it will help me if you consider these *theory lemmas*:

$$s = s' \wedge i = j \Rightarrow s = t, \quad s = s' \wedge i \neq j \Rightarrow s = r(a, j) "$$

To model the generation of theory lemmas for case splits, add the rule

T-Learn

$$\frac{\models_T \exists \mathbf{v}(l_1 \vee \cdots \vee l_n) \quad l_1, \dots, l_n \in L_S \quad \mathbf{v} \text{ vars not in } F}{F := F \cup \{l_1 \vee \cdots \vee l_n\}}$$

where L_S is a finite set of literals dependent on the initial set of clauses (see [BNOT06] for a formal definition of L_S)

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where L_S is a finite set of literals dependent on the initial set of clauses (see [BNOT06] for a formal definition of L_S)

Note: For many theories with a theory solver, there exists an appropriate finite L_S for every input F

The set L_S does not need to be computed explicitly

Now we can relax the requirement on the theory solver:

When $M \models_p F$, it must *either*

- determine whether $M \models_T \perp$ or
- generate a new clause by *T-Learn* containing at least one literal of L_S undefined in M

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The T -solver is *required* to determine whether $M \models_T \perp$ *only* if all literals in L_S are defined in M

Note: In practice, to determine if $M \models_T \perp$, the T -solver only needs a small subset of L_S to be defined in M

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

M	F	rule
$x = y \cup z$	F	by Propagate⁺

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

M	F	rule
$x = y \cup z$	F	by Propagate ⁺
$x = y \cup z \bullet y = \emptyset$	F	by Decide

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

M	F	rule
$x = y \cup z$	F	by Propagate⁺
$x = y \cup z \bullet y = \emptyset$	F	by Decide
$x = y \cup z \bullet y = \emptyset x \neq z$	F	by Propagate

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

	M	F	rule
	$x = y \cup z$	F	by Propagate ⁺
	$x = y \cup z \bullet y = \emptyset$	F	by Decide
	$x = y \cup z \bullet y = \emptyset \ x \neq z$	F	by Propagate
	$x = y \cup z \bullet y = \emptyset \ x \neq z \quad F, (x = z \vee e \in x \vee e \in z),$		by T-Learn

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

M	F	rule
$x = y \cup z$	F	by Propagate ⁺
$x = y \cup z \bullet y = \emptyset$	F	by Decide
$x = y \cup z \bullet y = \emptyset \ x \neq z$	F	by Propagate
$x = y \cup z \bullet y = \emptyset \ x \neq z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by T-Learn

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

	M	F	rule
	$x = y \cup z$	F	by Propagate⁺
	$x = y \cup z \bullet y = \emptyset$	F	by Decide
	$x = y \cup z \bullet y = \emptyset x \neq z$	F	by Propagate
	$x = y \cup z \bullet y = \emptyset x \neq z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by T-Learn
	$x = y \cup z \bullet y = \emptyset x \neq z \bullet e \in x$	$F, (x = z \vee e \in x \vee e \in z),$	by Decide

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

	M	F	rule
	$x = y \cup z$	F	by Propagate ⁺
	$x = y \cup z \bullet y = \emptyset$	F	by Decide
	$x = y \cup z \bullet y = \emptyset x \neq z$	F	by Propagate
	$x = y \cup z \bullet y = \emptyset x \neq z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by T-Learn
	$x = y \cup z \bullet y = \emptyset x \neq z \bullet e \in x$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by Decide

$$F: x = y \cup z \wedge y \neq \emptyset \vee x \neq z$$

	M	F	rule
	$x = y \cup z$	F	by Propagate ⁺
	$x = y \cup z \bullet y = \emptyset$	F	by Decide
	$x = y \cup z \bullet y = \emptyset x \neq z$	F	by Propagate
	$x = y \cup z \bullet y = \emptyset x \neq z$	$F, (x = z \vee e \in x \vee e \in z),$ $(x = z \vee e \notin x \vee e \notin z)$	by T-Learn
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	$x = y \cup z \bullet y = \emptyset x \neq z \bullet e \in x e \notin z$	$F, (x = z \vee e \in x \vee e \in z),$	by Propagate

