

Using Theorem Proving with Algorithmic Techniques for Large-scale System Verification

Ph.D. Oral Proposal

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Motivation

- Design of modern computing systems is error-prone.
 - Simulation and testing cannot catch all the bugs.
 - Bugs discovered after manufacture can be extremely expensive.
 - Can we mathematically prove that systems behave correctly?
 - McCarthy's Dream (1962) [2]:
“Instead of debugging a program, one should prove that it meets its specification, and this proof should be checked by a computer program.”

Formal Verification

Formal Verification is a practical approach to realizing McCarthy's dream.

- Model the executions of the system under interest as formal objects in some logic.
- Prove the desired properties as formal theorems about the models in the logic.
 - Use a (trusted) computer program to assist in the proof generation process.

Formal Verification: Approaches

- Deductive Verification (theorem proving)
 - Logic used is expressive but undecidable.
 - A “theorem prover” is responsible for finding and checking proofs.
 - A user “guides” the theorem prover in proof search.
- Algorithmic Verification (Model Checking)
 - The logic used is decidable.
 - Checking properties in the logic is automatic (at least in principle).
- **Our Goal:** Combine the two approaches “effectively” for verifying large systems.

Why Combine Two Techniques?

Neither technique is effective as is for verification of large systems!

- Deductive Verification:
 - Requires substantial interaction from a “knowledgable” user.
 - The proof might change considerably as the design evolves!
- Algorithmic Verification:
 - Involves an intelligent but exhaustive search of the states of the underlying system.
 - For large systems, these techniques suffer from *state explosion*.

Approach to Combination

- Use Theorem proving to verify the correlation between the “concrete system” and an “abstract model”.
 - The abstract system should have much fewer states.
 - Apply algorithmic verification techniques to verify such abstract models.
- Use the correlation proof and the algorithmic proof to conclude that the concrete system has the desired properties.

Basic Requirements

- Theorem proving aspect of the work must focus on lessening the manual effort.
 - Automatic (possibly heuristic) tools need to work with the theorem prover to help in verification of correlation.
- Algorithmic techniques should be carefully used so that state explosion can be avoided.
- The integration of the two techniques should be sound and efficient.

Domains of Interest

Our principal focus is on verification of implementations of multiprocessor system models:

- Synchronization protocols.
- Pipelined architectures.
- Cache coherence.

ACL2

ACL2 is the theorem proving system and logic that we use for our work.

- ACL2 is a programming language, logic, and a theorem prover for the logic.
 - It is relatively easy to code up decision procedures and tools in the ACL2 language, and possible to verify them.
- ACL2 has been successfully used for verification of large-scale system models.

We later discuss why we need to integrate “external” tools with ACL2 and how we propose to do it.

Modeling Systems

- Computing systems are traditionally modeled operationally in ACL2.
 - “*The meaning of a program is defined by its effect on the state vector.*” (McCarthy, 1962) [2]
- A system model in ACL2 is defined by three functions:
 - A state function `step` that takes a “current state” s and “current input” i and returns the “next state” s' .
 - A predicate `init?` that recognizes if a state is an “initial state”.
 - A function `label` that maps a state s to a collection of “observations” at s .

Framework: Well-founded Refinements

Introduced by Sumners [4] and follows from work by Manolios, Namjoshi and Sumners [1] on WEBs.

- Relates executions of two system models `impl` and `spec` at different levels of abstraction.
 - Define a function `rep` that maps a state of `impl` to a state of `spec`.
 - Show that for every “step” of `impl`, `spec` takes a “step” or “stutters”.
 - Show that the observations are preserved by `rep`, i.e., a state of `impl` and the corresponding `spec` state have “equivalent” observations.
 - Use an argument based on well-foundedness to show that “stuttering” is finite.

Examining Well-founded Refinements

We used the framework to verify several distributed protocols, including a (simplified) Bakery Algorithm.

- Joint work with Rob Sumners.
- Our Observations:
 - We needed to extend the framework to incorporate the notion of “fair executions”.
 - The “fairness constraints” have been subsequently extended and improved by Sumners [5].
 - Most of the human effort is expended in the process of defining and proving “invariants”.
 - Similar conclusion has been reached by others (independently) in verifying computing models.

Inductive Invariants

Inductive invariants are predicates that are true along every “step” of the system model.

