

# Satisfiability Checking

## 13 Gauß and Fourier-Motzkin variable elimination for linear real arithmetic

↳ only "+"

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# The Xmas problem

There are three types of Xmas presents Santa Claus can make.

- Santa Claus wants to reduce the overhead by making only two types.
- He needs at least 100 presents.
- He needs at least 5 of either type 1 or type 2.
- He needs at least 10 of the third type.
- Each present of type 1, 2, and 3 need 1, 2, resp. 5 minutes to make.
- Santa Claus is late, and he has only 3 hours left.
- Each present of type 1, 2, and 3 costs 3, 2, resp. 1 EUR.
- He has 300 EUR for presents in total.

$$\begin{aligned} & (p_1 = 0 \vee p_2 = 0 \vee p_3 = 0) \wedge p_1 + p_2 + p_3 \geq 100 \wedge \\ & (p_1 \geq 5 \vee p_2 \geq 5) \wedge p_3 \geq 10 \wedge p_1 + 2p_2 + 5p_3 \leq 180 \wedge \\ & 3p_1 + 2p_2 + p_3 \leq 300 \end{aligned}$$

# Quantifier-free linear real arithmetic (QFLRA)

signature 为一阶理论的概念, 一阶理论为一阶逻辑的受限形式, signature 为所允许出现的非逻辑符号

Linear real arithmetic is a first-order theory whose signature contains integer constants  $const$ , real-valued variables  $x$  and addition  $+$  to build (theory) terms, and strict comparison  $<$  to build (atomic) constraints:

term 为一阶逻辑概念

小于号  $<$  为 predicate

## Syntax of linear real arithmetic

Predicate (terms)  
terms:  $const$ , variable

Terms:  $t ::= const \mid x \mid t + t$

和他们构成的函数

Constraints:  $c ::= t < t$

Formulas:  $\varphi ::= c \mid \neg \varphi \mid \varphi \wedge \varphi \mid \exists x. \varphi$

- Syntactic sugar:  $t_1 = t_2$ ,  $t_1 \leq t_2$ ,  $5 \cdot x + \frac{2}{3} \cdot y < 4$   
都是 Linear, 只要没有变量 • 变量就行  
乘常数  $\rightarrow$  可以转化为  $+$  (同  $\times 3$ )  
 $\exists x, \varphi(x) \text{ true} \Leftrightarrow \forall x, \varphi(x) \text{ unsat}$
- The semantics is standard.
- Linear real arithmetic is also called linear real algebra.
- We consider the satisfiability problem for the quantifier-free fragment QFLRA (or equivalently the existential fragment, i.e., no universal quantifiers and no negation of expressions containing existential quantifiers).

# Reduced theory checks for lazy SMT

- **Reminder:** In (full/less) lazy SMT solving, the theory solver needs to check the satisfiability of **sets** of constraints (instead of arbitrary Boolean combinations).
- For **QFLRA**, we convert the input CNF **not to negate any constraint** by **transforming**
  - $\neg(t < t')$  to  $t \geq t'$ ,
  - $\neg(t \leq t')$  to  $t > t'$ ,
  - $\neg(t \geq t')$  to  $t < t'$ ,
  - $\neg(t > t')$  to  $t \leq t'$  and
  - $\neg(t = t')$  to  $(t < t' \vee t > t')$ .
- **Reduced theory checks:** If the input CNF contains **no negated constraints then**, for each Boolean solution, it is sufficient to check theory consistency for the **true constraints**.

# Correctness of reduced theory checks

If the input CNF contains no negated constraints then, for each Boolean solution, it is sufficient to check theory consistency for the **true** constraints.

**Proof:** Assume QFLRA CNF formula  $\varphi$  with constraints  $\mathcal{C}$ , none negated. Let  $\varphi_{abs}$  abstract  $\varphi$  by replacing constraints  $c \in \mathcal{C}$  by propositions  $b_c$ .