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T-solver can make the following deductions at this point:

$$e \in x \ \dots \Rightarrow e \in y \cup z \ \dots \Rightarrow e \in y \ \dots \Rightarrow e \in \emptyset \Rightarrow \perp$$

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$x = y \cup z$	F	by Propagate ⁺
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$$e \in x \ \dots \Rightarrow e \in y \cup z \ \dots \Rightarrow e \in y \ \dots \Rightarrow e \in \emptyset \Rightarrow \perp$$

This enables an application of **T-Conflict** with clause

$$x \neq y \cup z \vee y \neq \emptyset \vee x = z \vee e \notin x \vee e \in z$$

Correctness results can be extended to the new rule.

Soundness: The new **T-Learn** rule maintains satisfiability of the clause set.

Completeness: As long as the theory solver can decide $M \models_T \perp$ when all literals in L_S are determined, the system is still complete.

Termination: The system terminates under the same conditions as before. Roughly:

- Any lemma is (re)learned only finitely many times
- **Restart** is applied with increased periodicity

COMBINING THEORIES AND THEIR SOLVERS

Many applications give rise to **mixed-theory** formulas like:

$$a \approx b + 2 \wedge A = \text{store}(B, a + 1, 4) \wedge \\ A[b + 3] = 2 \vee f(a - 1) \neq f(b + 1)$$

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Question: Given solvers for each theory, can we **combine them modularly** into one for $T_{\text{LA}} \cup T_{\text{A}} \cup T_{\text{UF}}$?

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Question: Given solvers for each theory, can we **combine them modularly** into one for $T_{\text{LA}} \cup T_{\text{A}} \cup T_{\text{UF}}$?

Under certain conditions, we can do it with the **Nelson-Oppen combination method** [NO79, Opp80]

MOTIVATING EXAMPLE (CONVEX CASE)

Consider the following set of literals over $T_{\text{LRA}} \cup T_{\text{UF}}$
(T_{LRA} , linear **real** arithmetic):

$$\begin{aligned}f(f(x) - f(y)) &= a \\f(0) &> a + 2 \\x &= y\end{aligned}$$

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$$\begin{aligned} f(f(x) - f(y)) = a &\implies f(e_1) = a &&\implies f(e_1) = a \\ e_1 &= f(x) - f(y) &&e_1 = e_2 - e_3 \\ & &&e_2 = f(x) \\ & &&e_3 = f(y) \end{aligned}$$

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$$\begin{array}{lll} f(0) > a + 2 & \implies & f(e_4) > a + 2 \implies f(e_4) = e_5 \\ & & e_4 = 0 & e_4 = 0 \\ & & & e_5 > a + 2 \end{array}$$

Second step: exchange entailed *interface equalities*, equalities over shared constants $e_1, e_2, e_3, e_4, e_5, a$

L_1	L_2
$f(e_1) = a$	$e_2 - e_3 = e_1$
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Third step: check for

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$$L_1 \not\models_{\text{UF}} \perp$$

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Report **unsatisfiable**

MOTIVATING EXAMPLE (NON-CONVEX CASE)

Consider the following **unsatisfiable** set of literals over $T_{LIA} \cup T_{UF}$ (T_{LIA} , linear **integer** arithmetic):

$$1 \leq x \leq 2$$

$$f(1) = a$$

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$$f(1) = a \implies f(e_1) = a$$
$$e_1 = 1$$

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$$\begin{aligned}1 &\leq x \leq 2 \\ f(1) &= a \\ f(2) &= f(1) + 3 \\ a &= b + 2\end{aligned}$$

First step: *purify* literals so that each belongs to a single theory

$$\begin{aligned}f(2) = f(1) + 3 &\implies e_2 = 2 \\ f(e_2) &= e_3 \\ f(e_1) &= e_4 \\ e_3 &= e_4 + 3\end{aligned}$$

Second step: exchange entailed *interface equalities* over shared constants $x, e_1, a, b, e_2, e_3, e_4$

