



Constraint Solving

René Thiemann and Fabian Mitterwallner

based on a previous course by Aart Middeldorp

Outline

- 1. Summary of Previous Lecture**
- 2. Tightening**
- 3. Cubes**
- 4. Equality Detection**
- 5. Equality Elimination**
- 6. Further Reading**

Example (Application of Linear Integer Arithmetic: Termination Proving)

- consider program
- model loop-iteration as formula φ using pre-variables \vec{x} and post-variables \vec{x}'
- prove termination by choosing expression e and integer constant f and show that two LIA problems are unsatisfiable
 - $\varphi \wedge e(\vec{x}) < e(\vec{x}') + 1$ (\neg decrease)
 - $\varphi \wedge e(\vec{x}) < f$ (\neg bounded)
- for certain programs, reasoning over integers is essential

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Branch-and-Bound Algorithm

- core idea for finding integral solution
 - simplex algorithm is used to find real solution v or detect unsat in \mathbb{R}
 - whenever $q := v(x) \notin \mathbb{Z}$, consider two possibilities: add $x \leq \lfloor q \rfloor$ or $\lceil q \rceil \leq x$
- small model property is required for termination: obtain finite search space

Theorem (Small Model Property)

if LIA formula ψ has solution over \mathbb{Z} then it has a solution v with

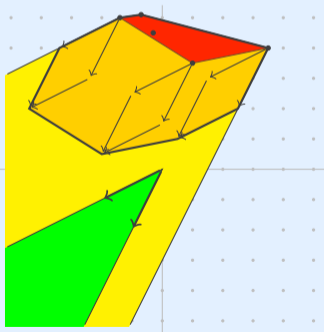
$$|v(x)| \leq \text{bound}(\psi) := (n + 1) \cdot \sqrt{n^n} \cdot c^n$$

for all x where

- n : number of variables in ψ
- c : maximal absolute value of numbers in ψ

Proof Idea of Small Model Property

- 1 convert conjunctive LIA formula ψ into form $A\vec{x} \leq \vec{b}$
- 2 represent polyhedron $\{\vec{x} \mid A\vec{x} \leq \vec{b}\}$ as polyhedron $P = \underbrace{\text{hull}(X)}_{\text{yellow}} + \underbrace{\text{cone}(V)}_{\text{green}}$
- 3 show that P has small integral solutions (orange), depending on X and V
- 4 approximate entries of vectors in X and V to obtain small model property



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- consider inequality $\sum a_i x_i \bowtie b$ with $\bowtie \in \{<, \leq\}$, and $a_1, \dots, a_n, b \in \mathbb{Z}$
- **tightening** preprocesses such inequality in a way that **preserves integer solutions** (but removes solutions in \mathbb{R})
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- last lecture: **tightening of strict inequalities** $\sum a_i x_i < b$ to $\sum a_i x_i \leq b - 1$
 - $2 < 517x + 2y < 3$ can be tightened to $3 \leq 517x + 2y \leq 2$, trivially unsat via simplex

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- example from last lecture
 - tighten $1 \leq 3x - 3y \leq 2$ to $\lceil \frac{1}{3} \rceil \leq x - y \leq \lfloor \frac{2}{3} \rfloor$
 - result $1 \leq x - y \leq 0$ is unsat by simplex

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- consider polyhedron P , the triangle
- none of the corners is integral, hence BB will require some iterations

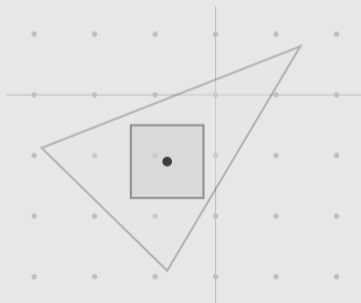


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Example

- consider polyhedron P , the triangle
- none of the corners is integral, hence BB will require some iterations
- cube C , the square, is contained in P and has edge-length 1.2
- hence C contains an integral solution, which can be calculated from the center point of C



Definition of Cubes

$\text{cube}_s(\vec{z})$ is the cube with center $\vec{z} \in \mathbb{R}^n$ and size $s \in \mathbb{R}_{\geq 0}$

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Proof

choose \vec{p} where $p_i = \lfloor z_i \rceil$, i.e., rounding z_i to the nearest integer

Example

the center of the cube on slide 9 is $\vec{z} = (-0.8, -1.1)^t$, so we compute $\vec{p} = (-1, -1)^t$

