

Fast Approximations of Quantifier Elimination



Isabel Garcia-Contreras*

Sharon Shoham



Hari Govind V K*

Arie Gurfinkel



@CAV 2023



* equal contribution

What is existential quantifier elimination (qelim)?

Given a formula $\varphi \triangleq \exists \mathbf{v} \cdot A(\mathbf{v})$, find a quantifier-free ψ that is equivalent to φ .

original

$$\exists x \cdot f(x) > 5 \wedge x \approx y$$

$$\exists x \cdot x > 5 \wedge y > x$$

$$\exists a \cdot a[i] \approx w \wedge a[j] \approx x \wedge a[k] \approx y \wedge a[l] \approx z$$

$$\exists x \cdot f(x) > 5$$

qelim

$$f(y) > 5$$

$$y > 6$$

$$(i \approx j \rightarrow w \approx x) \wedge (i \approx k \rightarrow w \approx y) \wedge \\ (i \approx l \rightarrow w \approx z) \wedge (j \approx k \rightarrow x \approx y) \wedge \\ (j \approx l \rightarrow y \approx z) \wedge (i \approx j \rightarrow y \approx z)$$

Does not exist

Why qelim?

Widely used in automated reasoning tasks

Playing with Quantified Satisfaction

Nikolaj Bjørner¹ and Mikoláš Janota²

¹ Microsoft Research, Redmond, USA

² Microsoft Research, Cambridge, UK

Abstract

We develop an algorithm for satisfiability of quantified formulas. The algorithm is based on recent progress in solving Quantified Boolean Formulas, but it generalizes beyond propositional logic to theories, such as linear arithmetic over integers (Presburger arithmetic), linear arithmetic over reals, algebraic data-types and arrays. Compared with previous algorithms for satisfiability of quantified arithmetical formulas our new implementation outperforms previous implementations in Z3 by a significant margin.

Solving Exists/Forall Problems With Yices

Extended Abstract

Bruno Dutertre

Computer Science Laboratory
SRI International
Bruno.Dutertre@sri.com

Abstract

Yices now includes a solver for Exists/Forall problem. We describe the problem, a general solving algorithm, and a key model-based generalization procedure. We explain the Yices implementation of these algorithms and survey a few applications.

1 Introduction

The traditional SMT problem is to determine whether a quantifier-free formula $\Phi(x)$ is satisfiable. Some solvers can also handle first-order formulas with arbitrary quantifiers. We are concerned with a simpler case, namely, formulas of the form $\forall y. \Phi(x, y)$, where $\Phi(x, y)$ is quantifier-free. Implicitly, the variables x are existentially quantified: we are checking the

Complete Functional Synthesis

Viktor Kuncak Mikaël Mayer Ruzica Piskac Philippe Suter*

School of Computer and Communication Sciences (I&C) - Swiss Federal Institute of Technology (EPFL), Switzerland
firstname.lastname@epfl.ch

Abstract

Synthesis of program fragments from specifications can make programs easier to write and easier to reason about. To integrate synthesis into programming languages, synthesis algorithms should behave in a predictable way—they should succeed for a

requires detailed specifications, which for large programs become difficult to write.

We therefore expect that practical applications of synthesis lie in its integration into the compilers of general-purpose programming languages. To make this integration feasible, we aim to identify

SMT-Based Model Checking for Recursive Programs

Anvesh Komuravelli, Arie Gurfinkel, and Sagar Chaki

Carnegie Mellon University, Pittsburgh, PA, USA

Abstract. We present an SMT-based symbolic model checking algorithm for safety verification of recursive programs. The algorithm is modular and analyzes procedures individually. Unlike other SMT-based approaches, it maintains both *over-* and *under-approximations* of procedure summaries. Under-approximations are used to analyze procedure calls without inlining. Over-approximations are used to block infeasible

Dealing with qelim

Expensive in general

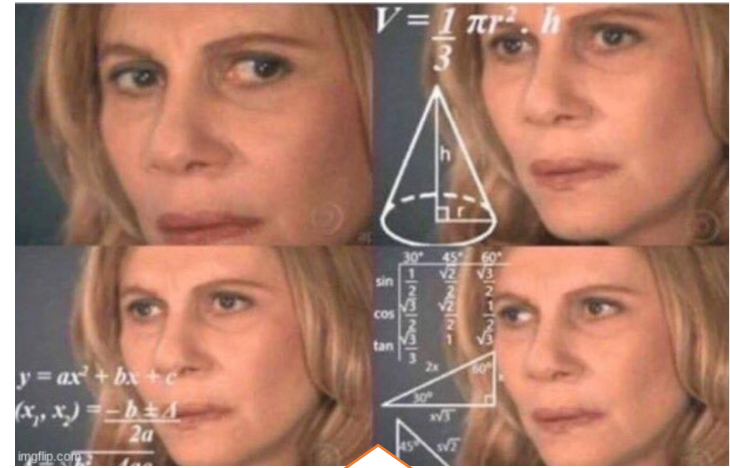
propositional: PSPACE-complete

LIA: $O(m^{2^n})$

... but cheap in some cases

Existing solvers try their luck with a light preprocess based on the **variable substitution** rule:

$\exists x \cdot x \approx t \wedge \varphi \equiv \varphi[x \mapsto t]$ (*) provided t is x-free



Btw, they should replace the math in this image by qelim problems...

Dealing with qelim by substitution

$$\exists x \cdot x \approx t \wedge \varphi \equiv \varphi[x \mapsto t]$$

Let's try: $\exists x, y \cdot A(x, y)$ with

$$A(x, y) \triangleq y \approx f(x) \wedge x \approx g(y) \wedge f(x) \approx 6$$

Trial #1: $A[y \rightarrow f(x)] : x \approx g(f(x)) \wedge f(x) \approx 6$

No more defs

Trial #2: $A[x \rightarrow g(y)] : y \approx f(g(y)) \wedge f(g(y)) \approx 6$

No more defs

Trial #3: $A[y \rightarrow 6][x \rightarrow g(6)] : 6 \approx f(g(6))$

qelim!

by transitivity

Relies on definitions **syntactically existing** in the formula
Depends on **substitution order**
Difficult to deal with **circular equalities**

Our aim: fast quantifier reduction

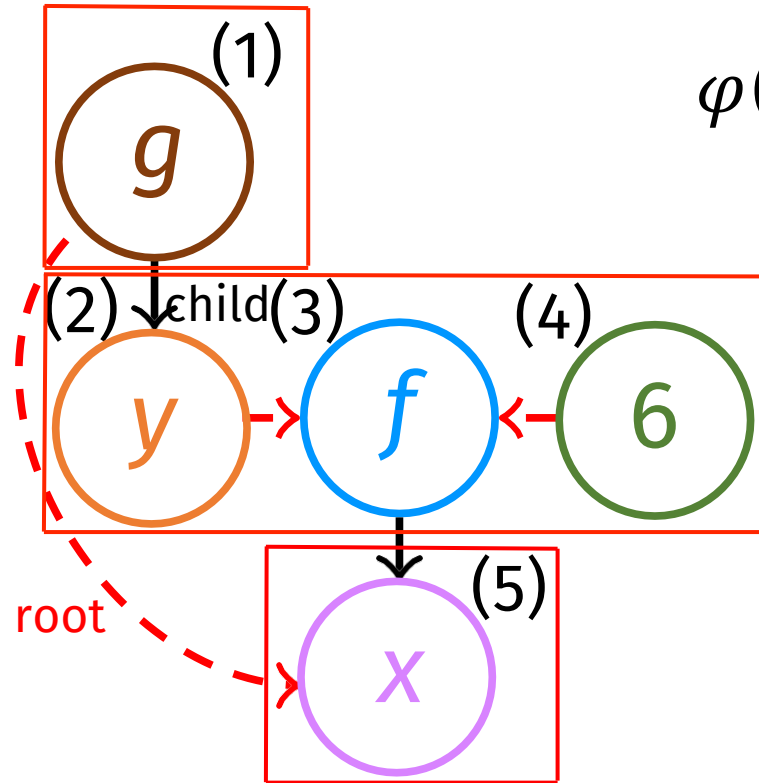
Quickly try to remove variables (reduction of variables)

Consider all definitions



Egraphs!