L_1	L_2
$1 \leq x$	$f(e_1) = a$
$x \leq 2$	$f(x) = b$
$e_1 = 1$	$f(e_2) = e_3$
$a = b + 2$	$f(e_1) = e_4$
$e_2 = 2$	
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No more entailed equalities, but $L_1 \models_{LIA} x = e_1 \vee x = e_2$

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Consider each case of $x = e_1 \vee x = e_2$ separately

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Case 1) $x = e_1$

Second step: exchange entailed *interface equalities* over shared constants $x, e_1, a, b, e_2, e_3, e_4$

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$L_2 \models_{\text{UF}} a = b$, which entails \perp when sent to L_1

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Case 2) $x = e_2$

Second step: exchange entailed *interface equalities* over shared constants $x, e_1, a, b, e_2, e_3, e_4$

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MOTIVATING EXAMPLE (NON-CONVEX CASE)

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$L_2 \models_{\text{UF}} e_3 = b$, which entails \perp when sent to L_1

- For $i = 1, 2$, let T_i be a first-order theory of *signature* Σ_i (set of function and predicate symbols in T_i other than $=$)
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We consider only input problems of the form

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Note: Because of purification, there is *no loss of generality* in considering only ground $(\Sigma_i \cup \mathcal{C})$ -literals

Bare-bones, non-deterministic, non-incremental version

[Opp80, Rin96, TH96]:

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1. Guess an *arrangement* A , i.e., a set of equalities and disequalities over \mathcal{C} such that

$$c = d \in A \text{ or } c \neq d \in A \text{ for all } c, d \in \mathcal{C}$$

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3. Otherwise, return **sat**

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Proposition (Soundness) If the method returns **unsat** for **every** arrangement, the input is $(T_1 \cup T_2)$ -unsatisfiable.

(Because satisfiability in $(T_1 \cup T_2)$ is always preserved)

Proposition (Completeness) If $\Sigma_1 \cap \Sigma_2 = \emptyset$ and T_1 and T_2 are **stably infinite**, when the method returns **sat** for **some** arrangement, the input is $(T_1 \cup T_2)$ -satisfiable.

Def. A theory T is *stably infinite* iff every quantifier-free T -satisfiable formula is satisfiable in an *infinite* model of T

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Many *interesting* theories are stably infinite:

- Theories of an *infinite structure* (e.g., integer arithmetic)
- *Complete* theories with an infinite model (e.g., theory of dense linear orders, theory of lists)
- *Convex* theories (e.g., EUF, linear real arithmetic)

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- *Convex* theories (e.g., EUF, linear real arithmetic)

Def. A theory T is *convex* iff, for any set L of literals
 $L \models_T s_1 = t_1 \vee \cdots \vee s_n = t_n \implies L \models_T s_i = t_i$ for some i

Note: With *convex theories*, *arrangements* do not need to be guessed, they can be computed by (theory) propagation

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Other interesting theories are *not* stably infinite:

- Theories of a finite structure (e.g., theory of bit vectors of finite size, arithmetic modulo n)
- Theories with models of bounded cardinality (e.g., theory of strings of bounded length)
- Some equational/Horn theories

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The Nelson-Oppen method has been *extended to* some classes of *non-stably infinite theories* [TZ05, RRZ05, JB10]

Let T_1, \dots, T_n be theories with respective solvers S_1, \dots, S_n

How can we integrate all of them *cooperatively* into a single SMT solver for $T = T_1 \cup \dots \cup T_n$?

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Quick Solution:

1. Combine S_1, \dots, S_n with Nelson-Oppen into a theory solver for T
2. Build a DPLL(T) solver as usual

Let T_1, \dots, T_n be theories with respective solvers S_1, \dots, S_n

How can we integrate all of them **cooperatively** into a single SMT solver for $T = T_1 \cup \dots \cup T_n$?

