

Satisfiability Checking

Summary III

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Informatik 2
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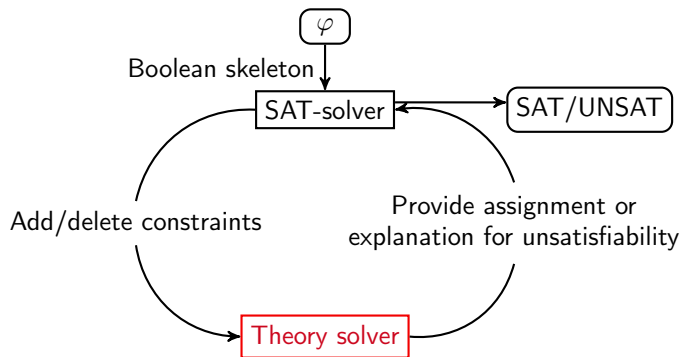
WS 19/20

We consider input formulae φ from the theory of **quantifier-free nonlinear real arithmetic (QFNRA)**:

$$\begin{array}{ll} p & := \text{const} \mid x \mid (p + p) \mid (p - p) \mid (p \cdot p) & \text{polynomials} \\ c & := p < 0 \mid p = 0 & \text{(polynomial) constraints} \\ \varphi & := c \mid (\varphi \wedge \varphi) \mid \neg\varphi & \text{QFNRA formulas} \end{array}$$

where constants *const* and variables x take real values from \mathbb{R} .

Connection to SMT



Solves $\begin{pmatrix} p_1 \sim_1 0 \\ \vdots \\ p_m \sim_m 0 \end{pmatrix}$ where $p_i \in \mathbb{Z}[x_1, \dots, x_n]$,
 $\sim_i \in \{<, \leq, =, \neq, \geq, >\}$ for $1 \leq i \leq m$.

- 1 Interval constraint propagation
- 2 Subtropical satisfiability
- 3 Virtual substitution
- 4 Cylindrical algebraic decomposition

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Interval constraint propagation (ICP)

- Incomplete but cheap method.
- **Basic idea:**
Start with a list containing a single initial **box** (value domain).
Use the input constraints to **contract** a non-empty box from the list.
If no contraction possible, **split** a non-empty box.
- **Termination:** all boxes are empty (UNSAT) or there is a sufficiently small non-empty box (possibly SAT).

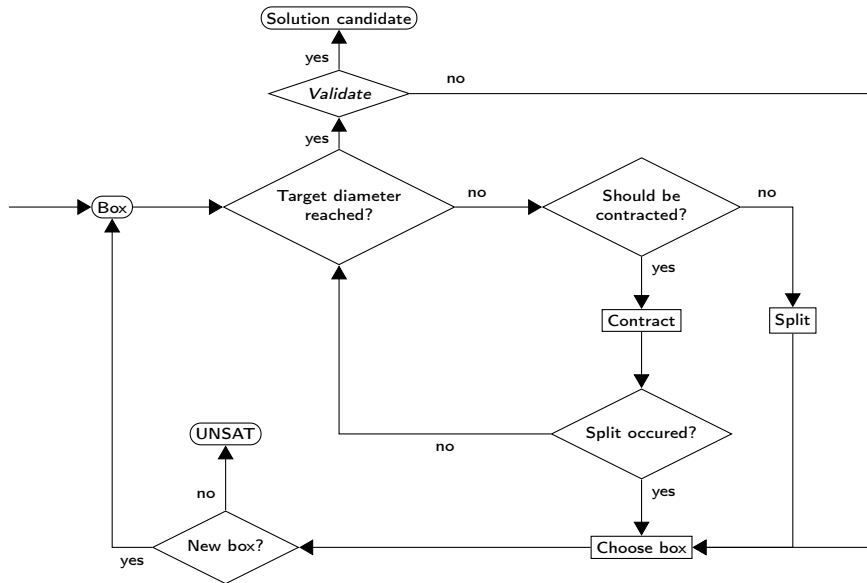
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First contraction approach: Interval arithmetic

Second contraction approach: Interval Newton method

Algorithm overview



Contraction I: Preprocessing

- Set $C' := C$ and $C := \emptyset$.
- Repeat as long as C' is not empty:
 - Take a constraint $e_1 \sim e_2$, $\sim \in \{<, \leq, =, \geq, >\}$, from C' .
 - Bring $e_1 \sim e_2$ to the normal form $r_1 \cdot m_1 + \dots + r_k \cdot m_k \sim 0$, where $r_i \in \mathbb{R}$ and m_i are monomials (either 1 or a product of variables) for each $i = 1, \dots, k$.
 - Replace each non-linear monomial m_i in $r_1 \cdot m_1 + \dots + r_k \cdot m_k \sim 0$ by a fresh variable h_i and add the result to C .
 - For each newly added variable h_i replacing m_i in the previous step, add an equation $h_i - m_i = 0$ to C , and initialize the bounds of h_i to the interval we get when we substitute the variable bounds in m_i and evaluate the result using interval arithmetic (note: the result will always be a single interval because there is no division or square root in m_i).

