On Theorem Proving for Program Checking

Historical perspective and recent developments

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Program checking and theorem proving

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  Design computer programs that (help to) check whether computer programs satisfy desired properties

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- **Theorem proving**:
  Design computer programs that (help to) check whether formulæ follow from other formulæ

Some motivation for program checking

- Software is everywhere
- Needed: *Reliability*
- Difficult goal: Software may be
  - Artful
  - Complex
  - Huge
  - Varied
  - Old (and undocumented)
  - Less standardized than hardware

Historical roots: program checking

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- Robert W. Floyd. *Assigning meanings to programs*. 1967.
- C. Anthony R. Hoare. *An axiomatic basis for computer programming*. 1969.

Historical roots: theorem proving

- J. Alan Robinson. *A machine oriented logic based on the resolution principle*. 1965.
- G. Robinson and Larry Wos. *Paramodulation and theorem-proving in first-order theories with equality*. 1969.
- Donald E. Knuth and Peter B. Bendix. *Simple word problems in universal algebras*. 1970.

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- John McCarthy, Marvin Minsky, Nathaniel Rochester, Claude Shannon. Proposal for the 1956 Dartmouth Conference on AI.

After four decades of research ...

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- *Software model checking*: BMC, CEGAR, SMT-MC ...

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- *Testing*: automated test case generation, (semi-)automated testing ...
- *Static analysis*: type systems, data-flow analysis, control-flow analysis, pointer analysis, symbolic execution, abstract interpretation ...
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- *Software model checking*: BMC, CEGAR, SMT-MC ...
- *Deductive verification*: weakest precondition calculi, verification conditions generation and proof ...

First summary

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  - Problems of increasing difficulty are attacked by
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- Theorem proving *is* artificial intelligence
- Theorem proving for program checking *is* artificial intelligence

Program checking and theorem proving

Software model checking with predicate abstraction

- Original model checking: finite state machine
- Software: infinitely many states
- How to finitize? Abstraction
- Model check *abstract program*
- Abstract counter-example + formula φ sat iff also concrete counter-example
- Apply theorem prover: if φ unsat refine abstraction with predicates from proof

More theorem proving in model checking

- No abstraction: finite representation by formulæ with quantifiers
- Backward reachability: from set of error states towards initial states
- Does pre-image of error states intersect with set of initial state?
- Did the computation of the pre-image reach a fixed point?
- Reduced to satisfiability of formulæ with *quantifiers*

Deductive verification

- The program is annotated with *assertions*
- Program variables appear in assertions as *free variables* (*constants* in refutational theorem proving)
- Program *state*: an assignment to free variables, hence an *interpretation*

Verifying compiler + theorem prover

- Given: annotated program
- Decomposition into basic paths
- Backward propagation by computing weakest pre-conditions
- Verification condition: the given pre-condition implies the computed one
- If the verification conditions are valid, the annotations are *invariants*
- Otherwise, counter-model is useful to find error in program or annotations

From invariant checking to invariant generation

- Manual annotation of programs is tedious and expensive
- Programmers may appreciate writing functional specifications, not loop invariants, run-time assertions, function call assertions
- Automated annotation
- Automated generation of *valid* annotations, that is, *invariants*

Static analysis for invariant generation

- Given: partially annotated program
- Decomposition into basic paths
- Forward propagation by computing strongest post-conditions
- Does the computed post-condition imply the given one?
- Answer by theorem proving
- If not, update the post-condition

Abstract interpretation

- Trade-off between precision and termination: abstraction
- Abstract interpretation: restrict language of admissible formulæ to an *abstract domain* (syntactically restricted class of formulæ)

Second summary

- There is much theorem proving in SW model checking
- Program checking use theorem prover as back-end reasoner
- Theorem prover must be decision procedure
- Model building as important as proof building
- Abstraction as a way to make satisfiability decidable
- However, problems may contain quantifiers: tension between *expressivity* and *decidability*

Inside theorem proving

Decision procedures

- Davis-Putnam-Logemann-Loveland (DPLL) procedure for SAT
- $\mathcal{T}$-solver: *Satisfiability procedure* for $\mathcal{T}$
  Equality: congruence closure (CC)
- DPLL($\mathcal{T}$)-based SMT-solver: *Decision procedure* for $\mathcal{T} = \bigcup_{i=1}^{n} \mathcal{T}_i$ with
- *Nelson-Oppen combination* of $\mathcal{T}_i$-sat procedures