Cube Inclusion

- consider some polyhedron $P = \{\vec{x} \mid A\vec{x} \leq \vec{b}\}$ for some $A \in \mathbb{R}^{m \times n}$ and $\vec{b} \in \mathbb{R}^m$
- we are interested in whether P contains a cube of size s , formally:

$$\exists \vec{z}. \text{cube}_s(\vec{z}) \subseteq \{\vec{x} \mid A\vec{x} \leq \vec{b}\} \quad (1)$$

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Lemma (Cube Inclusion for Single Inequality)

For a single inequality $\vec{a} \cdot \vec{x} \leq c$ with $\vec{a} \in \mathbb{R}^n$, $c \in \mathbb{R}$ there is the equivalence:

$$\text{cube}_s(\vec{z}) \subseteq \{\vec{x} \mid \vec{a} \cdot \vec{x} \leq c\} \quad \text{iff} \quad \vec{a} \cdot \vec{z} \leq c - s \sum_{i=1}^n |a_i| \quad (2)$$

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Corollary

Cube inclusion (1) is satisfied iff $A\vec{z} \leq \vec{b} - s \cdot \begin{pmatrix} |A_{11}| + \dots + |A_{1n}| \\ \dots \\ |A_{m1}| + \dots + |A_{mn}| \end{pmatrix}$ has solution $\vec{z} \in \mathbb{R}^n$

Proof of One Direction of the Cube Inclusion for Single Inequalities

We assume

$$(A) \quad \vec{x} \in \text{cube}_s(\vec{z}) \quad \text{and} \quad (B) \quad \vec{a} \cdot \vec{z} \leq c - s \sum_{i=1}^n |a_i|$$

and prove $\vec{a} \cdot \vec{x} \leq c$ as follows:

$$\vec{a} \cdot \vec{x}$$

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The Unit-Cube Test of Bromberger and Weidenbach

- a **unit-cube** has edge-length 1, i.e., $s = 1/2$
- testing the unit-cube inclusion is possible via one invocation of simplex by checking existence of \vec{z} for inequalities

$$A\vec{z} \leq \vec{b} - \frac{1}{2} \cdot \begin{pmatrix} |A_{11}| + \dots + |A_{1n}| \\ \dots \\ |A_{m1}| + \dots + |A_{mn}| \end{pmatrix}$$

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- increase applicability as follows
 - first **detect** all implied **equalities**
 - then **eliminate equalities** (or detect unsat purely from equalities, e.g., from $2x = 1$)
 - afterwards achieve higher success rate of unit-cube test (and lower bounds for BB)

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- $A\vec{x} \leq \vec{b}$ **implies equality** $\vec{c} \cdot \vec{x} = d$ if every solution $\vec{x} \in \mathbb{R}^n$ of $A\vec{x} \leq \vec{b}$ satisfies $\vec{c} \cdot \vec{x} = d$

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Further Results

- interestingly, also the other direction is satisfied
 - assume $A\vec{x} < \vec{b}$ is unsatisfiable
 - then there is some **minimal unsatisfiable subset** I such that $\bigwedge_{i \in I} \vec{a}_i \cdot \vec{x} < b_i$ is unsatisfiable (obtained by a single simplex invocation)
 - **lemma: for every $i \in I$, the i -th equality $\vec{a}_i \cdot \vec{x} = b_i$ is implied**

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 - **lemma: for every $i \in I$, the i -th equality $\vec{a}_i \cdot \vec{x} = b_i$ is implied**
- overall: given $A\vec{x} \leq \vec{b}$ with one simplex invocation it is possible to
 - either get access to an implied equality
 - or figure out that no such equality exists

Equality Basis Algorithm of Bromberger and Weidenbach

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- delete all trivial inequalities from $A\vec{x} \leq \vec{b}$, the result is the initial A' and \vec{b}'

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 - choose any equality $\vec{a}'_i \cdot \vec{x} = b'_i$ for $i \in I$

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 - choose any equality $\vec{a}'_i \cdot \vec{x} = b'_i$ for $i \in I$
 - choose any j such that $A'_{ij} \neq 0$
 - reorder equation to obtain equation (e_j) of the form $x_j = \frac{b'_i}{A'_{ij}} - \sum_{k \neq j} \frac{A'_{ik}}{A'_{ij}} x_k$

