Implementation of Saturating Theorem Provers



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Abstract

We will describe several aspects of the implementation of a saturating, superposition-based theorem prover for first-order logic. We discuss the basic architecture of a prover, and the organization of the actual proof search via the given-clause algorithm. Simplification and redundancy elimination are, in practice, critical, and we describe how these can be integrated into the basic proof procedure.

Another topic is that of terms and clauses, the most basic data objects in any prover, and how they can be implemented. We discuss bottlenecks of naive implementations, as well as the principle and some examples of indexing techniques to overcome these bottlenecks. The presentation concludes with a look at the influence of search heuristics.

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- 2 Basic Data Types
- Saturation Algorithm
- 4 Term and Clause Indexing

5 Search Control



6 Conclusion

Introduction

Black Box View



Black Box View (zoom)





Proving by Saturation

- ► Goal: Show unsatisfiability of a set of clauses S
- Approach:
 - Systematically enrich S with clauses derived via inferences from clauses in S (Saturation)
 - Optionally: Remove/simplify redundant clauses
- Outcome:
 - Derivation of the empty clause \Box (explicit witness of unsatisfiability)
 - ▶ Successful saturation (up to redundancy): *S* is satisfiable
 - ... or infinite sequence of derivations
- Properties:
 - Correctness: Only logical consequences are derived
 - Completeness: Every unsatisfiable S will eventually lead to the derivation of □





 $\begin{array}{ll} \text{(Equality Resolution)}\\ \underline{u \not\simeq v \lor R}\\ \overline{\sigma(R)} & \text{if } \sigma \ = \ \mathrm{mgu}(u,v), \end{array}$

Generating inference rules

- Necessary for completeness
- Increase size of proof state

(Superposition into negative literals)

 $\begin{array}{ll} \underline{s\simeq t\vee S} & \underline{u\not\simeq v\vee R} \\ \overline{\sigma(u[p\leftarrow t]\not\simeq v\vee S\vee R)} & \text{if } \sigma=mgu(u|_p,s), \\ \sigma(s)\not<\sigma(t),\ldots \end{array}$

(Deletion of resolved literals)

$$\frac{s \not\simeq s \lor R}{R}$$

Simplification rules

- Critical for performance
- Reduce size of proof state





(Equality Resolution)

$$\underline{u \neq v \lor R}$$
 if $\sigma = mgu(u, v)$,
 $\underline{\sigma(R)}$...

$$\frac{s \not\simeq s \lor R}{R}$$

(Superposition into negative literals)		(Rewriting of negative literals)		
$s\simeq t\vee S u\not\simeq v\vee R$	$\text{if } \sigma = mgu(u _p, s),$	$s\simeq t u \not\simeq v \vee R$	if $u _p = \sigma(s)$ and	
$\sigma(u[p \leftarrow t] \not\simeq v \lor S \lor R)$	$\sigma(s) \not< \sigma(t), \ldots$	$s \simeq t u[p \leftarrow \sigma(t)] \not\simeq v \lor R$	$\sigma(s) > \sigma(t)$	

$$(Equality Resolution) (Deletion of resolved literals)
$$\underbrace{u \neq v \lor R}{\sigma(R)} if \sigma = mgu(u, v), \underbrace{\frac{s \neq s \lor R}{R}}$$$$

Local (single premise)

- Easy to keep track of
- Cheap to implement

Non-local (multiple premises)

- Harder to keep track of (pairs of clauses!)
- Expensive to implement (find partners)

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$$(Equality Resolution)
$$\frac{u \neq v \lor R}{\sigma(R)} \quad \text{if } \sigma = mgu(u, v), \\
\frac{w \neq v \lor R}{\sigma(R)} \quad \dots \qquad (Deletion of resolved literals)
\frac{s \neq s \lor R}{R}
(Deletion of resolved literals)
$$\frac{s \neq s \lor R}{R}
>90% of generated clauses
(Superposition into negative literals)
$$\frac{s \approx t \lor S \quad u \neq v \lor R}{\sigma(u[p \leftarrow t] \neq v \lor S \lor R)} \quad \text{if } \sigma = mgu(u|_p, s), \\
\frac{s \approx t \quad u \neq v \lor R}{s \approx t \quad u[p \leftarrow \sigma(t)] \neq v \lor R} \quad \text{if } u|_p = \sigma(s) \text{ and } s \approx t \quad u[p \leftarrow \sigma(t)] \neq v \lor R$$$$$$$$

