

Satisfiability Modulo Linear Arithmetic

Combinatorial Problem Solving (CPS)

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Linear Arithmetic Theories

- In linear arithmetic theories, atoms are of the form:

$$a_1x_1 + \dots + a_nx_n \bowtie b$$

where \bowtie is one of: $=, \neq, <, >, \leq, \geq$

All symbols are interpreted with their usual meaning in arithmetic

- Example of atom:

$$x + y + 2z \geq 10$$

- Example of formula:

$$x \geq 0 \wedge (x + y \leq 2 \vee x - y \geq 6) \wedge (x + y \geq 1 \vee x - y \geq 4)$$

- Variables can be of real sort (\mathbb{R}) or integer sort (\mathbb{Z})
- If all vars are \mathbb{R} we have a problem of **Linear Real Arithmetic (LRA)**
- If all vars are \mathbb{Z} we have a problem of **Linear Integer Arithmetic (LIA)**

Overview of the Lecture

- De Moura & Dutertre's Algorithm for LRA
- LIA

De Moura & Dutertre's Algorithm

- Problem: given an input formula ϕ of LRA, is ϕ SAT?
- Assume for the time being ϕ only contains linear constraints of the form $c^T x \leq d$

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- **Preprocessing:** transform ϕ into $\hat{\phi} \wedge Ax = 0$, where:
 1. $\hat{\phi}$ is obtained from ϕ by replacing each $c^T x \leq d$ by $s \leq d$, where s is fresh variable
 2. $Ax = 0$ consists of all definitions $s = c^T x$

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- Example:

$$x \geq 0 \wedge (x + y \leq 2 \vee x - y \geq 6) \wedge (x + y \geq 1 \vee x - y \geq 4)$$

$$x \geq 0 \wedge (s_1 \leq 2 \vee s_2 \geq 6) \wedge (s_1 \geq 1 \vee s_2 \geq 4) \wedge \\ (s_1 = x + y \wedge s_2 = x - y)$$

De Moura & Dutertre's Algorithm

- Consistency checking is based on **dual bounded simplex**
- Theory solver handles feasibility problems of the form

$$Ax = 0 \wedge \ell \leq x \leq u$$

- **Only bounds asserted during search:**

$Ax = 0$ is asserted before any decision

There is no addition/deletion of rows!

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- **Free variables** (those without any bound in the formula) **can be eliminated** before starting search by means of Gaussian elimination
- E.g.: if y is free then equation $y = x - s_2$ is not asserted

$$x \geq 0 \wedge (s_1 \leq 2 \vee s_2 \geq 6) \wedge (s_1 \geq 1 \vee s_2 \geq 4) \wedge \\ (s_1 = 2x - s_2 \wedge y = x - s_2)$$

Basic Solver

- For solving $Ax = 0 \wedge \ell \leq x \leq u$, theory solver stores:
 - ◆ A tableau: $x_i = \sum_{x_j \in \mathcal{R}} \alpha_{ij} x_j$, $x_i \in \mathcal{B}$
 - ◆ For each variable x_i ,
the strongest asserted lower bound ℓ_i
the strongest asserted upper bound u_i
 - ◆ An assignment β such that
 - $A\beta = 0$
 - For each $x_j \in \mathcal{R}$: $\ell_j \leq \beta(x_j) \leq u_j$

Basic Solver

■ For solving $Ax = 0 \wedge \ell \leq x \leq u$, theory solver stores:

◆ A tableau: $x_i = \sum_{x_j \in \mathcal{R}} \alpha_{ij} x_j$, $x_i \in \mathcal{B}$

◆ For each variable x_i ,
the strongest asserted lower bound ℓ_i
the strongest asserted upper bound u_i

◆ An assignment β such that

■ $A\beta = 0$

■ For each $x_j \in \mathcal{R}$: $\ell_j \leq \beta(x_j) \leq u_j$

■ Maybe for some $x_i \in \mathcal{B}$, $\ell_i > \beta(x_i)$ or $\beta(x_i) > u_j$