**First** we show that if  $\varphi_{abs}$  has no solution then  $\varphi$  is also unsatisfiable (i.e. **UNSAT answers are correct**), even if we restrict all theory checks to those constraints that are true under the current Boolean solution.

- 1 The statement holds for the initial Boolean abstraction  $\varphi_{abs}$ .
- 2 The abstraction  $\varphi_{abs}$  is changed only when the constraints sent to the theory solver are inconsistent; the returned infeasible subset  $E$  is used for the abstraction refinement to  $\varphi_{abs} \wedge (\bigvee_{e \in E} \neg b_e)$ . Since  $E$  is infeasible, the refinement is still conservative (i.e. unsatisfiability of  $\varphi_{abs}$  implies unsatisfiability of  $\varphi$ ).

# Correctness of reduced theory checks

**Second** we show that if  $\varphi_{abs}$  has a solution  $\alpha$  and the true constraints  $C_T = \{c \in \mathcal{C} \mid \alpha(b_c) = \text{true}\}$  are consistent then  $\varphi$  has a solution (i.e. **SAT answers are correct**). So assume a  $\varphi_{abs}$ -solution  $\alpha$  with consistent  $C := C_T$ . We generate a solution for  $\varphi$  by stepwise considering each constraint  $c \in \mathcal{C} \setminus C$  as follows:

- 1 If  $C \cup \{\neg c\}$  is consistent then set  $C := C \cup \{\neg c\}$ .
- 2 Otherwise, set  $C := C \cup \{c\}$  and modify  $\alpha$  to assign *true* to  $c$ . Note that since  $\varphi$  is constraint-negation-free,  $c$  appears only positively in  $\varphi$ , therefore the updated assignment still satisfies  $\varphi_{abs}$ .

After completing the above procedure for each constraint, we get a satisfying assignment  $\alpha$  for  $\varphi_{abs}$  such that the true constraints and the negated false constraints build a consistent set, i.e.  $\varphi$  is satisfiable.

# Problem statement

Thus the theory solver needs to consider sets  $\mathcal{C} = \{c_1, \dots, c_M\}$  of constraints of the form

$$c_i : \sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i,$$

where

- $a_{ik}$  and  $b_i$  are integer (or rational) constants,
- $x_k$  are real-valued variables and
- $\sim_i \in \{=, \leq, <\}$  are comparison operators.

for  $k=1, \dots, N$  and  $i=1, \dots, M$ .

Note:  $t > b$  is equivalent to  $-t < -b$  and similarly for  $\geq$ .

# Gauß variable elimination (based on equations)

处理等式

- Assume that the  $i$ th constraint is an **equation** with  $a_{ij} \neq 0$  for some  $1 \leq j \leq N$ :

$$\sum_{k=1}^N a_{ik} \cdot x_k = b_i \quad (a_{ik}, b_i: \text{integer/rational constants}, x_k: \text{variables})$$

$$\Rightarrow a_{ij} \cdot x_j = b_i - \sum_{k \in \{1, \dots, j-1, j+1, \dots, N\}} a_{ik} \cdot x_k$$

$$\Rightarrow x_j = \frac{b_i}{a_{ij}} - \sum_{k \in \{1, \dots, j-1, j+1, \dots, N\}} \frac{a_{ik}}{a_{ij}} \cdot x_k := \beta_j$$

- Replace  $x_j$  by  $\beta_j$  in all constraints (and multiply the involved constraints by  $a_{ij}$  if integer coefficients are wanted).
- After removing tautologies, this **substitution** leads to an equisatisfiable problem with (at most)  $M - 1$  constraints in (at most)  $N - 1$  variables (at least the  $i$ th constraint and  $x_j$  are eliminated).



