

# Model-Constructing Satisfiability Calculus

A Model-Based Approach to SMT

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# Outline

- 1 Introduction
- 2 Arithmetic
- 3 Theory Combination
- 4 Conclusion

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# Satisfiability Modulo Theories and DPLL(T)

## Problem

Check a given formula for satisfiability modulo the union of background theories.

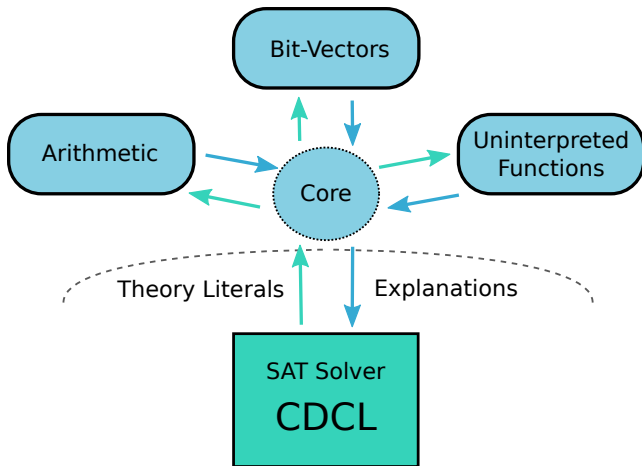
## Example (QF\_UFLRA)

$$(z = 1 \vee z = 0) \wedge (x - y + z = 1) \wedge (f(x) > f(y))$$

Main idea behind DPLL(T):

- 1 Use a SAT solver to enumerate the Boolean structure.
- 2 Check Boolean assignments with a decision procedure.

## DPLL(T) Architecture



## DPLL(T): Pros and Cons

### Good

- Simple interface
- Only conjunctions of constraints
- General combination methods

# DPLL(T): Pros and Cons

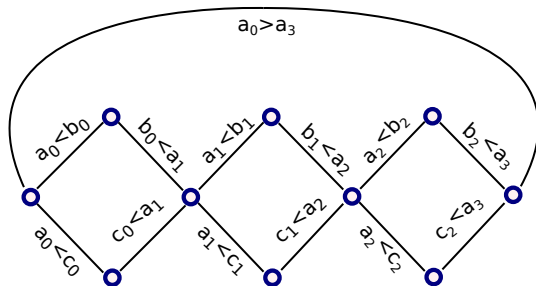
## Good

- Simple interface
- Only conjunctions of constraints
- General combination methods

## What can be improved?

- Simple interface can be restrictive
- Arbitrary conjunctions of constraints
- General combination methods

## DPLL(T): Simple Interface Issues

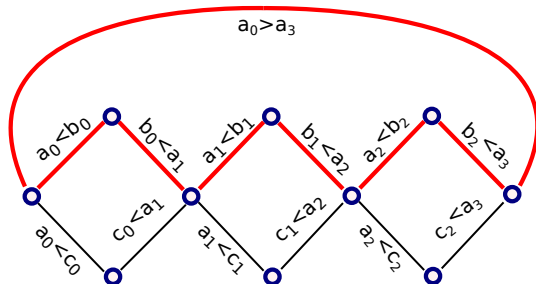


### Example (Diamonds)

$$a_0 > a_n \wedge \bigwedge_{k=0}^{n-1} ((a_k < b_k \wedge b_k < a_{k+1}) \vee (a_k < c_k \wedge c_k < a_{k+1}))$$



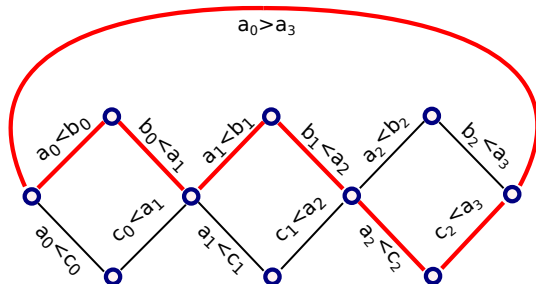
## DPLL(T): Simple Interface Issues



### Example (Diamonds)

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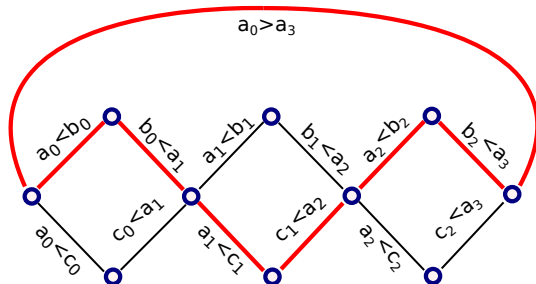
## DPLL(T): Simple Interface Issues



### Example (Diamonds)

$$a_0 > a_n \wedge \bigwedge_{k=0}^{n-1} ((a_k < b_k \wedge b_k < a_{k+1}) \vee (a_k < c_k \wedge c_k < a_{k+1}))$$

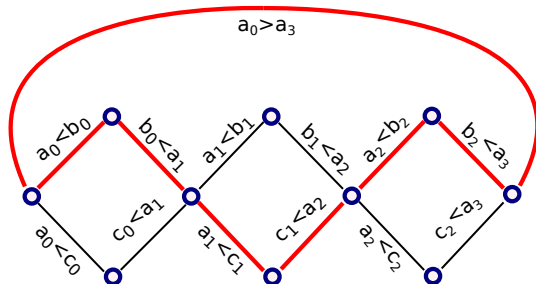
## DPLL(T): Simple Interface Issues



### Example (Diamonds)

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## DPLL(T): Simple Interface Issues

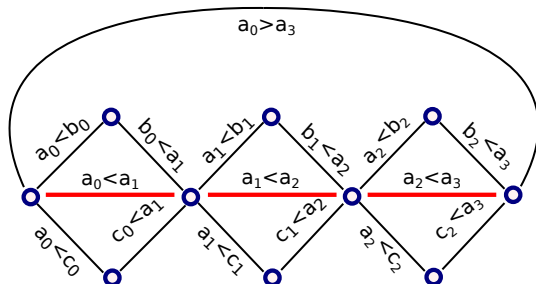


And so on...

Example (1) Exponential enumeration of paths.

$$a_0 > a_n \wedge \bigwedge_{k=0}^{n-1} ((a_k < b_k \wedge b_k < a_{k+1}) \vee (a_k < c_k \wedge c_k < a_{k+1}))$$

## DPLL(T): Simple Interface Issues



### Example (Diamonds)

$$a_0 > a_n \wedge \bigwedge_{k=0}^{n-1} ((a_k < b_k \wedge b_k < a_{k+1}) \vee (a_k < c_k \wedge c_k < a_{k+1}))$$

# DPPL(T): Simple Interface Issues

## How to fix this?

- Extensions of DPLL(T) can add new literals [BNOT06].
- Magic needed to discover these literals [BDdM08, HAMM14].
- More pragmatic approach would be desirable.

## Rethink the Architecture!

- Why is SAT solver special?
- Why the restriction on the interface?
- Let's dig deeper into how SAT solvers work.