DPLL

- Propositional logic
- Build candidate model M
- Decision procedure:
  model found: return *sat*;
  failure: return *unsat*
- Depth-first search with backtracking

DPLL

State of derivation: $M \parallel F$

- *Decide*: guess L is true, add it to M (decided literal)
- *UnitPropagate*: propagate consequences of assignment (implied literals)
- *Conflict*: detect $L_1 \vee \ldots \vee L_n$ all false
- *Explain*: unfold implied literals in conflict clause by resolution
- *Learn* conflict clause $C \vee L$
- *Backjump*: when only L assigned at current decision level, jump back to least recent level where C false and L unassigned, undo at least one decision, make L true (implied by $C \vee L$)
- *Unsat*: conflict clause is $\square$ (nothing else to try)

DPLL($\mathcal{T}$)

State of derivation: $M \parallel F$

- $\mathcal{T}$-*Propagate*: add to M an L that is $\mathcal{T}$-consequence of M
- $\mathcal{T}$-*Conflict*: detect that $L_1, \ldots, L_n$ in M are $\mathcal{T}$-inconsistent

Equality sharing method (Nelson-Oppen)

- $\mathcal{T}_i$'s *disjoint*: no shared function/predicate symbols beside $\simeq$
- Mixed terms *separated* by introducing new constants
- $\mathcal{T}_i$-solvers generate and propagate all entailed (disjunctions of) equalities between shared constants
- $\mathcal{T}_i$'s *stably infinite*: every $\mathcal{T}_i$-sat ground formula has $\mathcal{T}_i$-model with infinite cardinality
  (ensures existence of quantifier-free interpolants hence that propagation suffices in completeness proof)

Model-based theory combination

A variant of equality sharing (rule *PropagateEq*):

- Generating (disjunctions of) equalities true in *all* $\mathcal{T}_i$-models consistent with M may be expensive
- If each $\mathcal{T}_i$-solver builds a candidate $\mathcal{T}_i$-model M_i
- Generate and propagate equalities true in M_i
- Optimistic: if equality turns out to be inconsistent, backtrack

[Leonardo de Moura and Nikolaj Bjørner 2007]

Third summary

- SMT-solvers *are* theorem provers
- They do *model building*: both DPLL and CC
- Model-driven or *context-driven* deduction and simplification
- Especially good at theories such as *linear arithmetic* and *bit-vectors*, and integrating them with SAT
- Conceived for SAT and ground problems, not for quantifiers

Superposition-based inference system Γ

- Generic, FOL+=, axiomatized theories
- Deduce clauses from clauses (*expansion*)
- Remove redundant clauses (*contraction*)
- Well-founded *ordering* $\succ$ on terms and literals to restrict expansion and define contraction
- Semi-decision procedure
- No backtracking

Inference system Γ

State of derivation: set of clauses F

- *Resolution*
- *Superposition/Paramodulation*: resolution with equality built-in
- *Simplification*: by well-founded rewriting
- *Subsumption*: eliminate less general clauses
- Other rules: e.g., *Factoring* rules, *Deletion* of trivial clauses

Big engines as little engines

- *Termination* results by analysis of inferences: Γ is $\mathcal{T}$-satisfiability procedure
- Covered theories include: *lists*, *arrays* and *records* with or without extensionality, *recursive data structures*

Joint works with Alessandro Armando, Mnacho Echenim, Michaël Rusinowitch, Silvio Ranise and Stephan Schulz

Also for combination of theories

- **Theorem (Modularity of termination)**: if Γ terminates on $\mathcal{R}_i$-sat problems, it terminates also on $\mathcal{R}$-sat problems for $\mathcal{R} = \bigcup_{i=1}^{n} \mathcal{R}_i$, if the $\mathcal{R}_i$'s are *disjoint* and *variable-inactive*
- Variable-inactivity: no maximal literals of the form $t \simeq x$ where $x \notin Var(t)$ (no paramodulation from variables)
- The only inferences across theories are *superpositions from shared constants* (correspond to equalities between shared constants in equality sharing)

Joint work with Alessandro Armando, Silvio Ranise and Stephan Schulz

Variable inactivity implies stable infiniteness

- **Theorem:** if $\mathcal{R}$ is variable-inactive, then it is stably infinite
- Γ reveals lack of stable infiniteness by generating a *cardinality constraint* (e.g., $y \simeq x \vee y \simeq z$) which is not variable-inactive

