

Handling Polynomial and Transcendental Functions in SMT via Unconstrained Optimisation and Topological Degree Test

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① Introduction

② State-of-the-art

③ Satisfiability of sets of constraints

④ From set of constraints to formulas

⑤ Experiments

⑥ Conclusion

Problem definition

Context: Satisfiability Modulo Theories (SMT)

Theories of interest:

- \mathcal{NRA} : Quantifier-free theory of non linear arithmetic over the reals
- \mathcal{NTA} : \mathcal{NRA} augmented with exponential and trigonometric functions

Example of formula in \mathcal{NTA} :

$$(b_1 \rightarrow (\sin(y)^2 = x^3 - 2x^2y + 1 \wedge y > 0.5)) \wedge \\ (\neg b_1 \rightarrow (b_2 \rightarrow (y = \exp(0.1) \vee x = 0.1)))$$

where b_1, b_2 are Boolean variables and x, y are real-valued individual variables

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Focus on: proving satisfiability for \mathcal{NTA} formulas

Strategy

- 1 Translate a satisfiability problem into a numerical minimization one (*Logic-To-Optimisation*)
- 2 Use unconstrained optimisation to generate a set of candidate approximate solutions
- 3 Given a candidate approximate solution, restrict the original formula to a sub-formula in a specific form (variables are bounded, the number of equations is equal to the number of variables, ...)
- 4 Apply a procedure based on the topological degree test and interval arithmetic to witness the existence of a solution for the sub-formula

① Introduction

② State-of-the-art

Proving satisfiability in SMT(NTA)

Solving systems of equations with the Topological degree

③ Satisfiability of sets of constraints

④ From set of constraints to formulas

⑤ Experiments

⑥ Conclusion

① Introduction

② State-of-the-art

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Solving systems of equations with the Topological degree

③ Satisfiability of sets of constraints

④ From set of constraints to formulas

⑤ Experiments

⑥ Conclusion

Proving satisfiability in SMT(NTA)

\mathcal{NTA} has been proved undecidable by Richardson (1969).

Existing methods:

- Interval Constrain Propagation (ICP): based on interval arithmetic, can prove "sat" if it finds a box in which every point is a solution
- Incremental Linearization: starts from an abstract model and tries to check whether the formula is satisfiable under all possible interpretations of the transcendental functions involved
- δ -satisfiability: based on ICP, tries to prove that there exists a δ -perturbation of the original formula that is satisfiable

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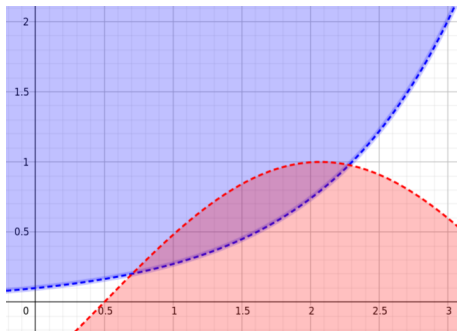


Figure 1: Graph of $(\exp(x) < 10y) \wedge (y < \sin(x - 1/2))$

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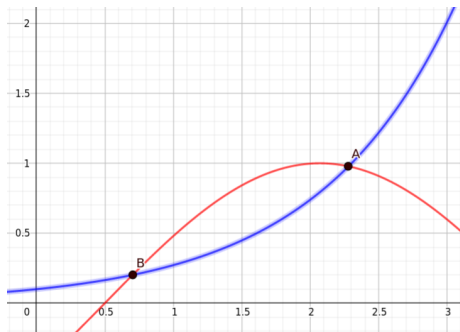


Figure 2: Graph of $(\exp(x) = 10y) \wedge (y = \sin(x - 1/2))$

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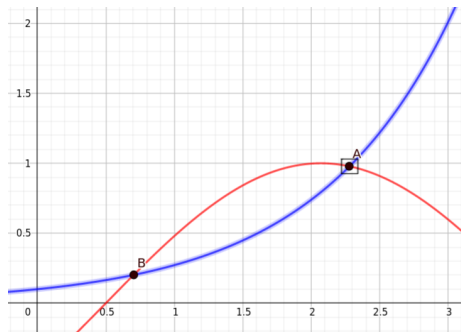


Figure 3: Graph of $(\exp(x) = 10y) \wedge (y = \sin(x - 1/2))$

① Introduction

② State-of-the-art

Proving satisfiability in SMT(NTA)

Solving systems of equations with the Topological degree

③ Satisfiability of sets of constraints

④ From set of constraints to formulas

⑤ Experiments

⑥ Conclusion

Topological degree test

Theorem (Bolzano)

If a continuous function $f : [a, b] \subseteq \mathbb{R} \rightarrow \mathbb{R}$ is s.t. $f(a) < 0$ and $f(b) > 0$ (or vice-versa), then f has a zero in $[a, b]$.