# Boolean Satisfiability

$$x_n \vee \cdots \vee x_1 \vee \overline{y_m} \vee \cdots \vee \overline{y_1}$$

- **Resolution procedure** by Davis, Putnam [DP60]
- **Search procedure** by Davis, Logemann, Loveland [DLL62]

## Resolution (DP)

- Find a proof
- Saturation
- Exponential

## Search (DLL)

- Find a model
- Search and backtracking
- Exponential

## Resolution (DP)

$$x \vee y \vee \bar{z}$$

$$x \vee \bar{y} \vee \bar{z}$$

$$\bar{x} \vee y \vee \bar{z}$$

$$\bar{x} \vee \bar{y} \vee \bar{z}$$

$$x \vee y \vee z$$

$$x \vee \bar{y} \vee z$$

$$\bar{x} \vee y \vee z$$

$$\bar{x} \vee \bar{y} \vee z$$



## Resolution (DP)

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$$\bar{x} \vee \bar{y} \vee \bar{z}$$

$$x \vee y \vee z$$

$$x \vee \bar{y} \vee z$$

$$\bar{x} \vee y \vee z$$

$$\bar{x} \vee \bar{y} \vee z$$

### Boolean Resolution

$$\frac{C \vee x \quad D \vee \bar{x}}{C \vee D}$$

# Resolution (DP)

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$

1 Eliminate  $z$

# Resolution (DP)

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$
$x \vee y$	$x \vee \bar{y}$	$\bar{x} \vee y$	$\bar{x} \vee \bar{y}$

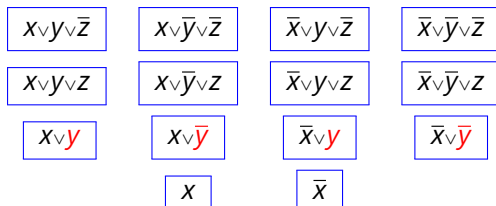
1 Eliminate  $z$

# Resolution (DP)

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$
$x \vee y$	$x \vee \bar{y}$	$\bar{x} \vee y$	$\bar{x} \vee \bar{y}$

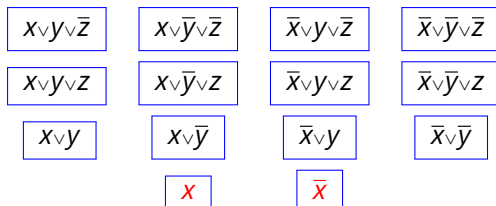
- 1 Eliminate  $z$
- 2 Eliminate  $y$

# Resolution (DP)



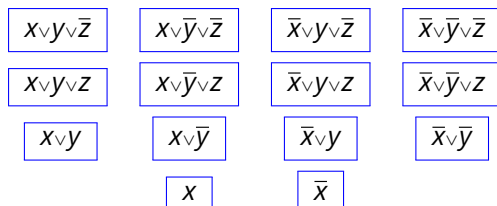
- 1 Eliminate  $z$
- 2 Eliminate  $y$

# Resolution (DP)



- 1 Eliminate  $z$
- 2 Eliminate  $y$
- 3 Eliminate  $x$

# Resolution (DP)



- 1 Eliminate  $z$
- 2 Eliminate  $y$
- 3 Eliminate  $x$
- 4 **unsat**

## Search (DLL)

$$x \vee y \vee \bar{z}$$

$$x \vee \bar{y} \vee \bar{z}$$

$$\bar{x} \vee y \vee \bar{z}$$

$$\bar{x} \vee \bar{y} \vee \bar{z}$$

$$x \vee y \vee z$$

$$x \vee \bar{y} \vee z$$

$$\bar{x} \vee y \vee z$$

$$\bar{x} \vee \bar{y} \vee z$$



# Search (DLL)

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$

1 try  $x \mapsto \top$

# Search (DLL)

$y \vee \bar{z}$	$\bar{y} \vee \bar{z}$
$y \vee z$	$\bar{y} \vee z$

① try  $x \mapsto \top$

① try  $y \mapsto \top$

# Search (DLL)

$\bar{z}$

$z$

①  $\text{try } x \mapsto \top$

①  $\text{try } y \mapsto \top$

# Search (DLL)

$\bar{z}$

$z$

① try  $x \mapsto \top$

① try  $y \mapsto \top$ , unsat

# Search (DLL)

$$y \vee \bar{z}$$

$$\bar{y} \vee \bar{z}$$

$$y \vee z$$

$$\bar{y} \vee z$$

① try  $x \mapsto \top$

① try  $y \mapsto \top$ , unsat

② try  $y \mapsto \perp$

# Search (DLL)

$\bar{z}$

$z$

① try  $x \mapsto \top$

① try  $y \mapsto \top$ , unsat

② try  $y \mapsto \perp$

# Search (DLL)

$\bar{z}$

$z$

① try  $x \mapsto \top$

① try  $y \mapsto \top$ , unsat

② try  $y \mapsto \perp$ , unsat

# Search (DLL)

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$

- ① try  $x \mapsto \top$ 
  - ① try  $y \mapsto \top$ , unsat
  - ② try  $y \mapsto \perp$ , unsat
- ② try  $x \mapsto \perp$  ...



# Boolean Satisfiability: CDCL

Marques-Silva, Sakallah

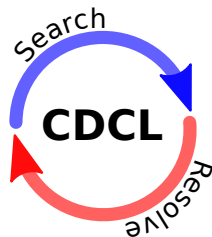
[SS97] GRASP: A new search algorithm for satisfiability

Moskewicz, Madigan, Zhao, Zhang, Malik

[MMZ<sup>+</sup>01] CHAFF: Engineering an efficient SAT solver

## Conflict-Directed Clause Learning

- Use the search to guide resolution
- Use resolution to guide the search



# Boolean Satisfiability: CDCL

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$



# Boolean Satisfiability: CDCL

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$

$\llbracket x \rrbracket$

# Boolean Satisfiability: CDCL

$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$

$\llbracket x, y \rrbracket$

# Boolean Satisfiability: CDCL

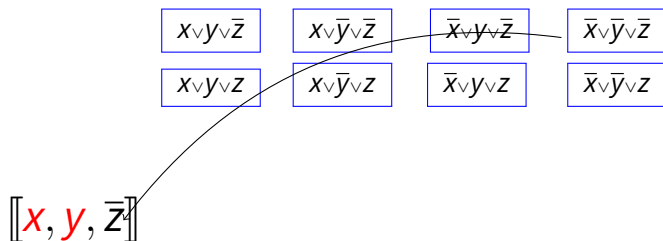
$x \vee y \vee \bar{z}$	$x \vee \bar{y} \vee \bar{z}$	$\bar{x} \vee y \vee \bar{z}$	$\bar{x} \vee \bar{y} \vee \bar{z}$
$x \vee y \vee z$	$x \vee \bar{y} \vee z$	$\bar{x} \vee y \vee z$	$\bar{x} \vee \bar{y} \vee z$

$\llbracket x, y \rrbracket$

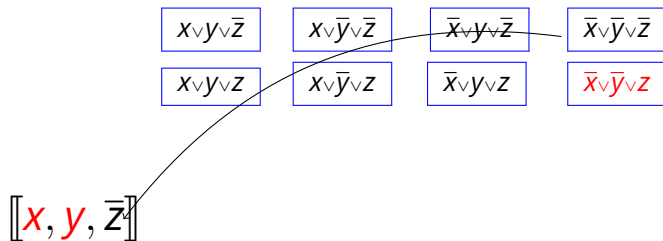
## Unit Propagation

$(\bar{x} \vee \bar{y} \vee \bar{z})$  is unit, propagate  $\bar{z}$ .

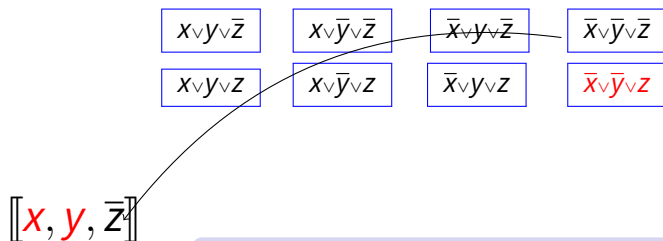
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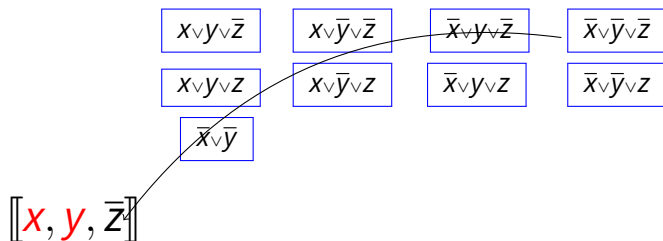


Resolve Conflict

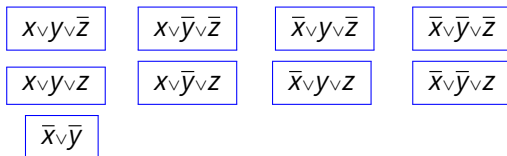
$$\frac{\bar{x} \vee \bar{y} \vee z \quad \bar{x} \vee \bar{y} \vee \bar{z}}{\bar{x} \vee \bar{y}}$$