Contraction I: Interval arithmetic

- Step 1: Partially extend real arithmetic operations to $\mathbb{R} \cup \{-\infty, +\infty\}$.
- Step 2: Extend real arithmetic operations to intervals (interval arithmetic).

Definition (Interval arithmetic)

Assume real intervals $A = [\underline{A}, \overline{A}]$ and $B = [\underline{B}, \overline{B}]$.

$$A + B = [\underline{A} + \underline{B}; \overline{A} + \overline{B}]$$

$$A - B = [\underline{A} - \overline{B}; \overline{A} - \underline{B}]$$

$$A \cdot B = [\min(\underline{A} \cdot \underline{B}, \underline{A} \cdot \overline{B}, \overline{A} \cdot \underline{B}, \overline{A} \cdot \overline{B}); \max(\underline{A} \cdot \underline{B}, \underline{A} \cdot \overline{B}, \overline{A} \cdot \underline{B}, \overline{A} \cdot \overline{B})]$$

$$A^2 = (A \cdot A) \cap [0; +\infty)$$

$$A \div B = A \cdot \frac{1}{B} = A \cdot [\frac{1}{\overline{B}}; \frac{1}{\underline{B}}] \text{ if } 0 \notin B \text{ (extended interval division if } 0 \in B)$$

Contraction I: Method

- Choose a constraint $c \in C$ and a variable x appearing in c .
We call such a pair (c, x) a **contraction candidate (CC)**.
- Bring c to a form $x \sim e$, $\sim \in \{<, \leq, =, \geq, >\}$, where e does not contain x .
(Note: due to preprocessing, if c is non-linear then it is of the form $h - m = 0$ with h a variable and m a monomial.)
- Replace all variables in e by their current bounds.
- Apply interval arithmetic to evaluate the right-hand-side (e with the variables substituted by their bounds) to a union of intervals.
- Make a case distinction for each interval B in that union.
- For each case, derive from the current bound A for x and the computed bound B for e a new bound on x , depending on the type of \sim , as follows:

$$\begin{array}{ll} x < e & \text{if } \underline{A} \geq \overline{B} \text{ then } \emptyset \text{ else } [\underline{A}, \min\{\overline{A}, \overline{B}\}] \\ x \leq e & [\underline{A}, \min\{\overline{A}, \overline{B}\}] \\ x = e & [\max\{\underline{A}, \underline{B}\}, \min\{\overline{A}, \overline{B}\}] \\ x \geq e & [\max\{\underline{A}, \underline{B}\}, \overline{A}] \\ x > e & \text{if } \overline{A} \leq \underline{B} \text{ then } \emptyset \text{ else } [\max\{\underline{A}, \underline{B}\}, \overline{A}] \end{array}$$

Contraction II: Preprocessing

- This second method is called the **interval Newton method**.
- Also this second propagation method needs some lightweight **preprocessing**:
 - Transform each constraint $e_1 \sim e_2$ in C to $e_1 - e_2 \sim 0$.
 - For each **inequation** $p \sim 0$ with $\sim \in \{<, \leq, \geq, >\}$ in C , replace p by a fresh variable h , add an equation $h - p = 0$ to C , and initialize the bounds of h to the interval we get when we substitute the variable bounds in p and evaluate the result using interval arithmetic (note: the result will always be a single interval because there is no division or square root in p).
- After this preprocessing, the constraint set contains equations $p = 0$ stating that a polynomial equals to zero, and inequations of the form $x \sim 0$ with x a variable and $\sim \in \{<, \leq, \geq, >\}$.
- Assume in the following a constraint $c \in C$ and a variable x in c as a contraction candidate.

Contraction II: Method

If the constraint c is an **inequation** then it has the form $x \sim 0$ (where x is a variable). Contraction (assuming that the current interval for x is A):

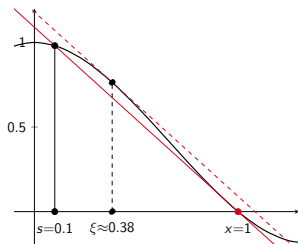
$$\begin{array}{ll} x < 0 & \text{if } \underline{A} \geq 0 \text{ then } \emptyset \text{ else } [\underline{A}, \min\{\bar{A}, 0\}] \\ x \leq 0 & [\underline{A}, \min\{\bar{A}, 0\}] \\ x \geq 0 & [\max\{\underline{A}, 0\}, \bar{A}] \\ x > 0 & \text{if } \bar{A} \leq 0 \text{ then } \emptyset \text{ else } [\max\{\underline{A}, 0\}, \bar{A}] \end{array}$$

Contraction II: Method

Assume now that the constraint c is an equation $f(x) = 0$ (with f being a polynomial).