What are egraphs?



$$\varphi(x, y) \triangleq \underset{(2)}{y} \approx \underset{(3)}{f(x)} \wedge \underset{(5)}{x} \approx \underset{(1)}{g(y)} \wedge \underset{(4)}{f(x)} \approx \underset{(4)}{6}$$

$$\text{root}(N(2)) = N(3)$$

$$\text{root}(N(4)) = N(5)$$

$$\text{root}(N(1)) = N(5)$$

$$\text{class}(N(1)) = \{N(1), N(5)\}$$

$$\text{class}(N(2)) = \{N(3), N(2), N(4)\}$$

Notation

$$G = \text{egraph}(\varphi)$$

Extracting terms from an egraph

Find one desired node per class \rightarrow representative (rep)

To extract a term of a node, use the terms of reps of its children

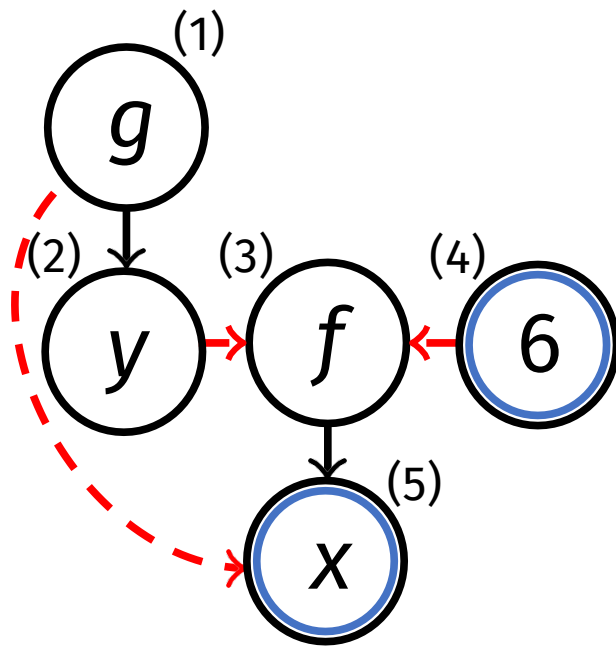
Notation

$\text{repr} : N \rightarrow N$ (representative function)

$\text{repr} = \{N(i), N(j)\}$ (we describe rep functions by the set of representatives)

$\text{ntt}(\text{node}, \text{repr})$ (node-to-term, we omit repr if obvious)

Extracting terms from an egraph



Example $\text{repr} = \{N(4), N(5)\}$

$\text{ntt}(N(5), \text{repr}) = x$

$\text{ntt}(N(3), \text{repr}) = f(x)$

$\text{ntt}(N(1), \text{repr}) = g(6)$

$\text{repr}(N(2)) = N(4)$
 $\text{repr}(N(3)) = N(4)$
 $\text{repr}(N(4)) = N(4)$
 $\text{repr}(N(1)) = N(5)$
 $\text{repr}(N(5)) = N(5)$

Extracting formulas from an egraph



Given a rep function `repr`, produce for each node:

$$\text{ntt}(\text{repr}(n)) \approx \text{ntt}(n)$$

Notation

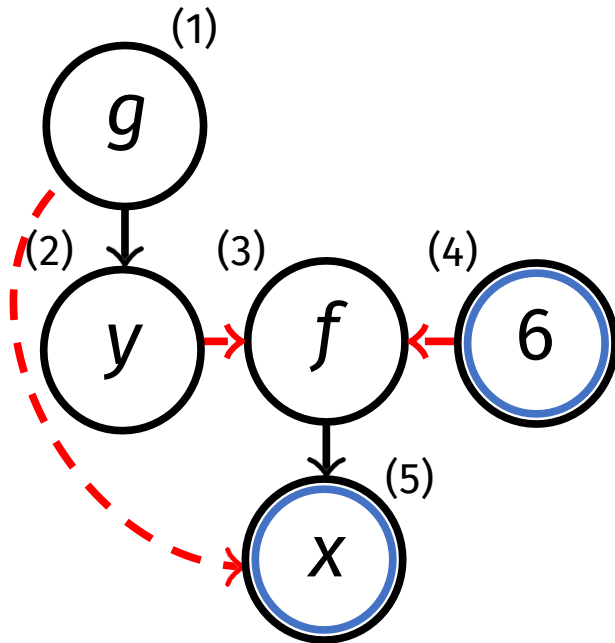
`G.to_formula(repr)` (extract a formula from an egraph)

$(\cdot)^\exists$ existential closure

Guarantee : $G = \text{egraph}(\varphi)$, $\varphi^\exists \equiv (\text{G.to_formula}(\text{repr}))^\exists$

Extracting formulas from an egraph

$G = \text{egraph}(y \approx f(x) \wedge x \approx g(y) \wedge f(x) \approx 6)$



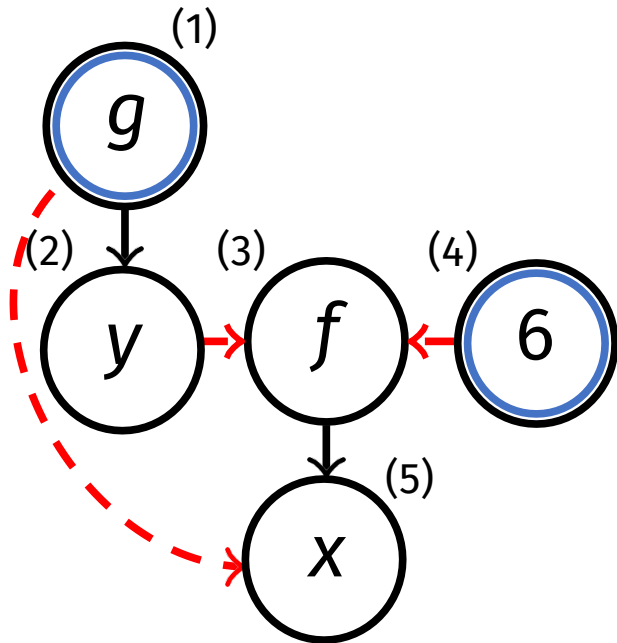
$\text{repr} = \{N(4), N(5)\}$

$G.\text{to_formula}(\text{repr}) =$

$$\underbrace{6 \approx f(x) \wedge 6 \approx y}_{\text{class}(N(4))} \wedge \underbrace{x \approx g(6)}_{\text{class}(N(5))}$$

Extracting for qelim

$G = \text{egraph}(y \approx f(x) \wedge x \approx g(y) \wedge f(x) \approx 6)$



$\text{repr} = \{N(4), N(1)\}$

$G.\text{to_formula}(\text{repr}) =$

$$\underbrace{6 \approx f(g(6))}_{\text{class}(N(4))} \wedge 6 \approx y \wedge \underbrace{g(f(6)) \approx x}_{\text{class}(N(5))}$$

$\text{class}(N(4))$

$\text{class}(N(5))$

bigger formula but more suitable for qelim!
just drop $(6 \approx y \wedge g(f(6)) \approx x)$

QEL – Quantifier reduction using egraphs



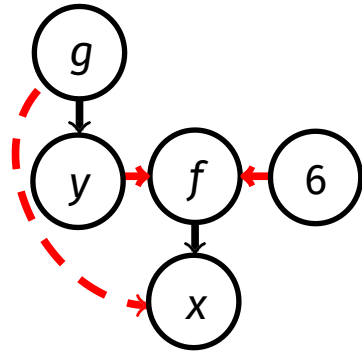
Problem that we are trying to solve:

Given a quantifier free formula $A(\mathbf{v})$,
find $B(\mathbf{u})$ with $\mathbf{u} \subseteq \mathbf{v}$ and $B(\mathbf{u})^{\exists} \equiv A(\mathbf{v})^{\exists}$
if \mathbf{u} is empty, we have qelim

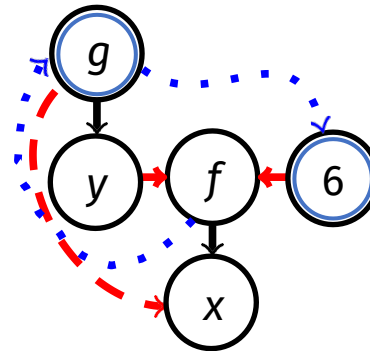
Using **transitivity** & **congruence** axioms