# Boolean Satisfiability: CDCL

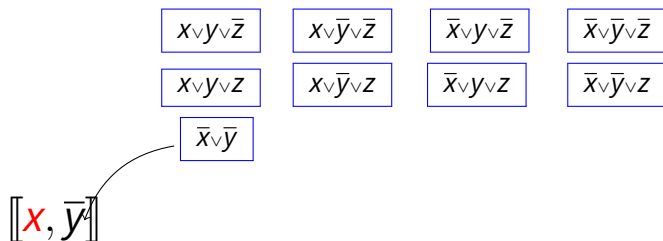


# Boolean Satisfiability: CDCL

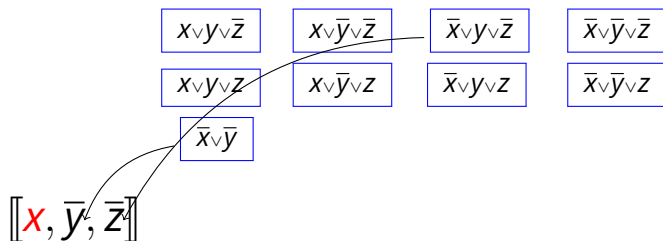


$\llbracket x \rrbracket$

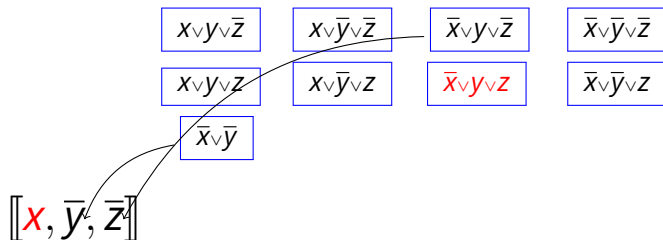
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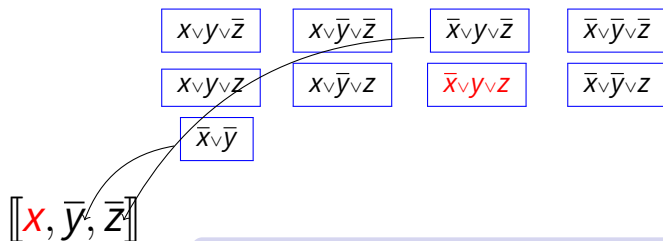
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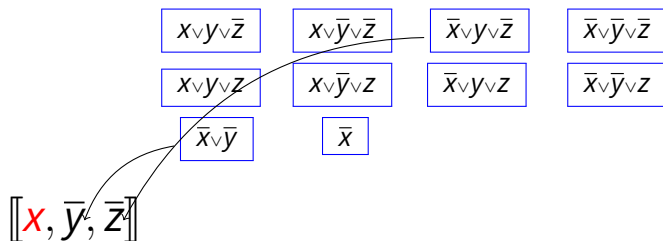
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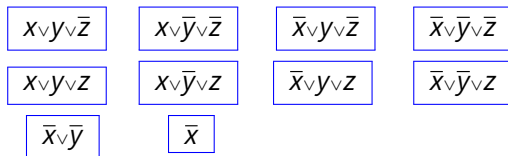
## Resolve Conflict

$$\frac{\frac{\bar{x} \vee y \vee z \quad \bar{x} \vee y \vee \bar{z}}{\bar{x} \vee y} \quad \bar{x} \vee \bar{y}}{\bar{x}}$$

# Boolean Satisfiability: CDCL

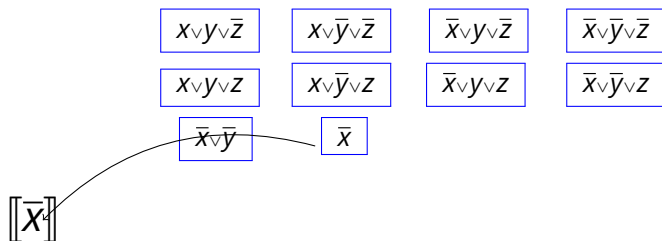


# Boolean Satisfiability: CDCL





# Boolean Satisfiability: CDCL



# Boolean Satisfiability: CDCL

## Model Construction

Build partial model by assigning variables to values

$$[\dots, x, \dots, \bar{y}, \dots, z, \dots] \text{ .}$$

## Unit Reasoning

Reason about unit constraints

$$(\bar{x} \vee y \vee \bar{z} \vee w) \text{ .}$$

## Explain Conflicts

Explain conflicts using clausal reasons

$$(\bar{x} \vee y \vee \bar{z}) \text{ .}$$

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# Linear Real Arithmetic

## Linear Arithmetic

$$a_1x_1 + \cdots + a_nx_n \geq b$$

$$a_1x_1 + \cdots + a_nx_n = b$$

## DPLL(T): Simplex

A model builder for a conjunction of linear constraints.

- Search for a model
- Escape conflicts through pivoting
- Built for the DPLL(T) framework

[DDM06] A fast linear-arithmetic solver for DPLL(T)

# Linear Real Arithmetic

## Linear Arithmetic

$$a_1x_1 + \cdots + a_nx_n \geq b$$

$$a_1x_1 + \cdots + a_nx_n = b$$

## Fourier-Motzkin Resolution

$$\frac{\begin{array}{r} 2x + 3y - z \geq -1 \\ 6x + 9y - 3z \geq -3 \end{array}}{5y + 5z \geq 1} \quad \frac{\begin{array}{r} -3x - 2y + 4z \geq 2 \\ -6x - 4y + 8z \geq 4 \end{array}}$$

- Feels like Boolean resolution (elimination).
- Behaves like Boolean resolution (exponential).

# Linear Real Arithmetic

## Model Construction

Build partial model by assigning variables to values

$$\llbracket \dots, C_1, C_2, \dots, x \mapsto 1/2, \dots, y \mapsto 1/2, \dots, z \mapsto -1, \dots \rrbracket .$$

## Unit Reasoning

Reason about unit constraints

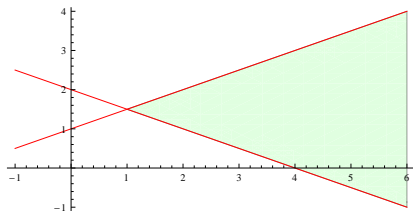
$$C_1 \equiv (x + y + z + w \geq 0) \quad C_2 \equiv (x + y + z - w > 0) .$$

## Explain Conflicts

Explain conflicts using valid clausal reasons

$$(\overline{C_1} \vee \overline{C_2} \vee x + y + z > 0) .$$

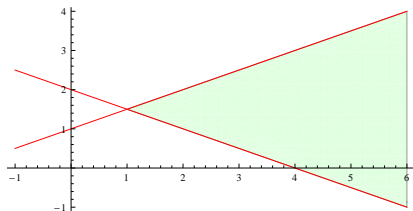
# Linear Real Arithmetic



$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$



# Linear Real Arithmetic

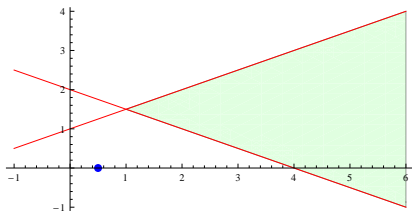


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$[[C_1, C_2]]$$



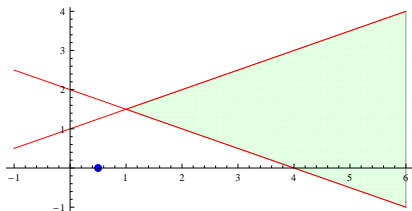
# Linear Real Arithmetic



$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$[[C_1, C_2, x \mapsto 0.5]]$$

