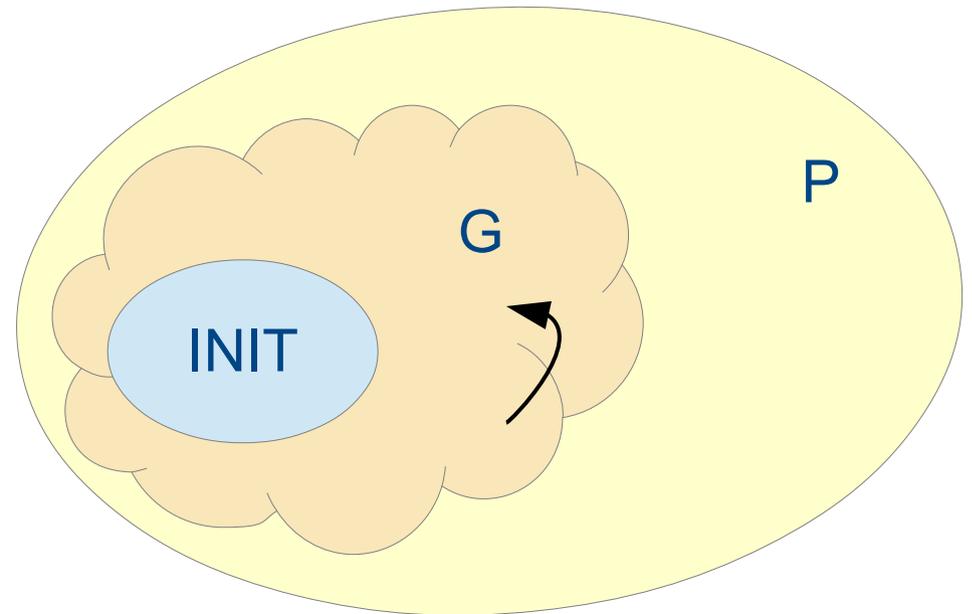


# Small Inductive Safe Invariants

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:
  - $\text{INIT} \Rightarrow G$
  - $\text{TR} \wedge G \Rightarrow G'$  (**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  $C_6$