QEL – Quantifier reduction using egraphs

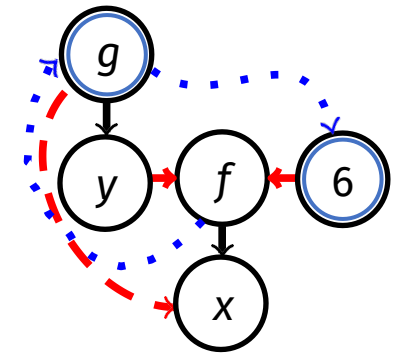
1. Build egraph



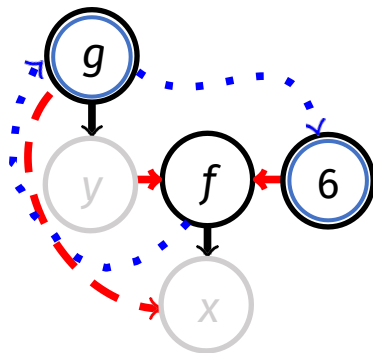
2. Find (ground) definitions



3. [Opt] Refine



4. Find a core

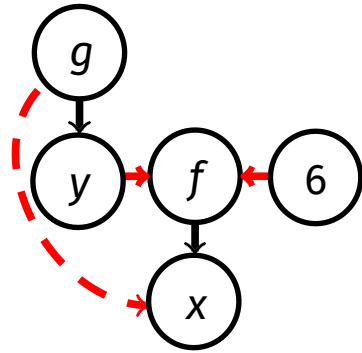


5. Output core

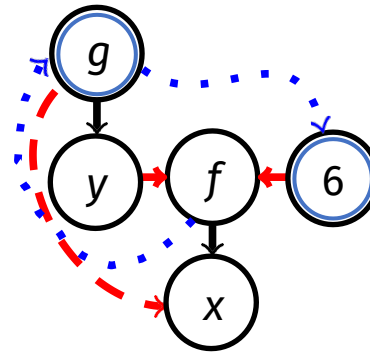
$$6 \approx f(g(6))$$

QEL – Quantifier reduction using egraphs

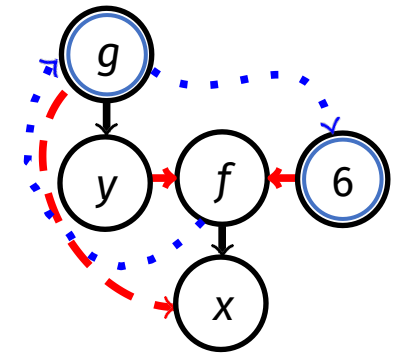
1. Build egraph



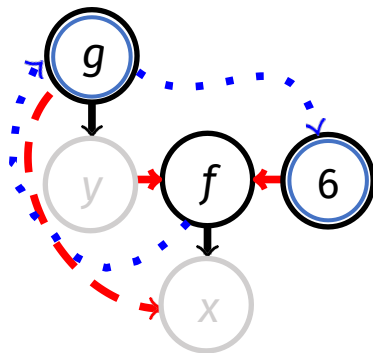
2. Find (ground) definitions



3. [Opt] Refine



4. Find a core



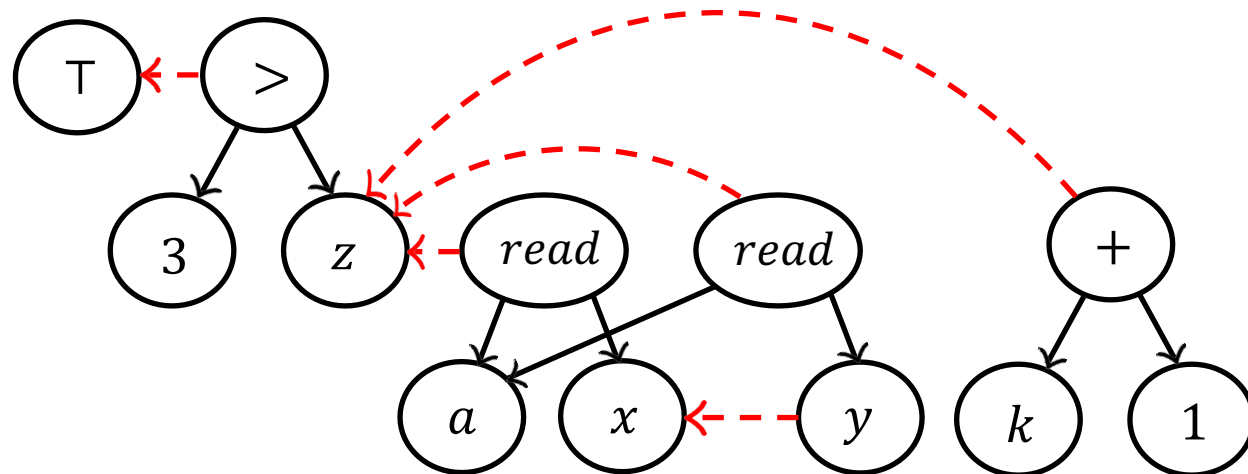
5. Output core

$$6 \approx f(g(6))$$

Constructively ground

A class is *ground* if it contains a node that is *constructively ground*

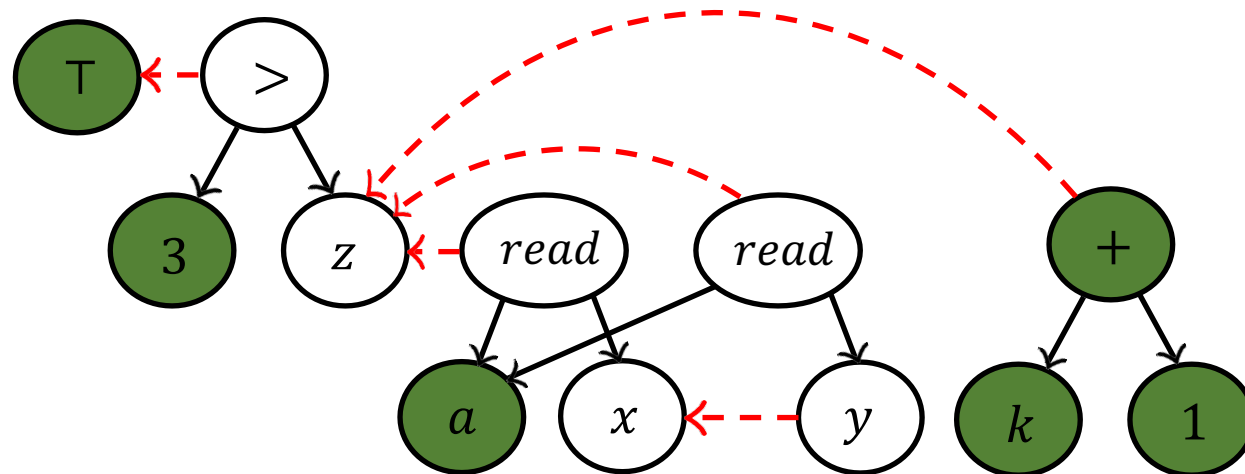
$$\gamma(x, y, z) \triangleq z \approx \text{read}(a, x) \wedge k + 1 \approx \text{read}(a, y) \wedge x \approx y \wedge 3 > z$$



Constructively ground

A class is *ground* if it contains a node that is *constructively ground*

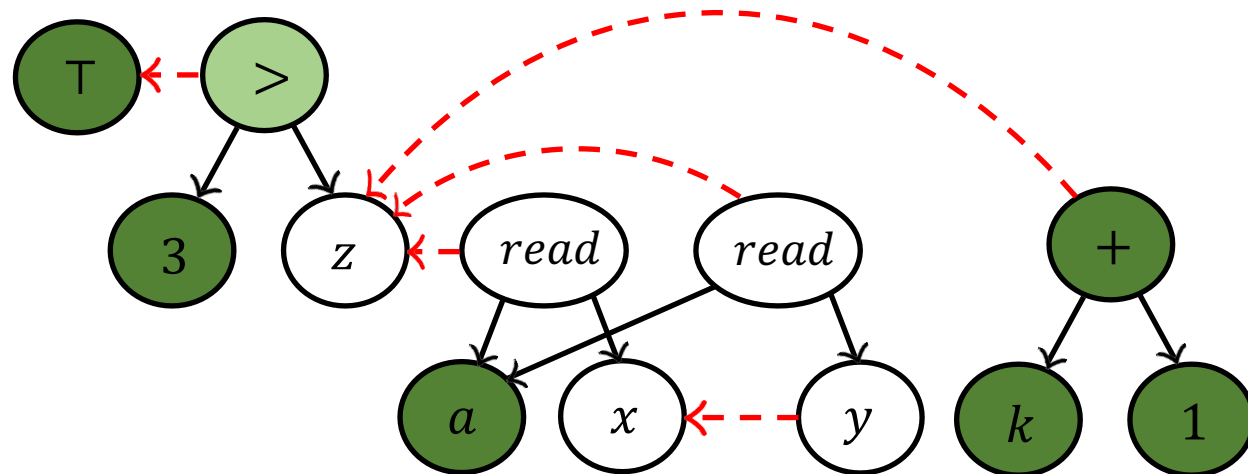
$$\gamma(x, y, z) \triangleq z \approx \text{read}(a, x) \wedge k + 1 \approx \text{read}(a, y) \wedge x \approx y \wedge 3 > z$$