# Linear Real Arithmetic



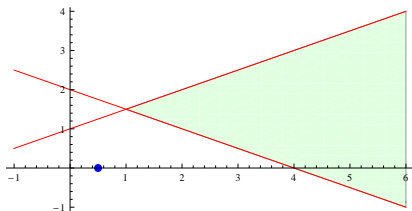
## Unit Constraint Reasoning

$$2y - x - 2 < 0 \Rightarrow (y < 1.25)$$

$$-2y - x + 4 < 0 \Rightarrow (y > 1.75)$$

$$\llbracket \mathcal{C}_1, \mathcal{C}_2, x \mapsto 0.5 \rrbracket$$

# Linear Real Arithmetic

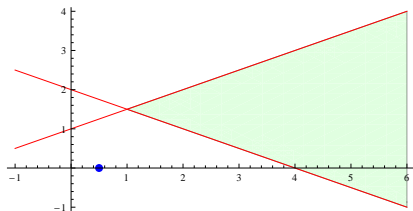


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$\llbracket C_1, C_2, x \mapsto 0.5 \rrbracket$$

Explanation  $C_1 \wedge C_2 \Rightarrow x \neq 0.5$

# Linear Real Arithmetic

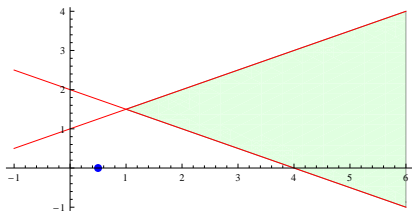


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$[[C_1, C_2, x \mapsto 0.5]]$$

Explanation  $C_1 \wedge C_2 \Rightarrow$

# Linear Real Arithmetic



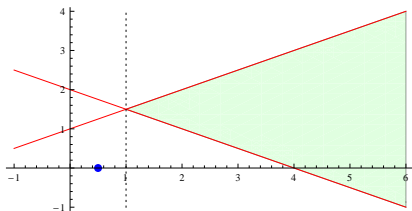
## Fourier-Motzkin

$$\frac{2y - x - 2 < 0 \quad -2y - x + 4 < 0}{-2x + 2 < 0}$$

$$[[C_1, C_2, x \mapsto 0.5]]$$

Explanation  $C_1 \wedge C_2 \Rightarrow$

# Linear Real Arithmetic



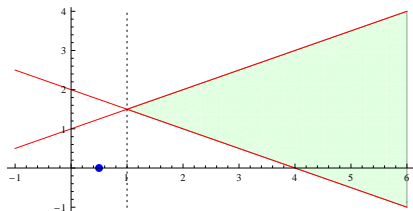
## Fourier-Motzkin

$$\frac{2y - x - 2 < 0 \quad -2y - x + 4 < 0}{-2x + 2 < 0}$$

$$[[C_1, C_2, x \mapsto 0.5]]$$

Explanation  $C_1 \wedge C_2 \Rightarrow x > 1$

# Linear Real Arithmetic

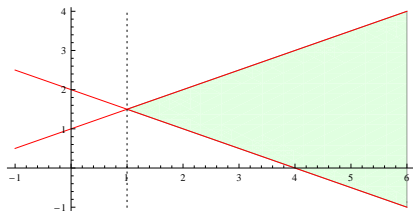


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$\llbracket C_1, C_2, x \mapsto 0.5 \rrbracket$$

$$\text{Explanation } \overline{C_1} \vee \overline{C_2} \vee (x > 1)$$

# Linear Real Arithmetic



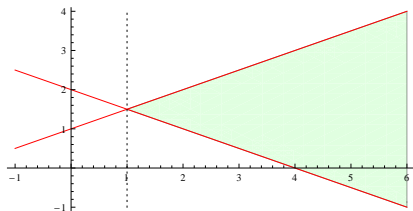
$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$[[C_1, C_2]]$$

Explanation  $\overline{C_1} \vee \overline{C_2} \vee (x > 1)$



# Linear Real Arithmetic

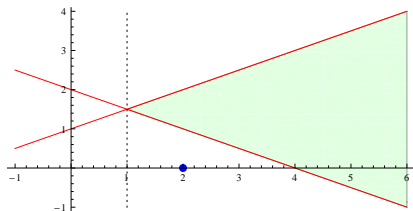


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$\llbracket C_1, C_2, x > 1 \rrbracket$$

Explanation  $\overline{C_1} \vee \overline{C_2} \vee (x > 1)$

# Linear Real Arithmetic

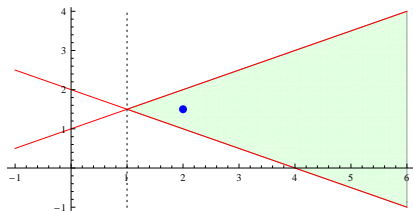


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$\llbracket C_1, C_2, x > 1, x \mapsto 2 \rrbracket$$

Explanation  $\overline{C_1} \vee \overline{C_2} \vee (x > 1)$

# Linear Real Arithmetic

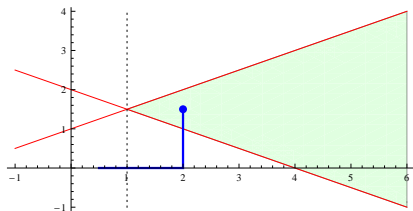


$$\overbrace{2y - x - 2 < 0}^{C_1} \wedge \overbrace{-2y - x + 4 < 0}^{C_2}$$

$$\llbracket C_1, C_2, x > 1, x \mapsto 2, y \mapsto 1.5 \rrbracket$$

$$\text{Explanation } \overline{C_1} \vee \overline{C_2} \vee (x > 1)$$

# Linear Real Arithmetic

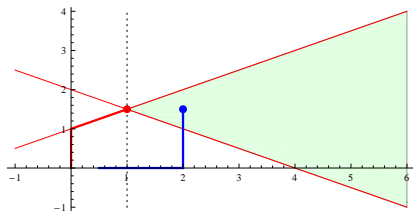


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# Linear Real Arithmetic

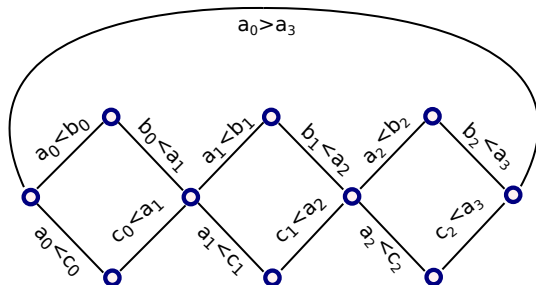


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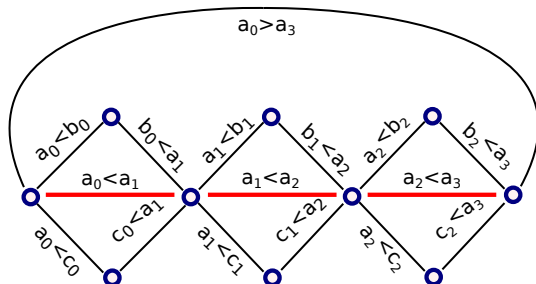
# Linear Real Arithmetic: Comparison to DPLL(T)



## Example (Diamonds)

$$a_0 > a_n \wedge \bigwedge_{k=0}^{n-1} ((a_k < b_k \wedge b_k < a_{k+1}) \vee (a_k < c_k \wedge c_k < a_{k+1}))$$

# Linear Real Arithmetic: Comparison to DPLL(T)



## Example (Diamonds)

$$a_0 > a_n \wedge \bigwedge_{k=0}^{n-1} ((a_k < b_k \wedge b_k < a_{k+1}) \vee (a_k < c_k \wedge c_k < a_{k+1}))$$