- $(\text{init? } s) \Rightarrow (\text{inv } s)$, and
- $(\text{inv } s) \Rightarrow (\text{inv } (\text{step } s \ i))$

For example, assume that a component of a state is a counter that is incremented at each “step” starting from 0. Then, an inductive invariant is:

- *The value stored in the “counter variable”, is a natural number.*

Invariants and Refinement

In well-founded refinement, we need to show that when `impl` takes a “step”, `spec` either takes a “step” or “stutters”.

$$(\text{rep } (\text{impl } s \ i)) = \begin{cases} (\text{rep } s) \\ (\text{spec } (\text{rep } s) \ (\text{pick } s \ i)) \end{cases}$$

If `inv` had been shown to be an inductive invariant, then we can assume `(inv s)` for this proof!

- Determine a predicate `inv` such that:
 1. `inv` is an inductive invariant.
 2. Assuming `inv` you can prove the conditions of well-founded refinement above.

An Invariant Prover

We have implemented a tool with the ACL2 system to generate and prove inductive invariants.

- Joint work with Rob Sumners.
- Basic Idea:
 - Start with a predicate `suff` that is strong enough to prove the conditions of well-founded refinement.
 - “Strengthen” `suff` using a refinement procedure to get an inductive invariant `inv`.
 - Procedure uses *term rewriting*, and lightweight *model checking*.
 - If it cannot strengthen `suff` to `inv` it produces a “counterexample”.

Invariant Strengthening Example

PC	Impl Program	PC	Spec Program
1	j=init;	1	j=init+3;
2	j++;		
3	j++;		
4	j++;		

An invariant we might like is: *The value of j in the impl system in a state where PC is 4 is equal to (init + 2).*

- But this is not an inductive invariant!

Informally, to prove the property as an invariant for the states where PC is 4, we need to know something about the states for which PC is 3.

Example: Continued

We strengthen the invariant by a simple rewriting technique:

- $(PC = 4)$ implies $(j = \text{init} + 2)$ simplifies to:
- $(PC = 3)$ implies $(j = \text{init} + 1)$.

The “simplification” is obtained using ACL2’s simplification engine along with built-in rewrite rules verified by the theorem prover.

- **Note:** Our tool critically depends on the availability of libraries of rewrite rules to help in the simplification process.

Applications of Invariant Prover

We have modeled a fairly complex multiprocessor memory system with caches and directories.

- We can prove well-founded refinement between the memory system and a simple spec.
- The invariant prover is used to generate invariants for this proof.
- Experience shows that the prover is useful.
- Observations:
 - Invariant prover critically depends on built-in libraries of “good” rewrite rules.
 - The system model we have used is at protocol level.

Decision Procedures

- Decision procedures (like model checking) implement some (decidable) logic.
 - Logic of the theorem prover (ACL2) might not be compatible with the logic of the decision procedure.
 - The semantics of LTL model checking is specified in terms of infinite sequence of states.
 - If sequences are modeled as lists, it is easy to prove in ACL2 that all sequences are finite.
- How do we then use decision procedures on abstract models and compose them with the refinement proof relating concrete and abstract models?
- Note: Ruben Gamboa faced similar problems trying to verify square root algorithms in ACL2.

Verifying Decision Procedures

- Decision procedures are programs too!
 - You can model them in ACL2, and prove properties about them.
 - We refer to such theorems about decision procedures as *characterizing theorems*.
 - They tell you exactly what can be derived in ACL2 if a decision procedure returns true or false on some verification problem.
 - The characterizing theorems sometimes turn out to be different from the traditional semantics of the decision procedure, in order to be expressible to the theorem prover.

Verifying Decision Procedures: Feasibility

Is verification of decision procedures to generate characterizing theorems feasible?

- We have verified two decision procedures:
 1. A (simple) compositional model checking procedure.
 2. A (simple) implementation of Generalized Symbolic Trajectory Evaluation, using *strong satisfiability*.
- Observations:
 - The approach is feasible, though non-trivial.
 - Characterizing theorems and their proofs can be very different from traditional ones.
 - But, the proof is “once-off” per decision procedure to be integrated.

