

# Formal Methods in Software Development

## Gauß and Fourier-Motzkin Variable Elimination for Linear Real Arithmetic

Mădălina Eraşcu

West University of Timișoara  
Faculty of Mathematics and Informatics  
Department of Computer Science

Based on slides of the lecture Satisfiability Checking (Erika Ábrahám), RTWH Aachen

WS 2019/2020

# 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.

# 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.

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

# Quantifier-free linear real arithmetic (QFLRA)

# Quantifier-free linear real arithmetic (QFLRA)

Linear real arithmetic is the first-order theory with signature  $\{0, 1, +, <\}$  and the domain being the reals  $\mathbb{R}$ .

# Quantifier-free linear real arithmetic (QFLRA)

Linear real arithmetic is the first-order theory with signature  $\{0, 1, +, <\}$  and the domain being the reals  $\mathbb{R}$ .

## Syntax of linear real arithmetic

Terms:  $t ::= 0 \mid 1 \mid x \mid t + t$

Constraints:  $c ::= t < t$

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

where  $x$  stays for a variable.

# Quantifier-free linear real arithmetic (QFLRA)

**Linear real arithmetic** is the first-order theory with signature  $\{0, 1, +, <\}$  and the domain being the reals  $\mathbb{R}$ .

## Syntax of linear real arithmetic

Terms:  $t ::= 0 \mid 1 \mid x \mid t + t$

Constraints:  $c ::= t < t$

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

where  $x$  stays for a variable.

- Syntactic sugar for constraints:  $a - b$ ,  $t_1 \leq t_2$ ,  $t_1 = t_2$ ,  $t_1 \neq t_2$ .

# Quantifier-free linear real arithmetic (QFLRA)

**Linear real arithmetic** is the first-order theory with signature  $\{0, 1, +, <\}$  and the domain being the reals  $\mathbb{R}$ .

## Syntax of linear real arithmetic

Terms:  $t ::= 0 \mid 1 \mid x \mid t + t$

Constraints:  $c ::= t < t$

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

where  $x$  stays for a variable.

- Syntactic sugar for constraints:  $a - b$ ,  $t_1 \leq t_2$ ,  $t_1 = t_2$ ,  $t_1 \neq t_2$ .
- The semantics is standard.



# Quantifier-free linear real arithmetic (QFLRA)

**Linear real arithmetic** is the first-order theory with signature  $\{0, 1, +, <\}$  and the domain being the reals  $\mathbb{R}$ .

## Syntax of linear real arithmetic

Terms:	$t ::=$	$0$		$1$		$x$		$t + t$
Constraints:	$c ::=$	$t < t$						
Formulas:	$\varphi ::=$	$c$		$\neg\varphi$		$\varphi \wedge \varphi$		$\exists x. \varphi$

where  $x$  stays for a variable.

- Syntactic sugar for constraints:  $a - b$ ,  $t_1 \leq t_2$ ,  $t_1 = t_2$ ,  $t_1 \neq t_2$ .
- The semantics is standard.
- Linear real arithmetic is also called **linear real algebra**.

# Quantifier-free linear real arithmetic (QFLRA)

**Linear real arithmetic** is the first-order theory with signature  $\{0, 1, +, <\}$  and the domain being the reals  $\mathbb{R}$ .

## Syntax of linear real arithmetic

Terms:	$t ::=$	$0$		$1$		$x$		$t + t$
Constraints:	$c ::=$	$t < t$						
Formulas:	$\varphi ::=$	$c$		$\neg\varphi$		$\varphi \wedge \varphi$		$\exists x. \varphi$

where  $x$  stays for a variable.

- Syntactic sugar for constraints:  $a - b$ ,  $t_1 \leq t_2$ ,  $t_1 = t_2$ ,  $t_1 \neq t_2$ .
- 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).

# Gauß variable elimination (based on equations)

Reminder: In an SMT solver for QFLRA, the theory solver needs to check the satisfiability of **sets** of constraints  $\sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i$ , where  $a_{i,k}$  and  $b_i$  are integer (or rational) constants,  $x_k$  are real-valued variables, and  $\sim_i \in \{=, \leq, <\}$  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)

Reminder: In an SMT solver for QFLRA, the theory solver needs to check the satisfiability of **sets** of constraints  $\sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i$ , where  $a_{i,k}$  and  $b_i$  are integer (or rational) constants,  $x_k$  are real-valued variables, and  $\sim_i \in \{=, \leq, <\}$  for  $k=1, \dots, N$  and  $i=1, \dots, M$ . (Note:  $t > b$  is equivalent to  $-t < -b$  and similarly for  $\geq$ .)

