Loop Acceleration for C Programs ^{3 4}

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December 14, 2017

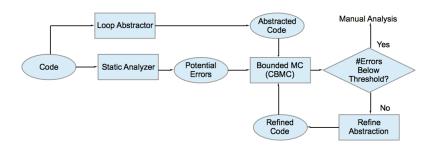
³ Joint Work with Charles M.B. and M. Praveen ⁴ Work supported in part by TCS Research

Outline

- Motivation and Context of Work
- ▶ Problem Definition: Loop Acceleration and our Restrictions
- Reduction of Problem to SMT
- Extension to handle overflows

Context of Work

► A Formal Verification Flow with Loop Abstraction



Abstract Loop Acceleration: Illustration

▶ An example where abstraction is adequate to prove property

```
int x, y = 0;
while (x <0x0ffffffff) {
    x++;
    y = y+2;
    z = x+y;
}
assert(!(y % 2);

int x, y = 0;
int k = *;
Int x0, y0 = 0;
assume (x <0x0ffffffff);
    x = x0 + k*1;
    y = y0 + k*2;
    z = *;
assume (!(x <0x0ffffffff));
assert(!(y % 2);</pre>
```

P. Darke, et. al [DATE2015] [FM2015]

Abstract Loop Acceleration: Illustration

► An example where abstraction is inadequate to prove property

```
int x = 0;
while (x <0x0fffffffff) {
   if (x < 0xfff0) x++;
   else x +=2;
}
assert(!(x % 2);
   int x = 0;
   int k, k1, k2 = *;
   int x0 = 0;
   assume (x < 0x0fffffffff);
   assume (k == k1+k2);
   x = x0 + k1*1 k2*2;
   assume (!(x < 0x0fffffffff));
   assert(!(x % 2);</pre>
```

Abstract Loop Acceleration: Illustration

► An example where abstraction is inadequate to prove property

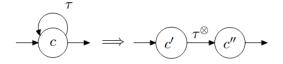
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while (x <0x0fffffffff) {
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   assume (k == k1+k2);
   x = x0 + k1*1 k2*2;
   assume (!(x <0x0fffffffff));
   assert(!(x % 2);</pre>
```

- ► Eliminates about 50% of spurious errors
- ▶ Was able to get 70% 80% of SVCOMP in loop (safe) category

Main Sources of Imprecision in LABMC Acceleration

- Conditional control flow in the body
- Presence Overflows
- ▶ Other forms of assignments: Nonlinear closed forms
- Nested loops

Loop Acceleration: The General Problem



Is as hard as the Model Checking problem!!

What is the sweet spot for effort investment?

Our Problem Definition: Precise Linear Acceleration

► INPUT: A single-loop code with restricted linear expressions

```
int x = x0;
while (x <0x0fffffffff) {
   if (x < 0xfff0) x++;
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Our Problem Definition: Precise Linear Acceleration

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int x = x0;
while (x <0x0fffffffff) {
   if (x < 0xfff0) x++;
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}
assert(!(x % 2);</pre>
```

- Every var assignment is of the form $x = x \pm \delta$
- Loop condition and input assumption is a conjunction of linear inequalities
- Conditional expressions in loop body is a single linear inequality

Problem Definition: Precise Linear Acceleration

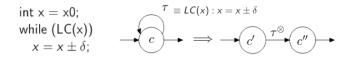
- ▶ OUTPUT: Precise acceleration for the loop as follows
- ► A sequence of guarded block assignments

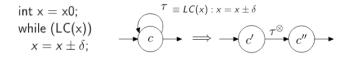
$$(C1(X) \Rightarrow X = X0 + k_1 * \Delta_1) \land (C2(X) \Rightarrow X = X0 + k_2 * \Delta_2) \land ...$$

Problem Definition: Precise Linear Acceleration

- ▶ OUTPUT: Precise acceleration for the loop as follows
- ► A sequence of guarded block assignments

```
ightharpoonup (C1(X) \Rightarrow X = X0 + k_1 * \Delta_1) \land
  (C2(X) \Rightarrow X = X0 + k_2 * \Delta_2) \wedge ...
          int x = 0;
          int k, k1, k2 = *;
          int x0 = 0;
          assume (x < 0x0ffffffff);
          assume (x < 0xfff0); //BlocksPath1
           assume (k1 == 0xfff0 - x0 + 1);
            x = x0 + k1*1;
          assume (!(x < 0xfff0)); //BlocksPath2</pre>
           assume (2*k2 == 0x0ffffffff - x+2);
           x = x + k2*2:
          assume (!(x <0x0ffffffff));
          assert(!(x % 2);
```





▶ Closed Form for value of x after k iterations: $\tau^k = x0 \pm k * \delta$

$$\begin{array}{ccc} \text{int } \mathbf{x} = \mathbf{x0}; & & & & & \\ \text{while (LC(x))} & & & & & \\ & \mathbf{x} = \mathbf{x} \pm \delta; & & & & \\ \end{array} \Longrightarrow \begin{array}{c} \tau \equiv \mathit{LC(x)} : \mathbf{x} = \mathbf{x} \pm \delta \\ & & & & \\ \end{array}$$

- ▶ Closed Form for value of x after k iterations: $\tau^k = x0 \pm k * \delta$
- ► Transitive Closure τ^* (at termination): $x0 \pm k * \delta$ (least) skolem solution to k s.t.
 - $\forall x_0 \exists k : ((LC(x_0) \pm k * \delta) \land \neg (LC(x_0) \pm (k+1) * \delta))$
 - ▶ Loop is non-terminating if no solution exists

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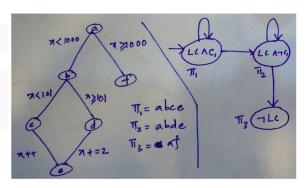
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 - Loop is non-terminating if no solution exists
- ► Naturally extends to multiple variables
- ► Synthesis problem decidable (for linear LC) as this belongs to accelerable loops [S. Bardin et al]
- ▶ Solution for τ^* expressible as a linear function of x_0 and δ