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- delete all trivial inequalities from $A\vec{x} \leq \vec{b}$, the result is the initial A' and \vec{b}'
- while $\text{simplex}(A'\vec{x} < \vec{b}')$ delivers minimal unsat core I
 - choose any equality $\vec{a}'_i \cdot \vec{x} = b'_i$ for $i \in I$
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 - remove trivial equations from $A'\vec{x} \leq \vec{b}'$ after simplifications

Equality Basis Algorithm of Bromberger and Weidenbach

- the task is to iteratively **deduce all equalities** for (satisfiable) constraints $A\vec{x} \leq \vec{b}$
- in each iteration, one equality is added and one variable is eliminated
- throughout the algorithm E stores equalities, and $A'\vec{x} \leq \vec{b}'$ is a set of inequalities
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- return E and the final $A'\vec{x} \leq \vec{b}'$

Lemma

- $A\vec{x} \leq \vec{b}$ is equivalent to $E \cup A'\vec{x} \leq \vec{b}'$ throughout the algorithm
- the final set E is even an **equality basis**, i.e., every implied equality $\vec{c} \cdot \vec{x} = d$ of $A\vec{x} \leq \vec{b}$ is a linear combination of equations in E

Example

- initial constraints

$$y' - y - x_1 \leq -1$$

$$y - y' + x_1 \leq 1$$

$$2x_2 - x_1 \leq 0$$

$$x_1 - x_2 \leq 0$$

$$-x_1 \leq 0$$

$$z - y + 2y' - x_2 \leq 5$$

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- simplex on strict inequalities detects equality $y' - y - x_1 = -1$, so $y' = y + x_1 - 1$
- hence $E = \{y' = y + x_1 - 1\}$ and inequalities become

$$(y + x_1 - 1) - y - x_1 \leq -1 \quad \text{is simplified and deleted} \quad -1 \leq -1$$

$$y - (y + x_1 - 1) + x_1 \leq 1 \quad \text{is simplified and deleted} \quad 1 \leq 1$$

$$2x_2 - x_1 \leq 0$$

$$x_1 - x_2 \leq 0$$

$$-x_1 \leq 0$$

$$z - y + 2(y + x_1 - 1) - x_2 \leq 5 \quad \text{is simplified to} \quad z + y + 2x_1 - x_2 \leq 7$$

Example (continued)

- $E = \{y' = y + x_1 - 1\}$ and
inequalities $\{2x_2 - x_1 \leq 0, x_1 - x_2 \leq 0, -x_1 \leq 0, z + y + 2x_1 - x_2 \leq 7\}$

Example (continued)

- $E = \{y' = y + x_1 - 1\}$ and inequalities $\{2x_2 - x_1 \leq 0, x_1 - x_2 \leq 0, -x_1 \leq 0, z + y + 2x_1 - x_2 \leq 7\}$
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- after substitution and simplification get $E = \{y' = y + x_2 - 1, x_1 = x_2\}$ and inequalities $\{x_2 \leq 0, -x_2 \leq 0, z + y + x_2 \leq 7\}$

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- after substitution and simplification get $E = \{y' = y - 1, x_1 = 0, x_2 = 0\}$ and inequalities $\{z + y \leq 7\}$
- no further equalities are detected
- finally simplex can be used to find solution of $z + y \leq 7$ for variables y and z , and E determines the values for all other variables

Final Remarks

- the algorithm detects equalities over \mathbb{R}
- given the final set E and final inequalities I , one can always easily transform real solutions of I to real solutions of the initial constraints by using E

Outline

1. Summary of Previous Lecture
2. Tightening
3. Cubes
4. Equality Detection
- 5. Equality Elimination**
6. Further Reading

Current Situation

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 - $X = Y \uplus Z$

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 - for every $y \in Y$ there is exactly one equation $y = e_y$ in E

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 - if $v(z) = 1$ then $v(y) = \frac{3}{2} + 9 \notin \mathbb{Z}$
- upcoming: algorithm to convert E in a way that
 - integer solutions can always be extended via resulting equations, or
 - it is detected that E itself is not solvable in the integers
 - remark: additional variables may be required, hence only obtain equisatisfiability

Diophantine Equation Solver of Griggio – Setup

- input: set E of linear equalities
- output: “unsat” or “sat” with list of equations in solved form
- **solved form**
 - list of equations of the shape $x = e_x$ with e_x linear expression with integer coefficients
 - no cyclic dependencies: e_x in list $[\dots, x = e_x, \dots, y = e_y, \dots]$ does neither contain x nor y