# What remains to be solved

- Let us assume first that after applying Gauß variable elimination as long as possible,  $m$  weak inequalities in  $n$  variables are left (i.e., there are no strict inequalities):

weak inequality: 可以包含等号, 即  $\leq, \geq$

strong inequality: 不包含等号, 即  $<, >$

$$\bigwedge_{1 \leq i \leq m} \sum_{1 \leq j \leq n} a_{ij} x_j \leq b_i$$

- Input in matrix form:  $A\vec{x} \leq \vec{b}$

$$\begin{matrix} m \text{ constraints} & \begin{pmatrix} a_{11} & a_{12} & \cdots & \cdots & a_{1n} \\ a_{21} & a_{22} & \ddots & & \vdots \\ \vdots & & & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & \cdots & a_{mn} \end{pmatrix} & \begin{pmatrix} x_1 \\ \vdots \\ \vdots \\ x_n \end{pmatrix} & \leq & \begin{pmatrix} b_1 \\ \vdots \\ \vdots \\ b_m \end{pmatrix} \\ & n \text{ variables} & & & & & \end{matrix}$$

# Fourier-Motzkin variable elimination

处理高斯不等式

- Earliest method for solving **linear inequality systems**:  
discovered in 1826 by Fourier, re-discovered by Motzkin in 1936
- Basic idea of **variable elimination**:
  - Pick a variable and eliminate it, yielding an **equisatisfiable formula** that does not refer to the eliminated variable any more.
  - Continue until all variables are eliminated.
- **Fourier-Motzkin** repeats iteratively for each variable  $x_i$ ,  $i = n, \dots, 1$ :
  - For each inequality, bring  $x_i$  to stay alone at one side.
  - Collect all **lower bounds**  $\beta_{\ell 1}, \dots, \beta_{\ell L}$  from inequalities  $\beta_{\ell j} \leq x_i$ .
  - Collect all **upper bounds**  $\beta_{u 1}, \dots, \beta_{u U}$  from inequalities  $x_i \leq \beta_{u j}$ .
  - $\mathbb{R}$  is dense  $\leadsto$  Constraint set is **solvable** iff the **interval** between the **highest lower bound** and the **lowest upper bound** is **not empty**.
  - **BUT: bounds are symbolic terms, i.e. no order is given!**
  - **Solution:** Require **each lower bound** to be **less or equal** **each upper bound**.

# Variable bounds

- For a variable  $x_n$ , we can partition the constraints according to the coefficients of  $x_n$ :
  - $a_{in} = 0$ : constraint  $i$  puts **no bound** on  $x_n$
  - $a_{in} > 0$ : constraint  $i$  puts an **upper bound** on  $x_n$
  - $a_{in} < 0$ : constraint  $i$  puts a **lower bound** on  $x_n$

$$\sum_{j=1}^n a_{ij} \cdot x_j \leq b_i$$

$$\Rightarrow a_{in} \cdot x_n \leq b_i - \sum_{j=1}^{n-1} a_{ij} \cdot x_j$$

$$(a) \quad a_{in} > 0 \Rightarrow x_n \leq \frac{b_i}{a_{in}} - \sum_{j=1}^{n-1} \frac{a_{ij}}{a_{in}} \cdot x_j \quad \text{upper bound}$$

$$(b) \quad a_{in} < 0 \Rightarrow x_n \geq \frac{b_i}{a_{in}} - \sum_{j=1}^{n-1} \frac{a_{ij}}{a_{in}} \cdot x_j \quad \text{lower bound}$$

# Example for upper and lower bounds

$$1 \cdot x_1 + (-1) \cdot x_2 + 0 \cdot x_3 \leq 0$$



- (1)  $x_1 - x_2 \leq 0$
- (2)  $x_1 - x_3 \leq 0$
- (3)  $-x_1 + x_2 + 2x_3 \leq 0$
- (4)  $-x_3 \leq -1$

Category for  $x_1$ ?

Upper bound  $x_1 \leq x_2$

Upper bound  $x_1 \leq x_3$

Lower bound  $x_1 \geq x_2 + 2x_3$

No bound  $0 \cdot x_1 \leq x_2 - 1$

# Eliminating unbounded variables

- Iteratively remove variables that are not bounded in both ways (and all the constraints that use them).
- The new problem has a solution iff the old problem has one!