■ Maybe for some $x_i \in \mathcal{R}$, $\ell_i < \beta(x_i) < u_j$

■ Supports two types of consistency checks:

light-weight and heavy-weight

Light-Weight Consistency Check

- Ensures non basic vars satisfy bounds and $A\beta = 0$
 - ◆ If returns **SAT** : Then model is consistent
 - ◆ If returns **UNSAT**: Then model is inconsistent
 - ◆ If returns **UNKNOWN**: Don't know

assert_lower($x_j \geq c_j$)

if $c_j \leq \ell_j$ then return **SAT**

if $c_j > u_j$ then return **UNSAT**

$\ell_j := c_j$;

if $x_j \in \mathcal{R} \wedge \beta(x_j) < \ell_j$ then update(x_j, ℓ_j)

return **UNKNOWN**

update(x_j, v)

for each $x_i \in \mathcal{B}$, $\beta(x_i) := \beta(x_i) + \alpha_{ij}(v - \beta(x_j))$

$\beta(x_j) := v$

Light-Weight Consistency Check

■ Ensures non basic vars satisfy bounds and $A\beta = 0$

- ◆ If returns **SAT** : Then model is consistent
- ◆ If returns **UNSAT**: Then model is inconsistent
- ◆ If returns **UNKNOWN**: Don't know

assert_upper($x_j \leq c_j$)

if $c_j \geq u_j$ then return **SAT**

if $c_j < \ell_j$ then return **UNSAT**

$u_j := c_j$;

if $x_j \in \mathcal{R} \wedge \beta(x_j) > u_j$ then update(x_j, u_j)

return **UNKNOWN**

update(x_j, v)

for each $x_i \in \mathcal{B}$, $\beta(x_i) := \beta(x_i) + \alpha_{ij}(v - \beta(x_j))$

$\beta(x_j) := v$

Heavy-Weight Consistency Check

- Light-weight consistency check is performed first (since it is cheaper)
- The only possible cases of unfeasibility that are left: bounds of basic vars
- **Dual Bounded Simplex** (with null objective function) is employed to get feasible basis
- Constraints are handled **in blocks** (as opposed to one at a time)

Heavy-Weight Consistency Check

check()

 loop

 select basic variable x_i such that $\beta_i < \ell_i$ or $\beta_i > u_i$

 if there is no such x_i then return SAT

 if $\beta_i < \ell_i$ then

 select non-basic variable x_j such that

$(\alpha_{ij} > 0 \wedge \beta(x_j) < u_j) \vee (\alpha_{ij} < 0 \wedge \beta(x_j) > \ell_j)$

 if there is no such x_j then return UNSAT

 pivot_and_update(x_i, x_j, ℓ_i)

 if $\beta_i > u_i$ then

 select non-basic variable x_j such that

$(\alpha_{ij} < 0 \wedge \beta(x_j) < u_j) \vee (\alpha_{ij} > 0 \wedge \beta(x_j) > \ell_j)$

 if there is no such x_j then return UNSAT

 pivot_and_update(x_i, x_j, u_i)

/* pivot_and_update(x_i, x_j, v):

 set basic x_i to v , adjust non-basic x_j and other basic vars as needed,

 swap x_i and x_j in the basis */

Heavy-Weight Consistency Check

pivot_and_update(x_i, x_j, v)

/* set basic x_i to v , adjust non-basic x_j and other basic vars as needed,
swap x_i and x_j in the basis */

$$\Theta := \frac{v - \beta(x_i)}{\alpha_{ij}}$$

$$\beta(x_i) := v$$

$$\beta(x_j) := \beta(x_j) + \Theta$$

$$\text{for each } x_k \in \mathcal{B} \wedge x_k \neq x_i, \beta(x_k) := \beta(x_k) + \alpha_{kj}\Theta$$

pivot(x_i, x_j)

- Recall **Bland's anticycling rule** in dual pricing and dual ratio test:
 - ◆ Set an order between variables
 - ◆ Always take the least possible variable
- **THEOREM**. This strategy guarantees termination