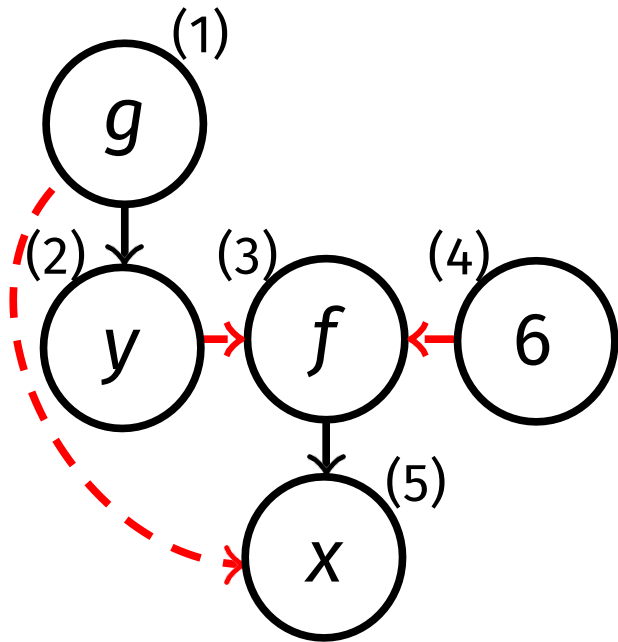
Constructively ground

A class is *ground* if it contains a node that is *constructively ground*

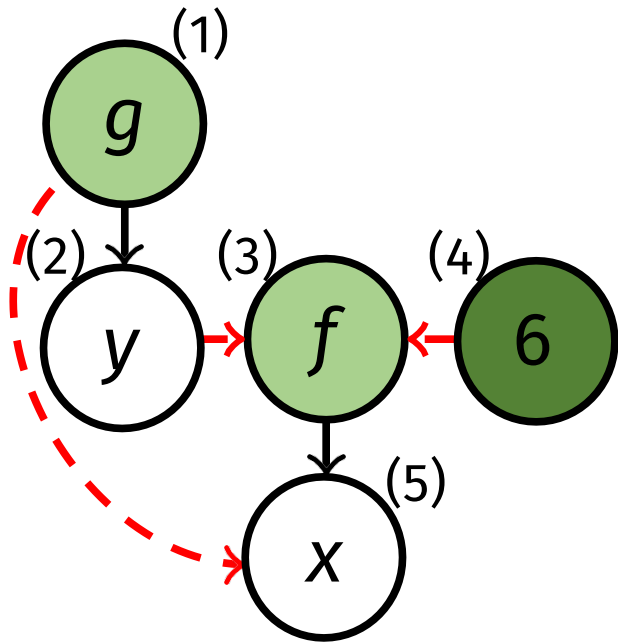
$$\gamma(x, y, z) \triangleq z \approx \text{read}(a, x) \wedge k + 1 \approx \text{read}(a, y) \wedge x \approx y \wedge 3 > z$$



Find repr maximizing constr. ground



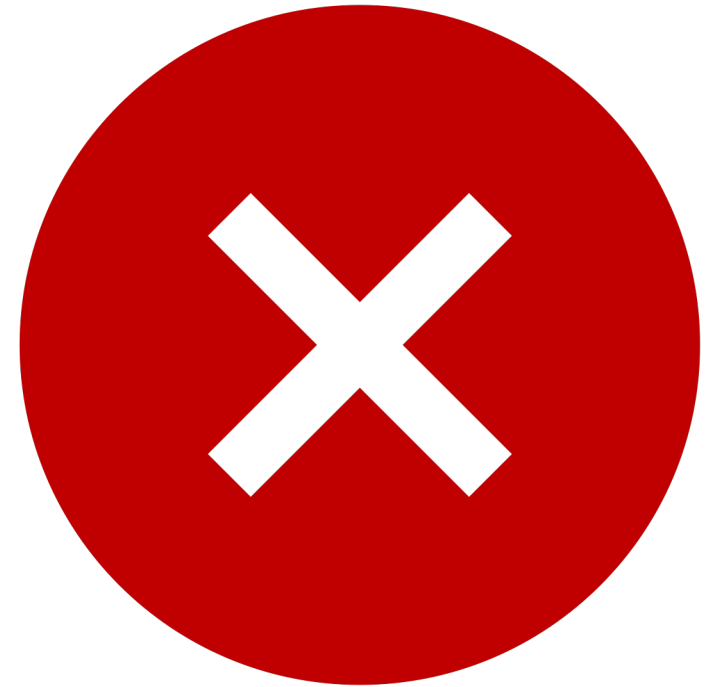
Find repr maximizing constr. ground



Let's choose constr. ground nodes as representatives!

Wait...
can we extract using
any representative
function?

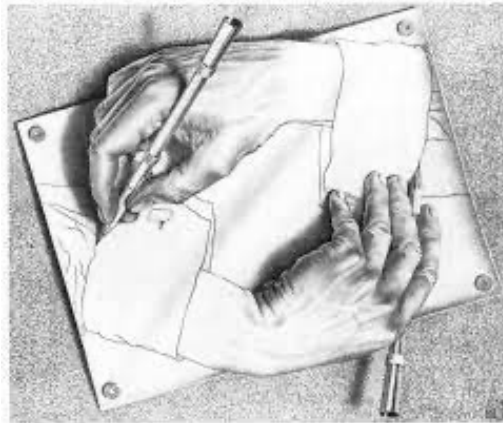
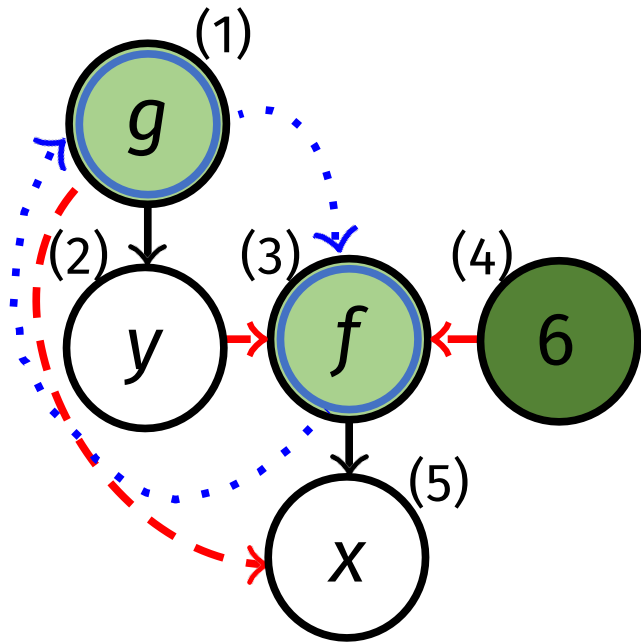
Not all representative functions
guarantee that ntt-extraction
terminates



Inadmissible representative function

Example $\text{repr} = \{N(1), N(3)\}$

$\text{ntt}(N(1)) = \text{ntt}(g(\text{ntt}(N(3)))) = \text{ntt}(g(f(\text{ntt}(N(1)))) \dots$



Admissible representative functions



A representative function $repr$ is **admissible** iff:

- **unique** rep per class
- $repr$ defines the **same classes** as **root**
- a node is not a representative of any of its $repr$ -descendants

Intuitively, the term of a node is not necessary to produce its own term

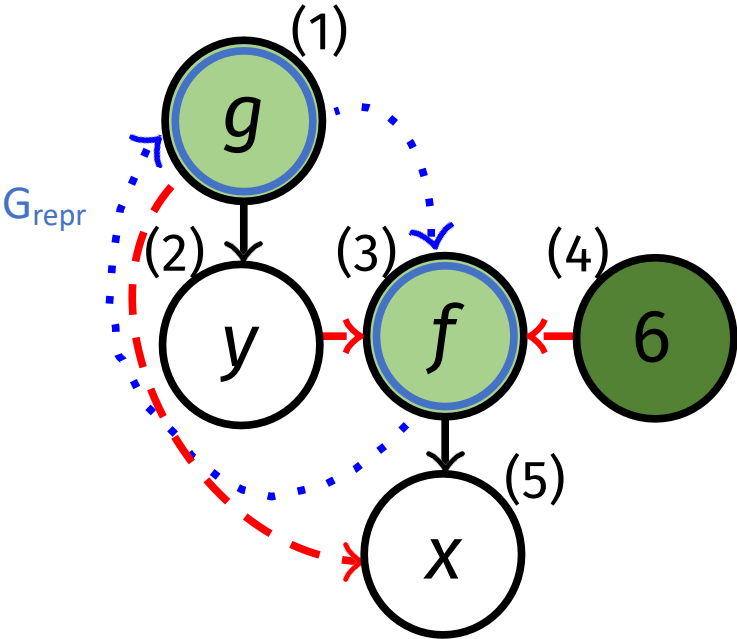
Formally, the graph $G_{repr} = \langle N, E_{repr} \rangle$ with

$E_{repr} \triangleq \{ (n, repr(c)) \mid c \in children(n), n \in N \}$ is acyclic

Admissible representative functions

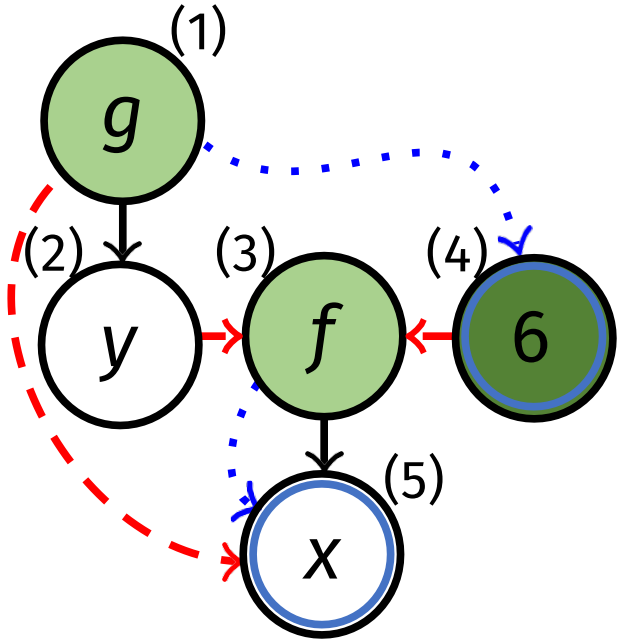
Example $\text{repr} = \{N(1), N(3)\}$

$$\text{ntt}(N(1)) = \text{ntt}(g(\text{ntt}(N(3)))) = \text{ntt}(g(f(\text{ntt}(N(1)))) \dots$$



Not
admissible!

