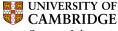
Integrating SAT and SMT Solvers with Interactive Theorem Provers

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Motivation LCF-Style Proof Checking

Motivation

- Interactive theorem proving needs automation.
 - Use SAT solvers to decide formulas of propositional logic.
 - Use SMT solvers to decide SMT formulas.
 - Can we do this without increasing the trusted code base?

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 - Can we do this without increasing the trusted code base?
- SAT and SMT solvers frequently contain bugs. How can we verify their results?
 - Proofs (of unsatisfiability) can be checked independently.
 - Can we keep the proof checker small?

Motivation LCF-Style Proof Checking

LCF-Style Proof Checking

Theorems are implemented as an abstract data type. They can be constructed only through a fixed set of functions provided by this data type.

Each constructor function implements an axiom or inference rule of the logic.

Advanced proof procedures must (ultimately) employ combinations of primitive inferences.

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

An LCF-Style Integration of Proof-Producing SAT Solvers

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Logic

Propositional logic:

- Boolean variables: p, q, ...
- Formulas: $\varphi ::= p \mid \neg \varphi \mid \varphi \lor \varphi \mid \varphi \land \varphi$

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

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Conjunctive normal form (CNF): a conjunction of clauses, where each clause is a disjunction of literals (i.e., possibly negated variables)

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Abstraction from higher-order to propositional logic: replace subterms by Boolean variables, e.g.,

$$(\forall x. P x) \lor \neg (\forall x. P x) \quad \mapsto \quad p \lor \neg p$$

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Resolution

$$\frac{P \cup \{x\}}{P \cup Q} = \frac{Q \cup \{\neg x\}}{Q \cup Q}$$

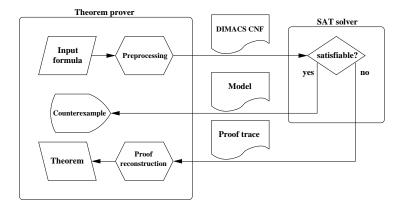
Theorem

Propositional resolution is sound and refutation complete.

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Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

System Overview



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DIMACS CNF Format

DIMACS CNF is the standard input format for SAT solvers.

Example: $(\neg x_1 \lor x_2) \land (\neg x_2 \lor \neg x_3) \land (x_1 \lor x_2) \land (\neg x_2 \lor x_3)$

DIMACS CNF File

c This is just a comment line. p cnf 3 4 -1 2 0 -2 -3 0 1 2 0 -2 3 0

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

zChaff Proof Format

No standard proof format for SAT solvers exists.

Example:

$$\underbrace{ \begin{array}{c} (3) \neg x_2 \lor x_3 \end{array}}_{\begin{array}{c} (3) \neg x_2 \lor x_3 \end{array}} \underbrace{ \begin{array}{c} (2) x_1 \lor x_2 & (0) \neg x_1 \lor x_2 \\ (4) x_2 & (1) \neg x_2 \lor \neg x_3 \end{array}}_{\begin{array}{c} (1) \neg x_2 \lor \neg x_3 \end{array}} \underbrace{ \begin{array}{c} x_1 \lor x_2 & \neg x_1 \lor x_2 \\ x_2 & & & \\ \hline & & & \\ \hline & & & \\ \end{array}}_{\begin{array}{c} (1) \neg x_2 \lor \neg x_3 \end{array}} \underbrace{ \begin{array}{c} x_1 \lor x_2 & \neg x_1 \lor x_2 \\ \hline & & & \\ \hline & & & \\ \end{array}}_{\begin{array}{c} (2) x_1 \lor x_2 & (0) \neg x_1 \lor x_2 \\ \hline & & & \\ \hline \end{array}}$$

zChaff Proof File	
CL: 4 <= 2 0	
VAR: 2 L: 0 V: 1 A: 4 Lits:	4
VAR: 3 L: 1 V: 0 A: 1 Lits:	5 7
CONF: 3 == 5 6	J

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Proof Reconstruction: Basics

The proof is a DAG. Each node represents an inference step and is connected to its premises.

