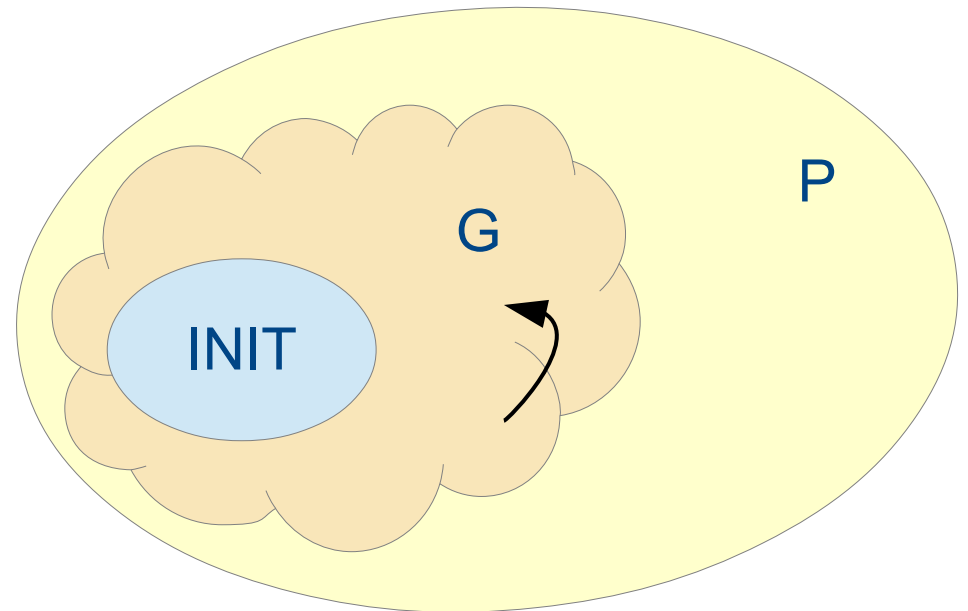


Small Inductive Safe Invariants

Alexander Ivrii, Arie Gurfinkel, Anton Belov

Introduction

- Consider a verification problem (INIT, TR, P)
- In the case that P holds, a Model Checker may produce a **proof** in terms of a **safe inductive invariant**
- A **safe inductive invariant** is a set of states G, satisfying:
 - G contains all the initial states
 - All the transitions from G lead back to G
 - G is contained in the set of states where P holds



Introduction

- Equivalently, a **safe inductive invariant** is a Boolean function G , satisfying:
 - $INIT \Rightarrow G$
 - $TR \wedge G \Rightarrow G'$ (**inductive**)
 - $G \Rightarrow P$ (**safe**)
- Following **IC3**, a recent trend is to produce such an invariant as a conjunction of **many simple lemmas** (such as clauses)
 - $G = C_1 \wedge \dots \wedge C_n$
- A typical invariant may contain **10,000s** of clauses

Introduction

- Our motivation is that **smaller** inductive invariants are **more useful**:
 - They are **relevant in the context of FAIR** [Bradley et al. 2011]
 - The cited paper introduces the problem and presents a solution
 - They **produce better abstractions**
 - A state variable not in the invariant is irrelevant for correctness
 - They **increase user comprehension**
 - They **improve regression verification**
- In this work we **minimize inductive invariants by removing clauses**
 - Look for **minimal** (or small) subsets
 - “**Minimal**” does not mean “**of minimum size**” (the latter is harder)

Problem Statement

- Following the standard (abuse of) notation for **CNFs**, we denote the conjunction of clauses as a set (and vice versa)
- **Minimal Safe Inductive Invariants (MSIS)**: Given a safe inductive invariant $\{C_1, \dots, C_n\}$, find a subset $\{C_{i_1}, \dots, C_{i_k}\}$ of $\{C_1, \dots, C_n\}$, so that:
 - $\{C_{i_1}, \dots, C_{i_k}\}$ is also a safe inductive invariant
 - $\{C_{i_1}, \dots, C_{i_k}\}$ is **minimal** (no proper subset of $\{C_{i_1}, \dots, C_{i_k}\}$ is safe and inductive)
- We want the solution to be **efficient** (ideally the time to minimize a safe inductive invariant should be much smaller than to compute it)