Admissible representative functions



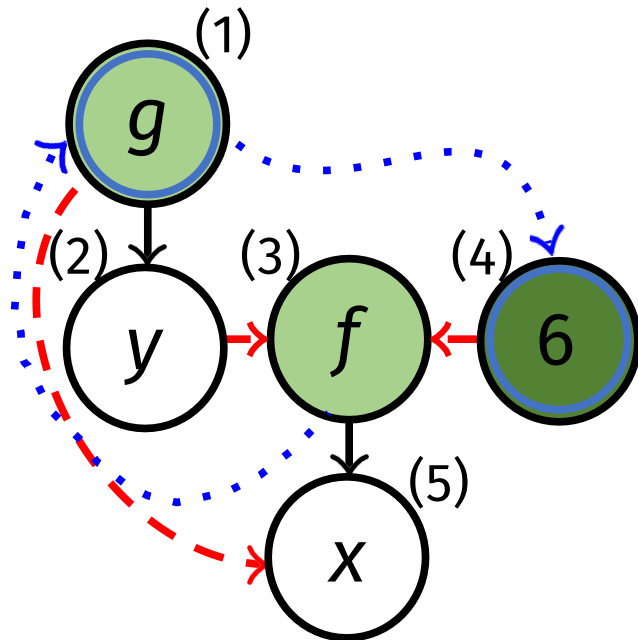
$\text{repr} = \{N(4), N(5)\}$

Admissible!

Admissibility is a *necessary and sufficient* condition for termination of `to_formula`

Choosing reps. based on (smaller) AST size guarantees admissibility...

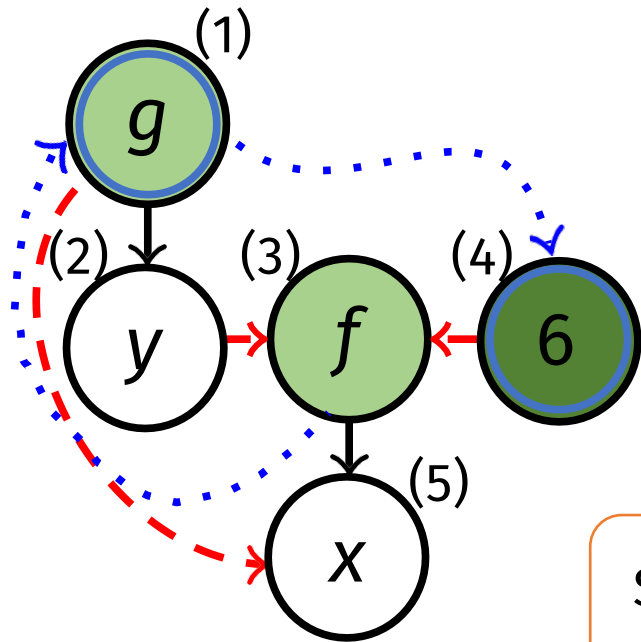
Find repr maximizing constr. ground



repr = {N(4), N(1)}

Admissible!

Find repr maximizing constr. ground



repr = {N(4), N(1)}

Why do constr. ground reps help?

Since terms of reps are used in ntt, **variables** with ground rep appear only **once** using to_formula:

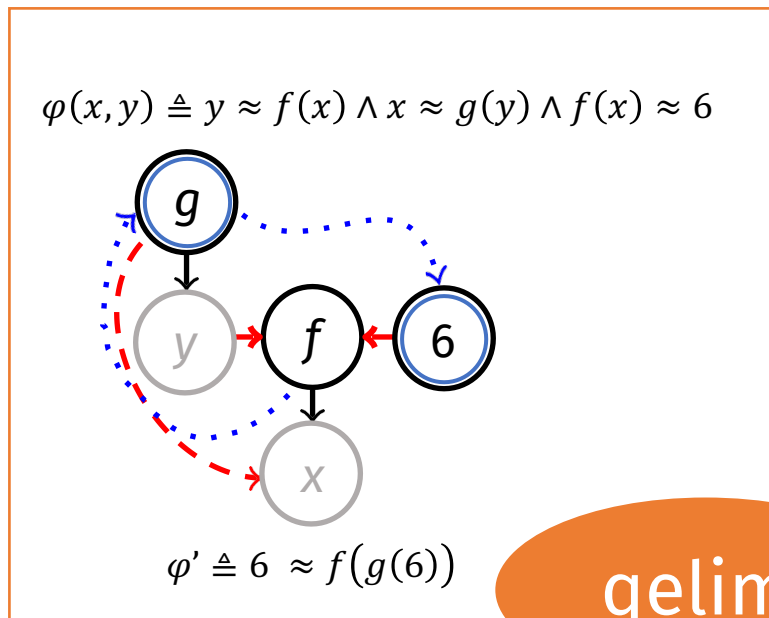
$G.to_formula(repr) = 6 \approx f(g(6)) \wedge 6 \approx y \wedge g(f(6)) \approx x$

Easy elimination!

QEL guarantees

A variable is eliminated if:

- It has a **ground definition**
- Its node is **not reachable** in G_{repr} by any of the nodes in the core*



If all variables meet the conditions, we find a **qelim**

QEL is **stronger than variable substitution**

$\exists x \cdot x \approx g(f(x)) \wedge f(x) \approx 6$ QEL finds $6 \approx g(f(6))$

For $\exists x, y \cdot f(x) \approx f(y) \wedge x \approx y$, QEL produces \top , which is a **qelim**

Model-Based Projection

Under-approximation if variables were not eliminated

Example: $\exists x \cdot f(x) > 5$: a projection is $f(0) > 5$

Presented as **rewrite rules**

Guarantees:

rewriting **terminates**
 result is an **under-approximation**
 output contains **no variables**

no approx.

approximate (split)
 based on a model

$$\begin{array}{c}
 \text{ELIMWRRD} \frac{\varphi[rd(wr(t, i, v), j)]}{(i = j \wedge \varphi[v]) \vee (i \neq j \wedge \varphi[rd(t, j)])} \qquad \text{ELIMWREQ} \frac{\varphi[wr(t_1, j, v) =_{\bar{i}} t_2]}{(j \in \bar{i} \wedge \varphi[t_1 =_{\bar{i}} t_2]) \vee (j \notin \bar{i} \wedge \varphi[t_1 =_{i, j} t_2 \wedge v = rd(t_2, j)])} \\
 \\
 \text{PARTIALEQ} \frac{\varphi[t_1 = t_2]}{\varphi[t_1 =_{\emptyset} t_2]} \text{ } t_i\text{'s have array sort} \qquad \text{TRIVEQ} \frac{\varphi[t =_{\bar{i}} t]}{\varphi[\top]} \qquad \text{SYMM} \frac{\varphi[t_1 =_{\bar{i}} t_2]}{\varphi[t_2 =_{\bar{i}} t_1]} \text{ } t_2 \text{ is a write term but } t_1 \text{ is not} \\
 \\
 \text{MBPLEFT} \frac{\varphi[\psi_1 \vee \psi_2]}{\varphi[\psi_1]} \quad M \models \varphi, \psi_1 \\
 \text{MBPRIGHT} \frac{\varphi[\psi_1 \vee \psi_2]}{\varphi[\psi_2]} \quad M \models \varphi, \psi_2 \\
 \text{MBPVAC} \frac{\varphi[\psi_1 \vee \psi_2]}{\varphi[\perp]} \quad M \models \varphi \quad M \not\models \psi_1, \psi_2
 \end{array}$$

Rules are defined for different theories separately

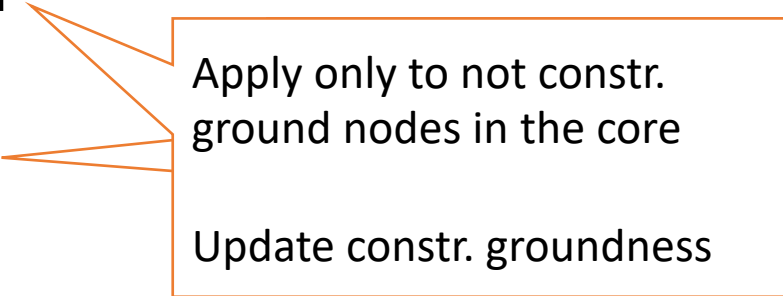
Implementing MBP using egraphs

Repeat until variables eliminated (out of the core*):

(1) Apply **equivalence-preserving MBP rules** until saturation

(2) Remove nodes from **core based on rules**

(3) If there are variables, apply **model-splitting MBP rules**



Apply only to not constr.
ground nodes in the core

Update constr. groundness

Very easy to combine theories, just as for SMT solving!