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- F is a set of equations that need to be processed, initially $F = E$
- processing is done equation by equation, and the selected equation is **marked**; initially no equation is marked
- *normalize* is a sub-algorithm that transforms each equation into form $\sum a_i x_i = b$ with $a_1, \dots, a_n, b \in \mathbb{Z}$, $\gcd(a_1, \dots, a_n, b) = 1$
 - $normalize(2x - 6y - 14 = 0) = (x - 3y = 7)$
 - $normalize(x = \frac{1}{2}y + \frac{2}{3}) = (6x - 3y = 4)$

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 - continue with step 1
- 7 otherwise, handle case $|a_k| > 1$ (cf. slide 26) and continue with step 1

Example

- $F = \{x = \frac{1}{6}y + 1, 4z + y = 2\}, S = []$

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- reordering: $y = 6x - 6$

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- reordering: $y = 6x - 6$
- updates: $S = [y = 6x - 6], F = \{4z + (6x - 6) = 2\}$

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- normalization: $F = \{2z + 3x = 4\}$

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- reordering: $y = 6x - 6$
- updates: $S = [y = 6x - 6], F = \{4z + (6x - 6) = 2\}$
- normalization: $F = \{2z + 3x = 4\}$
- selection: $2z + 3x = 4$ gets marked
- step 7 required

Diophantine Equation Solver – Case $|a_k| > 1$

- we can write any integer number c as $c^q a_k + c^r$ where c^q and c^r are quotient and remainder when dividing c by a_k

- hence

$$\sum a_i x_i = b \equiv a_k x_k + \sum_{i \neq k} a_i x_i = b$$

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Diophantine Equation Solver – Case $|a_k| > 1$

- we can write any integer number c as $c^q a_k + c^r$ where c^q and c^r are quotient and remainder when dividing c by a_k

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- update $S := (\text{eq}) : S$ and eliminate x_k in F by substituting with (eq) as in step 6 (note that the marker stays on the previously marked equation)

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 - on symbolic values: translate constraints (e.g., inequalities)
apply equations starting with end of S
 - on concrete value: translate solution (e.g., from a solution of inequalities)
compute values by equations starting at beginning of S

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Let E be a set of linear equations

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- if the result on input E is “sat” with answer S , then
 - E and S are equisatisfiable in \mathbb{Z} ,
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input: set of linear inequalities

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- 9 apply branch-and-bound on I and either store solution or return “unsat”
- 10 output solution where S is used to compute values of further variables

Example (for improved LIA solver)

- input

$$2x_1 \leq 5x_3$$

$$5x_3 \leq 2x_1$$

$$x_2 = 3x_4$$

$$2x_1 + x_2 + x_3 \geq 7$$

$$2x_1 + x_2 + x_3 \leq 8$$

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- equation detection splits constraints into

$$E = \{2x_1 = 5x_3, x_2 = 3x_4\}$$

and

$$I = \{7 \leq 2x_1 + x_2 + x_3 \leq 8\}$$

Example (continued)

- Diophantine equation solver on $E = \{2x_1 = 5x_3, x_2 = 3x_4\}$ delivers solved form S and introduces new variable x_5

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- (solution of I for x_4 and x_5 would be extensible to x_3, x_1, x_2 via S)

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- not discussed
 - how to make LIA-solver incremental in DPLL(T) approach
 - how to generate small unsatisfiable cores for DPLL(T) approach
- strategy used in DPLL(T) solvers for LIA
 - utilize (incomplete) LRA-reasoning as much as possible, and
 - only invoke full LIA-solver when a complete Boolean model has been found

Outline

1. Summary of Previous Lecture
2. Tightening
3. Cubes
4. Equality Detection
5. Equality Elimination
- 6. Further Reading**

Further Reading



Alberto Griggio

A Practical Approach to Satisfiability Modulo Linear Integer Arithmetic

Journal on Satisfiability, Boolean Modeling and Computation, volume 8, pages 1–27, 2012.



Martin Bromberger and Christoph Weidenbach

New techniques for linear arithmetic: cubes and equalities

Formal Methods in System Design, volume 51, pages 433–461, 2017.

Important Concepts

- cube
- Diophantine equation solver
- equality basis algorithm
- implied equality
- tightening