Conflict Explanations

- $\text{check}()$ detects an inconsistency when:

- ◆ If $\beta_i < \ell_i$ and for all non-basic x_j
 $(\alpha_{ij} > 0 \rightarrow \beta(x_j) \geq u_j) \wedge (\alpha_{ij} < 0 \rightarrow \beta(x_j) \leq \ell_j)$
- ◆ If $\beta_i > u_i$ and for all non-basic x_j
 $(\alpha_{ij} < 0 \rightarrow \beta(x_j) \geq u_j) \wedge (\alpha_{ij} > 0 \rightarrow \beta(x_j) \leq \ell_j)$

- Let $\mathcal{R}^+ = \{x_j \in \mathcal{R} \mid \alpha_{ij} > 0\}$ and $\mathcal{R}^- = \{x_j \in \mathcal{R} \mid \alpha_{ij} < 0\}$

- Since β satisfies all bounds on non-basic vars:

- ◆ If $\beta(x_i) < \ell_i$
 - for all $x_j \in \mathcal{R}^+, \beta(x_j) = u_j$
 - for all $x_j \in \mathcal{R}^-, \beta(x_j) = \ell_j$
- ◆ If $\beta(x_i) > u_i$
 - for all $x_j \in \mathcal{R}^+, \beta(x_j) = \ell_j$
 - for all $x_j \in \mathcal{R}^-, \beta(x_j) = u_j$

Conflict Explanations

- Assume $\beta(x_i) < \ell_i$.
- So for all $x_j \in \mathcal{R}^+$, $\beta(x_j) = u_j$ and for all $x_j \in \mathcal{R}^-$, $\beta(x_j) = \ell_j$
- Hence $\beta(x_i) = \sum_{x_j \in \mathcal{R}^+} \alpha_{ij} u_j + \sum_{x_j \in \mathcal{R}^-} \alpha_{ij} \ell_j$
- So for any x such that $Ax = b$

$$\beta(x_i) - x_i = \sum_{x_j \in \mathcal{R}^+} \alpha_{ij} (u_j - x_j) + \sum_{x_j \in \mathcal{R}^-} \alpha_{ij} (\ell_j - x_j)$$

- From this we can derive the implication

$$\bigwedge_{x_j \in \mathcal{R}^+} x_j \leq u_j \wedge \bigwedge_{x_j \in \mathcal{R}^-} x_j \geq \ell_j \Rightarrow x_i \leq \beta(x_i)$$

- Since $\beta(x_i) < \ell_i$ this is inconsistent with $x_i \geq \ell_i$
- The explanation of the conflict is

$$\{x_j \leq u_j \mid x_j \in \mathcal{R}^+\} \cup \{x_j \geq \ell_j \mid x_j \in \mathcal{R}^-\} \cup \{x_i \geq \ell_i\}$$

which is minimal (with respect to set inclusion)

Conflict Explanations

- Assume $\beta(x_i) > u_i$.
- So for all $x_j \in \mathcal{R}^+$, $\beta(x_j) = \ell_j$ and for all $x_j \in \mathcal{R}^-$, $\beta(x_j) = u_j$
- Hence $\beta(x_i) = \sum_{x_j \in \mathcal{R}^+} \alpha_{ij} \ell_j + \sum_{x_j \in \mathcal{R}^-} \alpha_{ij} u_j$
- So for any x such that $Ax = b$

$$\beta(x_i) - x_i = \sum_{x_j \in \mathcal{R}^+} \alpha_{ij} (\ell_j - x_j) + \sum_{x_j \in \mathcal{R}^-} \alpha_{ij} (u_j - x_j)$$

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$$\{x_j \geq \ell_j \mid x_j \in \mathcal{R}^+\} \cup \{x_j \leq u_j \mid x_j \in \mathcal{R}^-\} \cup \{x_i \leq u_i\}$$

which is minimal (with respect to set inclusion)

