# Loop Summarization with Rational Vector Addition Systems 

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## The Why

Invariant generation techniques are effective but can be unpredictable

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```
i = 0
while(i < 5) do
    i++
assert(i == 5)
```

Polyhedron domain with
widening / narrowing verifies
assertion

## The Why

Invariant generation techniques are effective but can be unpredictable

```
i \(=1\)
j \(=0\)
while(i < 5) do
    j \(=\) j \(+i\)
    i++
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Polyhedron domain with widening / narrowing fails to verify assertion

Not monotone: more information led to worse analysis

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Polyhedron domain with widening / narrowing fails to verify assertion

$$
\begin{aligned}
& \mathrm{i}=0 \\
& \mathrm{j}=0 \\
& \text { while(i<1000) do } \\
& \quad \mathrm{i}=\mathrm{i}+\text { step } \\
& j=j+\text { step } \\
& \text { assert }(\mathrm{i}==\mathrm{j})
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Ultimate Automizer verifies assertion

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Polyhedron domain with widening / narrowing fails to verify assertion

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& \text { while }(i<1000) \text { do } \\
& i=i+\text { step } \\
& j=j+s t e p \\
& \text { assert }(i==j)
\end{aligned}
$$

Ultimate Automizer fails to verify assertion within 1 hour

Not monotone: more information led to worse analysis

## The What

Want: invariant generation technique that is
predictable - can make theoretical guarantees about invariant quality (in particular, monotonicity)
precise - assertion verification capability comparable with state-of-the-art software model checkers

## The How

Exploit compositionality to compute transition formula that over-approximates reachability relation of input

