



# A Verified Decision Procedure for Univariate Real Arithmetic with the BKR Algorithm

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This material is based upon work supported by the NSF under Grant No. CNS-1739629, the NSF Graduate Research Fellowship Program under Grants Nos. DGE1252522 and DGE1745016, by A\*STAR Singapore, and by the AFOSR under grant number FA9550-16-1-0288. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the NSF, AFOSR, or A\*STAR Singapore.

# Problem

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- Real arithmetic questions involving the  $\exists$  (exists) and  $\forall$  (for all) **quantifiers** (ranging over the reals) are difficult for computers
- **Quantifier elimination (QE)**: The process of transforming a quantified statement into a *logically equivalent* quantifier-free statement

# Examples

## Example

$$\forall x. x^2 + 1 > 0$$



QE

True

## Example\*

$$\forall x \forall y. ((x^2 + ay^2 \leq 1) \Rightarrow (ax^2 - a^2xy + 2 \geq 0))$$



QE

$$(a \geq 0) \text{ and } (a^3 - 8a - 16 \leq 0)$$

QE is identifying exactly what conditions on  $a$  will make the original formula true!

\*This example is taken from some of Pablo Parrilo's lecture notes (Lecture 18 of his 2006 course, "Algebraic Techniques and Semidefinite Optimization"). Accessible through his webpage: <https://www.mit.edu/~parrilo/index.html>

# A Miraculous Result

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- Algorithms for QE exist (Tarski, 1930)
- Algorithms for QE are complicated



Alfred Tarski

# Terminology



- **Formulas:** Conjunctions and disjunctions of polynomial inequalities and equations (with rational coefficients)
- If a formula in a QE problem involves only one variable, we call it a **univariate** QE problem. Else it is a **multivariate** QE problem
- **Decision problems** are problems where all variables are quantified

# Examples, Revisited

## Example

$$\forall x. x^2 + 1 > 0$$



True

**A univariate decision problem**

## Example\*

$$\forall x \forall y. ((x^2 + ay^2 \leq 1) \Rightarrow (ax^2 - a^2xy + 2 \geq 0))$$



$$(a \geq 0) \text{ and } (a^3 - 8a - 16 \leq 0)$$

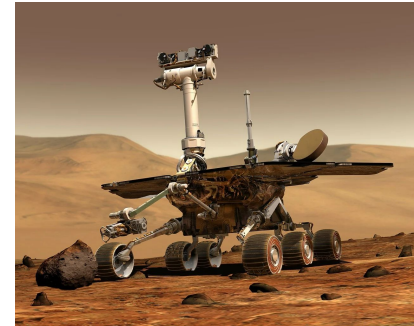
**A multivariate QE question**  
**Not a decision problem**

\*This example is taken from some of Pablo Parrilo's lecture notes (Lecture 18 of his 2006 course, "Algebraic Techniques and Semidefinite Optimization"). Accessible through his webpage: <https://www.mit.edu/~parrilo/index.html>

# Motivation

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- Quantified statements arise in a number of applications
  - Geometry proofs
  - Stability analysis
  - Verification of cyber-physical systems (like robots!)



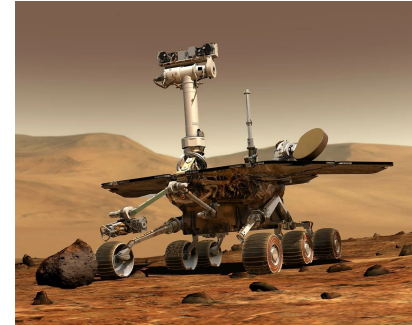
For more information, see:

Sturm, T. A Survey of Some Methods for Real Quantifier Elimination, Decision, and Satisfiability and Their Applications. *Math.Comput.Sci.* 11, 483–502 (2017).