Better Solution [Bar02, BBC⁺05b, BNOT06]:

1. Extend DPLL(T) to DPLL(T_1, \dots, T_n)
2. **Lift** Nelson-Oppen **to the** DPLL(X_1, \dots, X_n) **level**
3. Build a DPLL(T_1, \dots, T_n) solver

- Let $n = 2$, for simplicity
- Let T_i be of signature Σ_i for $i = 1, 2$, with $\Sigma_1 \cap \Sigma_2 = \emptyset$
- Let \mathcal{C} be a set of *free* constants
- Assume wlog that each input literal has signature $(\Sigma_1 \cup \mathcal{C})$ or $(\Sigma_2 \cup \mathcal{C})$ (no *mixed* literals)
- Let $M|_i \stackrel{\text{def}}{=} \{(\Sigma_i \cup \mathcal{C})\text{-literals of } M \text{ and their complement}\}$
- Let $I(M) \stackrel{\text{def}}{=} \{c = d \mid c, d \text{ occur in } \mathcal{C}, M|_1 \text{ and } M|_2\} \cup \{c \neq d \mid c, d \text{ occur in } \mathcal{C}, M|_1 \text{ and } M|_2\}$
(*interface literals*)

Propagate, Conflict, Explain, Backjump, Fail (unchanged)

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$$\text{Decide } \frac{l \in \text{Lit}(F) \cup I(M) \quad l, \bar{l} \notin M}{M := M \bullet l}$$

Only change: decide on interface equalities as well

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$$\text{Decide} \frac{l \in \text{Lit}(F) \cup I(M) \quad l, \bar{l} \notin M}{M := M \bullet l}$$

Only change: decide on interface equalities as well

$$T\text{-Propagate} \frac{l \in \text{Lit}(F) \cup I(M) \quad i \in \{1, 2\} \quad M \models_{T_i} l \quad l, \bar{l} \notin M}{M := M \bullet l}$$

Only change: propagate interface equalities as well, but reason locally in each T_i

T-Conflict

$$\frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_{T_i} \perp \quad i \in \{1, 2\}}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

T-Explain

$$\frac{C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_{T_i} \bar{l} \quad i \in \{1, 2\} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

Only change: reason locally in each T_i

T-Conflict

$$\frac{C = \text{no} \quad l_1, \dots, l_n \in M \quad l_1, \dots, l_n \models_{T_i} \perp \quad i \in \{1, 2\}}{C := \bar{l}_1 \vee \dots \vee \bar{l}_n}$$

T-Explain

$$\frac{C = l \vee D \quad \bar{l}_1, \dots, \bar{l}_n \models_{T_i} \bar{l} \quad i \in \{1, 2\} \quad \bar{l}_1, \dots, \bar{l}_n \prec_M \bar{l}}{C := l_1 \vee \dots \vee l_n \vee D}$$

Only change: reason locally in each T_i

I-Learn

$$\frac{\models_{T_i} l_1 \vee \dots \vee l_n \quad l_1, \dots, l_n \in M|_i \cup I(M) \quad i \in \{1, 2\}}{F := F \cup \{l_1 \vee \dots \vee l_n\}}$$

New rule: for entailed disjunctions of interface literals

EXAMPLE — CONVEX THEORIES

$$F := \underbrace{f(e_1) = a}_{0} \wedge \underbrace{f(x) = e_2}_{1} \wedge \underbrace{f(y) = e_3}_{2} \wedge \underbrace{f(e_4) = e_5}_{3} \wedge \underbrace{x = y}_{4} \wedge$$
$$\underbrace{e_2 - e_3 = e_1}_{5} \wedge \underbrace{e_4 = 0}_{6} \wedge \underbrace{e_5 > a + 2}_{7}$$
$$\underbrace{e_2 = e_3}_{8} \quad \underbrace{e_1 = e_4}_{9} \quad \underbrace{a = e_5}_{10}$$

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 \underbrace{e_2 - e_3 = e_1}_{5} \wedge \underbrace{e_4 = 0}_{6} \wedge \underbrace{e_5 > a + 2}_{7} \\
 \underbrace{e_2 = e_3}_{8} \quad \underbrace{e_1 = e_4}_{9} \quad \underbrace{a = e_5}_{10}
 \end{array}$$