Backtracking

- Number of backtrackings is often very large:
needs to be **efficiently implemented**
- The algorithm only requires, for each variable x_i ,
 - ◆ one stack for lower bounds ℓ_i
 - ◆ one stack for upper bounds u_i
- **No need to save successive β on a stack!**
Only one assignment β is kept
- Recall: for each $x_j \in \mathcal{R}$, then $\ell_j \leq \beta(x_j) \leq u_j$
Maybe for some $x_i \in \mathcal{R}$, $\ell_i < \beta(x_i) < u_j$
- Does **not require pivoting**: very cheap

Theory Propagation

- **Simple propagation:** $x \geq c$ implies $x \geq c'$ for all $c' \leq c$

- **Bound refinement:**

given an equation $x_i = \sum \alpha_j x_j$ that holds for any x solution to $Ax = 0$, then we can deduce bounds:

$$x_i \geq \sum_{\alpha_j > 0} \alpha_j \ell_j + \sum_{\alpha_j < 0} \alpha_j u_j$$

$$x_i \leq \sum_{\alpha_j > 0} \alpha_j u_j + \sum_{\alpha_j < 0} \alpha_j \ell_j$$

- **Might not be better bounds** than those already asserted
- Used with tableau rows
(but can be used with rows of original problem
or any linear combination of them)

Example

Assert literals $x \leq -4$, $x \geq -8$, $-x + y \leq 1$, $x + y \geq -3$

■ TABLEAU

$$\begin{cases} s_1 &= -x + y \\ s_2 &= x + y \end{cases}$$

■ ASSIGNMENT

$$x \rightarrow 0$$

$$y \rightarrow 0$$

$$s_1 \rightarrow 0$$

$$s_2 \rightarrow 0$$

■ BOUNDS

Example

Assert literals $x \leq -4$, $x \geq -8$, $-x + y \leq 1$, $x + y \geq -3$

■ TABLEAU

$$\begin{cases} s_1 &= -x + y \\ s_2 &= x + y \end{cases}$$

■ ASSIGNMENT

$$x \rightarrow -4$$

$$y \rightarrow 0$$

$$s_1 \rightarrow 4$$

$$s_2 \rightarrow -4$$

■ BOUNDS

$$x \leq -4$$

Example

Assert literals $x \leq -4$, $x \geq -8$, $-x + y \leq 1$, $x + y \geq -3$

■ TABLEAU

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■ ASSIGNMENT

$$x \rightarrow -4$$

$$y \rightarrow 0$$

$$s_1 \rightarrow 4$$

$$s_2 \rightarrow -4$$

■ BOUNDS

$$-8 \leq x \leq -4$$

Example

Assert literals $x \leq -4$, $x \geq -8$, $-x + y \leq 1$, $x + y \geq -3$

■ TABLEAU

$$\begin{cases} y = x + s_1 \\ s_2 = 2x + s_1 \end{cases}$$

■ ASSIGNMENT

$$x \rightarrow -4$$

$$y \rightarrow -3$$

$$s_1 \rightarrow 1$$

$$s_2 \rightarrow -7$$

■ BOUNDS

$$\begin{array}{ccccc} -8 & \leq & x & \leq & -4 \\ & & s_1 & \leq & 1 \end{array}$$

Example

Assert literals $x \leq -4$, $x \geq -8$, $-x + y \leq 1$, $x + y \geq -3$

■ TABLEAU

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■ ASSIGNMENT

$$x \rightarrow -4$$

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■ BOUNDS

$$-8 \leq x \leq -4$$

$$s_1 \leq 1$$

$$-3 \leq s_2$$

Example

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$$x \rightarrow -4$$

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$$s_1 \rightarrow 1$$

$$s_2 \rightarrow -7$$

■ BOUNDS

$$-8 \leq x \leq -4$$

$$s_1 \leq 1$$

$$-3 \leq s_2$$

Conflict between $x \leq -4$, $s_1 \leq 1$, $-3 \leq s_2$!