Joint work with Silvio Ghilardi, Enrica Nicolini, Daniele Zucchelli 2006

Fourth summary

- Resolution/superposition-based engines good for reasoning on formulæ with quantified variables: *automated* instantiation
- Not for large non-Horn clauses
- Not for theories such as linear arithmetic or bit-vectors
- Unexpected: they are satisfiability-procedures for theories such as lists, arrays, records and their combinations

Big and little engines together: a new theorem proving style

Problem statement

- Decide *satisfiability* of first-order formulæ generated by *verifying compilers* or *static analyzer*
- Satisfiability w.r.t. *background theories*
- With *quantifiers* to write, e.g.,
  - invariants about loops, heaps, data structures ...
  - axioms of *type systems* or *application-specific theories* without decision procedure
- Emphasis on *automation*: prover called by other tools

Typical verification problem

- Background theory $\mathcal{T}$
  - $\mathcal{T} = \bigcup_{i=1}^{n} \mathcal{T}_i$, e.g., linear arithmetic
- Set of formulæ: $\mathcal{R} \cup P$
  - $\mathcal{R}$: set of *non-ground* clauses without $\mathcal{T}$-symbols
  - P: large ground formula (set of ground clauses) with $\mathcal{T}$-symbols
- Determine whether $\mathcal{R} \cup P$ is *satisfiable* modulo $\mathcal{T}$ (Equivalently: determine whether $\mathcal{T} \cup \mathcal{R} \cup P$ is *satisfiable*)

A new theorem proving style

- Given the kind of problem
- Given the complementary strengths of SMT-solvers and resolution/superposition based theorem provers
- Put them together!
- A few approaches
  - DPLL(Γ+$\mathcal{T}$)
  - LASCA ([Konstatin Korovin and Andrei Voronkov 2007-09]), SUP(LA) ([Christoph Weidenbach et al. 2009]) ...

DPLL(Γ+$\mathcal{T}$): integrate Γ in DPLL($\mathcal{T}$)

- **Idea**: literals in M can be premises of Γ-inferences
- Stored as *hypotheses* in inferred clause
- *Hypothetical clause*: $(L_1 \wedge \ldots \wedge L_n) \rhd (L'_1 \vee \ldots L'_m)$ interpreted as $\neg L_1 \vee \ldots \vee \neg L_n \vee L'_1 \vee \ldots \vee L'_m$
- Inferred clauses inherit hypotheses from premises

Joint work with Leonardo de Moura and Chris Lynch
building on top of work by Nikolaj Bjørner and Leonardo de Moura

DPLL(Γ+$\mathcal{T}$) inferences

State of derivation: $M \parallel F$

- *Expansion*: take as pemises *non-ground* clauses from F and $\mathcal{R}$-literals (unit clauses) from M and add result to F
- *Backjump*: remove hypothetical clauses depending on undone assignments
- *Contraction*: as above + *scope level* to prevent situation where clause is deleted, but clauses that make it redundant are gone because of backjumping

Completeness of DPLL(Γ+$\mathcal{T}$)

- *Refutational completeness* of the inference system:
  - from that of Γ, DPLL($\mathcal{T}$) and equality sharing
  - made combinable by variable-inactivity
- *Fairness* of the search plan:
  - depth-first search fair only for ground SMT problems;
  - add *iterative deepening* on *inference depth*

Fifth summary

Use each engine for what is best at:

- DPLL($\mathcal{T}$) works on ground clauses
- Γ not involved with ground inferences and built-in theories
- Γ works on non-ground clauses and ground unit clauses taken from M: also Γ-inferences are context-driven
- Γ works on $\mathcal{R}$-sat problem
- *Completeness*: showed how to integrate Nelson-Oppen built-in theories and variable-inactive axiomatized theories

Decision procedures with speculative inferences

Problematic axioms do occur in relevant inputs

Example:

1. $\neg(x \sqsubseteq y) \vee f(x) \sqsubseteq f(y)$ (*Monotonicity*)
2. $a \sqsubseteq b$ generates by resolution
3. $\{f^i(a) \sqsubseteq f^i(b)\}_{i \geq 0}$

E.g. $f(a) \sqsubseteq f(b)$ or $f^2(a) \sqsubseteq f^2(b)$ often suffice to show satisfiability

Idea: Allow speculative inferences

1. $\neg(x \sqsubseteq y) \vee f(x) \sqsubseteq f(y)$
2. $a \sqsubseteq b$
3. $a \sqsubseteq f(c)$
4. $\neg(a \sqsubseteq c)$

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1. Add $f(x) \simeq x$
2. Rewrite $a \sqsubseteq f(c)$ into $a \sqsubseteq c$ and get $\square$: backtrack!