M	F	C	rule
	F	no	

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 F := \underbrace{f(e_1) = a}_{0} \wedge \underbrace{f(x) = e_2}_{1} \wedge \underbrace{f(y) = e_3}_{2} \wedge \underbrace{f(e_4) = e_5}_{3} \wedge \underbrace{x = y}_{4} \wedge \\
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 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺

EXAMPLE — CONVEX THEORIES

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 F := \underbrace{f(e_1) = a}_{0} \wedge \underbrace{f(x) = e_2}_{1} \wedge \underbrace{f(y) = e_3}_{2} \wedge \underbrace{f(e_4) = e_5}_{3} \wedge \underbrace{x = y}_{4} \wedge \\
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 \underbrace{e_2 = e_3}_{8} \quad \underbrace{e_1 = e_4}_{9} \quad \underbrace{a = e_5}_{10}
 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺
0 1 2 3 4 5 6 7 8		F	no	by T-Propagate (1, 2, 4 \models_{UF} 8)

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 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺
0 1 2 3 4 5 6 7 8		F	no	by T-Propagate (1, 2, 4 \models_{UF} 8)
0 1 2 3 4 5 6 7 8 9		F	no	by T-Propagate (5, 6, 8 \models_{LRA} 9)

EXAMPLE — CONVEX THEORIES

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 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺
0 1 2 3 4 5 6 7 8		F	no	by T-Propagate (1, 2, 4 \models_{UF} 8)
0 1 2 3 4 5 6 7 8 9		F	no	by T-Propagate (5, 6, 8 \models_{LRA} 9)
0 1 2 3 4 5 6 7 8 9 10		F	no	by T-Propagate (0, 3, 9 \models_{UF} 10)

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 F := \underbrace{f(e_1) = a}_{0} \wedge \underbrace{f(x) = e_2}_{1} \wedge \underbrace{f(y) = e_3}_{2} \wedge \underbrace{f(e_4) = e_5}_{3} \wedge \underbrace{x = y}_{4} \wedge \\
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 \underbrace{e_2 = e_3}_{8} \quad \underbrace{e_1 = e_4}_{9} \quad \underbrace{a = e_5}_{10}
 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺
0 1 2 3 4 5 6 7 8		F	no	by T-Propagate (1, 2, 4 \models_{UF} 8)
0 1 2 3 4 5 6 7 8 9		F	no	by T-Propagate (5, 6, 8 \models_{LRA} 9)
0 1 2 3 4 5 6 7 8 9 10		F	no	by T-Propagate (0, 3, 9 \models_{UF} 10)
0 1 2 3 4 5 6 7 8 9 10		F	$\bar{7} \vee \bar{10}$	by T-Conflict (7, 10 $\models_{\text{LRA}} \perp$)

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$$\begin{array}{c}
 F := \underbrace{f(e_1) = a}_{0} \wedge \underbrace{f(x) = e_2}_{1} \wedge \underbrace{f(y) = e_3}_{2} \wedge \underbrace{f(e_4) = e_5}_{3} \wedge \underbrace{x = y}_{4} \wedge \\
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 \underbrace{e_2 = e_3}_{8} \quad \underbrace{e_1 = e_4}_{9} \quad \underbrace{a = e_5}_{10}
 \end{array}$$

	M	F	C	rule
		F	no	
0 1 2 3 4 5 6 7		F	no	by Propagate ⁺
0 1 2 3 4 5 6 7 8		F	no	by T-Propagate (1, 2, 4 \models_{UF} 8)
0 1 2 3 4 5 6 7 8 9		F	no	by T-Propagate (5, 6, 8 \models_{LRA} 9)
0 1 2 3 4 5 6 7 8 9 10		F	no	by T-Propagate (0, 3, 9 \models_{UF} 10)
0 1 2 3 4 5 6 7 8 9 10		F	$\bar{7} \vee \bar{10}$	by T-Conflict (7, 10 $\models_{\text{LRA}} \perp$)
fail				by Fail

