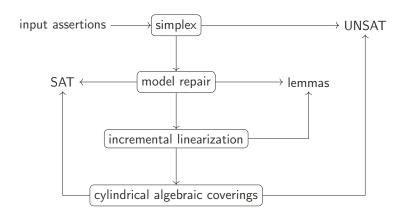
### Solving nonlinear real arithmetic in cvc5

Gereon Kremer

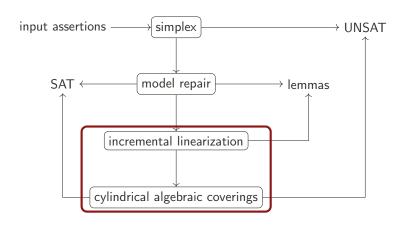
May 9, 2022



## Arithmetic solving in cvc5



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implicitly linearize:  $x \cdot y \leadsto a_{x \cdot y}$ 

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Lemma:  $y = 0 \Rightarrow x \cdot y = 0$ 

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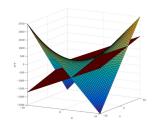
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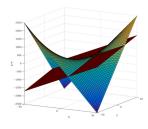
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Lemma: 
$$(x = 3 \land y = 1) \Rightarrow x \cdot y = 3$$

$$(x \le 3 \land y \le 1) \lor (x \ge 3 \land y \ge 1)$$

$$\Leftrightarrow (x \cdot y \ge 1 \cdot x + 3 \cdot y - 3 \cdot 1)$$



[Cimatti et al. 2018]

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#### Incremental linearization – schemas

► Guess partial assignment

$$s_1 \times \cdots \times s_k \times s_{k+1}$$

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► Refute partial assignment using intervals

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► Project covering to lower dimension

$$s_1 \times \cdots \times s_k \times \{(-\infty, a), [a, b], \dots (z, \infty)\} \rightarrow s_1 \times \cdots \times s_{k-1} \times (\alpha, \beta)$$

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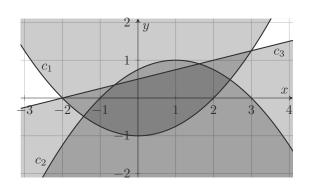
$$s_1 \times \cdots \times s_k \times \{(-\infty, a), [a, b], \dots (z, \infty)\} \to s_1 \times \cdots \times s_{k-1} \times (\alpha, \beta)$$

► Eventually get satisfying assignment or a covering in first dimension

$$s = s_1 \times \cdots \times s_n$$
 or  $s_1 \notin \{(-\infty, a), [a, b], \dots (z, \infty)\}$ 

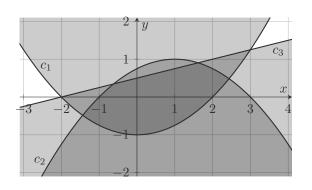
[Ábrahám et al. 2021] [Kremer et al. 2021]

$$c_1: 4 \cdot y < x^2 - 4$$
  $c_2: 4 \cdot y > 4 - (x - 1)^2$   $c_3: 4 \cdot y > x + 2$ 

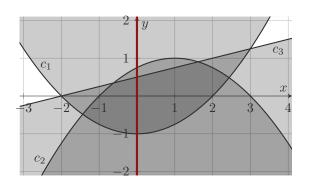


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No constraint for x

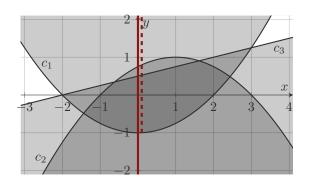


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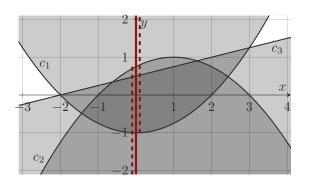
No constraint for x Guess  $x\mapsto 0$ 

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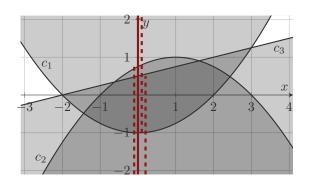
No constraint for xGuess  $x \mapsto 0$  $c_1 \to y \notin (-1, \infty)$ 

