

# A Complete Algorithm for Optimization Modulo Nonlinear Real Arithmetic

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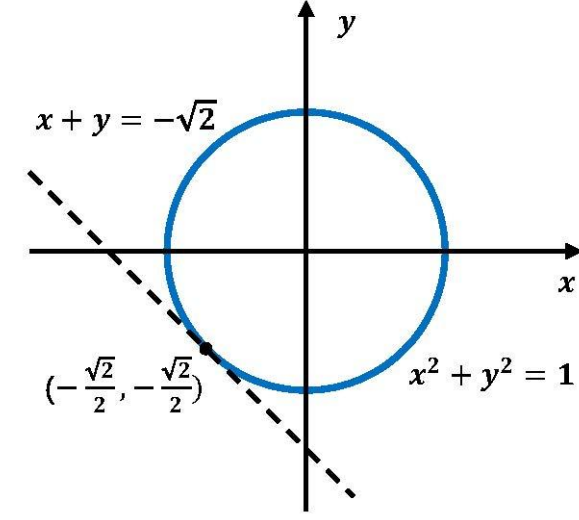
## Motivation

### Example

Consider an OMT(NRA) formula,

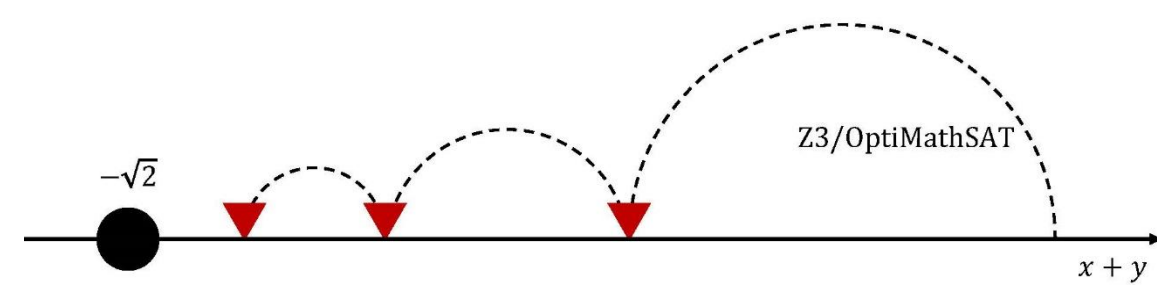
$$\min x + y, \text{ s. t. } x^2 + y^2 = 1,$$

the objective  $x + y$  has a minimum value of  $-\sqrt{2}$ .



Using 2 hours:

- Z3: timeout;
- OptiMathSAT: timeout.



## Contributions

- **Development of the first complete OMT solver for NRA:** Present OCAC and CDCL(OCAC), sound and complete algorithms, along with proofs of correctness and termination.

- **Investigation of Variants:** Explore two additional solving algorithm variants, using CAD and a first-order formulation.

- **Evaluation:** Integrated this algorithm into CVC5 and conducted empirical evaluations showing that it outperforms the leading OMT solver, OptiMathSAT.

## Basic Definitions

### Definition 1 (NRA Formula):

$p := x|c|p + p|p \cdot p$ , (polynomial)  
 $\phi := b|p \geq 0|p = 0|\neg\phi|\phi \wedge \phi|\phi \vee \phi$  (formula)  
 $x \in \mathbb{R}, c \in \mathbb{Q}, b \in \mathbb{B}$

### Definition 2 (Cell):

Given  $P = \{p_1, \dots, p_m\} \subseteq \mathbb{Q}[x_1, \dots, x_n]$  and  $s \in \mathbb{R}^n$ , a cell  $C(P, s)$  is a non-empty connected subset of  $\mathbb{R}^n$  that is sign-invariant for  $P$  and contains  $s$ .  
 For all  $i \in \{1, \dots, m\}$ ,

$$\forall s' \in C(P, s), \text{sign}(p_i(s)) = \text{sign}(p_i(s')).$$

### Definition 3 (Cell Interval):

$C(P, s)_{x_i}$  is an interval  $\{l, u\}$  over  $x_i$ , i.e.,  
 $\forall r \in \mathbb{R}. (l < r < u \rightarrow \bar{s} \in C(P, s))$ ,  
 where  $\bar{s} := (s^1, \dots, s^{i-1}, r, s^{i+1}, \dots, s^n)$ ,  
 replacing  $s_i$  with  $r$ .