Strict Inequalities and Disequalities

- **LEMMA.** A set of linear arithmetic literals Γ containing strict inequalities $S = \{p_1 > 0, \dots, p_n > 0\}$ is satisfiable iff there exists a rational number $\delta > 0$ s.t. for all δ' s.t. $0 < \delta' \leq \delta$, $\Gamma_\delta = (\Gamma \cup S_\delta) - S$ is satisfiable, where $S_\delta = \{p_1 \geq \delta, \dots, p_n \geq \delta\}$
- Strict inequalities are transformed into non-strict ones using an **infinitesimal** positive symbolic value δ :

$$\begin{aligned}x > a &\longrightarrow x \geq a + \delta \\x < a &\longrightarrow x \leq a - \delta\end{aligned}$$

- Disequalities $c^T x \neq d$ have to be split into $c^T x < d \vee c^T x > d$ while parsing
- Equalities $c^T x = d$ have to be split into $c^T x \leq d \wedge c^T x \geq d$ while parsing

Strict Inequalities and Disequalities

- δ is not given a concrete value

Just treated symbolically!

- Values are pairs of rationals with ordering:

- ◆ $a + b\delta \leq a' + b'\delta$ iff $a < a'$, or $a = a'$ and $b \leq b'$

- Arithmetic operations are defined pairwise:

- ◆ $(a + b\delta) + (a' + b'\delta) = (a + a') + (b + b')\delta$

- ◆ $c \cdot (a + b\delta) = (c \cdot a) + (c \cdot b)\delta$

- From now on let $\mathbb{Q}_\delta = \{a + b\delta \mid a, b \in \mathbb{Q}\}$

Strict Inequalities and Disequalities

LEMMA. Let $v_i = c_i + k_i \delta$, $w_i = d_i + h_i \delta$ ($i = 1 \dots m$) be such that $v_i \leq w_i$ hold. Then there is $\delta_0 \in \mathbb{Q}$ such that $\delta_0 > 0$ and

$$\begin{array}{ccc} c_1 + k_1 \epsilon & \leq & d_1 + h_1 \epsilon \\ & \vdots & \\ c_m + k_m \epsilon & \leq & d_m + h_m \epsilon \end{array}$$

are satisfied for any ϵ such that $0 < \epsilon \leq \delta_0$.

Strict Inequalities and Disequalities

PROOF: By definition

$$c_i + k_i \delta \leq d_i + h_i \delta \text{ iff } c_i < d_i, \text{ or } c_i = d_i \text{ and } k_i \leq h_i$$

We distinguish several cases:

- If $c_i = d_i$ and $k_i \leq h_i$ then $c_i + k_i \epsilon \leq d_i + h_i \epsilon$ for any $\epsilon > 0$
- If $c_i < d_i$ and $k_i \leq h_i$ then $c_i + k_i \epsilon \leq d_i + h_i \epsilon$ for any $\epsilon > 0$
- If $c_i < d_i$ and $k_i > h_i$ then $c_i + k_i \epsilon \leq d_i + h_i \epsilon$ for any ϵ such that $0 < \epsilon \leq \frac{d_i - c_i}{k_i - h_i}$

So for example take δ_0 such that

$$0 < \delta_0 < \min\left\{\frac{d_i - c_i}{k_i - h_i} \mid c_i < d_i \text{ and } k_i > h_i\right\}$$

Strict Inequalities and Disequalities

- Let S be a linear problem of the form

$$Ax = 0 \wedge \ell \bowtie^- x \bowtie^+ u$$

where $\ell_i, u_i \in \mathbb{Q}$ and \bowtie_i^-, \bowtie_i^+ are either $<$ or \leq

- S can be converted into a problem S' of the form

$$Ax = 0 \wedge \ell' \leq x \leq u'$$

where $\ell'_i, u'_i \in \mathbb{Q}_\delta$ as follows:

- ◆ $x_i > \ell_i \rightarrow x_i \geq \ell'_i$ with $\ell'_i = \ell_i + \delta$
- ◆ $x_i < u_i \rightarrow x_i \leq u'_i$ with $u'_i = u_i - \delta$

Strict Inequalities and Disequalities

■ **THEOREM.** S and S' are equisatisfiable.