Why finding an MSIS is not simple

- Recall that in particular we need to make sure that
 - $TR \wedge C_{i1} \wedge \dots \wedge C_{ik} \Rightarrow C_{i1}' \wedge \dots \wedge C_{ik}'$
- This query is **non-monotone**: each clause appears both as a premise and a conclusion
 - With fewer clauses, we need to **prove less**, but we can also **assume less**
- For example, it might be that:
 - $\{C_1, C_2, C_3, C_4\}$ is inductive,
 - $\{C_1, C_2, C_3\}$ is not inductive,
 - $\{C_1, C_2\}$ is inductive

Basic MSIS algorithm

- First, we present the approach described in [Bradley et al. 2011]
- The main idea is to tentatively remove a clause, and then to **iteratively** tentatively remove all no longer implied clauses, until:
 - Either **a smaller inductive invariant is obtained**
 - We can restrict to this smaller invariant
 - Or **the property itself is no longer implied**
 - We should restore all the tentatively removed clauses
- Repeat for every clause

Basic MSIS algorithm – Example

- Initially: $\{C_1, C_2, C_3, C_4, C_5, C_6\}$ is a safe inductive invariant for P
- Remove C_1 : $\{C_2, C_3, C_4, C_5, C_6\}$
 - Suppose that C_2' and C_4' are no longer implied
- Remove C_2 and C_4 as well (as they cannot be part of any MSIS of $\{C_2, C_3, C_4, C_5, C_6\}$) : $\{C_3, C_5, C_6\}$
 - Suppose that C_5' is no longer implied
- Remove C_5 as well : $\{C_3, C_6\}$
 - Suppose that C_6 and P are no longer implied
- It follows that C_1 cannot be removed (must be present in every MSIS of $\{C_1, C_2, C_3, C_4, C_5, C_6\}$)
- Restore all removed clauses

Basic MSIS algorithm – Example

- Currently:
 - $\{C_1, C_2, C_3, C_4, C_5, C_6\}$ is a safe inductive invariant for P
 - C_1 cannot be removed
- Remove C_2 : $\{C_1, C_3, C_4, C_5, C_6\}$
 - Suppose that C_3' and C_6' are no longer implied
- Remove C_3 and C_6 as well : $\{C_1, C_4, C_5\}$
 - Suppose that all remaining clauses and P are implied
- It follows that $\{C_1, C_4, C_5\}$ is a smaller safe inductive invariant

Basic MSIS algorithm – Example

- Currently:
 - $\{C_1, C_4, C_5\}$ is a safe inductive invariant for P
 - C_1 cannot be removed
- Proceed with the remaining clauses in a similar fashion

Basic MSIS algorithm

- Denote by **MaxInductiveSubset**(S , P) the procedure that computes the maximum inductive subset of S , aborting if it does not imply P
- Given a safe inductive invariant G for P , in the basic approach we
 - **Iteratively**
 - Choose a not-yet-considered clause C in G
 - Compute $X = \text{MaxInductiveSubset}(G \setminus C, P)$
 - If X is safe (X implies P), then replace G by X
- **Claim**: the described algorithm computes an **MSIS** of G
- Unfortunately, this algorithm **is not efficient**
 - A large number of SAT calls is required (\sim quadratic)
 - Does repeated work

What can we do better?

- Efficiently **under-approximate** an **MSIS**
 - Find clauses that must be present in any **MSIS** of **G**
- Efficiently **over-approximate** an **MSIS**
 - Remove clauses that are not part of some **MSIS** of **G**
- Optimize the basic **MSIS** algorithm
 - Minimizing the amount of wasted work
 - Taking clause dependency into account
- Combine under- and over- approximations with the optimized **MSIS** algorithm

Under-Approximation

- Given a safe inductive invariant $G = \{C_1, \dots, C_n\}$, we say that a clause C_i is **safe necessary** if C_i is present in **every MSIS** of G .
- We exploit the following observations:
 - Given a clause C in G , if $(G \setminus C) \wedge TR \Rightarrow P$ **does not hold** then C is safe necessary
 - Given a clause C in G and a safe necessary clause D (different from C), if $(G \setminus C) \wedge TR \Rightarrow D'$ **does not hold** then C is safe necessary
- The under-approximation algorithm iteratively applies the above two observations until fix-point
- The algorithm can be implemented very efficiently using **an incremental SAT-solver**