# Motivation

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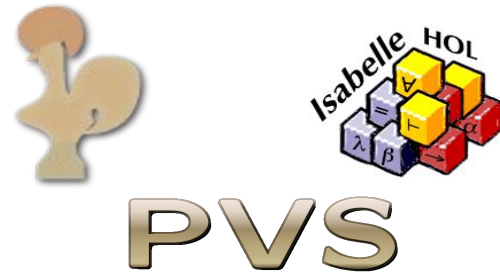
- Quantified statements arise in a number of applications
  - Geometry proofs
  - Stability analysis
  - Verification of cyber-physical systems (like robots!)
- Two conclusions
  - We want to know **how to do QE**
  - We want to be sure that we know **how to do QE correctly**





# Motivation

- **What we want:** Formally verified QE algorithms



# Motivation




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- **What we want:** Formally verified QE algorithms
- **Problem:** Dearth of efficient verified QE support
  - CPS theorem prover KeYmaera X outsources QE to unverified software
  - This can introduce bugs



# Related Work



-  = YES
-  = NO
-  = IN BETWEEN

	Efficient?	Verified?	Multivariate case builds directly on univariate?
Cohen-Hörmander			
Tarski			
CAD			

# Related Work



= YES



= NO



= IN BETWEEN

	Efficient?	Verified?	Multivariate case builds directly on univariate?
Cohen-Hörmander			
Tarski			
CAD			
BKR & Renegar Potential sweet spot!			

# BKR and Renegar



- Originally BKR\* was a decision procedure
- Renegar\*\* extended BKR to a general-purpose QE algorithm
  - Explains BKR in more detail
  - Fixes an error in BKR's multivariate complexity analysis

\*Michael Ben-Or, Dexter Kozen, and John H. Reif. The complexity of elementary algebra and geometry. *J. Comput. Syst. Sci.*, 32(2):251-264, 1986.

\*\*James Renegar. On the computational complexity and geometry of the first-order theory of the reals, part III: quantifier elimination. *J. Symb. Comput.*, 13(3):329-352, 1992.

# We formally verify\* the univariate cases of BKR and Renegar in Isabelle/HOL.



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\*Available on the Archive of Formal Proofs at:  
[https://www.isa-afp.org/entries/BenOr\\_Kozen\\_Reif.html](https://www.isa-afp.org/entries/BenOr_Kozen_Reif.html)

# High-level Context

- ~7000 LOC
  - Algorithm: ~110 LOC
  - Matrix library extensions: ~1800 LOC



# High-level Context

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- ~7000 LOC
  - Algorithm: ~110 LOC
  - Matrix library extensions: ~1800 LOC
- Why Isabelle/HOL?
  - Well-suited to formalizing mathematics
  - Strong math libraries
  - Sledgehammer





# Univariate BKR: Bird's Eye View



- Transform the problem:
  1. Decision problems to sign determination
  2. Sign determination to restricted sign determination
  3. To solve restricted sign determination, set up a matrix equation.

The main formalization challenge

# Step 1: Decision to Sign Determination

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- Solve decision problems by finding the *consistent sign assignments (CSAs)* for a set of polynomials (sign determination)

**Definition (sign assignment for  $\{g_1, \dots, g_n\}$ ).** A mapping  $\sigma: \{g_1, \dots, g_n\} \rightarrow \{+, -, 0\}$  is **consistent** if there is a real  $x$  where, for all  $i$ , the sign of  $g_i(x)$  matches  $\sigma(g_i)$ .

# Step 1: Decision to Sign Determination

- Solve decision problems by finding the *consistent sign assignments (CSAs)* for a set of polynomials (sign determination)

Decision Problem:  
 $\exists x. (x^2+1 \geq 0 \wedge 3x <$

$0)$

Find all consistent sign assignments for  $x^2 + 1$  and  $3x$

CSAs: (+, -), (+, 0), (+, +)

CSA (+, -) indicates the existence of a point  $k$  with  $(k^2+1 \geq 0 \wedge 3k < 0)$

# Correctness Results for Step 1



```
theorem decision_procedure:  
  "( $\forall x::real.$  fml_sem fml x)  $\longleftrightarrow$  decide_universal fml"  
  "( $\exists x::real.$  fml_sem fml x)  $\longleftrightarrow$  decide_existential fml"
```