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

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

# Gauß variable elimination (based on equations)

Reminder: In an SMT solver for QFLRA, the theory solver needs to check the satisfiability of **sets** of constraints  $\sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i$ , where  $a_{i,k}$  and  $b_i$  are integer (or rational) constants,  $x_k$  are real-valued variables, and  $\sim_i \in \{=, \leq, <\}$  for  $k=1, \dots, N$  and  $i=1, \dots, M$ . (Note:  $t > b$  is equivalent to  $-t < -b$  and similarly for  $\geq$ .)

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

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

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

# Gauß variable elimination (based on equations)

Reminder: In an SMT solver for QFLRA, the theory solver needs to check the satisfiability of **sets** of constraints  $\sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i$ , where  $a_{i,k}$  and  $b_i$  are integer (or rational) constants,  $x_k$  are real-valued variables, and  $\sim_i \in \{=, \leq, <\}$  for  $k=1, \dots, N$  and  $i=1, \dots, M$ . (Note:  $t > b$  is equivalent to  $-t < -b$  and similarly for  $\geq$ .)

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

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

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

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

# Gauß variable elimination (based on equations)

Reminder: In an SMT solver for QFLRA, the theory solver needs to check the satisfiability of **sets** of constraints  $\sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i$ , where  $a_{i,k}$  and  $b_i$  are integer (or rational) constants,  $x_k$  are real-valued variables, and  $\sim_i \in \{=, \leq, <\}$  for  $k=1, \dots, N$  and  $i=1, \dots, M$ . (Note:  $t > b$  is equivalent to  $-t < -b$  and similarly for  $\geq$ .)

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

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

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

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

- Replace  $x_j$  by  $\beta_j$  in all constraints (and multiply the involved constraints by  $a_{ij}$  if integer coefficients are wanted).

# Gauß variable elimination (based on equations)

Reminder: In an SMT solver for QFLRA, the theory solver needs to check the satisfiability of sets of constraints  $\sum_{k=1}^N a_{ik} \cdot x_k \sim_i b_i$ , where  $a_{i,k}$  and  $b_i$  are integer (or rational) constants,  $x_k$  are real-valued variables, and  $\sim_i \in \{=, \leq, <\}$  for  $k=1, \dots, N$  and  $i=1, \dots, M$ . (Note:  $t > b$  is equivalent to  $-t < -b$  and similarly for  $\geq$ .)

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

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

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

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

- 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 substitutiton 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,  $m$  non-strict inequalities in  $n$  variables are left (i.e., there are no strict inequalities):

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

# What remains to be solved

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

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

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

$m$  constraints

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

$n$  variables

- Earliest method for solving **linear inequality systems**:  
discovered in 1826 by Fourier, re-discovered by Motzkin in 1936

- 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.

- 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**: Collect requirements on the **lower and upper bounds** on the variable we want to eliminate.

# 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  $x_n$

# 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  $x_n$

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

# 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  $x_n$

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

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



# 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  $x_n$

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

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

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

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

# Example for upper and lower bounds

Category for  $x_1$ ?

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

# Example for upper and lower bounds

- Category for  $x_1$ ?  
Upper bound
- (1)  $x_1 - x_2 \leq 0$
  - (2)  $x_1 - x_3 \leq 0$
  - (3)  $-x_1 + x_2 + 2x_3 \leq 0$
  - (4)  $-x_3 \leq -1$

# Example for upper and lower bounds

- Category for  $x_1$ ?
- (1)  $x_1 - x_2 \leq 0$  Upper bound
  - (2)  $x_1 - x_3 \leq 0$  Upper bound
  - (3)  $-x_1 + x_2 + 2x_3 \leq 0$
  - (4)  $-x_3 \leq -1$

# Example for upper and lower bounds

- |                                | Category for $x_1$ ? |
|--------------------------------|----------------------|
| (1) $x_1 - x_2 \leq 0$         | Upper bound          |
| (2) $x_1 - x_3 \leq 0$         | Upper bound          |
| (3) $-x_1 + x_2 + 2x_3 \leq 0$ | Lower bound          |
| (4) $-x_3 \leq -1$             |                      |