Interval Newton method for the **univariate** case:

- Input:
 - interval A
 - univariate polynomial constraint $f(x) = 0$
 - sample point $s \in A$
- Output: contracted interval $A = s - \frac{f(s)}{f'(A)}$ (where $f'(x)$ is the first derivative of $f(x)$)



Componentwise multivariate interval Newton:

■ Input:

- interval $A = A_1 \times \dots \times A_n$
- multivariate polynomial constraint $f(x_1, \dots, x_n) = 0$
- sample point $s = (s_1, \dots, s_n) \in A$
- variable x_j

- Output: contracted interval $A = s - \frac{f(A_1, \dots, A_{j-1}, s_j, A_{j+1}, \dots, A_n)}{\frac{\partial f}{\partial x_j}(A_1, \dots, A_n)}$

■ Relative contraction

$$gain_{rel} = \frac{D_{old} - D_{new}}{D_{old}} = 1 - \frac{D_{new}}{D_{old}}$$

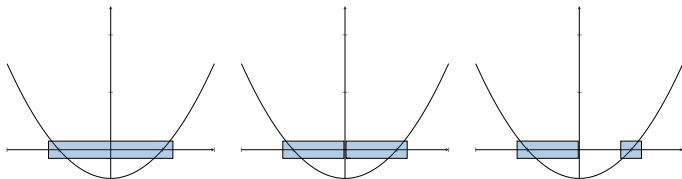
is in general **not predictable**.

■ Heuristics:

- assign a weight $W_k^{(ij)} \in [0; 1]$ to each CC
- select the next contraction candidate with the highest weight
- CCs with a weight less than some threshold ε are not considered for contraction
- let $r_{k+1}^{(ij)} \in [0; 1]$ be the achieved relative contraction
- update weight:

$$W_{k+1}^{(ij)} = W_k^{(ij)} + \alpha(r_{k+1}^{(ij)} - W_k^{(ij)})$$

When the weight of all CCs is below the threshold we do not make progress
→ split the box.



ICP is not well-suited for linear problems (slow convergence).

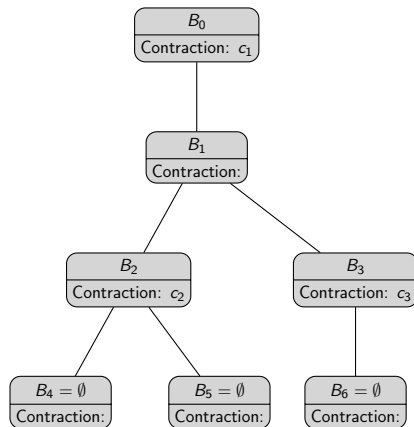
Make use of linear solvers (e.g. simplex) for linear constraints:

- Pre-process to separate linear and nonlinear constraints
- Use nonlinear constraints for contraction
- Validate resulting boxes against linear feasible region
- Box infeasible → add violated linear constraint for contraction

Incrementality and Explanations

We store the search history in a tree-structure. Each node stores information about one loop iteration:

- the box chosen and
- the constraint used for contraction if any.

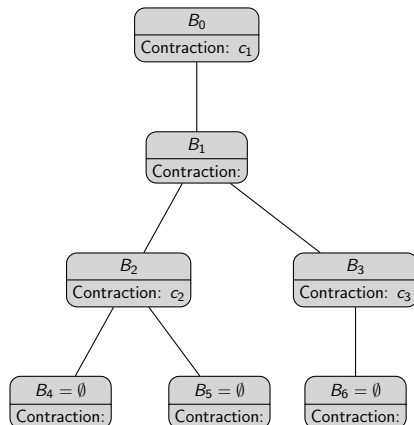


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Incrementality: Extend the tree.