Canonical semantics for formulas  
(defines what it means for a formula  
to hold at  $x$  in the standard way)

Our algorithms

## Step 2: Restricted Sign Determination

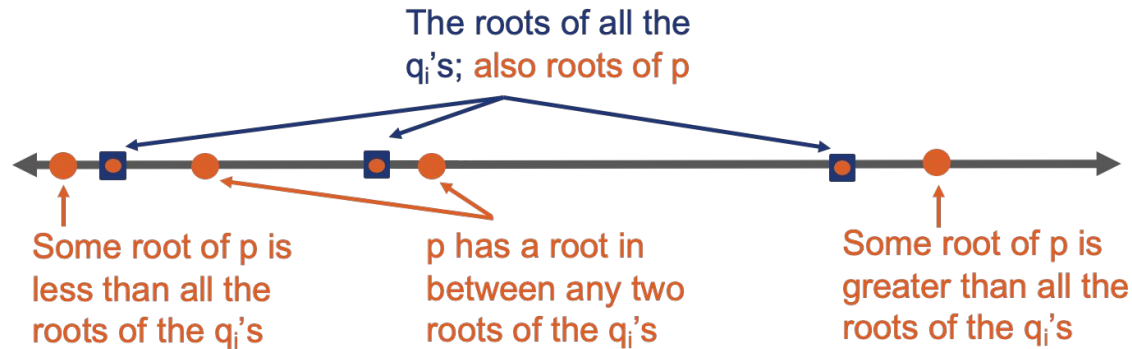
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- Restrict sign determination to finding all **CSAs** for a set of polynomials  $\{q_1, \dots, q_n\}$  *at the roots of* an auxiliary nonzero polynomial  $p$

Technical detail: BKR  
imposes some conditions on  
 $\{q_1, \dots, q_n\}, p$

## Step 2: Restricted Sign Determination

- Restrict sign determination to finding all **CSAs** for a set of polynomials  $\{q_1, \dots, q_n\}$  **at the roots of** an auxiliary nonzero polynomial  $p$



## Correctness Results for Step 2



definition *roots* :: "real poly  $\Rightarrow$  real set" where "roots p = {x. poly p x = 0}"

definition *consistent\_signs\_at\_roots* :: "real poly  $\Rightarrow$  real poly list  $\Rightarrow$  rat list set"  
where "*consistent\_signs\_at\_roots* p qs = (sgn\_vec qs) ` (roots p)"

Plug in the roots to the q\_i's,  
take the resulting signs

Solve for the roots of a  
polynomial

# Correctness Results for Step 2



definition *roots* :: "real poly  $\Rightarrow$  real set" where "roots p = {x. poly p x = 0}"

definition *consistent\_signs\_at\_roots* :: "real poly  $\Rightarrow$  real poly list  $\Rightarrow$  rat list set"  
where "consistent\_signs\_at\_roots p qs = (sgn\_vec qs) ` (roots p)"

theorem *find\_consistent\_signs\_at\_roots*:

assumes "p  $\neq$  0"

assumes " $\bigwedge q. q \in \text{set } qs \implies \text{coprime } p \ q$ "

shows "set (find\_consistent\_signs\_at\_roots p qs) = consistent\_signs\_at\_roots p qs"

our (constructive) algorithm

the nonconstructive definition



## Step 3: The Matrix Equation

- Stores all relevant information for sign determination
- Idea dates back to Tarski; similarities to Cohen and Mahboubi's formalization\*
- **But BKR does it efficiently**

\*Cyril Cohen and Assia Mahboubi. Formal proofs in real algebraic geometry: from ordered fields to quantifier elimination. Log. Methods Comput. Sci., 8(1), 2012. doi:10.2168/ LMCS-8(1:2)2012.