We implemented for **ADTs and Arrays**

Full elimination is guaranteed (under-approx.)

Implementation & evaluation – QSAT

Solving formulas **alternating exists** and **forall** quantifiers



| Category | Count | Z3EG | | Z3 | | YICESQS | |
|----------|-------|------------|------------|------------|------------|------------|-------------|
| | | SAT | UNSAT | SAT | UNSAT | SAT | UNSAT |
| LIA | 416 | 150 | 266 | 150 | 266 | 107 | 102 |
| LRA | 2 419 | 795 | 1 589 | 793 | 1 595 | 808 | 1610 |

Playing with Quantified Satisfaction

Nikolaj Bjørner¹ and Mikoláš Janota²

¹ Microsoft Research, Redmond, USA

² Microsoft Research, Cambridge, UK

Abstract

We develop an algorithm for satisfiability of quantified formulas. The algorithm is based on recent progress in solving Quantified Boolean Formulas, but it generalizes beyond propositional logic to theories, such as linear arithmetic over integers (Presburger arithmetic), linear arithmetic over reals, algebraic data-types and arrays. Compared with previous algorithms for satisfiability of quantified arithmetical formulas our new implementation outperforms previous implementations in Z3 by a significant margin.

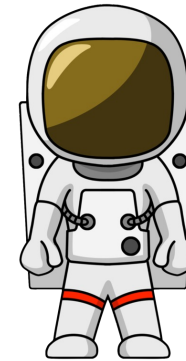
| Category | Count | Z3EG | | Z3 | |
|----------|-------|------------|--------------|------------|-------|
| | | SAT | UNSAT | SAT | UNSAT |
| LIA-ADT | 416 | 150 | 266 | 150 | 56 |
| LRA-ADT | 2 419 | 757 | 1 415 | 793 | 964 |

Implementation & evaluation – Spacer

CHC solving over ADTs, LIA and Arrays

Z3

| Category | Count | Z3EG | | Z3 | | ELDARICA | |
|----------------|-------|------------|-------|-------|-----------|--------------|--------------|
| | | SAT | UNSAT | SAT | UNSAT | SAT | UNSAT |
| Solidity | 3 468 | 2 324 | 1 133 | 2 314 | 1 114 | 2 329 | 1 134 |
| → abi | 127 | 19 | 108 | 19 | 88 | 19 | 108 |
| LIA-lin-Arrays | 488 | 214 | 72 | 212 | 75 | 147 | 68 |



SolCMC: Solidity Compiler's Model Checker



Leonardo Alt^{1(✉)}, Martin Blicha^{2,3},
Antti E. J. Hyvärinen², and Natasha Sharygina²

¹ Ethereum Foundation, Berlin, Germany
leo@ethereum.org

² Università della Svizzera italiana, Lugano, Switzerland
{martin.blicha, antti.hyvaerinen, natasha.sharygina}@usi.ch

³ Charles University, Prague, Czech Republic



Conclusion

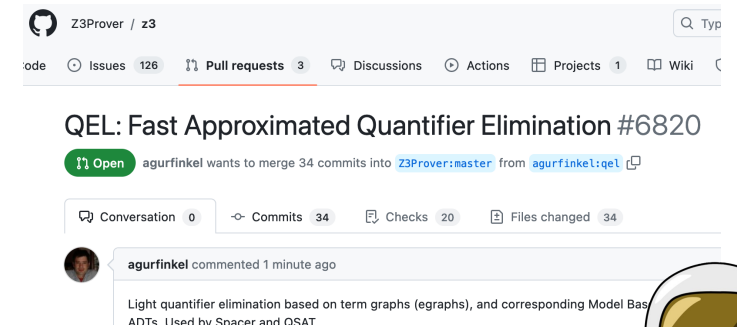
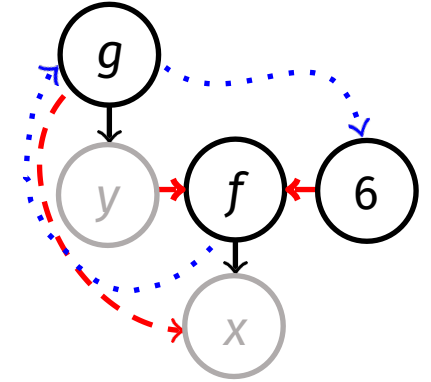
Characterized all possible extractions from egraphs via **admissible representative functions**

Presented QEL, an algorithm for quantifier reduction that is **complete relative to ground definitions** entailed by formulas

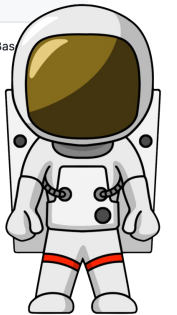
Use **theory rewrites** to under-approximate formulas (**Model-Based Projection**) when QEL was not complete

Implemented MBP for ADTs and Arrays

Implemented and evaluated within Z3: We used it to improve **QSAT** and the **Spacer CHC solver**



Z3



Fast Approximations of Quantifier Elimination



Isabel Garcia-Contreras

Sharon Shoham



Hari Govind V K

Arie Gurfinkel



@CAV 2023

Yes, I am the
representative!



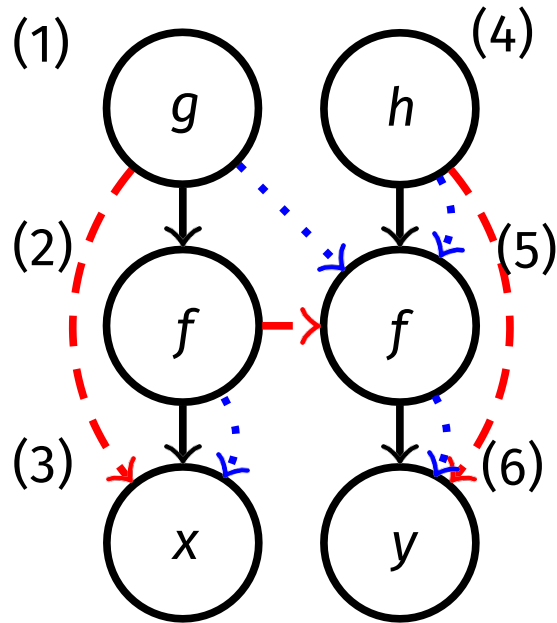
Thanks!

Refining repr

$$\psi(x, y) \triangleq x \approx g(f(x)) \wedge y \approx h(f(y)) \wedge f(x) \approx f(y)$$

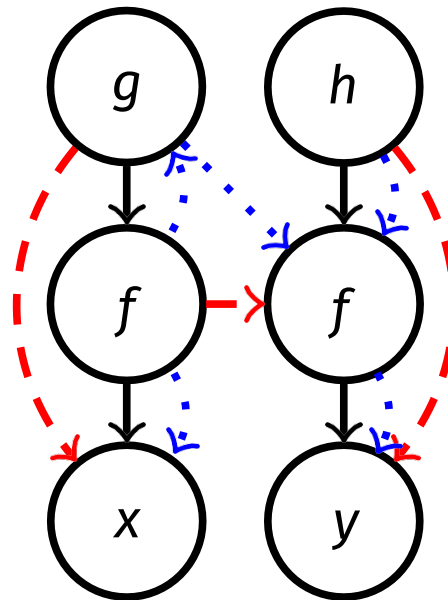
Nothing constr.
ground

repr = {N(3), N(5), N(6)}



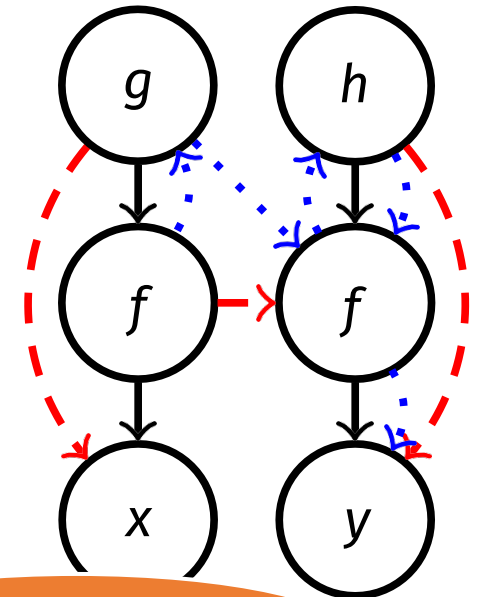
refine

repr = {N(1), N(5), N(6)}



try refine

repr = {N(1), N(5), N(4)}



Not admissible!