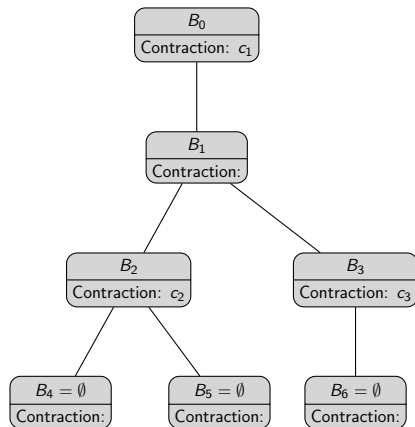
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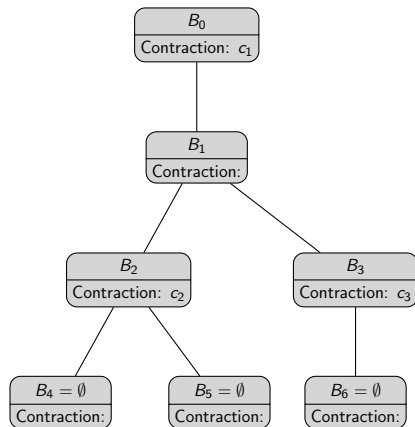
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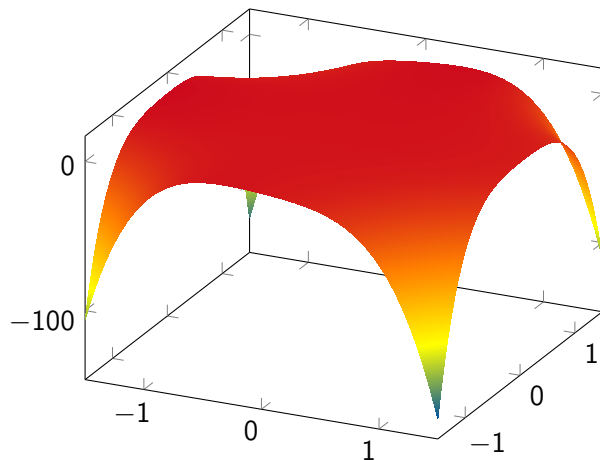
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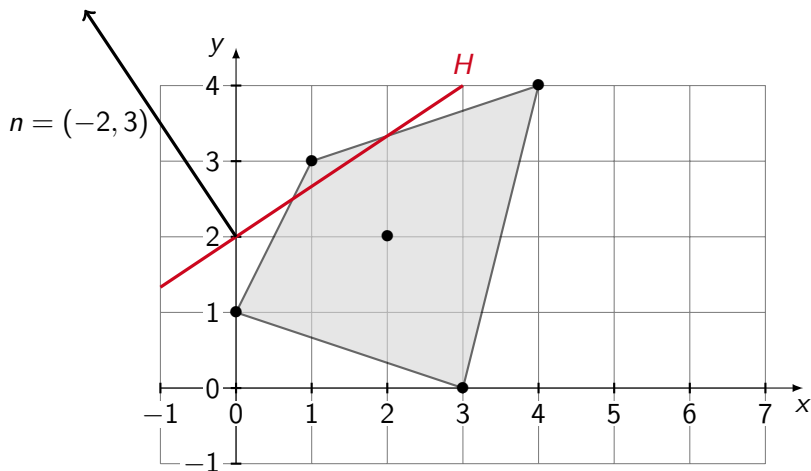


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$$f(x, y) = y + 2xy^3 - 3x^2y^2 - x^3 - 4x^4y^4$$

Hyperplanes separating vertices of the Newton polytope



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- **Virtual substitution method:** **quantifier elimination** procedure for real arithmetic formulas
- **Here:** only **existential** quantification, no free variables

$$\exists x_1 \dots \exists x_n. \varphi_n \equiv \exists x_1 \dots \exists x_{n-1}. \varphi_{n-1}$$

- **Restriction:** applicable only to variables that appear **at most quadratic** in the formula
- **Basic idea:** use **solution equation** to construct a finite set $T \subset \mathbb{R}$ of **test candidates** for x_n , and use **virtual substitution** to check whether one of the test candidates satisfies the formula:

$$\exists x_1 \dots \exists x_n. \varphi_n \equiv \exists x_1 \dots \exists x_{n-1}. \bigvee_{t \in T} \varphi_n[t // x_n].$$

Construction of the set of test candidates T

Given: A constraint $p \sim 0$, $p = ax^2 + bx + c$, $\sim \in \{=, <, >, \leq, \geq, \neq\}$.

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Linear in x :	$x_0 = -\frac{c}{b}$, if $a = 0 \wedge b \neq 0$
Quadratic in x :	$x_1 = \frac{-b + \sqrt{b^2 - 4ac}}{2a}$, if $a \neq 0 \wedge b^2 - 4ac \geq 0$
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The **finite endpoints** of possible solution intervals of $p \sim 0$ are the zeros of p (as the sign of p is invariant between its zeros).

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The **finite endpoints** of possible solution intervals of $p \sim 0$ are the zeros of p (as the sign of p is invariant between its zeros).

Note: If p has no zeros then the possible solution interval is $(-\infty, \infty)$.

Thus the **possible solution intervals** for x in $p \sim 0$ are:

constraints	possible solution intervals ($0 \leq i, j \leq 2, i \neq j$)			
$p = 0$	$[x_i, x_i]$			$(-\infty, \infty)$
$p < 0$ $p > 0$ $p \neq 0$	$(-\infty, x_i)$	(x_i, x_j)	(x_i, ∞)	$(-\infty, \infty)$
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$p \leq 0$ $p \geq 0$	$(-\infty, x_i]$		$[x_i, x_j]$ $[x_i, \infty)$	$(-\infty, \infty)$

We need to pick one **test candidate** from each of those intervals.

Note: In general, the zeros x_i are not constants but might contain other variables. Especially, it implies that they cannot be ordered by their values.

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We need to pick one **test candidate** from each of those intervals.