Alfred Tarski

# Step 3: The Matrix Equation

Find sign assignments to  $q_1, \dots, q_n$  at the roots of  $p$

## Tarski

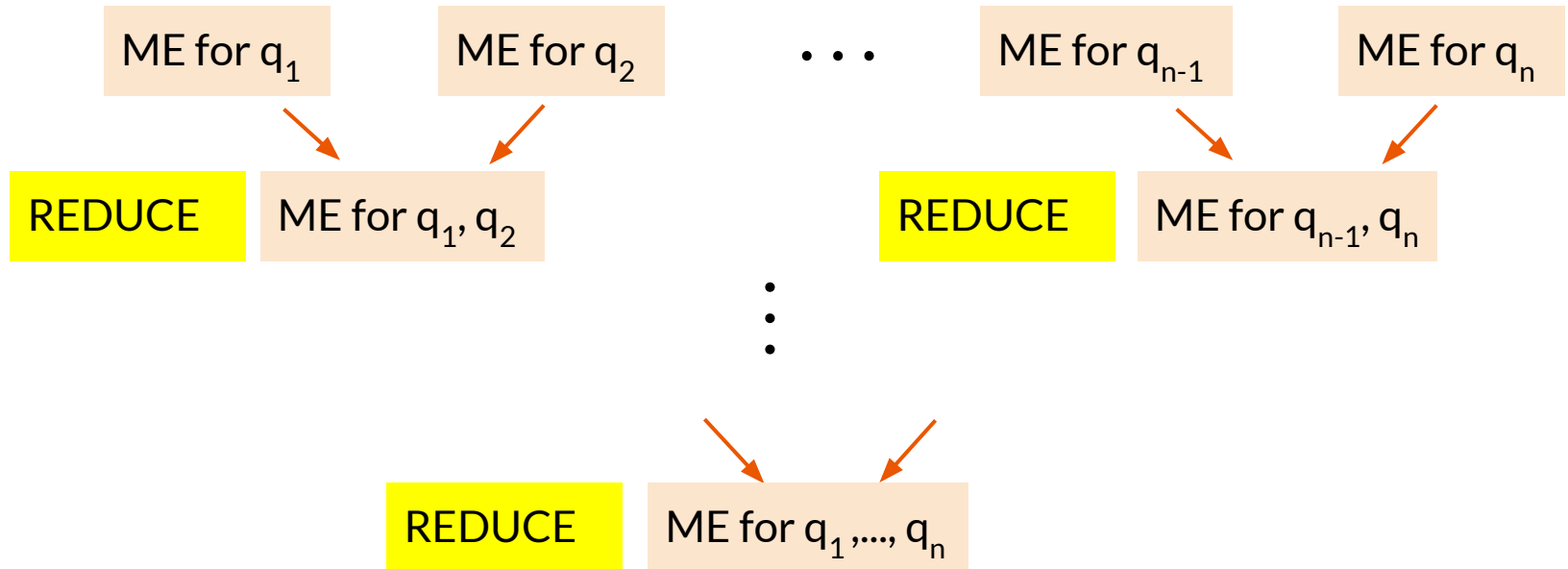
TQ stands for “Tarski query”, refers to invoking the (computational) Sturm-Tarski theorem

$$\begin{pmatrix} \# \text{ of } (+, \dots, +, +) \\ \# \text{ of } (+, \dots, +, -) \\ \vdots \\ \# \text{ of } (-, \dots, -, -) \end{pmatrix} = M^{-1} * \begin{pmatrix} \text{TQ subset 1} \\ \text{TQ subset 2} \\ \vdots \\ \text{TQ subset } 2^n \end{pmatrix}$$

Invertible matrix  
Size  $2^n \times 2^n$   
Can be computed

# Step 3: The Matrix Equation

Find sign assignments to  $q_1, \dots, q_n$  at the roots of  $p$   
BKR builds its matrix equation (ME) inductively



## Step 3: The Matrix Equation

After each combination, remove all inconsistent sign assignments  
(reduction step)

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \\ 1 \\ -1 \end{bmatrix} \quad \longrightarrow \quad \begin{bmatrix} 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & 1 & -1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}$$

Signs: ++, +-, -+, --

Signs: ++, +-, -+

# Reflections on Formalizing the Matrix Equation



- Inductive construction, inductive proof!
  - It took some work to identify the right inductive invariant
  - The reduction step poses the biggest challenge
- The reduction step requires extra proofs

\*Wenda Li. The Sturm-Tarski theorem. Archive of Formal Proofs, September 2014. [https://isa-afp.org/entries/Sturm\\_Tarski.html](https://isa-afp.org/entries/Sturm_Tarski.html), Formal proof development.