# Linear Real Arithmetic: Comparison to DPLL(T)

set	mcsat		cvc4		z3		mathsat5		yices	
	solved	time (s)	solved	time (s)	solved	time (s)	solved	time (s)	solved	time (s)
clocksynchro (36)	<b>36</b>	<b>123.11</b>	36	1166.55	36	1828.74	36	1732.59	36	1093.80
DTPScheduling (91)	<b>91</b>	<b>31.33</b>	91	72.92	91	100.55	89	1980.96	91	926.22
miplib (42)	8	97.16	<b>27</b>	<b>3359.40</b>	23	3307.92	19	5447.46	23	466.44
sal (107)	107	12.68	107	13.46	107	6.37	107	7.99	<b>107</b>	<b>2.45</b>
sc (144)	144	1655.06	144	1389.72	144	954.42	144	880.27	<b>144</b>	<b>401.64</b>
spiderbenchmarks (42)	42	2.38	42	2.47	42	1.66	42	1.22	<b>42</b>	<b>0.44</b>
TM (25)	25	1125.21	25	82.12	<b>25</b>	<b>51.64</b>	25	1142.98	25	55.32
ttastartup (72)	70	4443.72	72	1305.93	72	1647.94	72	2607.49	<b>72</b>	<b>1218.68</b>
uart (73)	73	5244.70	73	1439.89	73	1379.90	73	1481.86	<b>73</b>	<b>679.54</b>
	596	12735.35	<b>617</b>	<b>8832.46</b>	613	9279.14	607	15282.82	613	4844.53



# Linear Real Arithmetic: Comparison to DPLL(T)

## DPLL(T) Simplex (CVC4)

Total Physical Source Lines of Code (SLOC) = 22,597  
Development Effort Estimate, Person-Years (Person-Months) = 5.28 (63.38)  
(Basic COCOMO model, Person-Months =  $2.4 * (KSLOC ** 1.05)$ )  
Schedule Estimate, Years (Months) = 1.01 (12.10)  
(Basic COCOMO model, Months =  $2.5 * (person-months ** 0.38)$ )  
Estimated Average Number of Developers (Effort/Schedule) = 5.24  
Total Estimated Cost to Develop = \$ 713,502  
(average salary = \$56,286/year, overhead = 2.40).

## MCSAT Fourier-Motzkin (CVC4)

Total Physical Source Lines of Code (SLOC) = 1,966  
Development Effort Estimate, Person-Years (Person-Months) = 0.41 (4.88)  
(Basic COCOMO model, Person-Months =  $2.4 * (KSLOC ** 1.05)$ )  
Schedule Estimate, Years (Months) = 0.38 (4.57)  
(Basic COCOMO model, Months =  $2.5 * (person-months ** 0.38)$ )  
Estimated Average Number of Developers (Effort/Schedule) = 1.07  
Total Estimated Cost to Develop = \$ 54,942  
(average salary = \$56,286/year, overhead = 2.40).

Generated using David A. Wheeler's 'SLOCCount'.

# Non-Linear Arithmetic

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

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

## Examples

$$f(x, y) = (x^2 - 1)y^2 + (x + 1)y - 1 \in \mathbb{Z}[x, y]$$

$$g(x) = 16x^3 - 8x^2 + x + 16 \in \mathbb{Z}[x]$$

## Polynomial Constraints

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

# Non-Linear Arithmetic



Tarski [Tar48]

Quantifier elimination

Decidable, non-elementary

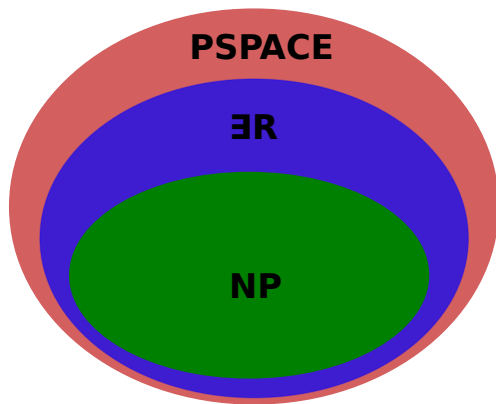


Collins [Col75]

Cylindrical Algebraic Decomposition

Doubly-exponential

# Non-Linear Real Arithmetic



Canny [Can88], Grigor'ev [Gri88]

# Non-Linear Real Arithmetic: CAD

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

## Projection (Saturation)

Project polynomials using a projection P

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

## 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.

# Non-Linear Real Arithmetic

## Model Construction

Build partial model by assigning variables to values

$$\llbracket \dots, C_1, C_2, \dots, x \mapsto \sqrt{2}/2, \dots \rrbracket .$$

## Unit Reasoning

Reason about unit constraints

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

## Explain Conflicts

Explain conflicts using valid clausal reasons

$$(\overline{C_1} \vee \overline{C_2} \vee x \leq 0 \vee x \geq 1) .$$

# Non-Linear Real Arithmetic

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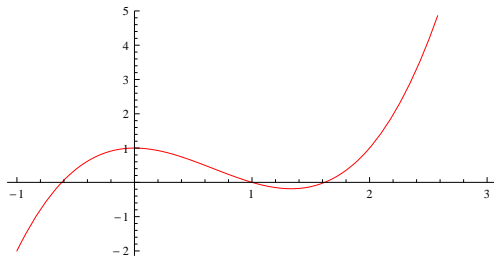
$$C_2 \equiv (xy > 1) \text{ .}$$

## Unit Non-Linear Constraints

$$x^3 - 2x^2 + 1 > 0$$

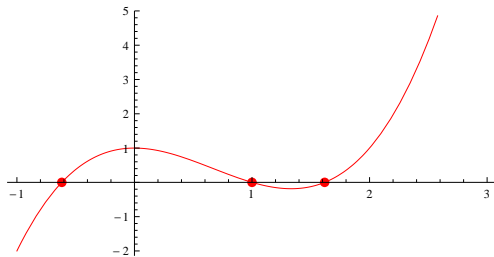


# Unit Non-Linear Constraints



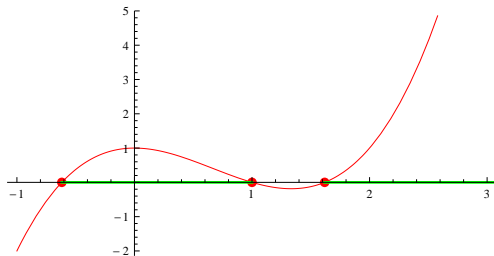
$$x^3 - 2x^2 + 1 > 0$$

# Unit Non-Linear Constraints



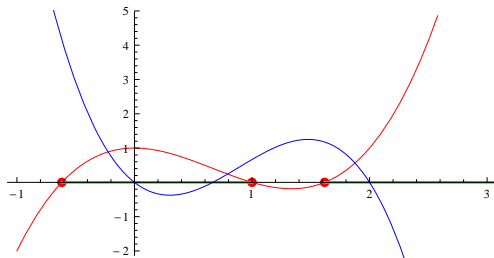
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# Unit Non-Linear Constraints



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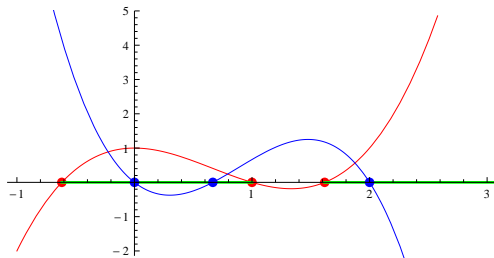
# Unit Non-Linear Constraints



$$x^3 - 2x^2 + 1 > 0$$

$$-3x^3 + 8x^2 - 4x > 0$$

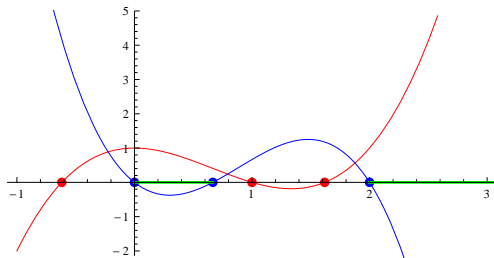
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# Unit Non-Linear Constraints



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# Non-Linear Real Arithmetic

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Build partial model by assigning variables to values

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Reason about unit constraints

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Explain conflicts using valid clausal reasons

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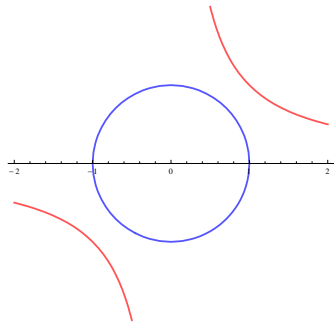
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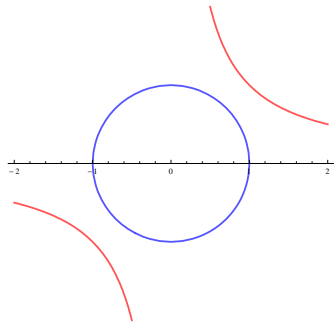
# Non-Linear Real Arithmetic