Nodes contain information parsed from the proof file (initially), or the derived theorem (after reconstruction).

A designated root node derives False.

Depth-first (postorder) traversal determines the order of proof reconstruction.

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Representation of SAT Problems

- Bad: use logical connectives \land , \lor
- Good: use sets of clauses and literals

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Representation of SAT Problems

Bad: use logical connectives $\wedge,\,\vee$

Good: use sets of clauses and literals

- The whole CNF problem is assumed: $\{\bigwedge_{i=1}^{k} C_i\} \vdash \bigwedge_{i=1}^{k} C_i$.
- **2** Each clause is derived: $\{\bigwedge_{i=1}^k C_i\} \vdash C_1, \ldots, \{\bigwedge_{i=1}^k C_i\} \vdash C_k$.
- On the as a sequent representation is used:

$$\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \ldots, \overline{p_n}\} \vdash \mathsf{False}.$$

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Representation of SAT Problems

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- On the as a sequent representation is used:

$$\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \ldots, \overline{p_n}\} \vdash \mathsf{False}.$$

The problem is an array of clauses. Clauses are sets of literals.

Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Resolution, LCF-Style

With the sequent representation, resolution is fast:

 $\{\bigwedge_{i=1}^{k} C_{i}, \overline{p_{1}}, \dots, \overline{p_{n}}\} \vdash \mathsf{False}, \{\bigwedge_{i=1}^{k} C_{i}, \overline{q_{1}}, \dots, \overline{q_{m}}\} \vdash \mathsf{False}$



Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Resolution, LCF-Style

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Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Resolution, LCF-Style

- $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \vdash \mathsf{False}, \{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \vdash \mathsf{False}$
 - IMPI: $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \setminus \{x\} \vdash x \Rightarrow \mathsf{False}$
 - **2** IMPI: $\{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \setminus \{\neg x\} \vdash \neg x \Rightarrow \mathsf{False}$



Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

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- $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \vdash \mathsf{False}, \{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \vdash \mathsf{False}$
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 - **2** IMPI: $\{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \setminus \{\neg x\} \vdash \neg x \Rightarrow \mathsf{False}$
 - $INST: \vdash (x \Rightarrow \mathsf{False}) \Rightarrow (\neg x \Rightarrow \mathsf{False}) \Rightarrow \mathsf{False}$



Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Resolution, LCF-Style

- $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \vdash \mathsf{False}, \{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \vdash \mathsf{False}$
 - IMPI: $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \setminus \{x\} \vdash x \Rightarrow \mathsf{False}$
 - **2** IMPI: $\{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \setminus \{\neg x\} \vdash \neg x \Rightarrow \mathsf{False}$
 - $INST: \vdash (x \Rightarrow \mathsf{False}) \Rightarrow (\neg x \Rightarrow \mathsf{False}) \Rightarrow \mathsf{False}$
 - MP: $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \setminus \{x\} \vdash (\neg x \Rightarrow \mathsf{False}) \Rightarrow \mathsf{False}$



Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance

Propositional Resolution, LCF-Style

$$\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \vdash \mathsf{False}, \{\bigwedge_{i=1}^k C_i, \overline{q_1}, \dots, \overline{q_m}\} \vdash \mathsf{False}$$

- IMPI: $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \setminus \{x\} \vdash x \Rightarrow \mathsf{False}$
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- $INST: \vdash (x \Rightarrow \mathsf{False}) \Rightarrow (\neg x \Rightarrow \mathsf{False}) \Rightarrow \mathsf{False}$
- MP: $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}\} \setminus \{x\} \vdash (\neg x \Rightarrow \mathsf{False}) \Rightarrow \mathsf{False}$
- MP: $\{\bigwedge_{i=1}^k C_i, \overline{p_1}, \dots, \overline{p_n}, \overline{q_1}, \dots, \overline{q_m}\} \setminus \{x, \neg x\} \vdash \mathsf{False}$