### Definition 4 (Extended Domain):

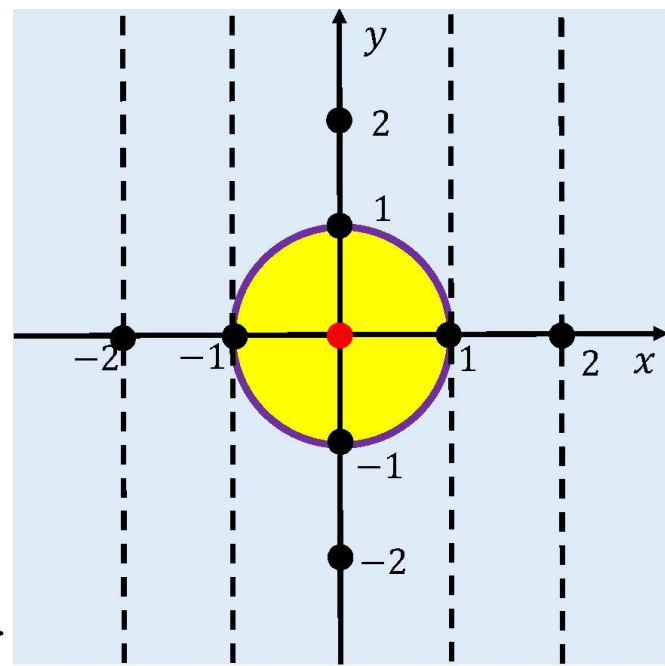
$$\bar{\mathbb{R}} := \mathbb{R} \cup \{\infty\} \cup \{-\infty\}$$

A Cylindrical Algebraic Decomposition (CAD) is a decomposition algorithm for a set of polynomials in  $\mathbb{R}^n$ ,  $n \in \mathbb{N}$  space resulting in a finite number of cells.

Consider  $\{x^2 + y^2 - 1\}$  and  $x < y$ :

- Projects  $y$  and obtains  $\{x^2 - 1\}$ ;
- Isolates roots, i.e.,  $x = -1$  and  $x = 1$ ;
- Divides  $\mathbb{R}$  to 5 segments;
- Samples  $x = \{-2, -1, 0, 1, 2\}$ , resulting in  $\{y^2 + 3\}$ ,  $\{y^2\}$ , and  $\{y^2 - 1\}$ ;
- Isolates roots for  $\{y^2 - 1\}$ , i.e.,  $\{-1, 1\}$ ;
- Divides  $\mathbb{R}$  to 5 segments;

Red point  $(0, 0)$  represents the sign-invariant region  $\{x^2 + y^2 - 1 < 0\}$ .

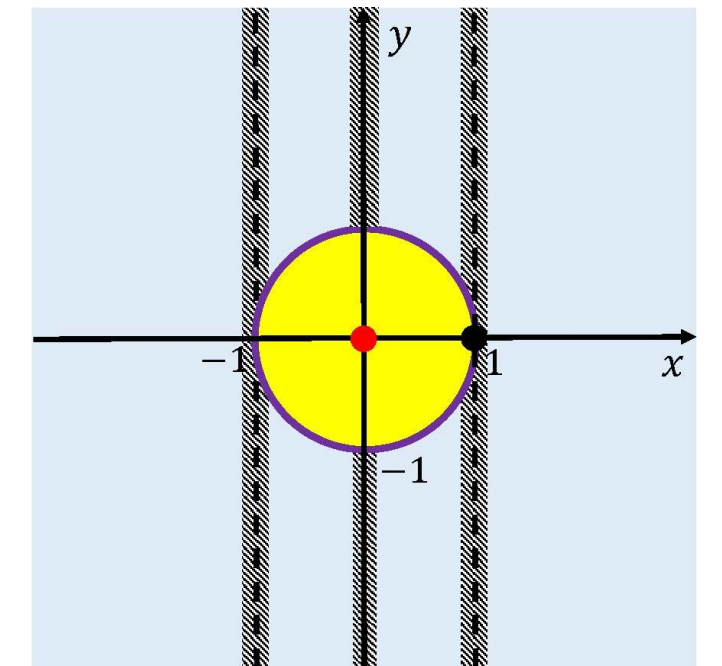


The Cylindrical Algebraic Covering Algorithm (CAC) is a variant of CAD, adjusted to the context of SMT.

Consider  $x^2 + y^2 < 1$  and  $x < y$ :

- Assigns  $x = 1$  and obtains  $y^2 < 0 \equiv \perp$ ;
- Identifies the characterizing set  $\{x^2 - 1\}$ ;
- Excludes unsatisfiable region  $\{-1, 1\} \times \mathbb{R}$ ;
- Assigns  $x = 0$  and obtains  $y^2 < 1$ ;
- Excludes  $\{0\} \times (-\infty, -1] \cup [1, +\infty)$ ;
- Assigns  $y = 0$  and reports  $\top$ .