# Example for upper and lower bounds

Category for $x_1$ ?	
(1) $x_1 - x_2 \leq 0$	Upper bound
(2) $x_1 - x_3 \leq 0$	Upper bound
(3) $-x_1 + x_2 + 2x_3 \leq 0$	Lower bound
(4) $-x_3 \leq -1$	No bound

# 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!

$$\begin{array}{rcl} -8x + 7y & \leq & 0 \\ -x & \leq & -3 \\ -y + z & \leq & 0 \\ -z & \leq & -10 \\ z & \leq & 20 \end{array}$$

# 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!

$$\begin{array}{rcl} \cancel{-8x + 7y} & \leq & \cancel{0} \\ \cancel{x} & \leq & \cancel{3} \\ -y + z & \leq & 0 \\ -z & \leq & -10 \\ z & \leq & 20 \end{array}$$



# 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!

$$\begin{array}{rcl} \cancel{-8x + 7y} & \leq & \cancel{0} \\ \cancel{x} & \leq & \cancel{3} \\ -y + z & \leq & 0 \\ -z & \leq & -10 \\ z & \leq & 20 \end{array} \quad \longrightarrow \quad \begin{array}{rcl} -y + z & \leq & 0 \\ -z & \leq & -10 \\ z & \leq & 20 \end{array}$$

# 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!

$$\begin{array}{rcl} \cancel{8x + 7y} & \leq & \cancel{0} \\ \cancel{x} & \leq & \cancel{3} \\ -y + z & \leq & 0 \\ -z & \leq & -10 \\ z & \leq & 20 \end{array} \quad \longrightarrow \quad \begin{array}{rcl} \cancel{y + z} & \leq & \cancel{0} \\ -z & \leq & -10 \\ z & \leq & 20 \end{array}$$

# 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!

$$\begin{array}{rcl} \cancel{-8x + 7y} & \leq & \cancel{0} \\ \cancel{x} & \leq & \cancel{3} \\ -y + z & \leq & 0 \\ -z & \leq & -10 \\ z & \leq & 20 \end{array} \quad \longrightarrow \quad \begin{array}{rcl} \cancel{y + z} & \leq & \cancel{0} \\ -z & \leq & -10 \\ z & \leq & 20 \end{array} \quad \longrightarrow \quad \begin{array}{rcl} -z & \leq & -10 \\ z & \leq & 20 \end{array}$$

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

$$\beta_l \leq x_n \leq \beta_u$$

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

$$\beta_l \leq x_n \leq \beta_u$$

- For each such pair, add the constraint

$$\beta_l \leq \beta_u$$

Category for  $x_1$ ?

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

# Fourier-Motzkin: Example

$$(1) \quad x_1 - x_2 \leq 0$$

$$(2) \quad x_1 - x_3 \leq 0$$

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

$$(4) \quad -x_3 \leq -1$$

---

Category for  $x_1$ ?

Upper bound

Upper bound

Lower bound

eliminate  $x_1$

# Fourier-Motzkin: Example

$$(1) \quad x_1 - x_2 \leq 0$$

$$(2) \quad x_1 - x_3 \leq 0$$

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

$$(4) \quad -x_3 \leq -1$$

---

$$(5) \quad 2x_3 \leq 0 \quad (\text{from 1,3})$$

Category for  $x_1$ ?

Upper bound

Upper bound

Lower bound

eliminate  $x_1$



# Fourier-Motzkin: Example

$$(1) \quad x_1 - x_2 \leq 0$$

$$(2) \quad x_1 - x_3 \leq 0$$

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

$$(4) \quad -x_3 \leq -1$$

Category for  $x_1$ ?

Upper bound

Upper bound

Lower bound

---

$$(5) \quad 2x_3 \leq 0 \quad (\text{from } 1,3)$$

$$(6) \quad x_2 + x_3 \leq 0 \quad (\text{from } 2,3)$$

eliminate  $x_1$

# Fourier-Motzkin: Example

Category for  $x_1$ ?

$$\begin{array}{l} \text{(1)} \quad x_1 - x_2 \leq 0 \\ \text{(2)} \quad x_1 - x_3 \leq 0 \\ \text{(3)} \quad x_1 + x_2 + 2x_3 \leq 0 \\ \text{(4)} \quad -x_3 \leq -1 \end{array}$$

eliminate  $x_1$

---

$$\begin{array}{ll} \text{(5)} \quad 2x_3 \leq 0 & \text{(from 1,3)} \\ \text{(6)} \quad x_2 + x_3 \leq 0 & \text{(from 2,3)} \end{array}$$