Propositional Logic, Resolution System Overview Proof Reconstruction Representation of SAT Problems Performance



Evaluation on SATLIB problems:

Problem	Variables	Clauses	Resolutions	zChaff (s)	lsabelle (s)
c7552mul.miter	11282	69529	242509	45	24
бріре	15800	394739	310813	137	55
6pipe_6_000	17064	545612	782903	265	156
7pipe	23910	751118	497019	440	169

Evaluation on pigeonhole instances:

Problem	Variables	Clauses	Resolutions	zChaff (s)	Isabelle (s)
pigeon-9	90	415	73472	1	2
pigeon-10	110	561	215718	6	4
pigeon-11	132	738	601745	24	12
pigeon-12	156	949	3186775	247	68



Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

An LCF-Style Integration of Proof-Producing SMT Solvers

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Satisfiability Modulo Theories

Goal: To decide the satisfiability of (quantifier-free) first-order formulas with respect to combinations of (decidable) background theories.

 $\varphi ::= \mathcal{A} \mid \neg \varphi \mid \varphi \lor \varphi \mid \varphi \land \varphi$

Applications:

- Formal verification
- Scheduling
- Compiler optimization

• . . .

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Example

Theories:

- \mathcal{I} : theory of integers $\Sigma_{\mathcal{I}} = \{ \leq, +, -, 0, 1 \}$
- \mathcal{L} : theory of lists $\Sigma_{\mathcal{L}} = \{=, \text{ hd, tl, nil, cons}\}$
- \mathcal{E} : theory of equality
 - $\Sigma:$ free function and predicate symbols

Problem: Is

 $x \leq y \land y \leq x + hd (\cos 0 nil) \land P (f x - f y) \land \neg P 0$ satisfiable in $\mathcal{I} \cup \mathcal{L} \cup \mathcal{E}$?

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Algorithms

SMT solvers typically use a combination of SAT solving and theory-specific decision procedures.

- DPLL: standard decision procedure for SAT (based on splitting and unit propagation)
- Nelson-Oppen: a decision procedure for the union of decidable theories (using variable abstraction and equality propagation)
- DPLL(T): tight integration of a theory-specific decision procedure with the DPLL algorithm

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SMT-LIB Format

SMT-LIB is the standard input format for SMT solvers.

- LISP-like syntax
- Based on first-order logic
- Modular: different "theories" and "logics"
- Version 2.0 is due real soon now
- http://goedel.cs.uiowa.edu/smtlib/

Greatly helped to unify the field!

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Translation from Higher-Order Logic

	SMT-LIB	Yices		SMT-LIB	Yices
int, real	\checkmark	\checkmark	let	(√)	\checkmark
nat, bool, $ ightarrow$		\checkmark	λ -terms		\checkmark
prop. logic	\checkmark	\checkmark	tuples		\checkmark
equality	\checkmark	\checkmark	records		\checkmark
FOL	\checkmark	\checkmark	data types		\checkmark
HOL		\checkmark	bit vectors	\checkmark	\checkmark
arithmetic	\checkmark	✓			

Abstraction is used to deal with unsupported terms/types.

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Z3's Proof Format

Term language: many-sorted first-order logic

- bool, int, real
- +, -, ·, \lor , \land , \neg , \top , \bot , \forall , \exists , distinct, select, store

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

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Proofs: natural deduction

- 34 axioms and inference rules, from simple (e.g., **mp**) to complex (e.g., **rewrite**, **th-lemma**)
- Contain: inference rule used, pointers to premises, conclusion

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Z3 Proof File (Example)

```
:
#57 := (iff #15 #34)
#58 := [rewrite]: #57
#61 := [monotonicity #58]: #60
:
```

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Proof Reconstruction: Basics

The proof is a DAG. Each node represents an inference step and is connected to its premises.

Nodes contain information parsed from the proof file (initially), or the derived theorem (after reconstruction).