$$
\begin{aligned}
& \boldsymbol{T R} \llbracket x:=a \rrbracket \triangleq x^{\prime}=a \wedge \bigwedge y^{\prime}=y \\
& y \neq x \\
& \operatorname{TR} \llbracket \text { if } b \text { then } S_{1} \text { else } S_{2} \rrbracket \triangleq b \wedge \operatorname{TR} \llbracket S_{1} \rrbracket \vee \neg b \wedge \operatorname{TR} \llbracket S_{2} \rrbracket \\
& \mathbf{T R} \llbracket S_{1} ; S_{2} \rrbracket \triangleq \exists \overrightarrow{x^{\prime \prime}} . \operatorname{TR} \llbracket S_{1} \rrbracket\left[\overrightarrow{x^{\prime \prime}} / \overrightarrow{x^{\prime}}\right] \wedge \operatorname{TR} \llbracket S_{2} \rrbracket\left[\overrightarrow{x^{\prime \prime}} / \vec{x}\right] \\
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Can encode loop-free segments without loss of information

```
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Reachability relation of loops needs to be over-approximated
If star operator is monotone, entire analysis in monotone

## This talk

1) Predictable loop summarization using rational vector addition system with resets (Q-VASR)
2) Precision improvement via capturing control flow using $\mathbb{Q}$-VASR with states (Q-VASRS)

## $\mathbb{Q}$-VASR

Key property:

Reachability relation is LIRA-definable and computable in polytime

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$\begin{array}{l}\text { Finite set of transformers. }\left\{\left[\begin{array}{l}x \\ y\end{array}\right] \rightarrow\left[\begin{array}{l}0 \\ y\end{array}\right]+\left[\begin{array}{c}1 \\ -1\end{array}\right]\right. \\ \text { Descrilbes reset/inc to } \\ \text { each dimension }\end{array} \overbrace{\left.\left[\begin{array}{c}x \\ y\end{array}\right] \rightarrow\left[\begin{array}{c}x \\ y\end{array}\right]+\left[\begin{array}{c}10 \\ -1\end{array}\right]\right\}}^{T^{1}}\}$

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Corresponds to transition formula of form
$\bigvee_{i \in T} \bigwedge_{j \in \text { vars }} x_{j}^{\prime}=\underbrace{r_{i j}}_{\{0,1\}} \cdot x_{j}+\underbrace{a_{i j}}_{\mathbb{Q}}$

$$
\begin{aligned}
& T^{1} \begin{array}{l}
\left(x^{\prime}=1 \wedge y^{\prime}=y-1\right) \vee \\
T^{2} \\
\left(x^{\prime}=x+10 \wedge y^{\prime}=y-1\right)
\end{array}
\end{aligned}
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*C. Haase and S. Halfon. Integer vector addition systems with states. in Proc: International Workshop on Reachability Problems (RP '14)

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## Proof Goal:

Amortized constant time operations
Achieved by representing queue as two lists (front and back)

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## dequeue()



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Amortized constant time operations
Achieved by representing queue as two lists (front and back)

## enqueue ( $-\ldots$,



## dequeue()


(2) If Front is


## Functional Queue

procedure enqueue(elt):
back_len := back_len + 1
size := size + 1
mem_ops := mem_ops + 1

Numeric abstraction reasoning about:

- length of back list
- length of front list
- total list size
- number of memory operations
procedure dequeue ():
if (front_len == 0) then
//Reverse back, append to front
while (back_len ! = 0) do
front_len : = front_len + 1
back_len := back_len - 1
mem_ops $=$ mem_ops +3
front_len := front_len - 1
size = size - 1
mem_ops $=$ mem_ops +2


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procedure dequeue():
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Most general harness
procedure harness():
nb_ops := 0
while nondet() do nb_ops := nb_ops + 1 if (size > 0 \&\& nondet()) enqueue() else
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\end{array} \\
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## Functional Queue Inner-Loop

$$
\begin{aligned}
& \text { while (back_len }!=0) \text { do } \\
& \text { Transition formula } \\
& \text { front } \bar{n} \text { : front en }+1 \text { for single iteration } \\
& \text { back_len := back_len - } 1 \\
& \text { mem_ops }=\text { mem_ops }+3 \\
& \text { back_len } \neq 0 \wedge\left(\begin{array}{l}
\text { front_len' }=\text { front_len }+1 \\
\wedge \text { back_len' }=\text { back_len }-1 \\
\wedge \text { mem_ops' }=\text { mem_ops }+3 \\
\wedge \text { size' }=\text { size }
\end{array}\right)
\end{aligned}
$$

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nb_ops := nb_ops + 1
$i f^{-}$(size > 0 \&\& nondet()) enqueue()
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dequeue()
procedure dequeue():

$$
\left.\begin{array}{l}
\text { if (front_len }==0) \text { then } \\
\text { back_len }^{\prime}=0 \wedge \\
\exists k \in \mathbb{N} .\left(\begin{array}{l}
\text { front_len }=\text { front_len }+k \wedge \\
\text { back_len' }=\text { back_len }-k \wedge \\
\text { mem_ops }^{\prime}=\text { mem_ops }+3 k \wedge \\
\text { size }=\text { size }
\end{array}\right.
\end{array}\right) .
$$

front_len := front_len - 1
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mem_ops $=$ mem_ops +2

## Functional Queue

procedure enqueue(elt):

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& \text { size }:=\text { size }+1 \\
& \text { mem_ops }:=\text { mem_ops }+1
\end{aligned}
$$

procedure harness():

$$
\text { nb_ops := } 0
$$

while nondet() do

$$
\text { nb_ops := nb_ops }+1
$$

$$
\text { if (size > } 0 \text { \&\& nondet()) }
$$ enqueue()

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procedure dequeue():
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| back_len $^{\prime}=0 \wedge$ |  |
| ---: | :--- |
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if ${ }^{-}$(size > $0^{-} \& \&$ nondet()) enqueue()
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procedure dequeue():
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```
front_len can increase by arbitrary
value
    back_len + front_len is always
    incremented or decremented by }
```

    front_len := front_len - 1
    size = size - 1
    mem_ops \(=\) mem_ops +2
    front_len can increase by arbitrary
value
back_len + front_len is always
incremented or decremented by 1
enqueue: (back_len + front_len $)++$
dequeue: (back_len + front_len $)--$


## State Space Transformation

$$
\begin{aligned}
& i=0 \\
& \text { while (*) do } \\
& \quad x=x+i+2 \\
& y=y+i \\
& i=j+1
\end{aligned}
$$

