Non-linear Arithmetic Solving for Termination Analysis

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Overview of the Talk

• Non-linear constraint solving

- Review of [JAR'12]
- Alternative Max-SMT approach

• Constraint-based termination analysis

- Review of program termination and constraint-based program analysis
- Using Max-SMT for termination analysis
- Implementation and experiments

• Conclusions & future work

Non-linear Constraint Solving

- Problem: Given a quantifier-free formula *F* containing polynomial inequality atoms, is *F* satisfiable?
- Applications: system analysis and verification, ... Here, focus will be on termination of imperative programs
- In Z: undecidable (Hilbert's 10th problem)
- In R: decidable, even with quantifiers (Tarski) But algorithms have prohibitive complexity
- Goal: Can we have a procedure that works "well" in practice?

Review of [JAR'12]

- Our method is aimed at proving satisfiability in the integers (as opposed to finding non-integer solutions, or proving unsatisfiability)
- Basic idea: use bounds on integer variables to linearize the formula
- Refinement: analyze unsatisfiable cores to enlarge bounds (and sometimes even prove unsatisfiability)

Translating into Linear Arithmetic

• For any formula there is an equisatisfiable one of the form

$$F \wedge (\bigwedge_i y_i = M_i)$$

where F is linear and each M_i is non-linear

• Example

$$u^4v^2 + 2u^2vw + w^2 \le 4 \land 1 \le u, v, w \le 2$$

$$\begin{aligned} x_{u^{4}v^{2}} + 2x_{u^{2}vw} + x_{w^{2}} &\leq 4 \land 1 \leq u, v, w \leq 2 \land \\ x_{u^{4}v^{2}} &= u^{4}v^{2} \land x_{u^{2}vw} = u^{2}vw \land x_{w^{2}} = w^{2} \end{aligned}$$

Translating into Linear Arithmetic

- Idea: linearize non-linear monomials with case analysis on some of the variables with finite domain
- Assume variables are in Z
- $F \land x_{u^4v^2} = u^4v^2 \land x_{u^2vw} = u^2vw \land x_{w^2} = w^2$ where F is $x_{u^4v^2} + 2x_{u^2vw} + x_{w^2} \le 4 \land 1 \le u, v, w \le 2$
- Since $1 \le w \le 2$, add $x_{u^2v} = u^2v$ and $w = 1 \rightarrow x_{u^2vw} = x_{u^2v}$ $w = 2 \rightarrow x_{u^2vw} = 2x_{u^2v}$

Translating into Linear Arithmetic

Applying the same idea recursively, the following linear formula is obtained: $x_{\mu^4\nu^2} + 2x_{\mu^2\nu\omega} + x_{\omega^2} \le 4$ $\wedge 1 < u. v. w < 2$ A model can be computed: $\wedge w = 1 \rightarrow x_{\mu^2 \nu \mu \nu} = x_{\mu^2 \nu}$ $\wedge w = 2 \rightarrow x_{\mu^2 \nu w} = 2 x_{\mu^2 \nu}$ $\mu = 1$ $\wedge u = 1 \rightarrow x_{u^2v} = v$ v = 1 $\wedge u = 2 \rightarrow x_{u^2v} = 4v$ w = 1 $\wedge w = 1 \rightarrow x_{w^2} = 1$ $x_{\mu^4\nu^2} = 1$ $\wedge w = 2 \rightarrow x_{w^2} = 4$ $x_{\mu 4} = 1$ $x_{\mu^2 \nu w} = 1$ $\wedge v = 1 \rightarrow x_{\mu^4 \nu^2} = x_{\mu^4}$ $x_{\mu^{2}\nu} = 1$ $\wedge v = 2 \rightarrow x_{u^4v^2} = 4x_{u^4}$ $x_{w^2} = 1$ $\wedge u = 1 \rightarrow x_{u^4} = 1$ $\wedge u = 2 \rightarrow x_{u^4} = 16$

Unsatisfiable Core Analysis

- If linearization achieves a linear formula then we have a sound and complete decision procedure
- If we don't have enough variables with finite domain...
 ... we can add bounds at cost of losing completeness
 We cannot trust UNSAT answers!
- But we can analyze why the CNF is UNSAT: an unsatisfiable core (= unsatisfiable subset of clauses) can be obtained from the trace of the DPLL execution [Zhang & Malik'03]
- If core contains no extra bound: truly UNSAT
 If core contains extra bound: guide to enlarge domains

Unsatisfiable Core Analysis

- $u^4v^2 + 2u^2vw + w^2 \le 3$ cannot be linearized
- Consider $u^4v^2 + 2u^2vw + w^2 \le 3 \land 1 \le u, v, w \le 2$
- The linearization is unsatisfiable:

$$\begin{array}{l} x_{u^{4}v^{2}} + 2x_{u^{2}vw} + x_{w^{2}} \leq 3 \\ \wedge 1 \leq x_{u^{4}v^{2}} \wedge x_{u^{4}v^{2}} \leq 64 \\ \wedge 1 \leq x_{u^{2}vw} \wedge x_{u^{2}vw} \leq 16 \\ \wedge 1 \leq x_{w^{2}} \wedge x_{w^{2}} \leq 4 \\ \wedge 1 \leq u \wedge u \leq 2 \\ \wedge 1 \leq v \wedge v \leq 2 \\ \wedge 1 \leq v \wedge v \leq 2 \\ \wedge 1 \leq w \wedge w \leq 2 \\ \cdots \end{array}$$

• Should decrease lower bounds for *u*, *v*, *w*

An Alternative Max-SMT Approach

- Max-STM(T): Given a set of weighted clauses, find a T-consistent assignment that minimizes cost (= sum of weights) of falsified clauses
- Assume we are given a non-linear formula and have computed a linearization (possibly with extra bounds).