PROOF: Let us see S' sat in \mathbb{Q}_δ implies S sat in \mathbb{Q} .

Let β' be a satisfying assignment for S' .

The inequalities $\ell'_j \leq \beta'(x_j) \leq u'_j$ are satisfied in \mathbb{Q}_δ .

Let $\beta'(x_j) = p_j + q_j \delta$, $\ell'_j = \ell_j + k_j \delta$, $u'_j = u_j + h_j \delta$ where

$k_j \in \{0, -1\}$, $k_j = 0$ iff \bowtie_i^- is \leq ,

$h_j \in \{0, -1\}$, $h_j = 0$ iff \bowtie_i^+ is \leq .

By the previous lemma, there is $\delta_0 \in \mathbb{R}$, $\delta_0 > 0$ such that

$$\ell_j + k_j \delta_0 \leq p_j + q_j \delta_0 \leq u_j + h_j \delta_0$$

Let us define $\beta(x_j) = p_j + q_j \delta_0$ for all x_j .

Then β satisfies both $\ell \bowtie x \bowtie u$ as well as $Ax = 0$

Strict Inequalities and Disequalities

PROOF (continued):

Let us see S sat in Q implies S' sat in Q_δ .

Trivial: any satisfying assignment β for S in Q is a satisfying assignment for S' in Q_δ

Overview of the Lecture

- De Moura & Dutertre's Algorithm for LRA
- LIA

SMT(LIA)

- State-of-the-art SMT solvers for LIA use:

- ◆ Branch & Bound
- ◆ Cutting Planes
- ◆ GCD Test

- Strict inequalities are transformed into non-strict ones:

$$\begin{aligned}x > a &\longrightarrow x \geq a + 1 \\x < a &\longrightarrow x \leq a - 1\end{aligned}$$

- So in what follows, all constraints will be non-strict

Branch & Bound (Feasibility)

```
 $S := \{P_0\}$  /* set of pending problems */  
while  $S \neq \emptyset$  do  
    remove  $P$  from  $S$ ; solve  $LP(P)$   
    if  $LP(P)$  is feasible then  
        Let  $\beta$  be basic solution obtained after solving  $LP(P)$   
        if  $\beta$  satisfies integrality constraints then  
            return SATISFIABLE  
        else  
            Let  $x_j$  be integer variable such that  $\beta_j \notin \mathbb{Z}$   
             $S := S \cup \{P \wedge x_j \leq \lfloor \beta_j \rfloor, \quad P \wedge x_j \geq \lceil \beta_j \rceil\}$   
return UNSATISFIABLE
```

Splitting on Demand

■ Two ways to implement Branch & Bound in SMT:

1. Branch & Bound is internal to the theory solver

✓ Modular and flexible

✗ Lots of code are repeated in SAT/theory solvers:
splitting heuristics, stack, etc.

2. Delegate splits to SAT solver: **splitting on demand**

◆ Whenever theory solver needs to split on x_j ,
it invents new lit l and asks SAT solver to split on it

◆ Internal meaning of the literal for theory solver:

■ $l \equiv x_j \leq \lfloor \beta_j \rfloor$

■ $\neg l \equiv x_j \geq \lceil \beta_j \rceil$

◆ Implementation of theory solver can be simplified

GCD Test

- Quick test which, if positive, ensures problem is UNSAT
- Let us consider an equation $\sum_{i=1}^n a_i x_i = b$ where $a_i, b \in \mathbb{Z}$
- Notation: $c \mid d$ means “ c divides d ”.
 $\text{GCD}(x, y)$ is the greatest common divisor of x and y .
- Let $g = \text{GCD}(a_1, \dots, a_n)$. If $g \nmid b$ then equation is UNSAT

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PROOF: If $x_i \in \mathbb{Z}$ satisfy the equation then $g \mid a_i$ implies $g \mid a_i x_i$, and hence $g \mid \sum_{i=1}^n a_i x_i = b$

GCD Test

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- In theory solver GCD test can be applied to tableau rows

Bibliography - Further Reading

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