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Transition formula for single iteration of loop
$x^{\prime}=x+i+2$
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Not representable as $\mathbb{Q}$-VASR

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# Predictable Analysis using Q-VASR Abstractions 

Key Result:

For any LRA transition formula $F$, we can compute a best $\mathbb{Q}$-VASR abstraction of $F$

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# Computing Best $\mathbb{Q}$-VASR Abstractions 

$\operatorname{DNF}(F)=C_{1} \vee C_{2} \vee \ldots C_{n}$

# Computing Best $\mathbb{Q}$-VASR Abstractions 

 Convert transitionformula to DNF

## Compute best $\mathbb{Q}$-VASR

for each LRA cube

# Computing Best $\mathbb{Q}$-VASR Abstractions 

 Convert transition formula to DNFCompute best $\mathbb{Q}$-VASR for each LRA cube abstraction of all Q-VASR abstractions


# Computing Best $\mathbb{Q}$-VASR Abstractions 

Compute best common abstraction of all $\mathbb{Q}$-VASR abstractions


Step 2 can only compute best $\mathbb{Q}$-VASR for LRA cube

Can use SMT solver to enumerate DNF lazily

## This talk

## 1) Predictable loop summarization using

 rational vector addition system with resets (Q-VASR)2) Precision improvement via capturing control flow using $\mathbb{Q}$-VASR with states (Q-VASRS)

# Q-VASRS Abstractions Example 

```
int x = 0, i = 0
while(*) do
    if(i%2 == 0)
        j = j + 1
    else
        x = x + 1
        i = i + 1
```


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A Best Q-VASR Abstraction Cannot show $2 x \leq i$

# $\mathbb{Q}$-VASRS Abstractions Example 

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\begin{gathered}
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\text { while(*) do } \\
\text { if(i\%2 = }=0) \\
i=j+1 \\
\text { else } \\
x=x+1 \\
i=i+1
\end{gathered}
$$

$$
\left\{\left[\begin{array}{c}
i \\
x
\end{array}\right] \rightarrow\left[\begin{array}{c}
i+1 \\
x
\end{array}\right],\left[\begin{array}{c}
i \\
x
\end{array}\right] \rightarrow\left[\begin{array}{c}
i+1 \\
x+1
\end{array}\right]\right\}
$$

$$
\mathrm{i} \% 2=0
$$

Q-VASRS Abstraction

Q-VASRS Abstraction can prove that loop maintains invariant $2 \mathrm{x} \leq \mathrm{i}$

## $\mathbb{Q}$-VASRS



## $\mathbb{Q}$-VASRS



Predicate Q-VASRS:
Control States are predicates over program variables. Predicates must partition state space.

## Best $\mathbb{Q}$-VASRS Abstractions

Key Result:
Can compute best $\mathbb{Q}$-VASRS abstraction of input LRA formula $F$ with a fixed set of predicates

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Finer set of predicates => potentially more precise abstraction

No best set of predicates, must settle for a good one

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That if $F \vDash G$, then $\operatorname{Predicates}(F)$ is at least as fine as $\operatorname{Predicates}(G)$

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## Solution

Use connected components of topological closure of $\exists x^{\prime} . F$ as predicates

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## Need monotonicity

That if $F \vDash G$, then $\operatorname{Predicates}(F)$ is at least as fine as $\operatorname{Predicates}(G)$

## Solution

Use connected components of topological closure of $\exists x^{\prime} . F$ as predicates


## Evaluation



Results newer than paper version:
$\mathbb{Q}$-VASR and $\mathbb{Q}-$ VASRS faster after optimization Q-VASR passes two more cases after bug fix

Timeout: 300 Seconds per case
SVCOMP-19 restricted to safe integer benchmarks from loops category
Most accurate tool in any given suite does not subsume all others

## Summary

- Developed predictable and compositional program analysis with $\mathbb{Q}-V A S R$
- Extended analysis with $\mathbb{Q}$-VASR with states to capture control flow information
- Shown improvements in both accuracy and speed over state-of-art-tools while providing guarantees about invariant quality