Then we transform the linear formula into a weighted one as follows:

- Clauses C of extra bounds are given finite weights ω_C (soft clauses)
- Rest of clauses are given weight ∞ (hard clauses)
- So we have a Max-SMT(LIA) problem, instead of an SMT(LIA) one
- If found model with null cost, we have a solution
- Else falsified soft clauses show bounds to relax

An Alternative Max-SMT Approach

- There exist simple Branch & Bound algorithms for Max-SMT [Nieuwenhuis & Oliveras, SAT'06], [Cimatti et al., TACAS'10]
- Advantages over the analysis of unsatisfiable cores
 - Max-SMT approach is easier to implement and maintain
 - Leads naturally to an extension to Max-SMT(NIA): Given a set of weighted clauses in NIA, linearize as usual but
 - Original clauses keep their weight
 - Clauses of case splits are given weight ∞
 - Clauses of extra bounds are given weights \u03c6 > W, where W is the sum of the weights of the original soft clauses

So models that violate original clauses are preferred over those violating case splits (that ensure a true model for NA can be reconstructed)

An Alternative Max-SMT Approach

- Example revisited
- $u^4v^2 + 2u^2vw + w^2 \le 3$ cannot be linearized
- Consider $u^4v^2 + 2u^2vw + w^2 \le 3 \land 1 \le u, v, w \le 2$, with extra bounds having weight 1
- Linearization does not have 0-cost solution: optimal solutions have weight 1, e.g. falsifying 1 ≤ w
- Should decrease lower bound of w

Current set of targeted programs:

- Imperative programs: iterative and recursive (ignoring return values)
- Integer variables and linear expressions (other constructions considered unknowns)

```
int gcd (int a, int b) {
  int tmp;
  while ( a \ge 0 \&\& b > 0 ) {
    tmp = b;
    if (a == b) b = 0;
    else {
      int z = a;
      while (z > b) z = b;
      b = z;
    }
    a = tmp;
  }
  return a;
```

As a transition system:



As a transition system:



Proving Termination

- Idea: prove that no transition can be executed infinitely many times.
- In order to discard a transition τ_i we need either:
 - an unfeasibility argument, or
 - a ranking function f over \mathbb{Z} such that

$$\begin{array}{l} \bullet & \tau_i \Longrightarrow f(x_1, \dots, x_n) \ge 0 & (bounded) \\ \bullet & \tau_i \Longrightarrow f(x_1, \dots, x_n) > f(x'_1, \dots, x'_n) & (strict-decreasing) \\ \bullet & \tau_j \Longrightarrow f(x_1, \dots, x_n) \ge f(x'_1, \dots, x'_n) \text{ for all } j & (non-increasing) \end{array}$$

Auxiliary Assertions: Invariants

- We may need invariant assertions to build our termination argument
- We consider inductive invariants:
 - Initiation condition (it holds the first time the location is reached)
 - Consecution condition
 - (it is preserved under every cycle back to the location)

Introduced in [Colon,Sankaranarayanan & Sipma, CAV'03]

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$$c_1x_1+\ldots+c_nx_n+d\leq 0$$

where c_1, \ldots, c_n, d are unknowns

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- Impose initiation and consecution conditions obtaining $\exists \forall \text{ problem}$
- Transform with Farkas' Lemma into ∃ problem over non-linear arith.
- Constraints can be solved with SMT(NA) solver, e.g. Barcelogic.

Following the ideas in [Bradley, Manna & Sipma, CAV'05]: constraint-based invariant gen. (IG) + linear ranking function gen. (RG)

Assume a single location:

- Templates
 - For the invariant: $I = c_1 x_1 + \ldots + c_n x_n + d \leq 0$
 - For the ranking function: $R = r_0 + r_1 x_1 + \ldots + r_n x_n x$
- Constraints
 - Initiation condition on I
 - Consecution condition on I
 - R is non-increasing for all transitions
 - Some transition τ_i can be discarded
 - $I \implies$ unfeasibility of τ_i , or
 - $I \implies$ strict decreasingness and boundedness of τ_i

Although this looks like the way to work, it is not that good in practice:

• Sometimes several invariants needed to generate ranking function Then the problem is unsatisfiable (no solution for ranking function)

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We can do it with Max-SMT

Using Max-SMT to combine IG and RG

We can assign weights to the termination conditions:

1
$$\land \tau_i \implies R \ge 0$$

2 $I \land \tau_i \implies R > R'$
3 $I \land \tau_j \implies R \ge R'$ for all j

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$$1 \wedge \tau_i \implies R \ge 0$$

$$2 \ I \ \land \ \tau_i \implies R > R'$$

(p_1, w_1)where p_1 represents the bound condition (1)(p_2, w_2)where p_2 represents the strict-decreasing condition (2)(p_3, w_3)where p_3 represents the non-increasing condition (3)

Once the problem is encoded in Max-SMT(NA):

- The Max-SMT solver looks for the best solution getting a ranking function if possible
- Otherwise, the weights can guide the search to get invariants and quasi-ranking functions that satisfy as many conditions as possible





Solver finds invariant $b \geq 1$ at l_8 and ranking function b for au_1



Solver finds invariant $b \ge 1$ at l_8 and ranking function b for τ_1



Nothing else can be done, but ...





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 $au_{5.3}: b \ge z, \quad b \ge 0, \quad b = b', \quad a' = tmp, \quad b' = z, \quad tmp' = tmp, \quad z' = z$

We can split τ_5 in three subcases and remove 5.1 by unfeasibility



 $au_{5.3}: b \ge z, b \ge 0, b = b', a' = tmp, b' = z, tmp' = tmp, z' = z$





Now, we cannot find a ranking function but get the invariant $a \ge z$ at l_8 .



Now, we cannot find a ranking function but get the invariant $a \ge z$ at l_8 . Next, again, we only generate the invariant tmp = b at l_8 .



With the invariant $a \ge 0$ at l_8 we have that function a + b fulfills for $\tau_{5,3}$:

 p_1 (bounded) and p_3 (non-increasing) but not p_2 (strict-decreasing)



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 p_1 (bounded) and p_3 (non-increasing) but not p_2 (strict-decreasing) The Max-SMT solver generates a + b





With ranking function a + b we can split $\tau_{5,3}$ into

 $\tau_{5.4}: \tau_{5.3} \land a+b > a'+b' \qquad \tau_{5.5}: \tau_{5.3} \land a+b = a'+b'$



With ranking function a + b we can split $\tau_{5,3}$ into

$$au_{5.4}: au_{5.3} \land a+b > a'+b' au_{5.5}: au_{5.3} \land a+b = a'+b'$$

Then $\tau_{5,4}$ can be removed and $\tau_{5,5}$ simplified: $\tau_{5,5}$: $\tau_{5,3} \land a = a'$





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 $au_0:$ a' = ?, b' = ?, tmp' = ?, z' = ? $au_4: b < z, a' = a, b' = b, tmp' = tmp, z' = z - b$

Solver generates ranking function z - b for τ_4



We are DONE!

Advantages of the method:

- Using Max-SMT we can characterize different ways of progress depending on whether p_1 , p_2 or p_3 are fulfilled.
- Using different weights we can encode which conditions are more important than others.

Implementation and experiments

- We have implemented these techniques
- The prototype reads C code
- Possible answers:
 - YES
 - NO (few cases)
 - Unknown

- Experiments:
 - Benchmarks used in the Termination Competition for Java programs. 111 instances of iterative programs and 41 instances of recursive programs where termination follows from scalar information.
- Results are very promising:
 - Our first implementation is already competitive compared with tools for Java programs that have been developed since many years ago.

Results from the TermComp full-run December 2011:

	Iterative			Recursive		
	YES	NO	MAYBE	YES	NO	MAYBE
AProVE	77	0	36	32	0	9
Costa	64	0	49	28	0	13
Julia	72	21	20	35	0	6
Max-SMT	76	22	18	32	0	9

- Experiments:
 - Programs made by students (can be ugly code). Obtained from an on-line learning environment (Jutge.org). 7924 instances coming from 12 different programming problems.
- Results are very promising:
 - These programs can be considered challenging. Most often they are not the most elegant solution but a working one with many more conditional statements than necessary.

	YES	NO	MAYBE
Max-SMT	6139	15	1770

- Experiments:
 - Benchmarks taken from [Cook et al., CAV'13] coming from Windows device drivers, the Apache web server, the PostgreSQL server, integer approximations of numerical programs from a book on numerical recipes, integer approximations of benchmarks from LLBMC, ... 260 instances known to be terminating.
- Results are very promising:

	YES
Cooperating-T2	245
Terminator	177
T2	189
ARMC	138
AproVE	197
AproVE+Interproc	185
KITTeL	196
Max-SMT	197

Conclusions

- Approach to SMT(NA) that directly extends to Max-SMT(NA)
- Approach to termination analysis relying on Max-SMT
- Our prototype is already a competitive tool

There is a very long list...

- Improve invariant generation techniques.
 (e.g., by combining with abstract interpretation)
- Improve termination of recursive functions.
- Termination in presence of other data types (arrays, etc.)
- Improve the NA solver combining Barcelogic solver with other methods that are much better proving unsatisfiability (like [Jovanovic and De Moura, IJCAR'12])

Thank you!