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1. Add $f(x) \simeq x$
2. Rewrite $a \sqsubseteq f(c)$ into $a \sqsubseteq c$ and get $\Box$: backtrack!
3. Add $f(f(x)) \simeq x$
4. $a \sqsubseteq b$ yields only $f(a) \sqsubseteq f(b)$
5. $a \sqsubseteq f(c)$ yields only $f(a) \sqsubseteq c$
6. Terminate and detect satisfiability

Speculative inferences in DPLL(Γ+$\mathcal{T}$)

- Speculative inference to induce termination on sat input
- What if it makes problem unsat?!
- Detect conflict and backjump:
  - Keep track by adding $\lceil C \rceil \triangleright C$
  - $\lceil C \rceil$: new propositional variable (a "name" for C)
  - Speculative inferences are *reversible*
- Rule *SpeculativeIntro* also bounded by iterative deepening

Example as done by system

1. $\neg(x \sqsubseteq y) \vee f(x) \sqsubseteq f(y)$
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2. Rewrite $a \sqsubseteq f(c)$ into $\lceil f(x) \simeq x \rceil \triangleright a \sqsubseteq c$
3. Generate $\lceil f(x) \simeq x \rceil \triangleright \square$; Backtrack, learn $\neg \lceil f(x) \simeq x \rceil$

Example as done by system

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3. Generate $\lceil f(x) \simeq x \rceil \triangleright \square$; Backtrack, learn $\neg\lceil f(x) \simeq x \rceil$
4. Add $\lceil f(f(x)) \simeq x \rceil \triangleright f(f(x)) \simeq x$
5. $a \sqsubseteq b$ yields only $f(a) \sqsubseteq f(b)$
6. $a \sqsubseteq f(c)$ yields only $f(a) \sqsubseteq f(f(c))$
   rewritten to $\lceil f(f(x)) = x \rceil \triangleright f(a) \sqsubseteq c$
7. Terminate and detect satisfiability

How to get decision procedures

To decide satisfiability modulo $\mathcal{T}$ of $\mathcal{R} \cup P$:

- Find sequence of "speculative axioms" U
- Show that there exists k s.t. k-bounded DPLL(Γ+$\mathcal{T}$) is guaranteed to terminate
  - with *Unsat* if $\mathcal{R} \cup P$ is $\mathcal{T}$-unsat
  - in a state which is not stuck at k if $\mathcal{R} \cup P$ is $\mathcal{T}$-sat

Axiomatizations of type systems

$$\begin{array}{rll} \text{Reflexivity} & x \sqsubseteq x & (1) \\ \text{Transitivity} & \neg(x \sqsubseteq y) \vee \neg(y \sqsubseteq z) \vee x \sqsubseteq z & (2) \\ \text{Anti-Symmetry} & \neg(x \sqsubseteq y) \vee \neg(y \sqsubseteq x) \vee x \simeq y & (3) \\ \text{Monotonicity} & \neg(x \sqsubseteq y) \vee f(x) \sqsubseteq f(y) & (4) \\ \text{Tree-Property} & \neg(z \sqsubseteq x) \vee \neg(z \sqsubseteq y) \vee x \sqsubseteq y \vee y \sqsubseteq x & (5) \end{array}$$

Multiple inheritance: $\mathsf{MI} = \{(1), (2), (3), (4)\}$
Single inheritance: $\mathsf{SI} = \mathsf{MI} \cup \{(5)\}$

Concrete examples of decision procedures

DPLL(Γ+$\mathcal{T}$) with *SpeculativeIntro* adding $f^j(x) \simeq f^k(x)$ for $j > k$ decides the satisfiability modulo $\mathcal{T}$ of problems

- MI $\cup$ P
- SI $\cup$ P

Joint work with Leonardo de Moura and Chris Lynch

Current and future challenges in program checking

- Improve *expressivity*, *scalability*, *precision* and *automation*

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- Improve *expressivity*, *scalability*, *precision* and *automation*
- Integration of model checking and theorem proving
- Integration of abstract interpretation and theorem proving
- Cooperation of verification and synthesis
- Software/hardware border: blurred, evolving

Current and future challenges in theorem proving

- For DPLL($\Gamma+\mathcal{T}$):
  - A top-notch implementation
  - More decision procedures

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- Automation and interaction
- Embedded theorem proving

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and

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