Under-Approximation – Example

- Initially:
 - $\{C_1, C_2, C_3, C_4, C_5, C_6\}$ is a safe inductive invariant for P
 - No clauses are marked as necessary
- Check if there is an unmarked clause without which P is not implied
 - Suppose that we find C_4
 - Mark C_4 as necessary
- Check if there is an unmarked clause without which P is not implied
 - Suppose that we find C_5
 - Mark C_5 as necessary
- Check if there is an unmarked clause without which P is not implied
 - Suppose that we find none

Under-Approximation – Example

- Check if there is an unmarked clause without which C_4' is not implied
 - Suppose that we find C_1
 - Mark C_1 as necessary
- Check if there is an unmarked clause without which C_4' is not implied
 - Suppose that we find none
- Check if there is an unmarked clause without which C_5' is not implied
 - Suppose that we find none
- Check if there is an unmarked clause without which C_1' is not implied
 - Suppose that we find none
- Therefore: C_1, C_4, C_5 belong to every MSIS of $\{C_1, C_2, C_3, C_4, C_5, C_6\}$

Under-Approximation

- **Claim:** the described algorithm computes a set of clauses that must be present in every **MSIS** of **G**
(however, it does not compute all such clauses)
- The algorithm makes only a **linear** number of SAT calls
(even in the size of the solution)
- The algorithm can be further improved if some clauses are initially known to be necessary
- For **IC3** proofs, the algorithm is very efficient and usually marks a large number of clauses

Over-Approximation

- Given a safe inductive invariant $G = \{C_1, \dots, C_n\}$ and two subsets A and B of G , we say that A **inductively supports** B (or equivalently that B is **supported** by A) if $TR \wedge A \wedge B \Rightarrow B'$
- **Greedy** compute a safe inductive subset of G as follows:
 - Choose any **minimal** subset A_1 of clauses needed to support P (and any necessary clauses, if known)
 - Choose any **minimal** subset A_2 of clauses needed to inductively support A_1
 - Choose any **minimal** subset A_3 of clauses needed to inductively support A_2
 - ...
 - Stop when the last computed set is empty
- The **over-approximation** is the union of all the sets considered

Over-Approximation

- **Claim:** the described algorithm computes a safe inductive subset of G (however, it is not guaranteed to be minimal)
- The algorithm makes only a **linear** number of **MUS** calls
- The quality and the run-time of the algorithm are greatly improved
 - If we compute **minimal** supporting sets
 - If we follow the presented **recursive** approach
 - Instead of computing a **global** unsatisfiable core as suggested in [Bradley et al. 2011]
 - If we consider all the clauses of A_i together, rather than 1-by-1
 - If some of the clauses are initially marked as necessary

Optimized MSIS algorithm

- An immediate optimization to the basic MSIS algorithm consists of
 - Marking necessary clauses as soon as they are discovered, and
 - Aborting the computation as soon as one of the necessary clauses becomes non-implied
- Given a safe inductive invariant G for P , in the optimized approach we
 - Keep track of necessary clauses N
 - Iteratively
 - Choose a not-yet-considered clause C in $G \setminus N$
 - Compute $X = \text{MaxInductiveSubset}(G \setminus C, P \cup N')$
 - If X is safe, then replace G by X
 - Otherwise, add C to N

Optimized MSIS algorithm – Example

- Consider the previous example:
 - $\{C_1, C_4, C_5\}$ is a safe inductive invariant for P
 - C_1 cannot be removed
- Remove C_4 : $\{C_1, C_5\}$
 - Suppose that C_1' is no longer implied
 - The basic algorithm removes C_1
 - The optimized algorithm aborts immediately
- Remove C_5 : $\{C_1, C_4\}$
 - Suppose that C_4' is no longer implied
 - The basic algorithm removes C_4 (and then possibly C_1 , etc)
 - The optimized algorithm aborts immediately

Optimized MSIS algorithm

- The optimized algorithm is significantly better than the basic algorithm
- Moreover, the optimized algorithm is significantly improved when some of the clauses are initially marked as necessary
- However, the optimized algorithm still requires a quadratic number of SAT queries in the worst case:
 - Queries of the form “which clauses become not implied if certain other clauses are removed?”
 - Each time that we remove a clause C_i from a safe inductive invariant, might need to make a linear number of such queries
 - Might need to process a linear number of clauses

