

# SC<sup>2</sup>: Satisfiability Checking and Symbolic Computation: [www.sc-square.org](http://www.sc-square.org)

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12 May 2017

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<sup>1</sup>Thanks to Erika Ábrahám, John Abbott, Bernd Becker, Anna M. Bigatti, Martin Brain, Bruno Buchberger, Alessandro Cimatti, Matthew England, Pascal Fontaine, Stephen Forrest, Alberto Griggio, Daniel Kroening, Werner M. Seiler & Thomas Sturm; H2020-FETOPEN-2016-2017-CSA project 712689

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The communities have their own technical terms, which we will distinguish as above

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#SAT (counting solutions) is a different problem from SAT

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- 1 There is no polynomial-time algorithm which will solve *all* SAT problems
-  But this doesn't necessarily imply exponential running time (though we don't know much better)
- 2 Any given SAT problem can be solved in polynomial time

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*“Despite substantial advances in verification technology, complexity issues with classical decision procedures are still a major obstacle for formal verification of real-world applications, e.g. in automotive and avionic industries.” [PQR09]*

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Over the integers it's undecidable anyway, so what's the point?

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**SMT** Starts from the Boolean structure, and dips into the theory, adding and retracting theory clauses as required

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**SAT** Frequently restarts [HH10], with some underpinning theory [LSZ93]

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**Also** No research in trying to make all the choices holistically.

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**Hard Problems?** Quite a challenge for SAT [Spe15]

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But There is symmetry, and we don't have to count the solutions one-by-one, so what is #SMT here?

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**So** We have a lot of work to do.

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**upper** A very significant improvement to [Dub90], again using  $r$  rather than  $n$  where possible

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Deduce weak worst-case complexity (i.e. apart from an exponentially-rare subset: [AL15]) of Gröbner bases is singly exponential

## There's a catch [Chi09]

### Theorem

$\forall n \geq n_0, d \geq d_0$  there are homogeneous  $f_1, \dots, f_\nu \in k[x_1, \dots, x_n]$  ( $\nu \leq n, \deg f_i \leq d$ ) and a prime ideal  $\mathfrak{p}$  such that

- 1 the zeros  $\mathcal{Z}(\mathfrak{p})$  coincides with a component, defined over  $k$ , of  $\mathcal{Z}(f_1, \dots, f_\nu)$ , and furthermore  $\mathcal{Z}(f_1, \dots, f_\nu)$  has exactly two components irreducible over  $\bar{k}$ :  $\mathcal{Z}(\mathfrak{p})$  and linear space;
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I don't fully understand the construction: it starts with [Yap91], as [MR13], but somehow builds a prime ideal inside this, with embedded high-multiplicity components

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**Can** Define  $(M, \mathfrak{D})$  analogously

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The induction (on  $n$ ) hypothesis is **order-invariant** decompositions

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Then [EBD15] we get  $O\left(m^{s2^{n-s}} d^{2^n}\right)$  behaviour

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ISSAC2017 [BDE<sup>+</sup>17] Can do Cylindrical Algebraic Decomposition in 12 variables with 11 equational constraints

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We can transpose this to the complexes, and get zero-dimensional QE examples in  $\mathbf{C}^n$  with  $2^{2^{O(n)}}$  isolated point solutions, even though the equations are all linear and the solution set is zero-dimensional.

## So let's not be mesmerised by the QE problem

Consider ([BDE<sup>+</sup>17]) a single semi-algebraic set defined by

$$f_1(x_1, \dots, x_{n-1}, k_1) = 0 \wedge f_2(x_1, \dots, x_{n-1}, k_1) = 0 \wedge \dots \\ f_{n-1}(x_1, \dots, x_{n-1}, k_1) = 0 \wedge x_1 > 0 \wedge \dots \wedge x_{n-1} > 0$$

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Of course, this doesn't guarantee that all the iterated resultants in [EBD15], or the Gröbner polynomials in [ED16], are primitive, but in practice they are.

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