$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$



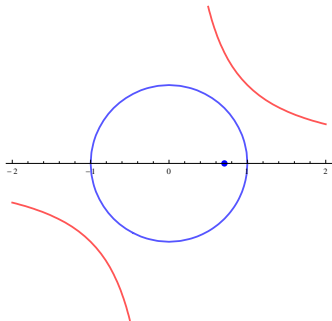
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$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$

$$[[C_1, C_2]]$$

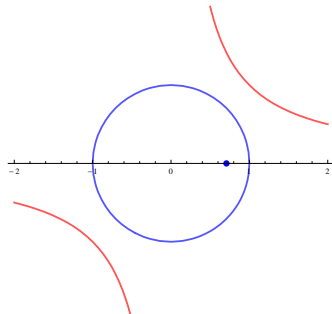
# Non-Linear Real Arithmetic



$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$

$$\llbracket C_1, C_2, x \mapsto \sqrt{2}/2 \rrbracket$$

# Non-Linear Real Arithmetic

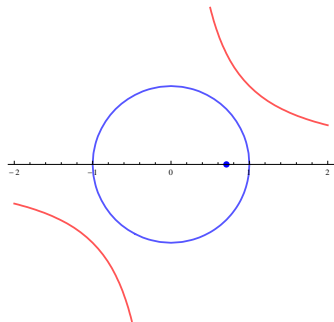


## Unit Constraint Reasoning

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

$$\llbracket C_1, C_2, x \mapsto \sqrt{2/2} \rrbracket$$

# Non-Linear Real Arithmetic

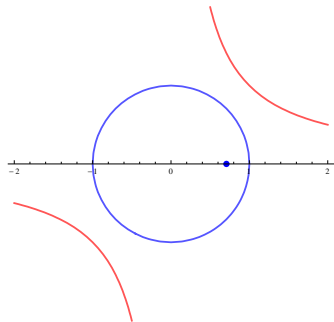


$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$

$$\llbracket C_1, C_2, x \mapsto \sqrt{2}/2 \rrbracket$$

Explanation  $C_1 \wedge C_2 \Rightarrow x \neq \sqrt{2}/2$

# Non-Linear Real Arithmetic

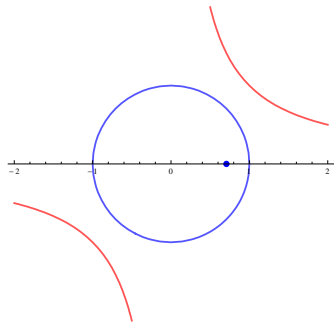


$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$

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Explanation  $C_1 \wedge C_2 \Rightarrow$

# Non-Linear Real Arithmetic

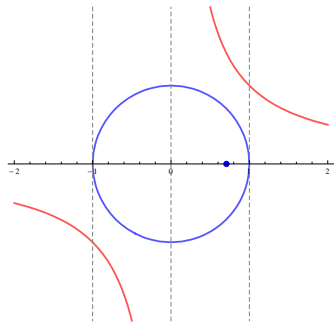


$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$

CAD Projection

$$P = \{x, -4 + 4x^2, 1 - x^2 + x^4\}$$

# Non-Linear Real Arithmetic



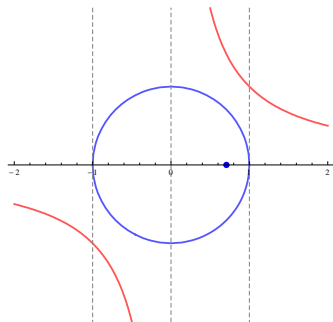
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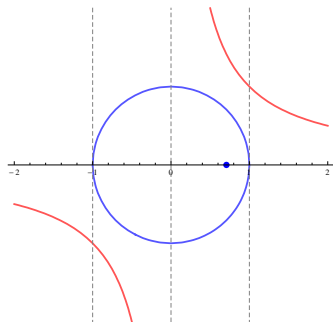


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$$\llbracket C_1, C_2, x \mapsto \sqrt{2}/2 \rrbracket$$

Explanation  $C_1 \wedge C_2 \Rightarrow x \leq 0 \vee x \geq 1$

# Non-Linear Real Arithmetic

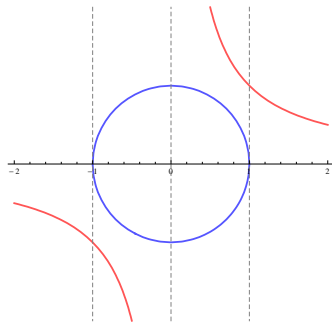


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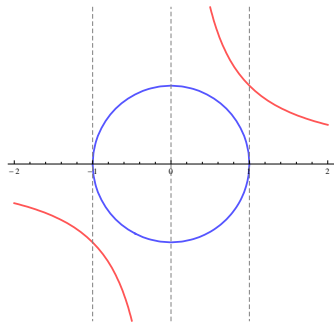


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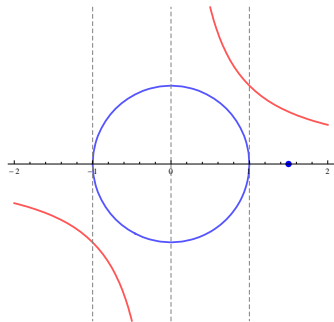


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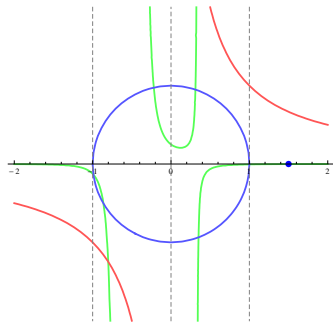


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# Non-Linear Real Arithmetic

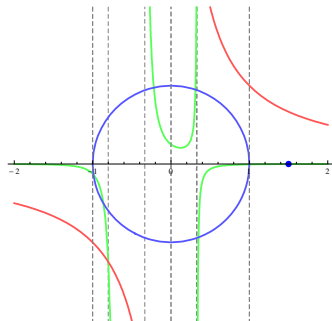


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$$[[C_1, C_2, x \geq 1, x \mapsto 3/2]]$$

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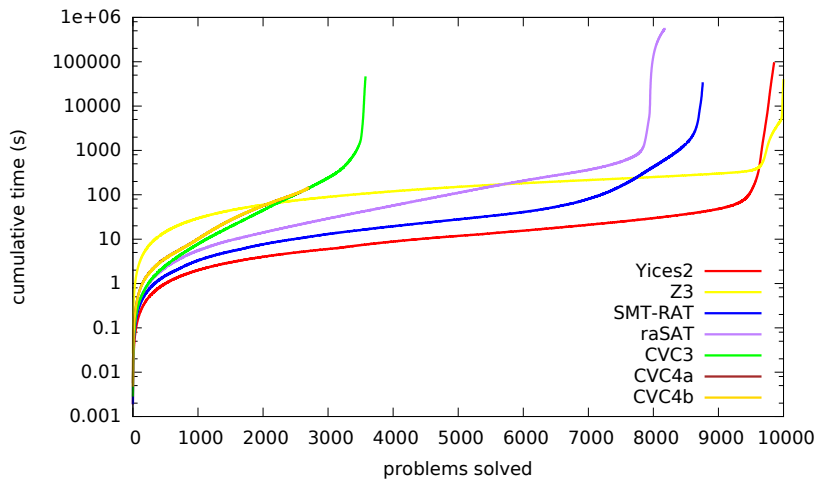


$$\overbrace{x^2 + y^2 < 1}^{C_1} \wedge \overbrace{xy > 1}^{C_2}$$

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# Non-Linear Real Arithmetic: SMT-COMP 2015





# Model-Based Procedures

## Linear Real Arithmetic

- Generalizing DPLL to Richer Logics [MKS09]
- Conflict Resolution [KTV09]
- Natural Domain SMT [Cot10]

## Linear Integer Arithmetic

- Cutting to the Chase: Solving Linear Integer Arithmetic [JDM11]

## Non-Linear Real Arithmetic

- Solving Non-Linear Arithmetic [JDM12, Jov12]

## General Framework

- Model-Constructing Satisfiability Calculus [DMJ13, JBdM13]

# Outline

- 1 Introduction
- 2 Arithmetic
- 3 Theory Combination**
- 4 Conclusion

# Theory Combination

## Combination of Theories

Given individual decision procedures for (quantifier free) first-order theories  $T_1$  and  $T_2$  how can we combine them in a modular fashion into a decision procedure for a theory  $T_1 \oplus T_2$ ?