A designated root node derives False.

Depth-first (postorder) traversal determines the order of proof reconstruction.

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

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Same as for SAT!

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Proof Reconstruction: Assumptions, Skolemization

Z3's proofs may contain local (cf. **hypothesis**) and global (cf. **asserted**) assumptions:

- Assume: $\{\varphi\} \vdash \varphi$
- At the very end, we check that all local assumptions have been discharged.

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

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Skolem functions (introduced by **sk**) are given hypothetical definitions in terms of Hilbert's choice operator.

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Rapid Prototyping

About one third of Z3's proof rules perform propositional reasoning. \rightarrow $T_{AUT}P_{ROVE}$

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Rapid Prototyping

About one third of Z3's proof rules perform propositional reasoning. \rightarrow TAUTPROVE

About one third of Z3's proof rules perform relatively simple first-order reasoning. \rightarrow $M\rm ETIS$

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

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Slow!

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Rapid Prototyping

About one third of Z3's proof rules perform propositional reasoning. \rightarrow $T_{AUT}P_{ROVE}$

About one third of Z3's proof rules perform relatively simple first-order reasoning. \rightarrow METIS

Slow!

Speedups of several orders of magnitude can be achieved through specialized implementations that perform the required inferences directly.

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning I

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning I

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning I

- **1** ASSUME: $\bigwedge_{i=1}^{n} \varphi_i \vdash \bigwedge_{i=1}^{n} \varphi_i$
- **2** Repeated CONJE: $\bigwedge_{i=1}^{n} \varphi_i \vdash \varphi_1, \ldots, \bigwedge_{i=1}^{n} \varphi_i \vdash \varphi_n$

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning I

• ASSUME:
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- **2** Repeated CONJE: $\bigwedge_{i=1}^{n} \varphi_i \vdash \varphi_1, \ldots, \bigwedge_{i=1}^{n} \varphi_i \vdash \varphi_n$
- Store these theorems in a red-black tree, indexed by their conclusion.

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning I

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- Store these theorems in a red-black tree, indexed by their conclusion.
- Repeated CONJI: $\bigwedge_{i=1}^{n} \varphi_i \vdash \bigwedge_{i=1}^{n} \varphi_{\pi(i)}$

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning II

Nested disjunctions: dual to nested conjunctions, but trickier

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Optimizations: Propositional and First-Order Reasoning II

Nested disjunctions: dual to nested conjunctions, but trickier

Unit resolution: $\frac{\Gamma \vdash \bigvee_{i \in I} \varphi_i}{\Gamma \cup \bigcup_{i \in J} \Gamma_i \vdash \bigvee_{i \in I \setminus J} \varphi_i}$ similar to nested disjunctions

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Propositional and First-Order Reasoning II

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Quantifier instantiations: determined by first-order term matching

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Theory-Specific Reasoning

Proforma theorems: more than 230 proforma theorems allow about 76% of all terms given to **rewrite** to be proved by instantiation

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Theory-Specific Reasoning

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Theorem caching: theorems proved by **rewrite** and **th-lemma** are cached (indexed by a term net) for later re-use

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Optimizations: Theory-Specific Reasoning

Proforma theorems: more than 230 proforma theorems allow about 76% of all terms given to **rewrite** to be proved by instantiation

Theorem caching: theorems proved by **rewrite** and **th-lemma** are cached (indexed by a term net) for later re-use

Generalization: terms passed to HOL4's arithmetic decision procedures are generalized first (for faster preprocessing)

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Implementation Techniques (Overview)

Primitive inference, proforma theorem: asserted, commutativity, hypothesis, iff-false, iff-true, mp, mp \sim , refl, symm, trans

Combination of primitive inferences/instantiations: and-elim, def-axiom, elim-unused, lemma, monotonicity, nnf-neg, nnf-pos, not-or-elim, pull-quant, quant-inst, quant-intro, sk, unit-resolution

