

# Solving Pseudo-Boolean Constraints with SMT

A few first steps

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March 2nd, 2018



# What was already done?

- Bachelor thesis in 2017
- Pseudo-Boolean problems:
  - Linear
  - No objectives
  - Only conjunctions

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  - No objectives
  - Only conjunctions
- Different strategies in SMT-RAT
- Comparison with MiniSat+



## Pseudo-Boolean constraint

Boolean variables  $x_i$ , integer coefficients  $a_i$ :

$$\sum_{i=1}^n a_i \cdot x_i \sim a_0$$

We assume  $true = 1$  and  $false = 0$ .

# Problem definition

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## Satisfiability

Given a set of pseudo-Boolean constraints  $C$  over variables  $x_i$ :  
Find Boolean values for all  $x_i$  such that all  $c \in C$  evaluate to *true*.

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Properties:

- Encoding grows with the coefficients
- Efficient for small constraints
- Large arithmetic constraints can be a problem



- Encode **easy** constraints in propositional logic

$$\begin{array}{lll} ax_1 \geq b, a > b > 0 & \Rightarrow & x_1 \\ x_1 - x_2 \geq 0 & \Rightarrow & x_2 \rightarrow x_1 \\ \sum a_i x_i \geq b, \sum a_i = b & \Rightarrow & \bigwedge x_i \\ \sum x_i \sim b & & \text{cardinality constraints} \end{array}$$

# Our SMT-based approach

- Encode **easy** constraints in propositional logic

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- Consider **remaining** constraints to be **linear integer** constraints
- Use (any) SMT solver for **linear integer arithmetic**
- **Simplex** and **Branch&Bound**

# Experimental results

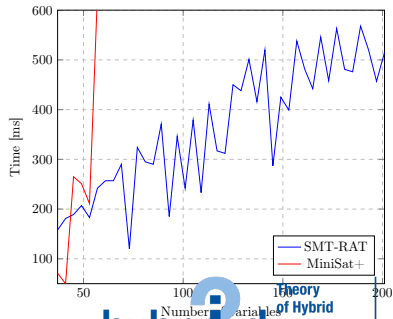
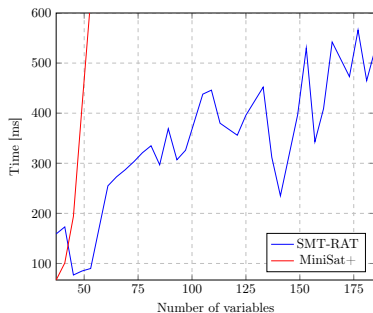
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- Ignore objective functions
- MiniSat+:  $\approx 60\%$  solved
- SMT-RAT:  $\approx 20\%$  solved
- But heavily depends on the structure of the benchmark



# Preprocessing: Gauss

Use Gaussian elimination to **simplify equations**. Use equations to **simplify inequations**.

$$2x_1 + 1x_2 + 1x_3 = 5$$

$$1x_1 - 2x_2 + 1x_3 = 4$$

$$1x_1 + 1x_2 = 2$$

$$-5x_1 + 1x_3 \geq 2$$

$$4x_1 + 1x_2 + 4x_4 \geq 1$$

$\Rightarrow$

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$$-2x_3 = -1$$

$$-1x_2 \geq -1$$

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**Less constraints**, **eliminate variables** from individual constraints.

**Detrimental**. Our guess: input constraints are **sparse** but **become dense**.



# Preprocessing: Residual Number Systems

Use an adaption of the Chinese Remainder Theorem to convert **one large** constraint to **several easy** constraints.

$$748x_1 + 936x_2 + 58x_3 + 493x_4 + 145x_5 + 85 + x_6 = 105$$

Choose primes in a clever way: 5, 17, 29

$$3x_1 + 3x_3 + 3x_4 = 0 \quad \text{mod } 5$$

$$7x_3 + 9x_5 = 3 \quad \text{mod } 17$$

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**No effect.** Our guess: Simplex performance depends on **total number of terms**, smaller coefficients do not matter.

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# Future work

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## Boolean combinations

Arbitrary **Boolean combinations** instead of pure conjunctions