# Theory Combination

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Given individual decision procedures for (quantifier free) first-order theories  $T_1$  and  $T_2$  how can we combine them in a modular fashion into a decision procedure for a theory  $T_1 \oplus T_2$ ?

- Nelson-Oppen Method [NO79]
- Allows one to decide the combination using decision procedures for  $T_1$  and  $T_2$  as black-boxes
- Most SMT Solvers that involve more than one theory use a combination method based on Nelson-Oppen

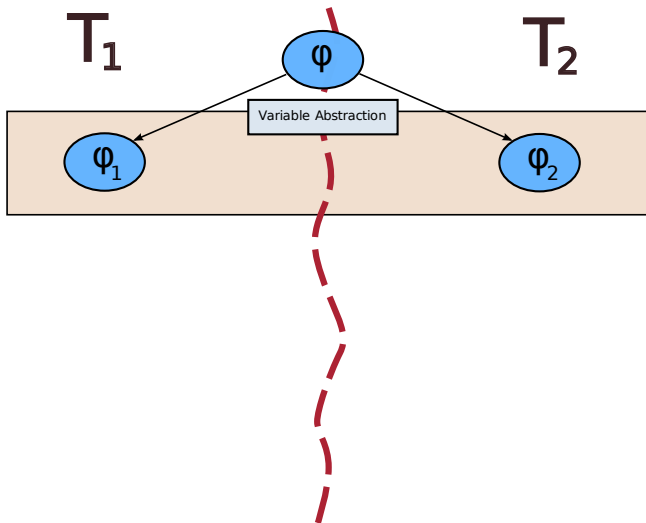
# Theory Combination: Nelson-Oppen

$T_1$

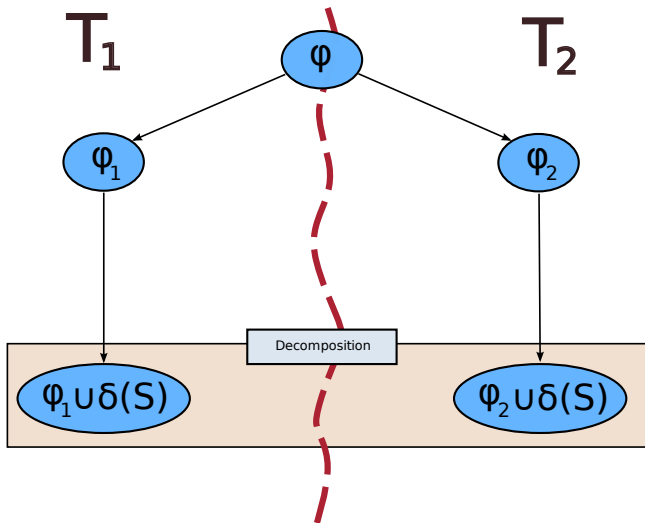


$T_2$

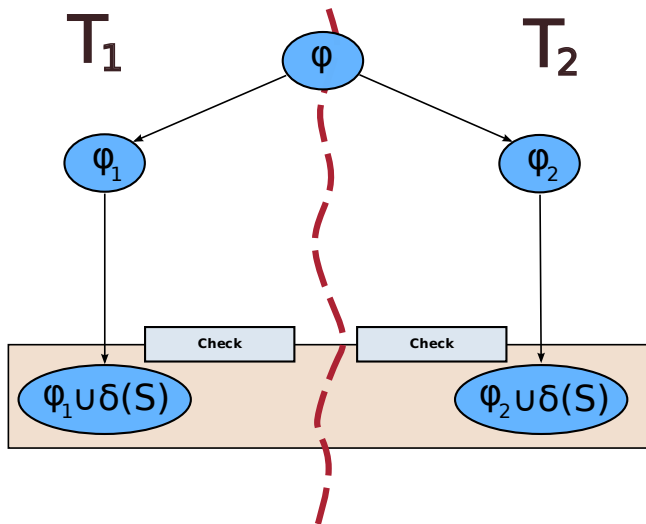
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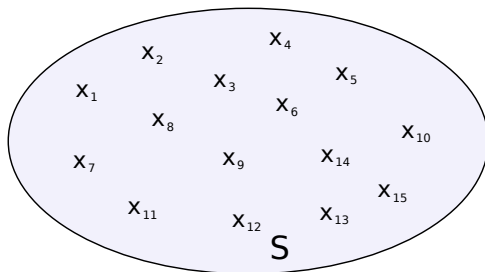


# Theory Combination: Nelson-Oppen Complexity

## Complexity

Search for a suitable arrangement over the shared variables introduces a heavy layer of complexity:

$$O(\mathcal{T}_1(n)) \oplus O(\mathcal{T}_2(n)) \Rightarrow O(2^{n^2} \times (\mathcal{T}_1(n) + \mathcal{T}_2(n))) .$$

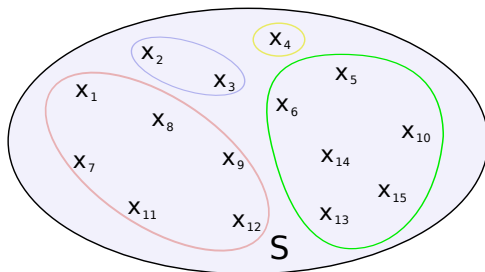


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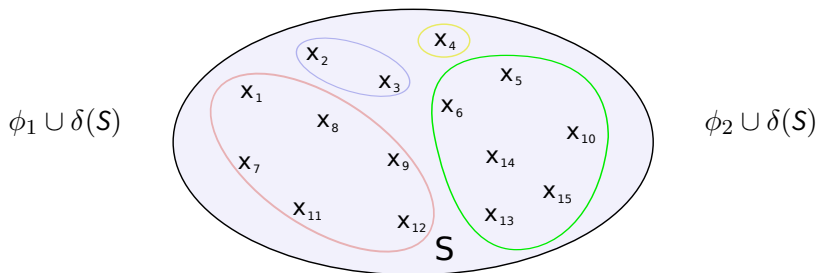


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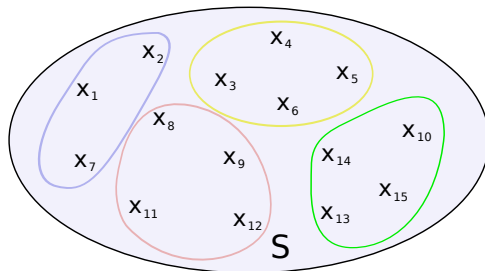


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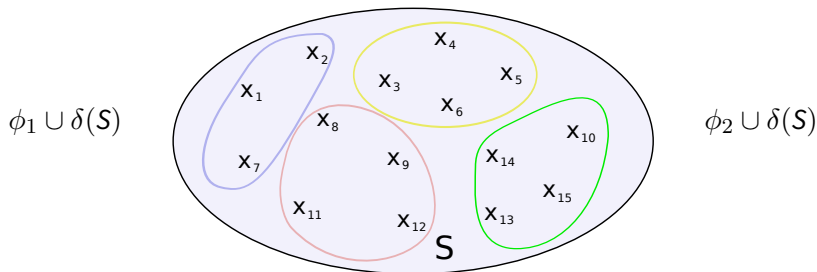


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## Complexity

Search for a suitable arrangement over the shared variables introduces a heavy layer of complexity:

$$O(\mathcal{T}_1(n)) \oplus O(\mathcal{T}_2(n)) \Rightarrow O(2^{n^2} \times (\mathcal{T}_1(n) + \mathcal{T}_2(n))) .$$



# Theory Combination: Nelson-Oppen Complexity

Again, two modules have a restricted interface to each other!