Proof.

Corollary of the Intermediate Value Theorem



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Corollary of the Intermediate Value Theorem



Given a continuous function $f : B \subseteq \mathbb{R}^n \rightarrow \mathbb{R}^n$, the *topological degree* $\deg(f, B)$ is an integer that can be defined in several (very technical) ways.

Property (Topological degree test)

If $\deg(f, B) \neq 0$, then the equation $f = 0$ has a solution in B .

The topological degree is computable if $0 \notin f(\partial B)$ (tool: TopDeg)

Limitations of the topological degree test

The topological degree test is applicable only for conjunctions of constraints in a specific form:

- i there are only equations (no inequalities)
- ii the number of equations is equal to the number of variables
- iii all variables are bounded
- iv (empirical limitation) the bounds on the variables are "sufficiently small" for the practical effectiveness of the test.

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Limitation (i) has been tackled by Franek, Ratschan, and Zgliczynski (2016), who proposed a quasi-decidability (i.e. that always terminates on robust instances) procedure that pairs the topological degree test with interval arithmetic to deal with inequalities.

- ① Introduction
- ② State-of-the-art
- ③ Satisfiability of sets of constraints**
- ④ From set of constraints to formulas
- ⑤ Experiments
- ⑥ Conclusion

Method

Can we use topological degree test, given a general formula with m variables, n equations, k inequalities?

Plan:

- Find candidate solutions through unconstrained optimisation
- Bound the variables in the formula to a box
- If $n < m$, try to instantiate $m - n$ variables
- Apply a procedure that uses interval arithmetic and the topological degree test

First, we work only with conjunctions of constraints. Then, we extend to full Boolean combinations.

Logic-to-Optimisation

Idea: try to find candidate solutions by translating a satisfiability problem into a numerical minimization problem (*Logic-To-Optimisation*)

We define an operator $\mathcal{L2O}$ that maps a formula in m variables to a non-negative real function from \mathbb{R}^m to $\mathbb{R}^{\geq 0}$ as follows:

$$\mathcal{L2O}(f \bowtie 0) : \mathbf{x} \mapsto \begin{cases} 0 & \text{if } \bowtie \in \{\leq, =\} \text{ and } [f](\mathbf{x}) \bowtie 0 \\ \mathcal{L2O}(f \leq 0) & \text{if } \bowtie \text{ is } < \\ [f]^2(\mathbf{x}) & \text{otherwise} \end{cases}$$

$$\mathcal{L2O}(b) \equiv \mathcal{L2O}(-x_b \leq 0)$$

$$\mathcal{L2O}(\neg b) \equiv \mathcal{L2O}(x_b + \epsilon \leq 0)$$

$$\mathcal{L2O}(\phi_1 \wedge \phi_2) \equiv \mathcal{L2O}(\phi_1) + \mathcal{L2O}(\phi_2)$$

$$\mathcal{L2O}(\phi_1 \vee \phi_2) \equiv \mathcal{L2O}(\phi_1) * \mathcal{L2O}(\phi_2),$$

Property: every model of ϕ is a zero of $\mathcal{L2O}(\phi)$ (but not the vice-versa)