如果只有 upper bound / lower bound, 暂时先把约束忽略, 回头再将其他变量的结果代入

$$\left\{ \begin{array}{l} \cancel{-8x + 7y \leq 0} \\ \cancel{x \leq 3} \\ -y + z \leq 0 \\ -z \leq -10 \\ z \leq 20 \end{array} \right. \xrightarrow{\text{忽略}} \begin{array}{l} \cancel{y + z \leq 0} \\ -z \leq -10 \\ z \leq 20 \end{array} \xrightarrow{\text{将 } z \text{ 代入}} \begin{array}{l} -z \leq -10 \\ z \leq 20 \end{array}$$

*y可取值为 < -z 的最小值*

*将 z 代入*

# Fourier-Motzkin variable elimination

- For each pair of a lower bound  $\beta_\ell$  and an upper bound  $\beta_u$ , we have

$$\beta_\ell \leq x_n \leq \beta_u$$

- For each such pair, add the constraint

$$\beta_\ell \leq \beta_u$$

此时变量 $x$ 被消去

# Fourier-Motzkin: Example

			Category for $x_1$ ?
<del>(1)</del>	<del><math>x_1 - x_2 \leq 0</math></del>		Upper bound
<del>(2)</del>	<del><math>x_1 - x_3 \leq 0</math></del>		Upper bound
<del>(3)</del>	<del><math>x_1 + x_2 + 2x_3 \leq 0</math></del>		Lower bound
(4)	$-x_3 \leq -1$		Lower bound
<hr/>			eliminate $x_1$
(5)	$2x_3 \leq 0$	(from 1,3)	Upper bound
(6)	$x_2 + x_3 \leq 0$	(from 2,3)	Upper bound
<hr/>			we eliminate $x_3$
(7)	$1 \leq 0$	(from 4,5)	

→ **Contradiction** (the system is UNSAT)

## Algorithm: satCheck (only weak inequalities, full lazy)

**Input:** Set  $C$  of weak linear real arithmetic inequalities over variables  $x_1, \dots, x_n$

**Output:** Satisfiability of  $\bigwedge_{c \in C} c$

- 1  $k := 0$ ;
- 2 if  $k = n$  then goto 8 else  $k := k + 1$
- 3  $C_\ell := \{(\sum_{j=1}^n a_j x_j \leq b) \in C \mid a_k < 0\}$ ;
- 4  $C_u := \{(\sum_{j=1}^n a_j x_j \leq b) \in C \mid a_k > 0\}$ ;
- 5  $C := C \setminus (C_\ell \cup C_u)$ ; //case for  $a_k = 0$
- 6  $C := C \cup \{ \frac{b_\ell}{a_{\ell k}} - \sum_{j=k+1}^n \frac{a_{\ell j}}{a_{\ell k}} \cdot x_j \leq \frac{b_u}{a_{uk}} - \sum_{j=k+1}^n \frac{a_{uj}}{a_{uk}} \cdot x_j \mid$   
 $\sum_{j=1}^n a_{\ell j} x_j \leq b_\ell \in C_\ell \wedge \sum_{j=1}^n a_{uj} x_j \leq b_u \in C_u \}$ ;
- 7 goto 2
- 8 if  $\ell \leq u$  for all  $(\ell \leq u) \in C$  then return SAT; //here,  $\ell$  and  $u$  are constants!
- 9 else return UNSAT;



# Strict inequalities

The approach works also if we have both weak and strict inequalities. All we need to change is that

- we distinguish between **strict** and **weak** lower and upper bounds (defined by strict respectively weak inequalities), and
- for each pair of lower and upper bounds, if any of them is strict then we add the constraint

$$\beta_l < \beta_u \quad \Leftarrow \quad \beta_l < x \wedge x \leq \beta_u$$

instead of

$$\beta_l \leq \beta_u \quad \Leftarrow \quad \beta_l \leq x \wedge x \leq \beta_u$$