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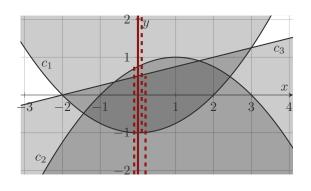
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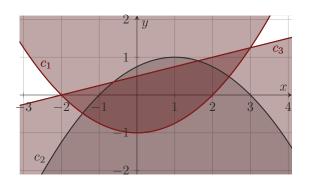
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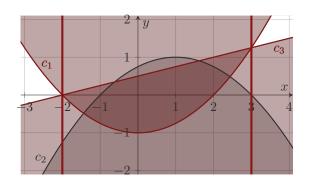
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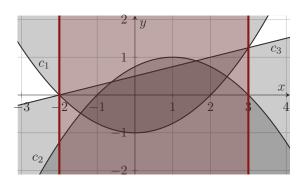
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```
function get_unsat_cover((s_1, \ldots, s_{i-1}))
I := get_unsat_intervals(s)
while \bigcup_{I \subset \mathbb{T}} I \neq \mathbb{R} do
  s_i := \mathtt{sample\_outside}(\mathbb{I})
  if i = n then return (SAT, (s_1, \ldots, s_{i-1}, s_i))
  (f, O) := get\_unsat\_cover((s_1, \ldots, s_{i-1}, s_i))
  if f = SAT then return (SAT, O)
  else if f = \text{UNSAT} then
    R := \text{construct\_characterization}((s_1, \dots, s_{i-1}, s_i), O)
    J := interval\_from\_characterization((s_1, ..., s_{i-1}), s_i, R)
    \mathbb{I} := \mathbb{I} \cup \{J\}
return (UNSAT, I)
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function get_unsat_cover((s_1, \ldots, s_{i-1}))
                                                                Real root isolation over a
                                                                partial sample point
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                                                                 Select sample from \mathbb{R} \setminus I
  s_i := sample_outside(\mathbb{I})
  if i = n then return (SAT, (s_1, \ldots, s_{i-1}, s_i))
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                                                                   Recurse to next variable
  (f,O) := \mathtt{get\_unsat\_cover}((s_1,\ldots,s_{i-1},s_i))
                                                                   CAD-style projection:
  if f = SAT then return (SAT, O)
                                                                   Roots of polynomials re-
  else if f = \text{UNSAT} then
    R := construct\_characterization((s_1, ..., \S))
                                                                   strict where covering is
                                                                   still applicable
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return (UNSAT, I)

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return (UNSAT, I)
                                                                 Extract interval from poly-
                                                                 nomials
```

### Some implementation details

- Heavily based on LibPoly [Jovanovic et al. 2017]
- ► Implements stuff beyond [Ábrahám et al. 2021]:
  - Different projection operators (McCallum, Lazard)
  - ► Lazard's lifting [Lazard 1994] [Kremer et al. 2021] using CoCoALib [Abbott et al. 2018]
  - ▶ Different variable orderings inspired by [England et al. 2014]
  - Generates infeasible subsets
     Store assertions with every interval
  - ► Supports mixed-integer problems using naive B&B-style intervals
  - Generation of formal proof skeletons
     Helps understanding, not detailed enough for automated verification
  - Experimental support for incremental checks
     No performance benefit observed, lives in a branch
- ► Arbitrary theory combination

Real algebraic numbers are first-class citizens of cvc5

# ${\sf Experiments}$

	QF_NRA	sat	unsat	solved
$\rightarrow$	cvc5	5137	5596	10733
	Yices2	4966	5450	10416
	z3	5136	5207	10343
$\rightarrow$	cvc5.cov	5001	5077	10078
	SMT-RAT	4828	5038	9866
	veriT	4522	5034	9556
	MathSAT	3645	5357	9002
$\rightarrow$	cvc5.inclin	3421	5376	8797

#### Incremental linearization – extensions

Also supports extended operators in the same style:

- ightharpoonup transcendentals  $(\pi, \sin, \cos, \tan, \dots)$
- ► exponentials (exp)
- bitwise and on integers (IAND, bvand in arithmetic)
- ▶ power of two (POW2, bit shift in arithmetic)

Easily integrates other solving techniques

- does one or more of the following:
  - ▶ generate a (preferably) linear lemma that rejects the current model
  - ► finds a proper model
- implemented: ICP-style propagations
- ▶ ideas: GB-style conflicts, subtropical satisfiability, . . .

#### Conclusion

- combines linearization and coverings
- conceptually simple strategy
- easily integrates other techniques
- ▶ there is more to do...

Any questions?

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