B.I.G. MSIS algorithm

- The **B.I.G.** algorithm makes use the following observation: given a safe inductive invariant G and a clause C
 - Either $G \setminus C$ remains a safe inductive invariant
 - Or C is safe necessary for P or for some other clause in G
- The **B.I.G.** algorithm makes only a linear number of SAT queries
- The technique is inspired by the **B**inary **I**mplication **G**raphs used in SAT-solvers
- Purely by coincidence, **B.I.G.** also represents the authors' initials ;-)

B.I.G. MSIS algorithm – Example

- Initially: $\{C_1, C_2, C_3, C_4, C_5, C_6\}$ is a safe inductive invariant for P
- Remove C_1 : $\{C_2, C_3, C_4, C_5, C_6\}$
 - Suppose that C_4' is no longer implied (and possibly other clauses)
 - We infer: C_1 is needed for C_4
 - Equivalently: if C_4 is in the invariant, then C_1 is in the invariant
 - Denote this graphically by $\{C_1\} \rightarrow \{C_4\}$
- Restore C_1 and remove C_4 : $\{C_1, C_2, C_3, C_5, C_6\}$
 - Suppose that C_5' is no longer implied (and possibly other clauses)
 - We infer: C_4 is needed for C_5
 - Denote this graphically by $\{C_1\} \rightarrow \{C_4\} \rightarrow \{C_5\}$ (note transitivity)
- Restore C_4 and remove C_5

B.I.G. MSIS algorithm – Example

- Currently:

- C_5 is tentatively removed:

$\{C_1, C_2, C_3, C_4, C_6\}$

- Know: $\{C_1\} \rightarrow \{C_4\} \rightarrow \{C_5\}$

- Case I: P and all remaining clauses are still implied

- In this case, can permanently remove the (last) clause C_5

- Know: $\{C_1\} \rightarrow \{C_4\}$

- Make the query for C_4

B.I.G. MSIS algorithm – Example

- Currently:
 - C_5 is tentatively removed: $\{C_1, C_2, C_3, C_4, C_6\}$
 - Know: $\{C_1\} \rightarrow \{C_4\} \rightarrow \{C_5\}$
- Case II: P (or one of known necessary clauses) is not implied
 - In this case, **all** of the clauses C_1, C_4, C_5 are necessary
 - Make the query for some new clause

B.I.G. MSIS algorithm – Example

- Currently:
 - C_5 is tentatively removed: $\{C_1, C_2, C_3, C_4, C_6\}$
 - Know: $\{C_1\} \rightarrow \{C_4\} \rightarrow \{C_5\}$
- Case III: A new clause (for example C_6) is not implied
 - Infer: C_5 is needed for C_6
 - Know: $\{C_1\} \rightarrow \{C_4\} \rightarrow \{C_5\} \rightarrow \{C_6\}$
 - Make the query for C_6