EXAMPLE — NON-CONVEX THEORIES

$$\begin{aligned}
 F := & \underbrace{f(e_1) = a}_{0} \wedge \underbrace{f(x) = b}_{1} \wedge \underbrace{f(e_2) = e_3}_{2} \wedge \underbrace{f(e_1) = e_4}_{3} \wedge \\
 & \underbrace{1 \leq x}_{4} \wedge \underbrace{x \leq 2}_{5} \wedge \underbrace{e_1 = 1}_{6} \wedge \underbrace{a = b + 2}_{7} \wedge \underbrace{e_2 = 2}_{8} \wedge \underbrace{e_3 = e_4 + 3}_{9} \\
 & \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{aligned}$$

EXAMPLE — NON-CONVEX THEORIES

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 F := \underbrace{f(e_1) = a}_0 \wedge \underbrace{f(x) = b}_1 \wedge \underbrace{f(e_2) = e_3}_2 \wedge \underbrace{f(e_1) = e_4}_3 \wedge \\
 \underbrace{1 \leq x}_4 \wedge \underbrace{x \leq 2}_5 \wedge \underbrace{e_1 = 1}_6 \wedge \underbrace{a = b + 2}_7 \wedge \underbrace{e_2 = 2}_8 \wedge \underbrace{e_3 = e_4 + 3}_9 \\
 \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{array}$$

M F

C rule

F

no

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 F := \underbrace{f(e_1) = a}_0 \wedge \underbrace{f(x) = b}_1 \wedge \underbrace{f(e_2) = e_3}_2 \wedge \underbrace{f(e_1) = e_4}_3 \wedge \\
 \underbrace{1 \leq x}_4 \wedge \underbrace{x \leq 2}_5 \wedge \underbrace{e_1 = 1}_6 \wedge \underbrace{a = b + 2}_7 \wedge \underbrace{e_2 = 2}_8 \wedge \underbrace{e_3 = e_4 + 3}_9 \\
 \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{array}$$

M F		C	rule
		no	
0 ... 9	F	no	by Propagate ⁺

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 F := \underbrace{f(e_1) = a}_0 \wedge \underbrace{f(x) = b}_1 \wedge \underbrace{f(e_2) = e_3}_2 \wedge \underbrace{f(e_1) = e_4}_3 \wedge \\
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 \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{\text{UF}} 10$)

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
 \underbrace{1 \leq x}_4 & \wedge & \underbrace{x \leq 2}_5 & \wedge & \underbrace{e_1 = 1}_6 & \wedge & \underbrace{a = b + 2}_7 & \wedge & \underbrace{e_2 = 2}_8 & \wedge & \underbrace{e_3 = e_4 + 3}_9
 \end{array} \\
 \begin{array}{cccc}
 \underbrace{a = e_4}_{10} & \underbrace{x = e_1}_{11} & \underbrace{x = e_2}_{12} & \underbrace{a = b}_{13}
 \end{array}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{\text{UF}} 10$)
$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{\text{LIA}} \bar{4} \vee \bar{5} \vee 11 \vee 12$)

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
 \underbrace{1 \leq x}_4 & \wedge & \underbrace{x \leq 2}_5 & \wedge & \underbrace{e_1 = 1}_6 & \wedge & \underbrace{a = b + 2}_7 & \wedge & \underbrace{e_2 = 2}_8 & \wedge & \underbrace{e_3 = e_4 + 3}_9
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 \end{array}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{UF} 10$)
$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
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 \end{array}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{\text{UF}} 10$)
$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{\text{LIA}} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, 11 \models_{\text{UF}} 13$)

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
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 \end{array} \\
 \begin{array}{cccc}
 \underbrace{a = e_4}_{10} & \underbrace{x = e_1}_{11} & \underbrace{x = e_2}_{12} & \underbrace{a = b}_{13}
 \end{array}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{\text{UF}} 10$)
$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{\text{LIA}} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, 11 \models_{\text{UF}} 13$)
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict ($7, 13 \models_{\text{UF}} \perp$)