Finally, it finds a model, i.e., the red point  $(0, 0)$ .



## Generalized Optimization Modulo Theories

### Definition 5 (GOMT problem):

A General Optimization Modulo Theories problem is a tuple  $\mathcal{GO} := (t, <, \phi)$ , where,

- $t$ , a  $\Sigma$ -term of some sort  $\sigma$ , is an objective term to optimize;
- $<$  is a strict partial order definable in  $\mathbb{T}$ , whose defining formula has two free variables, each of sort  $\sigma$ , and
- $\phi$  is a  $\Sigma$ -formula.

Example is a simple example of  $\mathcal{GO}(x + y, <, x^2 + y^2 = 1)$ .

For any GOMT problem  $\mathcal{GO}$  and  $\mathbb{T}$ -interpretations  $\mathbb{I}$  and  $\mathbb{I}'$ , we say that:

- $\mathbb{I}$  is  $\mathcal{GO}$ -consistent if  $\mathbb{I} \models \phi$ ;
- $\mathbb{I}$  dominates  $\mathbb{I}'$  denoted by  $\mathbb{I} <_{\mathcal{GO}} \mathbb{I}'$ , if  $\mathbb{I}$  and  $\mathbb{I}'$  are  $\mathcal{GO}$ -consistent and  $\mathbb{I}^1 < \mathbb{I}'^1$ ;
- $\mathbb{I}$  is a  $\mathcal{GO}$ -solution if  $\mathbb{I}$  is  $\mathcal{GO}$ -consistent and no  $\mathbb{T}$ -interpretation  $\mathcal{GO}$ -dominates  $\mathbb{I}$ .

### Definition 6 (GOMT over NRA):

A  $\mathcal{GO}_{\text{NRA}}$  is a tuple  $\mathcal{GO}_{\text{NRA}} := (t, <, \phi)$ , where,

- $t$  is the polynomial objective term to optimize;
- $<$  is a strict partial order definable in NRA, whose formula defines over  $\bar{\mathbb{R}} \times \bar{\mathbb{R}}$ ,
- $\phi$  is an SMT(NRA) formula.

## Optimization Cylindrical Algebraic Covering

### Algorithm 1 OCAC

**Input:**  $\psi \wedge t = x_t$ : The OMT branch formula, with  $n$  variables.  
**Output:**  $g, v, l$ : A flag that  $\psi \wedge t = x_t$  is satisfiable; The optimum value; The cutting lemma.  
 1:  $\mathbb{I} := \emptyset, g := \perp, v := \text{None}$   
 2: **while**  $\bigcup_{\mathbb{I} \in \mathbb{I}} l \neq \mathbb{R}$  **do**  
 3:  $s_t := \text{Sample\_Objective\_Value}(\mathbb{I})$   
 4:  $(T, O) := \text{Solve\_Internal}(\psi \wedge t = x_t \wedge x_t = s_t)$   
 5: **if**  $T = \top$  **then**  
 6:  $g := T, v := \text{Analyze\_Cell}(O)$   
 7:  $O := O \cup [s_t, +\infty)$   
 8: **end if**  
 9:  $\mathbb{I} := \mathbb{I} \cup \{O\}$   
 10: **end while**  
 11: **return**  $(g, v, \text{Lemma}(v))$

### Definition 7 (GOMT in OCAC):

A  $\mathcal{GO}_{\text{OCAC}}$  is a tuple  $\mathcal{GO}_{\text{OCAC}} := (x_t, <_{\text{Cell}}, \psi)$ , where,

- $x_t$ , a variable with  $x_t = t$ , is the objective variable where  $t$  is the polynomial objective term to optimize;
- $<_{\text{Cell}}$  is a strict partial order definable in NRA, whose formula defines over  $\bar{\mathbb{R}} \times \bar{\mathbb{R}}$ , and  $x_t^1 <_{\text{Cell}} x_t^1$  is equivalent to:  
 $\mathbb{I}(C(P(\psi), \mathbb{I}_{\mathbb{R}})_{x_t}) < \mathbb{I}(C(P(\psi), \mathbb{I}'_{\mathbb{R}})_{x_t})$ ,
- $\psi$  is a conjunction of polynomial atoms.