Basic Data Types

Running Example

► Clauses:

- Properties:
 - C_1 is a non-unit Horn clause (at most one positive
 - ▶ C₂ is a positive unit-clause (one literal) literal)
 - Both are purely equational

Running Example

- Clauses:
- Properties:
 - C_1 is a non-unit Horn clause (at most one positive
 - ▶ C₂ is a positive unit-clause (one literal) literal)
 - Both are purely equational
- Basic components:
 - add is a function symbol of arity 2
 - ▶ *s* is a function symbol of arity 1
 - ▶ 0 is a constant (function symbol of arity 0)
 - ▶ X, Y are variables (implicitly universally quantified)
 - \blacktriangleright $\ \simeq$ represents the equality relation
 - $\blacktriangleright \hspace{0.1 cm} \not\simeq \hspace{0.1 cm}$ represents the negated equality relation
 - \blacktriangleright \lor is the disjunctive operator (logical or)

Running Example

- ► Clauses:
 - $C_1 = X \not\simeq 0 \lor add(X, s(Y)) \simeq s(Y)$
- Properties:
 - C_1 is a non-unit Horn clause (at most one positive
 - ▶ C₂ is a positive unit-clause (one literal) literal)
 - Both are purely equational
- Terms and literals:
 - > X, Y, 0 are elementary terms
 - s(X) is a (composite) term
 - ▶ So are add(s(X), Y), s(add(X, Y)), add(X, s(Y)),...
 - $add(s(X), Y) \simeq s(add(X, Y))$ is a positive literal
 - ▶ $add(X, s(Y)) \simeq s(Y)$ is a positive literal
 - ▶ $X \neq 0$ is a negative literal

Enc.	Name	Arity	Remarks
0	-	-	Unused
1	0	0	
2	add	2	
3	S	1	
3			As needed

Signature table

- Associates function symbol with index (small integer)
- Can store additional information
- Implement as array: Fast look-up (O(1)) by index
- ► Add e.g. tree for fast mapping *name→index*

```
typedef struct funccell
\{ /* f_{-}code \text{ is implicit by position in the array } */
  char* name;
   int arity;
   . . .
  Type_p type; /* Simple type of the symbol */
   FunctionProperties properties;
}FuncCell, *Func_p;
typedef struct sigcell
ł
   long size; /* Size of the array */
   FunCode f_count; /* Largest used f_code */
   Func_p f_info; /* The array */
   StrTree_p f_index; /* Back-assoc: symbol=>index */
   . . .
   TypeTable_p type_table;
}SigCell, *Sig_p;
```

Variables

- Variables have no persistent names
 - Each clause is individually universally quantified
 - Scope of a variable name is one clause
- Frequent encoding: Small negative integers
 - ▶ −1 is the first variable in a clause
 - \blacktriangleright -3 is the second variable in a clause
 - -5 is the third variable in a clause
 - ► ...
 - ► Temporary association index↔name for parsing

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 - ► Temporary association index↔name for parsing
- ► This leaves even numbers for alternative variables
 - When two clauses interact via unification, the usually need disjoint variable sets!

Terms are Trees

- Terms are ordered trees
 - Leaves are labeled constants or variables
 - Internal nodes are labeled with non-constant symbols
 - ▶ Node with symbol of arity *n* has *n* children



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E.c.	NI	1	

Enc.	Name
-1	X
-3	Y
-5	Ζ



Term Representations

- ► Cool languages have s-expressions: Terms for free
 - ▶ LISP/Scheme: $add(s(Y), X) \implies (add (s Y) X)$
 - ▶ Python: add(s(Y),X) \implies [2 [3 -3] -1])
- Cooler languages have recursive data types
 - ML/Caml/OCaml
 - Haskell

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- Cooler languages have recursive data types
 - ▶ ML/Caml/OCaml
 - Haskell
- C has pointers

```
typedef struct termcell
ł
   FunCode
                  f_code ;
                                   /* Top symbol of term */
   TermProperties properties; /* Like basic, lhs, top
   int
                                   /* Redundant, but saves
                    arity;
                                      around the signature
                                      time */
                                   /* Pointer to array of
   struct termcell* *args;
                                      arguments */
   . . .
}TermCell, *Term_p, **TermRef;
```