3 A Quantified Satisfiability Game

Algorithm 1: QSAT

```
1  $j \leftarrow 1$ ;  
2  $M \leftarrow \text{null}$ ;  
3 while True do  
4   if  $F_j \wedge \text{strategy}(M, j)$  is unsat then  
5     if  $j = 1$  then  
6       return  $G$  is false  
7     if  $j = 2$  then  
8       return  $G$  is true  
9      $C \leftarrow \text{Core}(F_j, \text{strategy}(M, j))$ ;  
10     $J \leftarrow \text{Mbp}(M, \text{tail}(j), C)$ ;  
11     $j \leftarrow$  index of max variable in  $J \cup \{1, 2\}$  of same parity as  $j$ ;  
12     $F_j \leftarrow F_j \wedge \neg J$ ;  
13     $M \leftarrow \text{null}$ ;  
14  else  
15     $M \leftarrow$  the current model;  
16     $j \leftarrow j + 1$ ;
```

Playing with Quantified Satisfaction

Nikolaj Bjørner¹ and Mikoláš Janota²

¹ Microsoft Research, Redmond, USA

² Microsoft Research, Cambridge, UK

Abstract

We develop an algorithm for satisfiability of quantified formulas. The algorithm is based on recent progress in solving Quantified Boolean Formulas, but it generalizes beyond propositional logic to theories, such as linear arithmetic over integers (Presburger arithmetic), linear arithmetic over reals, algebraic data-types and arrays. Compared with previous algorithms for satisfiability of quantified arithmetical formulas our new implementation outperforms previous implementations in Z3 by a significant margin.

Bonus: Formulas with Minimal Variables Appearing?

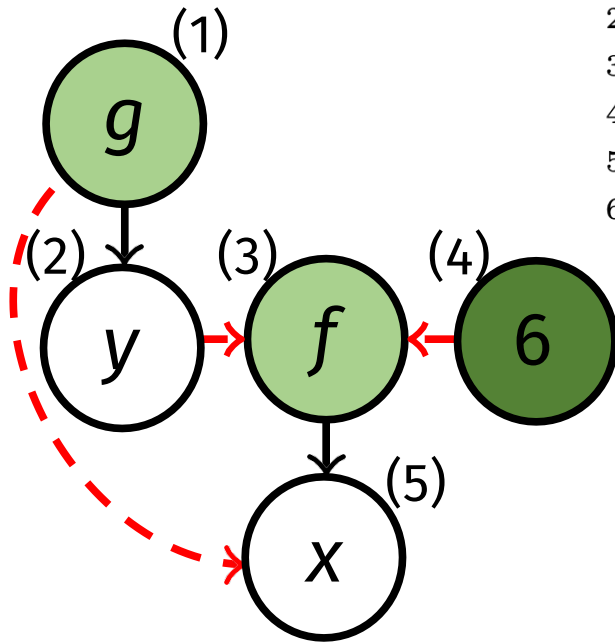
$$\varphi(x, y, z) \triangleq x \approx f(z) \wedge y \approx g(z) \wedge z \approx h(x, y)$$

Possible outputs, depending on **refinement order** of **repr**:

$$\begin{aligned}\varphi_1(z) &\triangleq z \approx h(f(z), g(z)) \\ \varphi_2(x, y) &\triangleq x \approx f(h(x, y)) \wedge y \approx g(h(x, y))\end{aligned}$$

Open question: hard due to sharing?

Find repr maximizing constr. ground



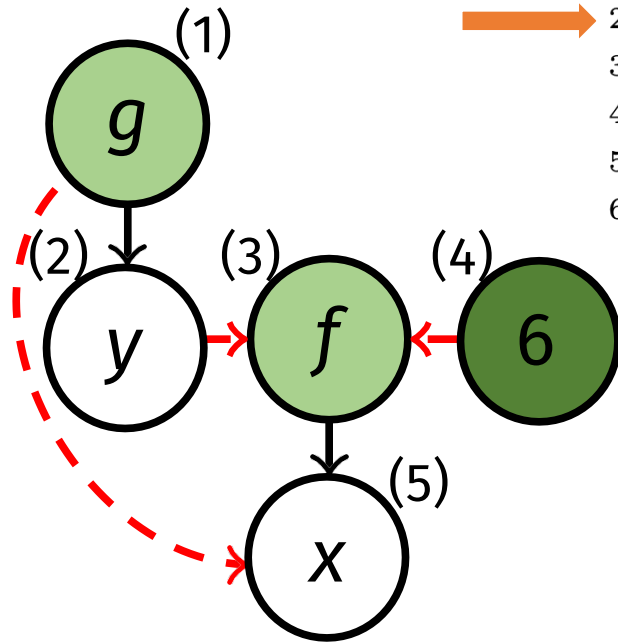
egraph :: find_defs(*v*)

```
1: for  $n \in N$  do repr( $n$ ) := ★  
2: todo := {leaf( $n$ ) |  $n \in N \wedge \text{ground}(n)$ }  
3: repr := process(repr, todo)  
4: todo := {leaf( $n$ ) |  $n \in N$ }  
5: repr := process(repr, todo)  
6: ret repr
```

egraph :: process(repr, todo)

```
7: while todo  $\neq \emptyset$  do  
8:    $n := \text{todo.pop}()$   
9:   if repr( $n$ )  $\neq \star$  then continue  
10:  for  $n' \in \text{class}(n)$  do repr( $n'$ ) :=  $n$   
11:  for  $n' \in \text{class}(n)$  do  
12:    for  $p \in \text{parents}(n')$  do  
13:      if  $\forall c \in \text{children}(p) \cdot \text{repr}(c) \neq \star$  then  
14:        todo.push( $p$ )  
15:  ret repr
```

Find repr maximizing constr. ground



egraph :: find_defs(*v*)

- 1: **for** $n \in N$ **do** repr(n) := ★
- 2: $todo := \{leaf(n) \mid n \in N \wedge ground(n)\}$
- 3: repr := process(repr, *todo*)
- 4: $todo := \{leaf(n) \mid n \in N\}$
- 5: repr := process(repr, *todo*)
- 6: **ret** repr

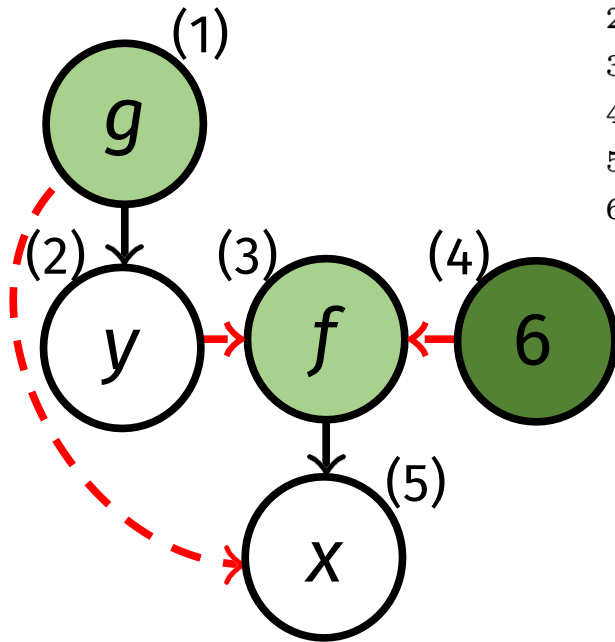
todo = [N(4)]

repr = {}

egraph :: process(repr, *todo*)

- 7: **while** $todo \neq \emptyset$ **do**
- 8: $n := todo.pop()$
- 9: **if** repr(n) \neq ★ **then continue**
- 10: **for** $n' \in class(n)$ **do** repr(n') := n
- 11: **for** $n' \in class(n)$ **do**
- 12: **for** $p \in parents(n')$ **do**
- 13: **if** $\forall c \in children(p) \cdot repr(c) \neq \star$ **then**
- 14: $todo.push(p)$
- 15: **ret** repr