## Correctness and Termination

### Theorem (Correctness of CAC):

Let  $\psi$  be a conjunction of polynomial atoms with  $x_1, \dots, x_n$ ,  $S = \{s_1, \dots, s_m\} \subseteq \mathbb{R}$ ,  $P^1, \dots, P^m \subseteq \mathbb{Q}[x_1, \dots, x_i]$  and  $s \in \mathbb{R}^{i-1}$  for  $1 < i \leq n$ . If  $\{s\} \times \mathbb{R} \subseteq \bigcup_{j=1}^m C(P_j, (s, s_j))$  and for  $1 \leq j \leq m$ ,  $C(P_j, (s, s_j))$  is unsatisfiable for  $\psi$ , then  $C(\text{proj}_{\text{cov}}, (P^1, \dots, P^m, s, S), s)$  is unsatisfiable.

### Theorem (Termination of OCAC):

Given an OMT branch formula  $\psi \wedge t = x_t$ , OCAC terminates.

### Theorem (Satisfiable Interval):

Given an OMT branch formula  $\psi \wedge t = x_t$ ,  $P$  denotes the set of polynomials in  $\psi \wedge t = x_t$ . If  $\psi \wedge t = x_t$  is satisfiable, that is, there exists a complete assignment  $s = (s_t, s_1, \dots, s_n)$  that satisfies  $\psi \wedge t = x_t$ , then  $\forall y_o \in C(\text{proj}_{\text{dec}}(P), s)_{x_t}, \psi \wedge t = y_o$  is satisfiable.

### Theorem (Correctness of OCAC):

Given an OMT branch formula  $\psi \wedge t = x_t$ , if  $\psi \wedge t = x_t$  is unsatisfiable, OCAC returns UNSAT; otherwise, OCAC can find the optimum.

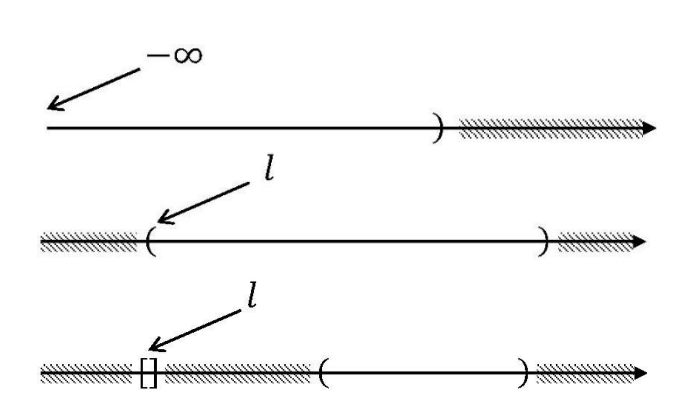
### Corollary

Given a conjunction of polynomial atoms  $\psi$ , let  $l_o$  of  $x_t$  represent the leftmost satisfiable interval for minimization, which can be characterized by three cases:

- $l_o = (-\infty, u)$  implies  $\min(x_t) = -\infty$ ;
- $l_o = (l, u)$  implies  $\min(x_t) = l + \epsilon$ ;
- $l_o = [l, \infty)$  implies  $\min(x_t) = l$ .

### Theorem (Termination and Correctness):

For a given instance  $\mathcal{GO}_{\text{NRA}} := (t, <, \phi)$ , the CDCL(OCAC) algorithm is guaranteed to terminate and produce correct results.



## Experiments

### • CAD-Based Variant.

Using CAD to precompute the candidate intervals and select representatives from the leftmost interval until finding the optimal.

### • First-Order Formulation.

Converting the optimization problem into the first-order formula and solving it.

### • OptiMathSAT.

The leading OMT solver using linear or binary search.

### • Benchmarks:

- Generated from QF\_NRA benchmarks of SMT-LIB.

### • For satisfiable instances:

- Randomly select from declared variables;
- Add minimization objectives:  
 $x, x + y, x^2 + y^2, xy, \text{ or } xy + z$ ;
- Randomly select 10000 instances.

- For unsatisfiable instances, use original benchmarks, a total of 5532 instances.

- Timeout: 1200 s.
- For FOL baseline: select the best results from CVC5, dReal, yicesQS, and Z3.

	#(RAN)	#(RAN + ε)	#(Q)	#(Q + ε)	#(∞)	#SAT	#UNSAT
CDCL(CAD)	246	551	802	2990	1129	5718	4568
FOL	304	610	1101	3545	1165	6725	4392
OptiMathSAT(Bin)	0	0	943	1870	353	3166	5040
OptiMathSAT(Lin)	0	0	928	1819	336	3083	5040
CDCL(OCAC) (Ours)	369	981	1084	4248	1250	7932	5019

Table 1: Performance on the number of solved instances, including 10000 satisfiable and 5532 unsatisfiable ones.

Figure 1: Performance on satisfiable instances over time.