Logic-to-Optimisation

Example: $\phi \equiv x^3 - 2x - \exp(x) + 5 = 0$

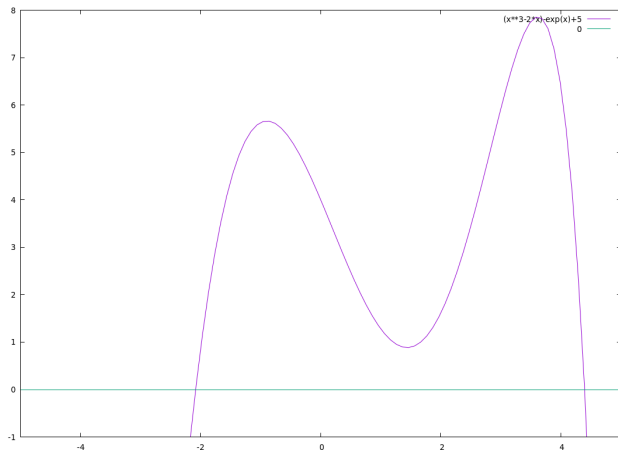


Figure 4: Graph of $x^3 - 2x - \exp(x) + 5$

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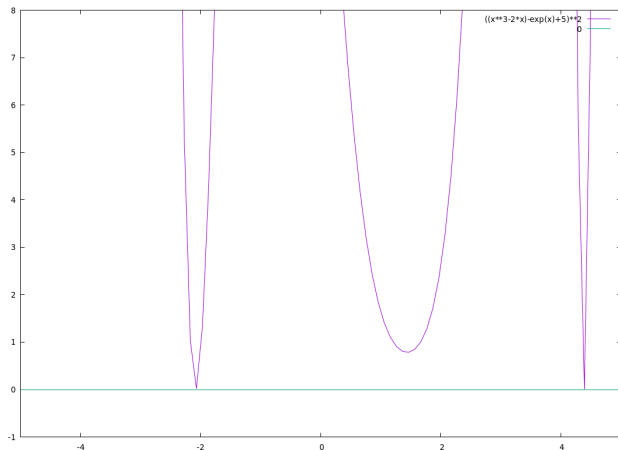


Figure 5: Graph of $\mathcal{L}2\mathcal{O}(\phi)$

Local search using L2O

$\phi \rightarrow \mathcal{L2O} + \text{Unconstr. Opt. (basin-hopping)} \rightarrow \text{several local minima}$
 $\tilde{\mathbf{x}}_1, \dots, \tilde{\mathbf{x}}_k$

Given a local minimum $\tilde{\mathbf{x}}$, we have 3 tactics:

- ① Check if it is a solution (just for \mathcal{NRA})
- ② Reduce to a linear under-approximation in UFLRA by forcing all the multiplications to be linear (just for \mathcal{NRA})
- ③ - Restrict to a sub-formula $\phi|_B$ by imposing that all the variables range over a box B containing $\tilde{\mathbf{x}}$.
 - Instantiate exceeding variables.
 - Apply the quasi-decidability procedure

Quasi-decidability procedure (Idea)

Input: A formula $\phi|_B$ in m variables, n equations, and k inequalities \leq s.t. $n = m$ or $n = 0$.

Loop: We start with a grid that initially contains only B . At each iteration, we split each box of the grid into sub-boxes and:

1. Remove sub-boxes that cannot contain a solution (using interval arithmetic)
2. Try to find a sub-box in which the inequalities are satisfied everywhere and the topological degree test witnesses the existence of a solution for the equations

Output:

- **False** if at some point the grid is emptied
- **True** if at some point test 2 of the Loop succeeds

A general procedure

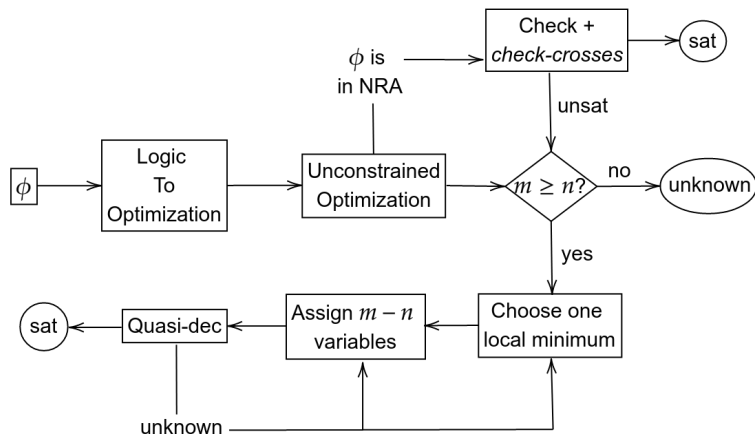


Figure 6: General procedure of UGOT (as in *Unconstrained Global Optimisation and Topological degree*)

- ① Introduction
- ② State-of-the-art
- ③ Satisfiability of sets of constraints
- ④ From set of constraints to formulas
- ⑤ Experiments
- ⑥ Conclusion

From constraints set to formulas

Let ϕ in CNF. Two approaches:

- *Eager* approach:
 - $\mathcal{L2O}(\phi) \rightarrow$ Unconstr. Opt. \rightarrow several local minima $\tilde{\mathbf{x}}_1, \dots, \tilde{\mathbf{x}}_k$
 - Given $\tilde{\mathbf{x}}$, we try to *decide* the disjunctions through the insight given by $\tilde{\mathbf{x}}$
Example: $\tilde{\mathbf{x}} = 1$, we replace in ϕ the clause " $x < 3 \vee x > 5$ " with " $x < 3$ "
- Apply UGOT