Automated proof procedure: ---

Combination of the above: rewrite, th-lemma

Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

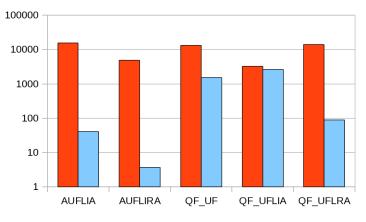
Experimental Results

Logic	Solved (Z3)		Reconstructed		Failed		Factor
	#	Time	#	Time	#T	#Z	
AUFLIA	100	0.180 s	100	0.407 s	0	0	2.3
AUFLIRA	100	0.051 s	97	0.038 s	0	3	0.7
QF_UF	96	2.992 s	74	16.618 s	1	21	5.6
QF_UFLIA	99	0.534 s	92	5.889 s	7	0	11.0
QF_UFLRA	100	0.189 s	100	1.673 s	0	0	8.9



Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Total Run-Times

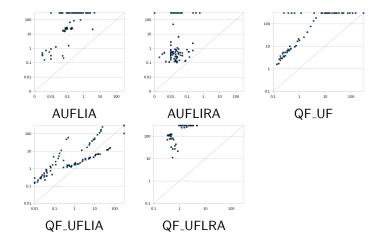


Isabelle/HOL (Böhme, SMT '09) vs. HOL4 (average speedup: 12.4)



Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

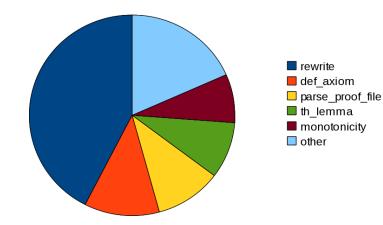
Run-Times: Individual Problems





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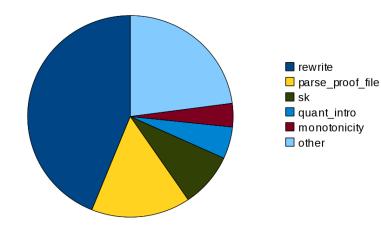
Profiling: AUFLIA





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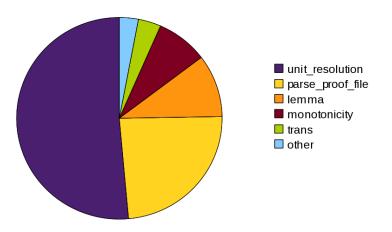
Profiling: AUFLIRA





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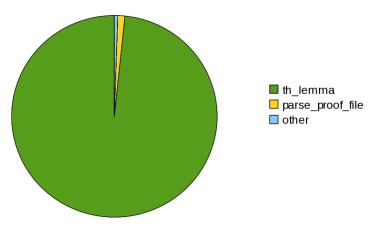
Profiling: QF_UF





Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

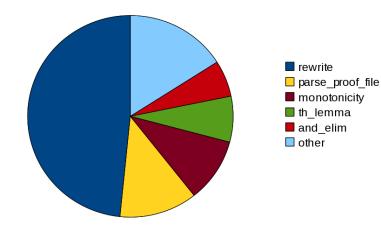
Profiling: QF_UFLIA





Overview Translation from Higher-Order Logic Proof Reconstruction Implementation, Optimizations Experimental Results

Profiling: QF_UFLRA





Conclusions Future Work Questions?

- LCF-style proof checking for SAT and SMT is feasible.
- In LCF-style theorem provers, specialized implementations can be much faster than automated generic proof procedures.
- Z3's proof format is reasonably easy to check. Only rewrite and th-lemma are overly complex.

Conclusions Future Work Questions?

Future Work

- A standard proof format for SAT solvers
- A standard proof format for SMT solvers
- Proof reconstruction for bit vectors
- Parallel proof checking
- Proof compression

Conclusions Future Work Questions?



Thank you for your attention.

Tjark Weber Integrating SAT and SMT Solvers with ITPs