Verifying Compositional Model Checking

- Joint work with John Matthews and Mark Tuttle.
- Uses a composition of two (trivial) model checking reductions.
 - Conjunctive reduction
 - Cone of Influence reduction
- The model checking logic used is LTL.
- Observations:
 - The reductions are really trivial, but their verification turned out to be complicated.

Notes on Our Proof

The chief road-block was in specifying the semantics of LTL!

- We could not specify the semantics in terms of infinite paths.
- We used *eventually periodic paths*, that is, infinite paths composed of a finite “prefix” followed by a finite “cycle”.
 - **Known Result:** If there is an infinite path violating an LTL property, then there also exists an eventually periodic path violating the property.
- All proofs had to be cast into this framework. Proofs turned out to be different and sometimes complicated.
- Full details in our paper [3].

Verification of GSTE

Joint work with Warren A. Hunt (Jr).

- GSTE is an efficient lattice-based automatic verification technique.
- Properties are specified not as formulas but in terms of *assertion graphs*.
- Several notions of correctness exist in the GSTE literature, namely *strong*, *terminal*, and *fair* satisfiabilities.
 - Fair satisfiability can express any ω -regular property.
- We verified an algorithm that implements *strong satisfiability* which is normally used for verification of safety properties.

Notes on Our Proof

- Our implementation is not terribly efficient, but we anticipate that a more efficient implementation will be verified along the same lines.
 - More efficient implementations have been done in ACL2 itself by Erik Reeber.
- Proof involves mechanically verifying results from lattices and partial order relations.
- To our knowledge, this is the first mechanical verification of GSTE in a general-purpose theorem prover.
 - We are on the way to verify a slightly more sophisticated implementation that satisfies *terminal satisfiability*.

Characterizing Theorems

- Compositional Model Checking:
 - The compositional algorithm decomposes a verification problem to a collection of “smaller” verification problems. (In this context, a verification problem is a pair $\langle M, \phi \rangle$.)
 - **Theorem** (proved by ACL2): The original verification problem returns true if and only if every verification problem produced by the compositional algorithm returns true.
 - Truth of a verification problem is defined according to model checking semantics for LTL (in terms of eventually periodic paths).

Using Characterizing Theorems

Given specific M and ϕ , we use the theorem to enable us to decompose the verification problem.

- Such verification has actually been done on relatively simple system models.
 - ACL2 has used the characterizing theorem to deduce that it can decompose the problem into a number of simpler problems.
 - Each of the simpler problems could now be verified by a model checker. (Some issues there, we will discuss them later.)
 - Soundness of the combination guaranteed by the soundness of ACL2 and characterizing theorem.

External Tools

Do we need to implement every decision procedure efficiently in ACL2?

- We did NOT implement an efficient model checker, but used an external model checker (Cadence SMV).
- ACL2 does not have a mechanism of hooking an external tool.
 - We had to hack into ACL2 to get the hook.
 - Guarantee from composition: *If the external tool satisfies the characterizing theorem, then the verification using the composite structure is correct.*
- Better approaches for integrating external tools with ACL2 are explored by Erik Reeber.

Summary of Proposals

- Use theorem proving to verify correlation between executions of a “concrete system” and an “abstract model”.
 - Abstract system must be “simpler” than the concrete system under consideration.
 - Design automatic tools to lessen manual effort in this task.
- Design a framework to integrate algorithmic procedures to verify properties of the abstract system.
 - The integration should be sound and efficient.
- Use the composite framework to verify models of multiprocessor systems of practical complexity.

Our Accomplishments

- Examined and extended well-founded refinements.
 - Verified several distributed synchronization protocols, including a (simplified) Bakery Algorithm.
 - Built a tool for checking and strengthening invariants in this framework.
- Used the framework and our tool to verify some properties of a complex multiprocessor cache system.
- Explored the feasibility of integrating decision procedures with ACL2, using characterizing theorems.
 - Verified a compositional model checking algorithm and an algorithm for GSTE.
- (Roughly) integrated external procedures (Cadence SMV) with ACL2.

Proposals

1. Verification of RTL level designs.
 - Such designs are much “lower level” than the models we verified.
 - We are looking to building better libraries for reasoning about such systems.
2. Verification of Pipelined systems.
 - We are working on an approach to verify pipelined machines using well-founded refinements.
 - We are using quantified first-order predicates to simplify the definition of invariants.
3. Building and integrating more efficient decision procedures and external tools.

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