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 - Apply UGOT
- *Lazy* approach:
 - Integration of UGOT as a theory solver inside the DPLL(T)-based SMT-solver MATHSAT
 - Pair UGOT (good for proving sat) with Incremental Linearization (good for proving unsat)
 - Naive implementation. For each full Boolean assignment:
 - We call Incremental Linearization at most k times.
 - If it does not find a model or produce a conflict, we call UGOT

- ① Introduction
- ② State-of-the-art
- ③ Satisfiability of sets of constraints
- ④ From set of constraints to formulas
- ⑤ Experiments**
- ⑥ Conclusion

Experiments

	Total	Sturn-MGC	ezsmt	Sturn-MBO	zankl	UltimateAut	Economics-M	meti-tarski	Heizmann	hycomp	kissing	LassoRanker
MATHSAT	3193	0	32	0	32	31	85	2718	3	11	18	263
UGOT _{EAGER}	4388	0	0	1	52	32	68	4042	0	0	36	157
MATHSAT+UGOT	4441	0	32	0	63	27	84	3948	3	13	35	236
RASAT	4285	0	0	0	45	0	0	4225	0	0	15	0
YICES	4946	0	32	0	58	39	91	4369	0	227	10	120
CVC5	5108	0	32	0	63	36	89	4342	2	226	18	300
Z3	5153	2	30	0	72	47	93	4391	6	280	35	198
DREAL	/(5021)	/(9)	/(0)	/(274)	/(153)	/(55)	/(126)	/(4079)	/(51)	/(45)	/(19)	/(210)

Fig. 2. Summary of results for SMT(NRA) sat cases. The results in parenthesis indicate "MAYBE SAT" answers.

	Total	dreal	bmc		Total	dreal	bmc
MATHSAT	94	39	57	MATHSAT	70	21	49
UGOT _{EAGER}	304	255	49	UGOT _{EAGER}	253	203	50
MATHSAT+UGOT	170	70	100	MATHSAT+UGOT	140	35	105
CVC5	94	40	54	CVC5	63	17	46
iSAT3	/	/	/	iSAT3	38 (828)	7 (599)	31 (229)
DREAL	/(578)	/(423)	/(155)	DREAL	/(137)	/(36)	/(101)

Fig. 3. Summary of results for SMT(NTA) sat cases. On the left the original instances; on the right the bounded instances. The results in parenthesis indicate "MAYBE SAT" answers.

Experiments (and what about unsat cases?)

	Total	Sturm-MGC	ezsmt	Sturm-MBO	zankl	UltimateAut	Economics-M	meti-tarski	Heizmann	hycomp	kissing	LassoRanker	hong
MATHSAT	5280	0	2	285	34	7	11	2251	1	2259	0	412	20
MATHSAT+UGOT	5043	0	2	163	32	5	11	2239	0	2253	0	323	20
RASAT	4094	0	0	2	14	0	0	2018	0	1950	0	0	20
YICES	5449	0	2	285	32	12	39	2587	12	2201	0	259	20
CVC5	5645	0	2	285	31	10	35	2581	7	2206	0	468	20
z3	5281	5	2	153	27	12	19	2578	3	2225	0	248	9
DREAL	3889	0	0	131	4	0	3	1784	0	1946	1	0	20

Fig. 4. Summary of results for SMT(NRA) unsat cases.

	Total	dreal	bmc		Total	dreal	bmc
MATHSAT	533	85	448	MATHSAT	524	88	436
MATHSAT+UGOT	522	85	437	MATHSAT+UGOT	521	88	433
CVC5	453	75	378	CVC5	465	85	380
iSAT3	/	/	/	iSAT3	449	63	386
DREAL	468	184	284	DREAL	446	156	290

Fig. 5. Summary of results for SMT(NTA) unsat cases. On the left the original instances; on the right the bounded instances.

- ① Introduction
- ② State-of-the-art
- ③ Satisfiability of sets of constraints
- ④ From set of constraints to formulas
- ⑤ Experiments
- ⑥ Conclusion**

Conclusion and Discussion

Recap:

- Unconstrained optimisation to find candidate solutions
- Quasi-dec procedure with topological degree test to check if a region contains a solution
- Eager approach: candidate solution insight used to decide disjunctions
- Lazy approach: pair the new method with incremental linearization

Open questions:

- How to choose which candidate solution to test? Which variables to instantiate? How to determine the box?
- Extend to unsat (generate lemmas / conflict clauses)?
- Extend to solve problems involving differential equations?

Thank you

Thank you for your attention!

Questions?