# Reflections on Formalizing the Matrix Equation



- Isabelle/HOL has well-developed libraries
  - The Sturm-Tarski theorem is already formalized\* (the key computational tool for the matrix equation)
  - A number of linear algebra libraries are available

\*Wenda Li. The Sturm-Tarski theorem. Archive of Formal Proofs, September 2014. [https://isa-afp.org/entries/Sturm\\_Tarski.html](https://isa-afp.org/entries/Sturm_Tarski.html), Formal proof development.

# Extending the Matrix Libraries



- We build on a matrix library by Thiemann and Yamada\*
- Our additions (~1800 LOC):
  - A computational notion of the Kronecker product
  - An algorithm to extract a basis from the rows of a matrix
    - Involved proving that row rank equals column rank

# Code Export and Experiments





# Experiments with SML code



- We export our formally verified algorithm to SML for experimentation
- Compare to:
  - A naive (unverified) version of Tarski's algorithm
  - Li, Passmore, and Paulson\*

\*Wenda Li, Grant Olney Passmore, and Lawrence C. Paulson. Deciding univariate polynomial problems using untrusted certificates in Isabelle/HOL. *J. Autom. Reason.*, 62(1):69–91, 2019.

# Experiments with SML code

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- We export our formally verified algorithm to SML for experimentation
- Compare to:
  - A naive (unverified) version of Tarski's algorithm
  - Li, Passmore, and Paulson\*
- Li et. al is faster:
  - CAD is generally faster than BKR
  - Their procedure is highly optimized
  - They use Mathematica as an untrusted oracle



\*Wenda Li, Grant Olney Passmore, and Lawrence C. Paulson. Deciding univariate polynomial problems using untrusted certificates in Isabelle/HOL. *J. Autom. Reason.*, 62(1):69–91, 2019.

# Experiments with SML code

\*Compiled with mlton  
\*Run on a laptop  
\*Dashes indicate timeout  
\*Times in seconds

Formula	#Poly	# $N(p, q)$ (Naive)	# $N(p, q)$ (BKR)	Time (Naive)	Time (BKR)	Time ([18])
ex1	4 (12)	20	31	0.003	0.006	3.020
ex2	5 (6)	576	180	5.780	0.442	3.407
ex3	4 (22)	112	120	1794.843	1865.313	3.580
ex4	5 (3)	112	95	0.461	0.261	3.828
ex5	8 (3)	576	219	28.608	8.333	3.806
ex6	22 (9)	50331648	-	-	-	6.187
ex7	10 (12)	6144	-	-	-	-
ex1 $\wedge$ 2	9 (12)	2816	298	317.432	3.027	3.033
ex1 $\wedge$ 2 $\wedge$ 4	13 (12)	28672	555	-	51.347	3.848
ex1 $\wedge$ 2 $\wedge$ 5	16 (12)	131072	826	-	436.575	3.711

Benchmarks  
from [18]

3s startup time for  
Mathematica

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# Future Work and Conclusion



# Future Work



- Optimizing univariate BKR
  - Add parallelism
  - Optimize Tarski queries
- Formally verified complexity analysis (ambitious!)
- Formalizing multivariate BKR

# Conclusion

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- We have formally verified the univariate case of BKR's QE algorithm
  - BKR hits a potential **sweet spot** in between practicality and ease of verification
  - Contributes to Isabelle/HOL's matrix libraries
  - Export code to SML for faster runtime
- Multivariate BKR is ongoing work



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≡ Questions?