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As test candidates we take **the smallest value** from each of those possible solution intervals:

Construction of the set of test candidates T

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As test candidates we take **the smallest value** from each of those possible solution intervals:

■ $p = 0, p \leq 0, p \geq 0$

1 Zeros of the polynomial p

2 $-\infty$ ($:=$ sufficiently small value)

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- $p = 0, p \leq 0, p \geq 0$

- 1 Zeros of the polynomial p
- 2 $-\infty$ (:= sufficiently small value)

- $p < 0, p > 0, p \neq 0$

- 1 Zeros of the polynomial p plus an infinitesimal ϵ
- 2 $-\infty$

Construction of the set of test candidates T

Example: $\exists y \exists x : (y = 0 \vee y^2 + 1 < 0) \wedge x - 3 \leq 0 \wedge xy + 1 < 0$

eliminate x
 \equiv

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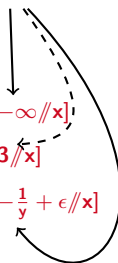
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 \equiv

$$\begin{aligned} \exists y : & \quad ((y = 0 \vee y^2 + 1 < 0) \wedge x - 3 \leq 0 \wedge xy + 1 < 0) [-\infty // x] \\ \vee & \quad ((y = 0 \vee y^2 + 1 < 0) \wedge x - 3 \leq 0 \wedge xy + 1 < 0) [3 // x] \\ \vee (\mathbf{y \neq 0} \wedge & \quad ((y = 0 \vee y^2 + 1 < 0) \wedge x - 3 \leq 0 \wedge xy + 1 < 0) [-\frac{1}{y} + \epsilon // x]) \end{aligned}$$

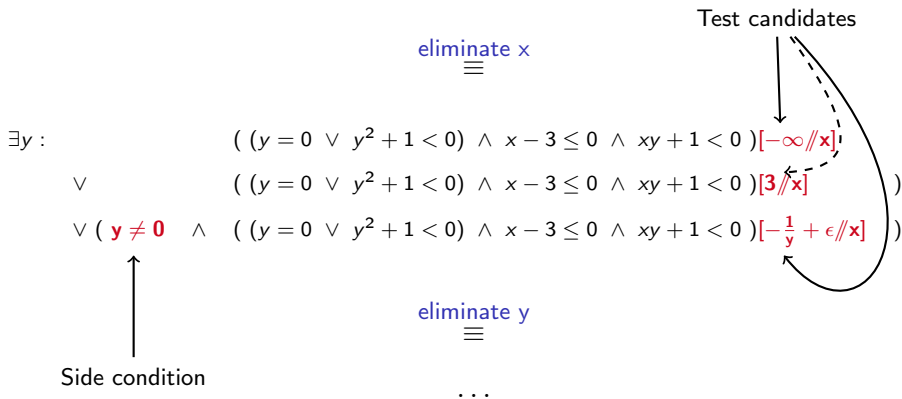
Side condition

Test candidates



Construction of the set of test candidates T

Example: $\exists y \exists x : (y = 0 \vee y^2 + 1 < 0) \wedge x - 3 \leq 0 \wedge xy + 1 < 0$



Virtual substitution of a variable by a test candidate

Example: $(g(x) = 0)[\frac{q+r\sqrt{t}}{s} // x]$

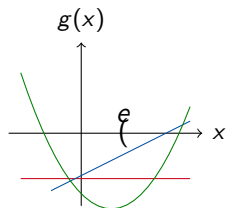
- 1 Substitute $\frac{q+r\sqrt{t}}{s}$ for x in $g(x) = 0$ in the common way.
- 2 Transform the result to $\frac{\hat{q}+\hat{r}\sqrt{t}}{\hat{s}} = 0$ where \hat{q} , \hat{r} , and \hat{s} are polynomials.

3
$$\frac{\hat{q}+\hat{r}\sqrt{t}}{\hat{s}} = 0 \quad \Leftrightarrow \quad \hat{q} + \hat{r}\sqrt{t} = 0$$
$$\Leftrightarrow \hat{q}\hat{r} \leq 0 \wedge \|\hat{q}\| = \|\hat{r}\sqrt{t}\| \quad \Leftrightarrow \quad \hat{q}\hat{r} \leq 0 \wedge \hat{q}^2 - \hat{r}^2t = 0$$

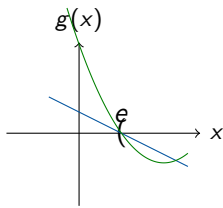
Result: $(g(x) = 0)[\frac{q+r\sqrt{t}}{s} // x] = (\hat{q}\hat{r} \leq 0 \wedge \hat{q}^2 - \hat{r}^2t = 0)$

Virtual substitution of a variable by a test candidate

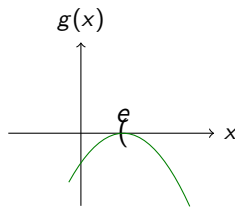
Example: $(g(x) < 0)[e + \epsilon // x]$



Case 1



Case 2



Case 3

Result:

$$\underbrace{g[e//x] < 0}_{\text{Case 1}} \vee \underbrace{g[e//x] = 0 \wedge g'[e//x] < 0}_{\text{Case 2}} \vee \underbrace{g[e//x] = 0 \wedge g'[e//x] = 0 \wedge g''[e//x] < 0}_{\text{Case 3}}$$

- 1 Interval constraint propagation
- 2 Subtropical satisfiability
- 3 Virtual substitution
- 4 Cylindrical algebraic decomposition

Cylindrical algebraic decomposition: Idea

- Assume a set P of polynomials in n variables together with a sign condition for each polynomial in P .
- The **cylindrical algebraic decomposition (CAD)** method produces a decomposition of \mathbb{R}^n into a finite number of P -sign-invariant regions (CAD cells).
- Take an arbitrary element (sample point) from each of the CAD cells.
- If all sign conditions are satisfied for at least one sample point then the problem is satisfiable.
- Otherwise the problem is unsatisfiable.