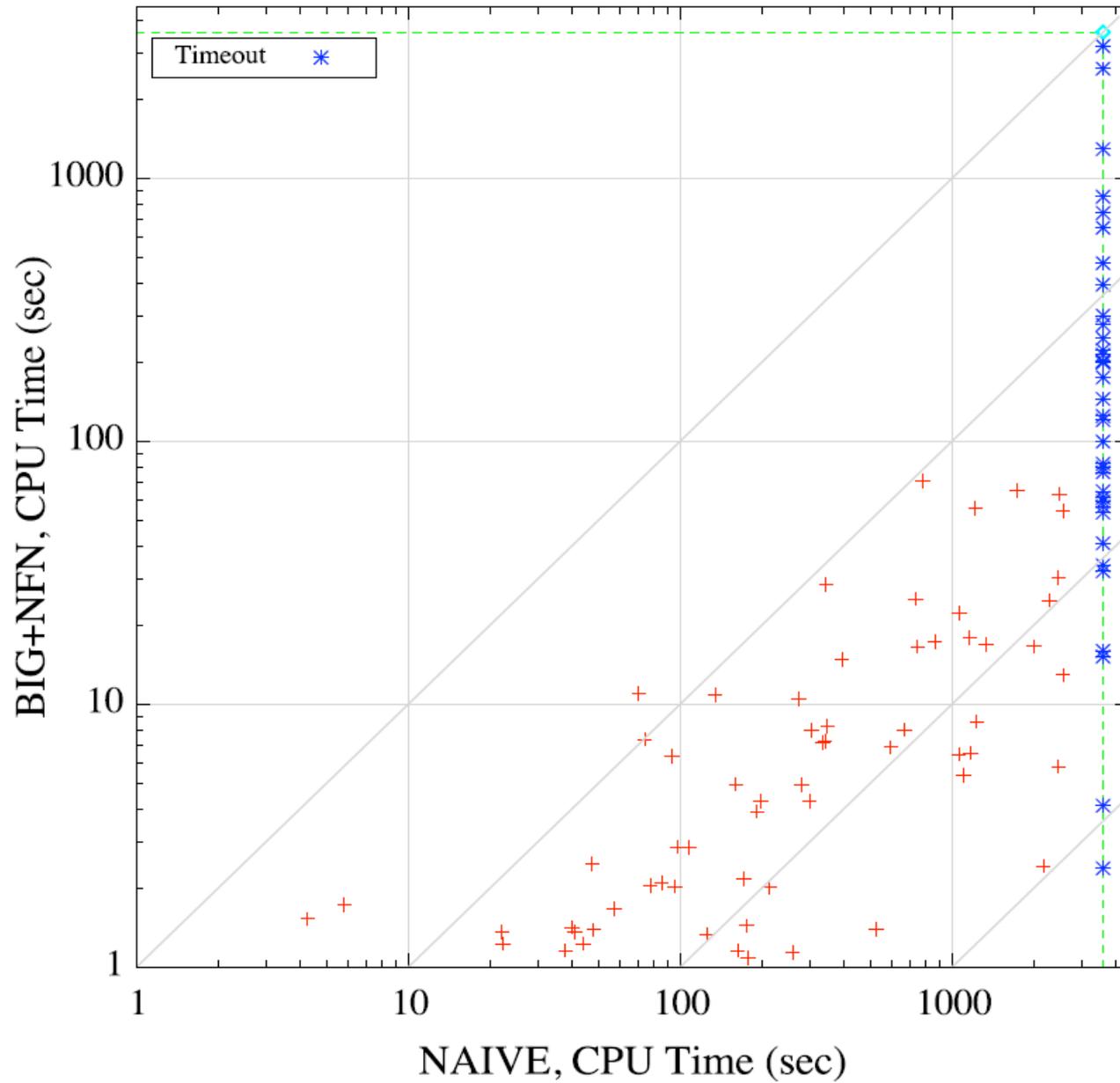
# B.I.G. MSIS algorithm – Example

- 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

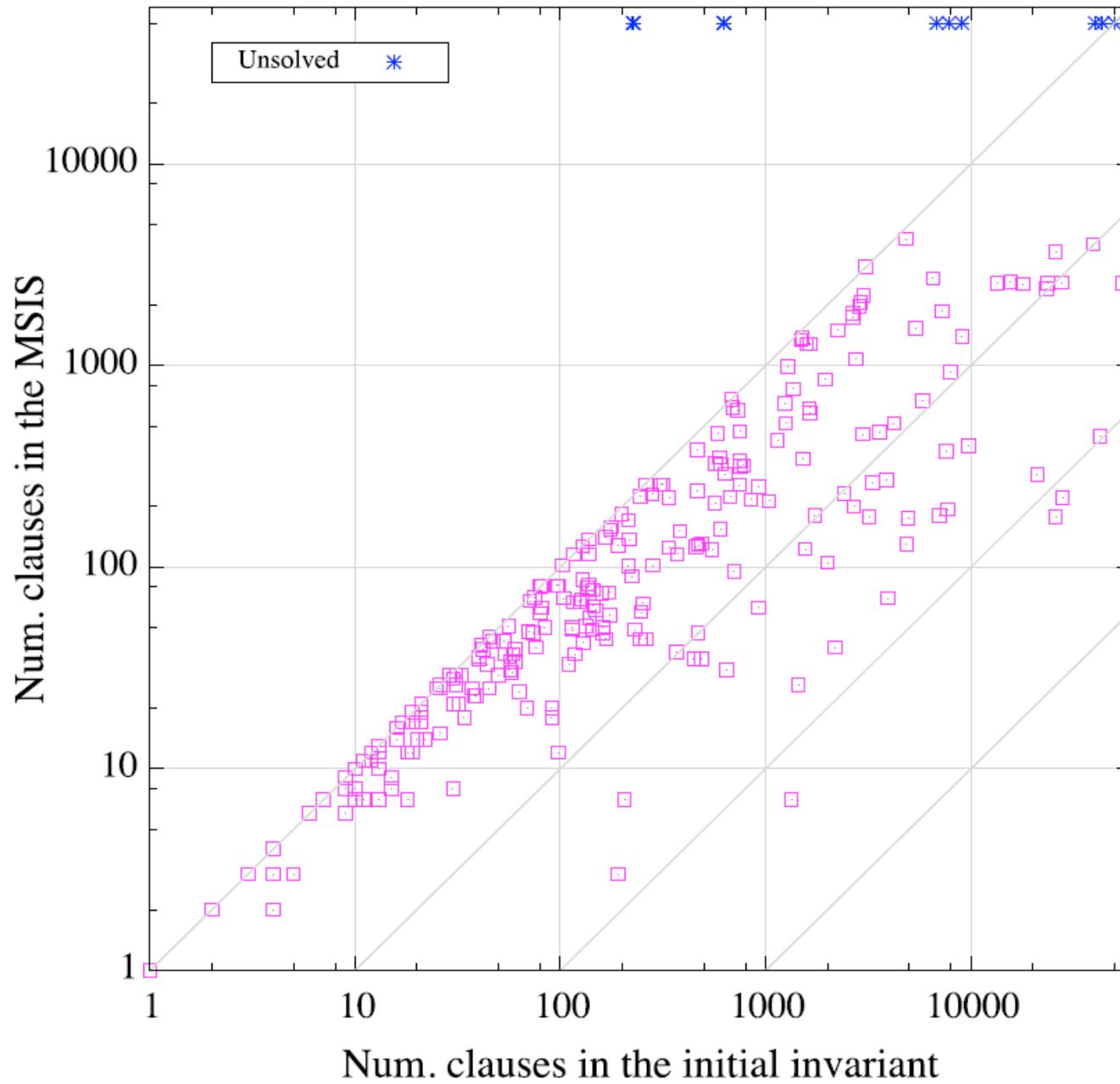
- **Experimentally** the following combination of the presented ideas works the best
  - 1) **Run under-approximation**
    - About **70%** of the final MSIS clauses are identified in this stage
  - 2) **Run over-approximation** (with marked necessary clauses)
    - After this stage over-approximates 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!

# Reduction in the Number of Clauses



# 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  $Q$