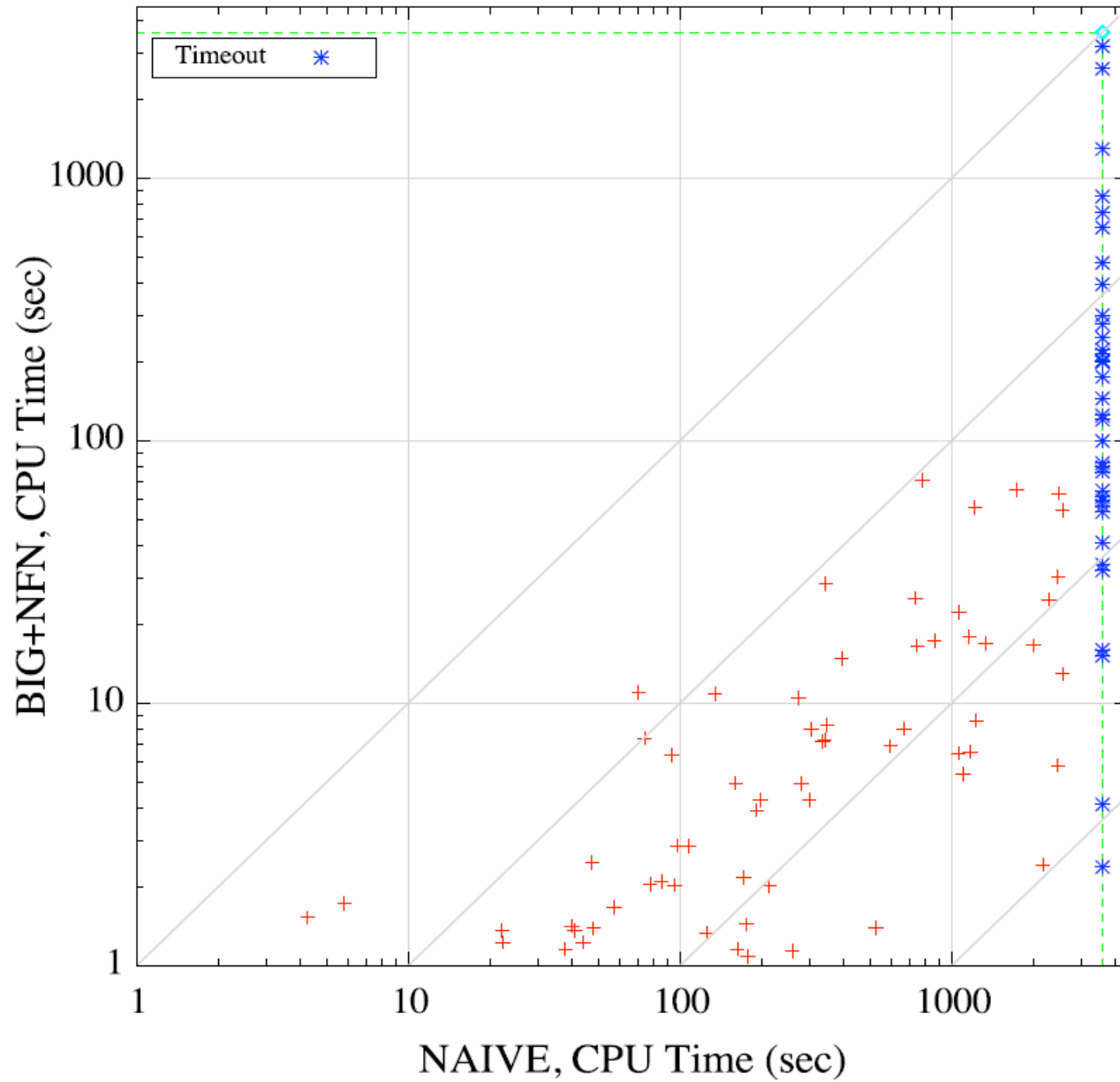
B.I.G. MSIS algorithm – Example

- Currently:
 - C_5 is tentatively removed: $\{C_1, C_2, C_3, C_4, C_6\}$
 - Know: $\{C_1\} \rightarrow \{C_4\} \rightarrow \{C_5\}$
- Case IV: A previous clause (for example C_4) is not implied:
 - Either
 - All clauses between C_4 and C_5 are in the final invariant
 - None of the clauses between C_4 and C_5 are in the invariant
 - Know: $\{C_1\} \rightarrow \{C_4, C_5\}$
 - Make the query for $\{C_4, C_5\}$

Combined MSIS algorithm

- **Experimentally** the following combination of the presented ideas works the best
 - 1) **Run under-approximation**
 - About **70%** of the final MSIS clauses are identified in this stage
 - 2) **Run over-approximation** (with marked necessary clauses)
 - After this stage over-approximates the final MSIS by only **4%**
 - In many cases already produces an MSIS
 - 3) **Run under-approximation** (on the reduced invariant)
 - About **90%** of the final MSIS clauses are identified
 - 4) **Run Optimized MSIS** or **B.I.G. MSIS** on the remaining clauses
 - On average improves the basic MSIS algorithm by **10 to 1000** times

Overall Improvement in Run-Time



Thank You!

Under-Approximation – Implementation

- Introduce an auxiliary variable a_i for every clause C_i of G
- Load $TR \wedge (a_1 \Leftrightarrow C_1) \wedge \dots \wedge (a_n \Leftrightarrow C_n)$ into the solver
- Encode the constraint “at most one out of $\neg a_1, \dots, \neg a_n$ is true”
- Keep unprocessed elements in a queue Q , initially $Q = \{P\}$
- Iteratively:
 - Consider the first element q in Q
 - Solve, passing $\neg q$ as assumptions
 - If SAT:
 - Exactly one of the a_i evaluates to false
 - Mark the corresponding C_i as necessary and set $a_i = \text{true}$
 - Add C_i' to Q
 - If UNSAT:
 - Proceed to the next element in Q