Let $R \subseteq \mathbb{R}^{n-1}$ be a region and $P = \{p_1, \dots, p_m\} \subset \mathbb{Z}[x_1, \dots, x_n]$, where $m \geq 1$ and $n \geq 2$.

Intuition: If P is delineable on R then the real roots of P vary continuously over R , while maintaining their number and order.

Definition

P is **delineable** on R if for $1 \leq i, j \leq m$ with $i \neq j$ and for all $a \in R$:

- 1 the number of roots of $p_i(a)$ is constant,
- 2 the number of different roots of $p_i(a)$ is constant,
- 3 the number of common roots of $p_i(a)$ and $p_j(a)$ is constant.

Cylindrical algebraic decomposition

Let $P = (p_1, \dots, p_m) \in \mathbb{Z}[x_1, \dots, x_n]^m$ and $\mathcal{C} \subseteq 2^{\mathbb{R}^n}$ finite with $m, n \geq 1$.

Definition

\mathcal{C} is called **cylindrical algebraic decomposition (CAD)** of \mathbb{R}^n for P if the following holds:

- 1 $\bigcup \mathcal{C} = \mathbb{R}^n$,
- 2 $C \cap C' = \emptyset$ for all $C, C' \in \mathcal{C}$ with $C \neq C'$,
- 3 If $n = 1$, then every $C \in \mathcal{C}$ is a P -sign invariant region.
- 4 If $n > 1$ then there exists a CAD \mathcal{C}' of \mathbb{R}^{n-1} such that for every $C \in \mathcal{C}$ there is a $C' \in \mathcal{C}'$ such that the projection of C to the first $n - 1$ dimensions is C' .

An element $C \in \mathcal{C}$ is called a **cell**.

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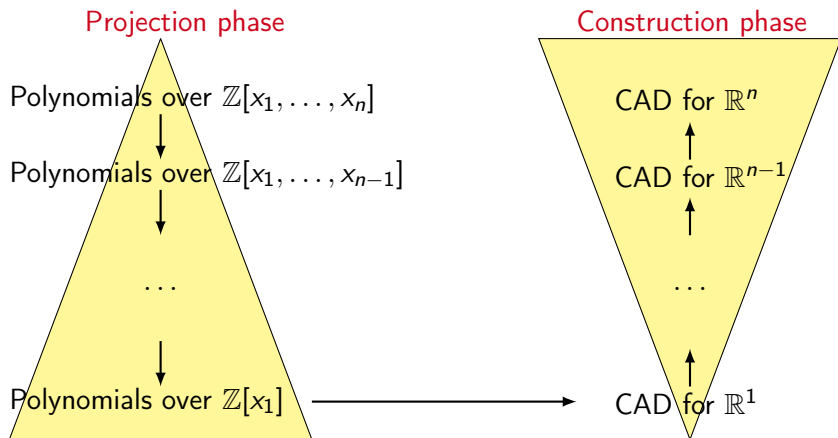
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An element $C \in \mathcal{C}$ is called a **cell**.

Remark

One **sample point** per cell is sufficient in order to represent a CAD.

A CAD for a set of polynomials from $\mathbb{Z}[x_1, \dots, x_n]$ splits \mathbb{R}^n into **sign-invariant** regions.



Let $P = \{p_1, \dots, p_m\} \in \mathbb{Z}[x_1, \dots, x_n]$ where $n \geq 2$ and $m \geq 1$.

Definition

A mapping

$$\text{proj} : \mathbb{Z}[x_1, \dots, x_n] \longrightarrow \mathbb{Z}[x_1, \dots, x_{n-1}]$$

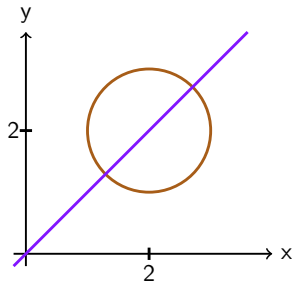
is called a **CAD-Projection**, if any region $R \subseteq \mathbb{R}^{n-1}$ is $\text{proj}(P)$ -sign invariant *iff* R is P -delineable.

Remark

- Usually, $|\text{proj}(P)| = |P|^2$. Thus, projecting recursively up to the univariate case is in $\mathcal{O}(|P|^{2^{n-1}})$.