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
 \underbrace{1 \leq x}_4 & \wedge & \underbrace{x \leq 2}_5 & \wedge & \underbrace{e_1 = 1}_6 & \wedge & \underbrace{a = b + 2}_7 & \wedge & \underbrace{e_2 = 2}_8 & \wedge & \underbrace{e_3 = e_4 + 3}_9
 \end{array} \\
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 \underbrace{a = e_4}_{10} & \underbrace{x = e_1}_{11} & \underbrace{x = e_2}_{12} & \underbrace{a = b}_{13}
 \end{array}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{UF} 10$)
$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, 11 \models_{UF} 13$)
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict ($7, 13 \models_{UF} \perp$)
$0 \dots 9 \ 10 \ \bar{13}$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
 \underbrace{1 \leq x}_{4} & \wedge & \underbrace{x \leq 2}_{5} & \wedge & \underbrace{e_1 = 1}_{6} & \wedge & \underbrace{a = b + 2}_{7} & \wedge & \underbrace{e_2 = 2}_{8} & \wedge & \underbrace{e_3 = e_4 + 3}_{9}
 \end{array} \\
 \begin{array}{cccc}
 \underbrace{a = e_4}_{10} & \underbrace{x = e_1}_{11} & \underbrace{x = e_2}_{12} & \underbrace{a = b}_{13}
 \end{array}
 \end{array}$$

M	F	C	rule
	F	no	
$0 \dots 9$	F	no	by Propagate ⁺
$0 \dots 9 \ 10$	F	no	by T-Propagate ($0, 3 \models_{UF} 10$)
$0 \dots 9 \ 10$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \overline{4} \vee \overline{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, 11 \models_{UF} 13$)
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	$\overline{7} \vee \overline{13}$	by T-Conflict ($7, 13 \models_{UF} \perp$)
$0 \dots 9 \ 10 \ \overline{13}$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Backjump
$0 \dots 9 \ 10 \ \overline{13} \ \overline{11}$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, \overline{13} \models_{UF} \overline{11}$)

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
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 \underbrace{a = e_4}_{10} \quad \underbrace{x = e_1}_{11} \quad \underbrace{x = e_2}_{12} \quad \underbrace{a = b}_{13}
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$0 \dots 9 \ 10$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by I-Learn ($\models_{LIA} \bar{4} \vee \bar{5} \vee 11 \vee 12$)
$0 \dots 9 \ 10 \bullet 11$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, 11 \models_{UF} 13$)
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	$\bar{7} \vee \bar{13}$	by T-Conflict ($7, 13 \models_{UF} \perp$)
$0 \dots 9 \ 10 \ \bar{13}$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
$0 \dots 9 \ 10 \ \bar{13} \ \bar{11}$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, \bar{13} \models_{UF} \bar{11}$)
$0 \dots 9 \ 10 \ \bar{13} \ \bar{11} \ 12$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate

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$0 \dots 9 \ 10 \ \bar{13}$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Backjump
$0 \dots 9 \ 10 \ \bar{13} \ \bar{11}$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, \bar{13} \models_{UF} \bar{11}$)
$0 \dots 9 \ 10 \ \bar{13} \ \bar{11} \ 12$	$F, \bar{4} \vee \bar{5} \vee 11 \vee 12$	no	by Propagate
...			(exercise)

EXAMPLE — NON-CONVEX THEORIES

$$\begin{array}{c}
 \begin{array}{ccccccc}
 \overbrace{f(e_1) = a}^0 & \wedge & \overbrace{f(x) = b}^1 & \wedge & \overbrace{f(e_2) = e_3}^2 & \wedge & \overbrace{f(e_1) = e_4}^3 \\
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M	F	C	rule
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$0 \dots 9 \ 10 \bullet 11$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Decide
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, 11 \models_{UF} 13$)
$0 \dots 9 \ 10 \bullet 11 \ 13$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	$\overline{7} \vee \overline{13}$	by T-Conflict ($7, 13 \models_{UF} \perp$)
$0 \dots 9 \ 10 \ \overline{13}$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Backjump
$0 \dots 9 \ 10 \ \overline{13} \ \overline{11}$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by T-Propagate ($0, 1, \overline{13} \models_{UF} \overline{11}$)
$0 \dots 9 \ 10 \ \overline{13} \ \overline{11} \ 12$	$F, \overline{4} \vee \overline{5} \vee 11 \vee 12$	no	by Propagate
...			(exercise)
fail	by Fail

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