```
#include "assert.h"
                                 71<1000
                                               7 71000
int main(){
                                                               LLAC.
    int x=0:
    while(x<1000){
         if(x<101)
              X++;
                             71(10)
                                         72101
         else
              X+=2;
    assert(!x%2)
                           **
   Fig1: Program with control flow
```

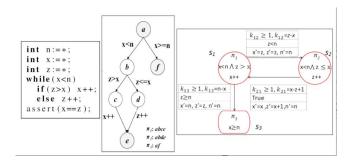
► Loop-head CFG: Each transition is a single iteration of loop

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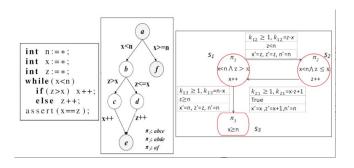
int main(){
    int x=0;
    while(x<1000){
        if(x<101)
            x++;
        else
            x+=2;
    }
    assert(!|x%2)
}
```



- ► Loop-head CFG: Each transition is a single iteration of loop
- ▶ The closed form of τ^* is defined by the set of all feasible looping-path patterns $(\langle \pi_{i_1}, \pi_{i_2}, ..., \pi_{i_n} \rangle)$ that account for all legal terminating execution traces of loop
- ▶ For this example: $\pi_1^*\pi_2^*\pi_3$ defines the entire feasible set



▶ CFG has a cycle besides self loops (in S_1 and S_2)



- ▶ CFG has a cycle besides self loops (in S_1 and S_2)
- ► The following looping-path patterns define all possible loop execution traces
 - $\pi_1^*(\pi_2\pi_1)^*\pi_3$
 - $\pi_2^*(\pi_1\pi_2)^+\pi_1\pi_3$
 - π₃

Problem Reduction to SMT: Loops with Conditionals

```
INPUT: Restricted Linear Program: X = X_0; assume (Init(X_0)); while (LC(X_0) { if c_1(X) X = X \pm \Delta_1; else if c_2(X) ..... else c_n(X) X = X \pm \Delta_n; }
```

Problem Reduction to SMT: Loops with Conditionals

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INPUT: Restricted Linear Program: X = X_0; assume (Init(X_0)); while (LC(X_0) { if c_1(X) X = X \pm \Delta_1; else if c_2(X) ..... else c_n(X) X = X \pm \Delta_n; } Step1 Synthesis of Loop-head CFG: Eliminate all (definitive)
```

▶ $UNSAT(LC(X^{pre}) \land X^{post} = X \pm \Delta_i \land LC(X^{post}) \land BoolCombo_i(c_1(X^{pre}), ..., c_n(X^{pre})) \land$

 $BoolCombo_j(c_1(X^{post}), ..., c_n(X^{post})))$

infeasible transitions

Problem Reduction to SMT: Loops with Conditionals