Example: CAD projection

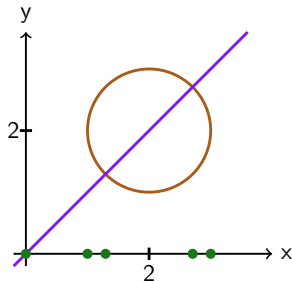
$$P = \begin{pmatrix} (x-2)^2 + \\ (y-2)^2 - 1, \\ x - y \end{pmatrix}$$



$$\text{proj}(P) = \left\{ \begin{aligned} &x^2 - 4x + 3, \\ &-4x + x^2 + \frac{7}{2}, \\ &x^4 - 8x^3 + 30x^2 - 56x + 49, \\ &x^2 - 4x + 7, \\ &x \end{aligned} \right\}$$

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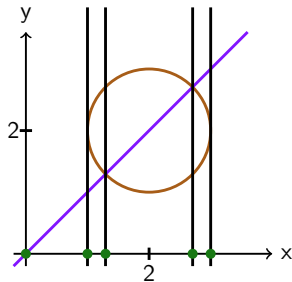


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$$\left\{ \begin{array}{l} \{1, 3\} \\ \{2 - \frac{\sqrt{2}}{2}, 2 + \frac{\sqrt{2}}{2}\} \\ \{ \} \\ \{ \} \\ \{0\} \end{array} \right\}$$

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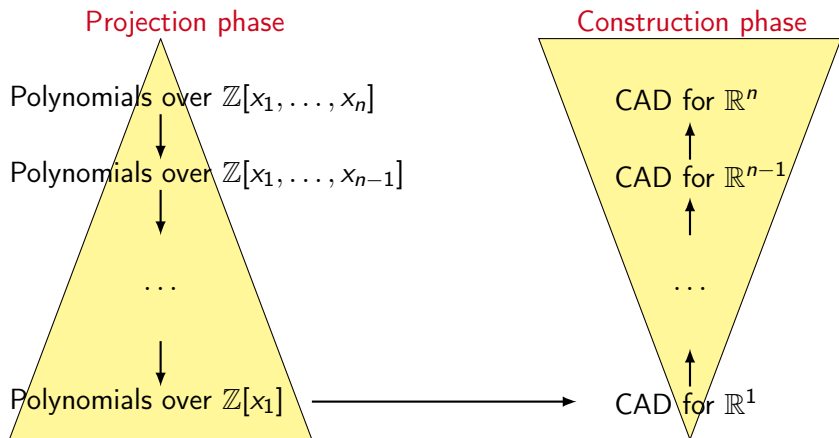
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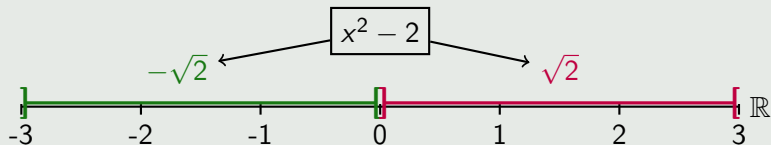
Representing real roots (real algebraic numbers)

Interval representation

$$\left(\underbrace{p,}_{\in \mathbb{Z}[x]} \quad \underbrace{(l, r)} \right)$$

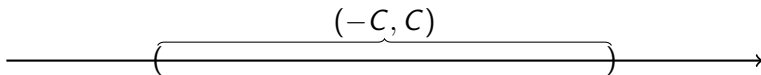
exactly one real root of p in (l, r)

Example



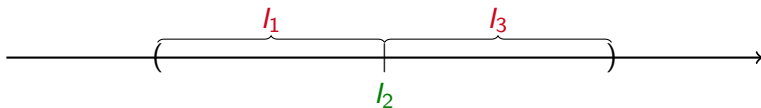
Real root isolation in \mathbb{R}

- Assume a set $P = \{p_1 \sim_1 0, \dots, p_k \sim_k\}$ of **univariate** polynomial constraints with $p_i \in \mathbb{Z}[x]$ and $\sim_i \in \{<, \leq, =, \neq, \geq, >\}$.
- **Cauchy bound** \Rightarrow Interval $(-C, C)$ containing all real roots of p_1, \dots, p_k .
- **Sturm sequence** \Rightarrow count the real roots of each p_i in an interval.
- Split C until each sub-interval I contains at most one real root.



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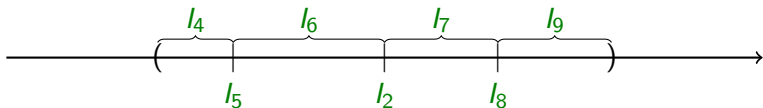
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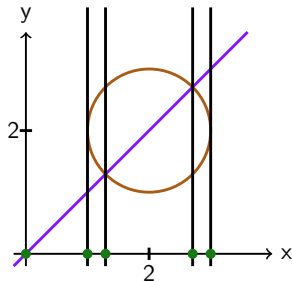


CAD for \mathbb{R} with respect to p_1, \dots, p_k :

$[(p_i, l_j), (p_i, l_j)]$ for each l_j containing a real root of a p_i and open intervals between them.