# Fourier-Motzkin: Example

Category for  $x_1$ ?

$$\begin{array}{l} \text{(1)} \quad x_1 - x_2 \leq 0 \\ \text{(2)} \quad x_1 - x_3 \leq 0 \\ \text{(3)} \quad x_1 + x_2 + 2x_3 \leq 0 \\ \text{(4)} \quad -x_3 \leq -1 \end{array}$$

eliminate  $x_1$

$$\begin{array}{l} \text{(5)} \quad 2x_3 \leq 0 \quad \text{(from 1,3)} \\ \text{(6)} \quad x_2 + x_3 \leq 0 \quad \text{(from 2,3)} \end{array}$$

eliminate  $x_2$

# Fourier-Motzkin: Example

Category for  $x_1$ ?

~~(1)  $x_1 - x_2 \leq 0$~~

~~(2)  $x_1 - x_3 \leq 0$~~

~~(3)  $x_1 + x_2 + 2x_3 \leq 0$~~

(4)  $-x_3 \leq -1$

---

(5)  $2x_3 \leq 0$  (from 1,3)

(6)  $x_2 + x_3 \leq 0$  (from 2,3)

---

Lower bound

eliminate  $x_1$

Upper bound

Upper bound

eliminate  $x_2$

# Fourier-Motzkin: Example

Category for  $x_1$ ?

~~(1)  $x_1 - x_2 \leq 0$~~

~~(2)  $x_1 - x_3 \leq 0$~~

~~(3)  $x_1 + x_2 + 2x_3 \leq 0$~~

(4)  $-x_3 \leq -1$

---

(5)  $2x_3 \leq 0$  (from 1,3)

~~(6)  $x_2 + x_3 \leq 0$  (from 2,3)~~

---

(7)  $1 \leq 0$  (from 4,5)

Lower bound

eliminate  $x_1$

Upper bound

Upper bound

eliminate  $x_2$

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

# Strict inequalities

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

# Strict inequalities

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

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

$$\beta_l < \beta_u$$

instead of

$$\beta_l \leq \beta_u .$$

- Question: Does this method work also for linear **integer** arithmetic, i.e., if the variables range over the integers (instead of the reals)?



# 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.

# 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.
- Reason: The integer domain is **not dense**.

# 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.
- 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?

# 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.
- 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: No.

# 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.
- 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: No.
- 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).

- Worst-case complexity:

$m \rightarrow$

- Worst-case complexity:

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

- Worst-case complexity:

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



- Worst-case complexity:

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

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

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

- Worst-case complexity:

$$\begin{aligned} m &\rightarrow \left(\frac{m}{2}\right)^2 = \frac{m^2}{4} \\ &\rightarrow \left(\frac{\frac{m^2}{4}}{2}\right)^2 = \frac{m^4}{4^3} \\ &\rightarrow \left(\frac{\frac{m^4}{4^3}}{2}\right)^2 = \frac{m^8}{4^7} \\ &\rightarrow \dots \end{aligned}$$

■ Worst-case complexity:

$$\begin{aligned} m &\rightarrow \left(\frac{m}{2}\right)^2 = \frac{m^2}{4} \\ &\rightarrow \left(\frac{\frac{m^2}{4}}{2}\right)^2 = \frac{m^4}{4^3} \\ &\rightarrow \left(\frac{\frac{m^4}{4^3}}{2}\right)^2 = \frac{m^8}{4^7} \\ &\rightarrow \dots \\ &\rightarrow 4 \cdot \left(\frac{m}{4}\right)^{2^d} \end{aligned}$$

- Worst-case complexity:

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

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

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

$$\rightarrow \dots$$

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

- Heavy!

- Worst-case complexity:

$$\begin{aligned}m &\rightarrow \left(\frac{m}{2}\right)^2 = \frac{m^2}{4} \\&\rightarrow \left(\frac{\frac{m^2}{4}}{2}\right)^2 = \frac{m^4}{4^3} \\&\rightarrow \left(\frac{\frac{m^4}{4^3}}{2}\right)^2 = \frac{m^8}{4^7} \\&\rightarrow \dots \\&\rightarrow 4 \cdot \left(\frac{m}{4}\right)^{2^d}\end{aligned}$$

- Heavy!
- The bottleneck: case-splitting

# Requirements on theory solver in the SMT context

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

# Next Lecture: Linear Integer Arithmetic