```
INPUT: Restricted Linear Program: X = X_0; assume (Init(X_0)); while (LC(X_0) { if c_1(X) \ X = X \pm \Delta_1; else if c_2(X) ..... else c_n(X) \ X = X \pm \Delta_n; }
```

- Step1 Synthesis of Loop-head CFG: Eliminate all (definitive) infeasible transitions
 - ▶ $UNSAT(LC(X^{pre}) \land X^{post} = X \pm \Delta_i \land LC(X^{post}) \land BoolCombo_i(c_1(X^{pre}), ..., c_n(X^{pre})) \land BoolCombo_j(c_1(X^{post}), ..., c_n(X^{post})))$
- Step2 Generate a complete set of all looping-path patterns in the CGF b/w every pair of start and terminal states.
 - Can be done performing a depth-first-search of CFG

▶ For each feasible *looping-path* pattern ($<\psi_1^*, \psi_2^*, ..., \psi_I^*>$ generated in Step-2), synthesize $k_{i_0}, k_{i_1}, ...k_{i_l}$ s.t. the following is valid:

$$\forall X_0 \exists k_{i_0}, k_{i_1}, ... k_{i_l} : Init(X_0) \land \\ (c_{i_0}(X_0) \land X_1 = X_0 + k_{i_0} * \Delta_0 \land \neg c_{i_0}(X_1)) \land \\ \\ (c_{i_l}(X_{l-1}) \land X_1 = X_{l-1} + k_{i_l} * \Delta_l \land LC(X_l))$$

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- Claim: A linear solution is guaranteed, if one exists (SAT)
 - ▶ Conditions are all conjunctive linear inequalities ⇒ monotonic

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- Final Loop Summary: Disjunction closed forms for all patterns in the complete set (Step 2)

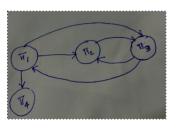
Restrictions on CFG for Decidability

- ► Flattable CFG: The loop-head CFG for our programs must be such that the set of all paths in the CFG can be defined by a finite set of looping patterns each of which is of the following for the following for
- $lackbox{ } <\psi_1^*,\psi_2^*,...,\psi_I^*>$, s.t. $\psi_I=<\pi_{i_0}^{c_1}...\pi_{i_n}^{c_n}>$ for constants c
- A sufficient condition for Flattable CFG:
 - Flat CFG: Every state in the CFG must be part of at most once cycle
 - CFG is reducible to an equivalent flat CFG
- Claim: The acceleration problem for a restricted linear program (defined earlier) with a flattable CFG is decidable [Bardin, Finkel, et.al]

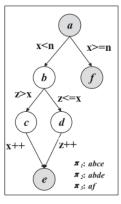
Loops with Non-periodic Cycles: A non-flattable CFG

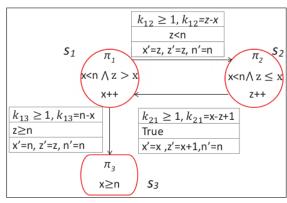
Below is the program where there are multiple interconnected circuits.

```
#include "assert.h"
int main(){
    int k=0; x=0; z=0;
    while(k<100)
         if(k < x)
             k=k+5:
         else
             if(z>x)
                 X++;
             else
                  Z++
    assert(z==x);
```



Solving for the sub-loop iteration numbers: k_i





$$\pi_1^*(\pi_2\pi_1)*\pi_3$$

Solving for the sub-loop iteration numbers: k_i

So.

In the previous example, consider $s_1 \rightarrow s_2$. Let k_{12} be the state counter. So.

$$X'_{k_{12}-1}: x'=x+k_{12}-1, n'=n, z'=z$$

$$X'_{k_{12}}: x'=x+k_{12}, n'=n, z'=z$$

$$cond=(x+k_{12}-1< n)\wedge \overline{(z>x+k_{12}-1)}\wedge (x+k_{12}< n)\wedge \overline{(z\leq x+k_{12})}$$
 So,
$$\phi_{12}: k_{12}=z-x$$

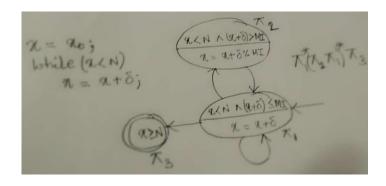
$$\psi_{12} : z < n$$
 $U_{12}(X, k_{12}) = \{z, z, n\}$

The moment we are in s_1 let us say that the initial values of $X = \{x, z, n\}$. After doing the transition $s_1 \to s_2$, the values of the variables will be $\{z, z, n\}$

Some Related Work

- ▶ What kind of loops are effectively and precisely Accelerable?
 - ► Flat acceleration in symbolic model checking [Sébastien Bardin, Alain Finkel, Jérôme Leroux, and Philippe Schnoebelen]
 - PROTEUS: Computing Disjunctive Loop Summary via Path Dependency Analysis. FSE'16 [Xiaofei Xie, Bihuan Chen, Yang Liu, Wei Le Xiaohong Li]
 - Simplifying Loop Invariant Generation Using Splitter Predicates. CAV'11 [Rahul Sharma, Isil Dillig, Thomas Dillig, and Alex Aiken]

Challenges of Handling Overflows



Problem Reduction to SMT: Conditionals with Overflows

 $ightharpoonup \exists I, k_{i_0}, k_{i_1}, ... k_{i_l} : Init(X_0) \land$

$$(LC(X_0) \wedge X_1 = (X_0 + k_0 * \Delta_0 \mod MAXINT) \wedge \neg LC(X_1)) \wedge \dots$$

$$(LC(X_{l-1}) \land X_1 = (X_{l-1} + k_{l-1} * \Delta_l \bmod MAXINT) \land \neg LC(X_l))$$

► Some Observations:

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- ► Some Observations:
 - Nonlinear SMT solving required due to mod operation
 - ▶ For a fixed bit-vector size and for given x_0 , δ it is possible to solve for k_i by model enumeration

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- Some Observations:
 - Nonlinear SMT solving required due to mod operation
 - ▶ For a fixed bit-vector size and for given x_0 , δ it is possible to solve for k_i by model enumeration
 - Can we exploit the cyclic algebraic properties of mod operation to eliminate or optimize enumeration?
 - ▶ Termination can be checked without enumeration
 - Synthesizing a closed form may require minimization search

Conclusions

- Precise loop acceleration is a hard problem
- ► Can recent advances in quantifier elimination and skolem function generation help?
- Are there other sub-classes that admit easy solutions?