Model-based theory combination [dMB08]

# Uninterpreted Functions

$$x = y$$

$$x \neq y$$

$$x = f(y, z)$$

## DPLL(T): Congruence Closure

- Incremental algorithms for congruence closure.
- Propagation of entailed equalities.
- Combination through Nelson-Oppen style procedures.

# Uninterpreted Functions

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$$x = f(y, z)$$

## DPLL(T): Congruence Closure

- Incremental algorithms for congruence closure.
- Propagation of entailed equalities.
- Combination through Nelson-Oppen style procedures.

## Alternative: Ackermannization [Ack54]

$$\frac{}{x_1 = y_1 \wedge x_2 = y_2 \Rightarrow f(x_1, x_2) = f(y_1, y_2)}$$



## Uninterpreted Functions: Example

$$f(x) < f(y)$$



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$$f(x) < f(y)$$

$$\llbracket f(x) < f(y) \rrbracket$$

## Uninterpreted Functions: Example

$$f(x) < f(y)$$

$$\llbracket f(x) < f(y), \textcolor{red}{f(x)} \mapsto 0 \rrbracket$$

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## Uninterpreted Functions: Example

$$f(x) < f(y)$$

$$\llbracket f(x) < f(y), \textcolor{red}{f(x)} \mapsto 0, \textcolor{red}{f(y)} \mapsto 1, \textcolor{red}{x} \mapsto 0, \textcolor{red}{y} \mapsto 0 \rrbracket$$

Explain Conflict: Ackermanization

$$\overline{x = y \Rightarrow f(x) = f(y)}$$

# Uninterpreted Functions: Example

$$f(x) < f(y)$$

$$\llbracket f(x) < f(y), \textcolor{red}{f(x)} \mapsto 0, \textcolor{red}{f(y)} \mapsto 1, \textcolor{red}{x} \mapsto 0, \textcolor{red}{y} \mapsto 0 \rrbracket$$

Explain Conflict: Ackermanization

$$\frac{}{x \neq y \vee f(x) = f(y)}$$



# Uninterpreted Functions: Example

$$f(x) < f(y)$$

$$\llbracket f(x) < f(y), \textcolor{red}{f(x)} \mapsto 0, \textcolor{red}{f(y)} \mapsto 1 \rrbracket$$

Explain Conflict: Ackermanization

$$\frac{}{x \neq y \vee f(x) = f(y)}$$

# Uninterpreted Functions: Example

$$f(x) < f(y)$$

$$\llbracket f(x) < f(y), \textcolor{red}{f(x)} \mapsto 0, \textcolor{red}{f(y)} \mapsto 1, x \neq y \rrbracket$$

Explain Conflict: Ackermanization

$$\frac{}{x \neq y \vee f(x) = f(y)}$$

# Uninterpreted Functions: Example

$$f(x) < f(y)$$

$$\llbracket f(x) < f(y), \textcolor{red}{f(x)} \mapsto 0, \textcolor{red}{f(y)} \mapsto 1, x \neq y, \textcolor{red}{x} \mapsto 0 \rrbracket$$

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$$f(x) < f(y)$$

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Explain Conflict: Ackermanization

$$\frac{}{x \neq y \vee f(x) = f(y)}$$

# Uninterpreted Functions: Comparison to DPLL(T)

	mcsat		cvc4		z3		mathsat5		yices	
set	solved time (s)		solved time (s)		solved time (s)		solved time (s)		solved	time (s)
EufLaArithmetic (33)	33	39.57	33	49.11	<b>33</b>	<b>2.53</b>	33	20.18	33	4.61
Hash (198)	198	34.81	198	10.60	198	7.18	198	1330.88	<b>198</b>	<b>2.64</b>
RandomCoupled (400)	400	68.04	400	35.90	400	31.44	<b>400</b>	<b>18.56</b>	384	39903.78
RandomDecoupled (500)	500	34.95	500	40.63	500	30.98	<b>500</b>	<b>21.86</b>	500	3863.79
Wisa (223)	223	9.18	223	87.35	223	10.80	223	65.27	<b>223</b>	<b>2.80</b>
wisas (108)	<b>108</b>	<b>40.17</b>	108	5221.37	108	443.36	106	1737.41	108	736.98
	<b>1462</b>	<b>226.72</b>	1462	5444.96	1462	526.29	1460	3194.16	1446	44514.60

# Uninterpreted Functions: Comparison to DPLL(T)

## DPLL(T) CC (CVC4)

Total Physical Source Lines of Code (SLOC) = 5,727  
Development Effort Estimate, Person-Years (Person-Months) = 1.25 (15.00)  
(Basic COCOMO model, Person-Months =  $2.4 * (KSLOC^{**}1.05)$ )  
Schedule Estimate, Years (Months) = 0.58 (7.00)  
(Basic COCOMO model, Months =  $2.5 * (person-months^{**}0.38)$ )  
Estimated Average Number of Developers (Effort/Schedule) = 2.14  
Total Estimated Cost to Develop = \$ 168,836  
(average salary = \$56,286/year, overhead = 2.40).

## MCSAT Ackermanization (CVC4)

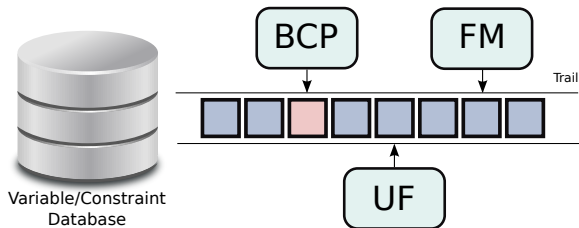
Total Physical Source Lines of Code (SLOC) = 247  
Development Effort Estimate, Person-Years (Person-Months) = 0.05 (0.55)  
(Basic COCOMO model, Person-Months =  $2.4 * (KSLOC^{**}1.05)$ )  
Schedule Estimate, Years (Months) = 0.17 (2.00)  
(Basic COCOMO model, Months =  $2.5 * (person-months^{**}0.38)$ )  
Estimated Average Number of Developers (Effort/Schedule) = 0.28  
Total Estimated Cost to Develop = \$ 6,223  
(average salary = \$56,286/year, overhead = 2.40).

Generated using David A. Wheeler's 'SLOCCount'.

# Outline

- 1 Introduction
- 2 Arithmetic
- 3 Theory Combination
- 4 Conclusion**

# MCSAT: Simple Architecture



Each plugins reasons about their domain:

- 1 Track when a constraint becomes unit or fully assigned [MMZ<sup>+</sup>01].
- 2 Unit constraints imply feasible sets of individual variables.
- 3 Propagate any constraints/variables whose value is implied.
- 4 Explain any unit conflicts with clausal explanations.
- 5 When asked, decide unassigned variable to feasible value.



# MCSAT: Implementations

- QF\_NRA in Yices2 [▶ Link](#)
- QF\_NRA in Z3 [▶ Link](#)
- QF\_UFLRA in CVC4 [▶ Link](#)
- LibPoly: Library for polynomial manipulation [▶ Link](#)

# MCSAT: Appeal

## Compared to DPPL(T)

- Only need to reason about constraints in one variable.
- Easy to add new decision procedures.
- Can generate new facts in a conflict directed manner.
- Simple combination mechanism: build the model.
- Performs well on practical problems.

# MCSAT: TODO & Research Problems

- Linear arithmetic
  - Simplex vs Fourier-Motzkin?
  - Add integer reasoning.
- Non-linear arithmetic:
  - More natural conflict explanations?
  - Integrate interval reasoning into MCSAT [BDG<sup>+</sup>14].
  - Add integer reasoning.
- Bit-vectors, floating point, strings...
  - How to reason about unit constraints?
  - How to represent feasible sets?
  - How to explain conflicts?
- Arrays
  - Adapt lemmas on demand to MCSAT [BB08]?

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