# Linear integer arithmetic, non-linear real arithmetic

- Question: Does this method work also for linear integer arithmetic, i.e., if the variables range over the integers (instead of the reals)?
- Answer: No. 因为给定区间内可能没有整数, e.g.  $2.5 \leq x \leq 2.7$
- Reason: The integer domain is not dense.
- Question: Does this method work also for non-linear real arithmetic, i.e., if the variables range over the reals but also multiplication is allowed?
- Answer: Sound but incomplete. e.g.  $x \cdot y \leq 2$   
 $y=0 \rightarrow 0 \leq 2$   
 $y \neq 0 \rightarrow \begin{cases} y > 0 & x \leq \frac{2}{y} \\ y < 0 & x \geq \frac{2}{y} \end{cases}$  not work in general
- Reason: In general it is not possible to transform constraints containing non-linear polynomial expressions such that we have a single variable on the left-hand-side and a real-arithmetic expression on the right-hand side (we would need complicated case distinctions, fractions and roots). 但是, 如果对于 non-linear, 变量是以 linear 的形式出现:  
e.g. 对于  $x$   $x + y \cdot z \leq 10 \Rightarrow x \leq 10 - y \cdot z$  可以转到一边 ✓

## Bonus exercise 19 (8 minutes)

Assume the following constraint set:

$$\{ \overset{(1)}{2x - y} \leq 8, \quad \overset{(3)}{2x + y} \leq -8, \quad \overset{(2)}{-2x - y} \leq 8, \quad \overset{(4)}{-2x + y} \leq 8 \}$$

Eliminating first  $y$  and then  $x$  using the method of Fourier-Motzkin will result in a single constraint. Which is it (without applying any simplification)?

<b>1</b> $4 \leq 0$	(1) $2x - 8 \leq y$	(3) $y \leq -2x - 8$	
<b>2</b> $-4 \leq 4$	(2) $-2x - 8 \leq y$	(4) $y \leq 2x + 8$	
<b>3</b> $0 \leq 4$	(1)+(3)	$2x - 8 \leq -2x - 8$	$4x \leq 0 \quad x \leq 0$
<b>4</b> $0 \leq 0$	(1)+(4)	$2x - 8 \leq 2x + 8$	$(0 \leq 16)$
<b>5</b> $-4 \leq 0$	(2)+(3)	$-2x - 8 \leq -2x - 8$	$(0 \leq 0)$
	(2)+(4)	$-2x - 8 \leq 2x + 8$	$-16 \leq 4x \quad -4 \leq x$

$-4 \leq 0$

- Worst-case complexity:

$$m \rightarrow \left(\frac{m}{2}\right)^2 = \frac{m^2}{4}$$

$$\rightarrow \left(\frac{\frac{m^2}{4}}{2}\right)^2 = \frac{m^4}{4^3}$$

$$\rightarrow \left(\frac{\frac{m^4}{4^3}}{2}\right)^2 = \frac{m^8}{4^7}$$

$\rightarrow \dots$

$$\rightarrow 4 \cdot \left(\frac{m}{4}\right)^{2^d}$$

最差情况: 一半的上界( $\frac{m}{2}$ ), 一半的下界( $\frac{m}{2}$ )

$$\hookrightarrow \left(\frac{m}{2}\right) \times \left(\frac{m}{2}\right) = \frac{m^2}{4}$$

$2^d$  dimension, 即变量的数量

- Heavy!

- The bottleneck: case-splitting

# Requirements on theory solver in the SMT context

- 1 Incrementality?
- 2 Minimal infeasible subsets?
- 3 Backtracking?

- How does the Gauß method eliminate a variable from a set of QFLRA constraints using an equality?
- How does the Fourier-Motzkin method eliminate a variable from a set of QFLRA constraints?
- What is the complexity of the methods?