Example: CAD projection

$$P = \begin{pmatrix} (x-2)^2 + \\ (y-2)^2 - 1, \\ x - y \end{pmatrix}$$

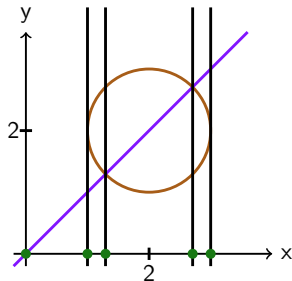


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1-dimensional CAD (dimension x):

$$\underbrace{\left(-\infty, 1 \right), \{1\}, \left(1, 2 - \frac{\sqrt{2}}{2} \right), \left\{ 2 - \frac{\sqrt{2}}{2} \right\}, \left(2 - \frac{\sqrt{2}}{2}, 2 + \frac{\sqrt{2}}{2} \right), \left\{ 2 + \frac{\sqrt{2}}{2} \right\}, \left(2 + \frac{\sqrt{2}}{2}, 3 \right), \{3\}, (3, \infty) \right)}_{\text{samples} \rightarrow Z_1}$$

The CAD sample construction in a nutshell

$$\rightarrow P_n \subseteq \mathbb{Z}[x_1, \dots, x_n]$$

eliminate

x_n



$$P_{n-1} \subseteq \mathbb{Z}[x_1, \dots, x_{n-1}]$$

eliminate

x_{n-1}



\vdots

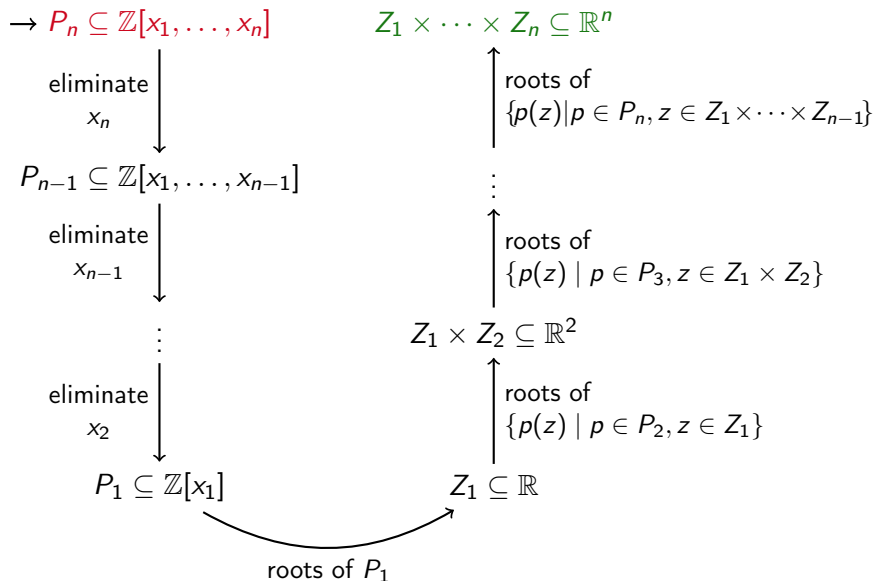
eliminate

x_2



$$P_1 \subseteq \mathbb{Z}[x_1]$$

The CAD sample construction in a nutshell



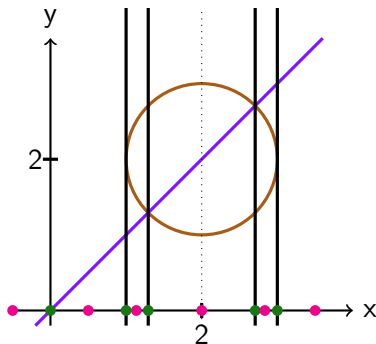
Example: CAD sample construction

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Samples for $\text{proj}(P)$:

$$\{0, 1, 2 - \frac{\sqrt{2}}{2}, 2 + \frac{\sqrt{2}}{2}, 3\}$$

$$\{-0.5, 0.5, 1.135, \\ 2, 2.835, 3.5\}$$



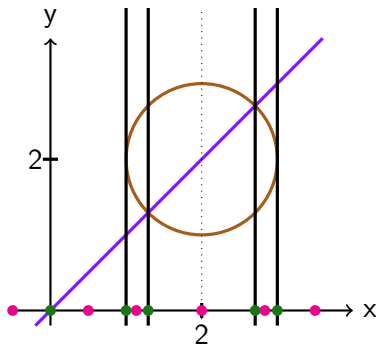
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Example sample constructions

- $(2-2)^2 + (y-2)^2 - 1$ has zeros 1 and 3.
- $2 - y$ has zero 2.
- Two-dimensional samples are $(2, s)$, one s taken from the each of $(-\infty, 1)$, $\{1\}$, $(1, 2)$, $\{2\}$, $(2, 3)$, $\{3\}$, $(3, \infty)$.