Find repr maximizing constr. ground



egraph :: find_defs(*v*)

```
1: for  $n \in N$  do repr( $n$ ) := ★  
2:  $todo := \{leaf(n) \mid n \in N \wedge ground(n)\}$   
3: repr := process(repr,  $todo$ )  
4:  $todo := \{leaf(n) \mid n \in N\}$   
5: repr := process(repr,  $todo$ )  
6: ret repr
```

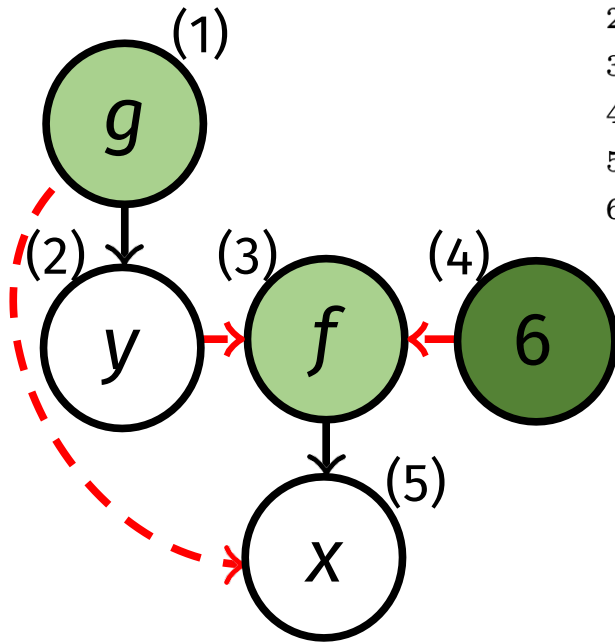
$todo = []$

$repr = \{N(4)\}$

egraph :: process(repr, $todo$)

```
7: while  $todo \neq \emptyset$  do  
8:  $n := todo.pop()$   
9: if repr( $n$ )  $\neq$  ★ then continue  
10: for  $n' \in class(n)$  do repr( $n'$ ) :=  $n$   
11: for  $n' \in class(n)$  do  
12: for  $p \in parents(n')$  do  
13: if  $\forall c \in children(p) \cdot repr(c) \neq \star$  then  
14:  $todo.push(p)$   
15: ret repr
```


Find repr maximizing constr. ground



egraph :: find_defs(*v*)

- 1: **for** $n \in N$ **do** repr(n) := ★
- 2: $todo := \{leaf(n) \mid n \in N \wedge ground(n)\}$
- 3: repr := process(repr, *todo*)
- 4: $todo := \{leaf(n) \mid n \in N\}$
- 5: repr := process(repr, *todo*)
- 6: **ret** repr

todo = [N(1)]

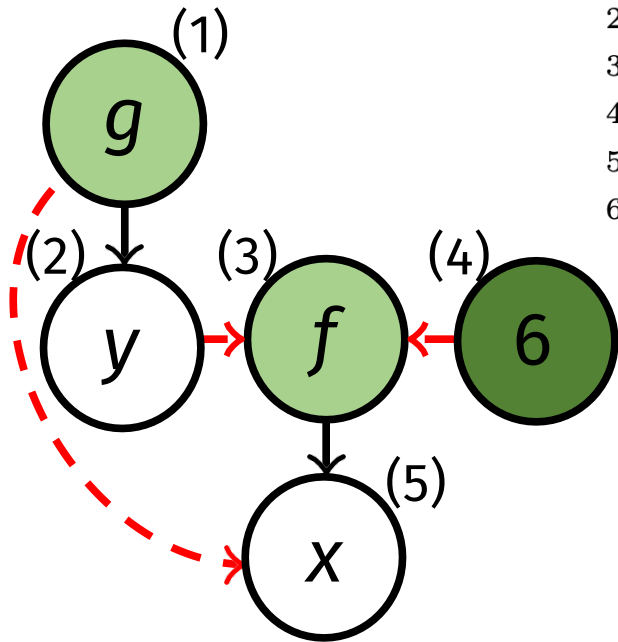
repr = {N(4)}

egraph :: process(repr, *todo*)

- 7: **while** $todo \neq \emptyset$ **do**
- 8: $n := todo.pop()$
- 9: **if** repr(n) \neq ★ **then continue**
- 10: **for** $n' \in class(n)$ **do** repr(n') := n
- 11: **for** $n' \in class(n)$ **do**
- 12: **for** $p \in parents(n')$ **do**
- 13: **if** $\forall c \in children(p) \cdot repr(c) \neq \star$ **then**
- 14: *todo.push(p)*
- 15: **ret** repr



Find repr maximizing constr. ground



egraph :: find_defs(*v*)

```
1: for  $n \in N$  do repr( $n$ ) := ★  
2: todo := {leaf( $n$ ) |  $n \in N \wedge \text{ground}(n)$ }  
3: repr := process(repr, todo)  
4: todo := {leaf( $n$ ) |  $n \in N$ }  
5: repr := process(repr, todo)  
6: ret repr
```

todo = []

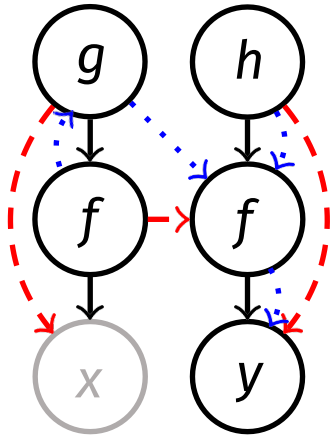
repr = {N(4), N(1)}

egraph :: process(repr, todo)

```
7: while todo  $\neq \emptyset$  do  
8:    $n := \text{todo.pop}()$   
9:   if repr( $n$ )  $\neq \star$  then continue  
10:  for  $n' \in \text{class}(n)$  do repr( $n'$ ) :=  $n$   
11:  for  $n' \in \text{class}(n)$  do  
12:    for  $p \in \text{parents}(n')$  do  
13:      if  $\forall c \in \text{children}(p) \cdot \text{repr}(c) \neq \star$  then  
14:        todo.push( $p$ )  
15:  ret repr
```

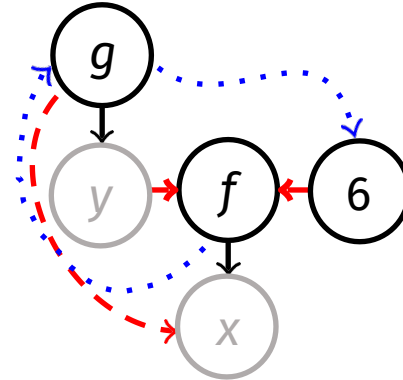
QEL examples

$$\psi(x, y) \triangleq x \approx g(f(x)) \wedge y \approx h(f(y)) \wedge f(x) \approx f(y)$$



$$\psi'(y) \triangleq y \approx h(f(y)) \wedge f(y) \approx f(g(f(y)))$$

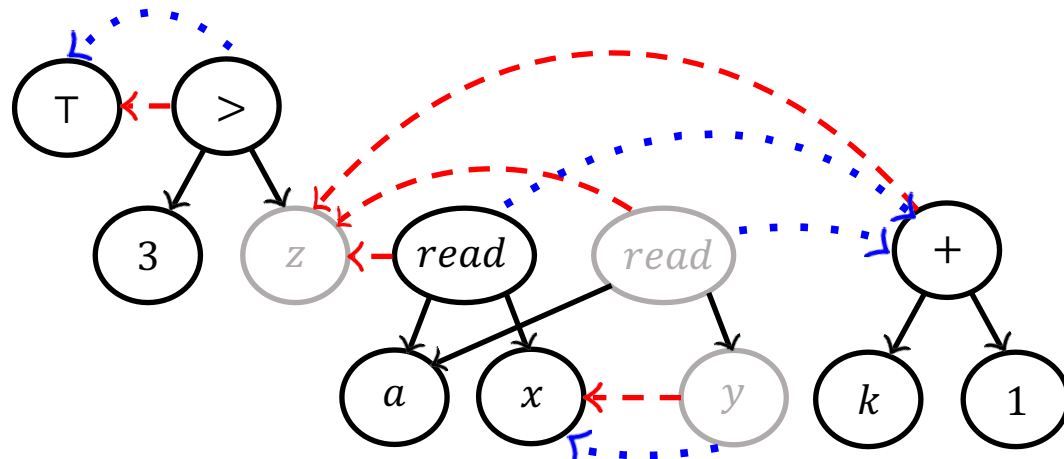
$$\varphi(x, y) \triangleq y \approx f(x) \wedge x \approx g(y) \wedge f(x) \approx 6$$



$$\varphi' \triangleq 6 \approx f(g(6))$$

qelim!

$$\gamma(x, y, z) \triangleq z \approx read(a, x) \wedge k + 1 \approx read(a, y) \wedge x \approx y \wedge 3 > z$$



$$\gamma'(x) \triangleq k + 1 \approx read(a, x) \wedge 3 > k + 1$$