Equations/Inequations/Literals

- Meta-information plus two terms plus next pointer
 - ▶ This reflects the pure equational paradigm of E
 - Alternative: Atoms are terms (with a predicate symbol as the top symbol)
 - Literals have only one term pointer
 - Equality is just a special predicate symbol

```
typedef struct eqncell
{
    EqnProperties properties; /* Positive, maximal, eq. */
    Term_p lterm;
    Term_p rterm;
    ...
    struct eqncell *next; /* For lists of equations */
}EqnCell, *Eqn_p, **EqnRef;
```

. . .

Meta-information plus list of literals

```
typedef struct clause_cell
                        ident; /* Hopefully unique ident
  long
                                      for all clauses created
                                      during a proof run */
                        literals; /* List of literals */
   Eqn_p
   FormulaProperties
                        properties;
   Eval_p
                        evaluations; /* List of evaluations */
   struct clausesetcell* set; /* Is the clause in a set?
   struct clause_cell* pred;
                                /* For clause sets = doubl
   struct clause_cell*
                                 /* linked lists */
                        succ;
```

$C_1 = X \not\simeq 0 \lor add(X, s(Y)) \simeq s(Y)$



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 $C_1 = ???$



 $C_1 = \Box$



Saturation Algorithm

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Simplification rules

- Critical for performance
- Reduce size of proof state
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Requirements for a Saturation Procedure

- Apply all non-redundant generating inferences (in the limit)
 - Necessary for completeness
 - Requires some form of tracking or book-keeping
- Integrate simplification/redundancy elimination
 - Reduces size of proof state and search space
 - Critical for performance
 - ▶ Ideal: No generating inferences involving redundant clauses
- Support search control
 - No blind saturation
 - Suitable choice point(s) for heuristics
- Low overhead
 - Efficient in time
 - Efficient in memory

We start with a blank slate



We represent the proof state *S* by two sets of clauses:

- P holds the processed clauses (originally empty)
- ► U holds the unprocessed clauses (originally all clauses in S)





 Aim: Move everything from U to P



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- Aim: Move everything from U to P
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- Invariant: P is interreduced



- Aim: Move everything from U to P
- Invariant: All generating inferences with premises from P have been performed
- Invariant: P is interreduced
- Clauses added to U are simplified with respect to P

The Given-Clause Loop in Fewer Words

```
while U \neq \{\}
  g = delete_best(U)
  g = simplify(g, P)
  if g == \Box
     SUCCESS, Proof found
  if g is not subsumed by any clause in P (or otherwise redundant w.r.t. P)
     P = P \setminus \{c \in P \mid c \text{ subsumed by (or otherwise redundant w.r.t.) } g\}
     T = \{c \in P \mid c \text{ can be simplified with } g\}
     P = (P \setminus T) \cup \{g\}
     T = T \cup \text{generate}(g, P)
     foreach c \in T
        c = \text{cheap}_{simplify}(c, P)
        if c is not trivial
           U = U \cup \{c\}
SUCCESS, original U is satisfiable
```

Compare and Contrast



while $U \neq \{\}$ $g = \text{delete_best}(U)$ q = simplify(q, P)if $q == \Box$ SUCCESS, Proof found if g is not redundant w.r.t. P $P = P \setminus \{c \in P \mid c \text{ redundant w.r.t. } g\}$ $T = \{ c \in P \mid c \text{ simplifiable with } g \}$ $P = (P \setminus T) \cup \{g\}$ $T = T \cup \text{generate}(q, P)$ foreach $c \in T$ $c = \text{cheap}_{\text{simplify}}(c, P)$ if c is not trivial $U = U \cup \{c\}$ SUCCESS, original U is satisfiable

"You can't handle the truth!"



Term and Clause Indexing

$$(Equality Resolution)$$

$$\frac{u \neq v \lor R}{\sigma(R)} \quad \text{if } \sigma = mgu(u, v),$$

$$\frac{s \neq s \lor R}{R}$$

$$(Superposition into negative literals)$$

$$\frac{s \simeq t \lor S}{\sigma(u[p \leftarrow t] \neq v \lor S \lor R)} \quad \text{if } \sigma = mgu(u|_p, s),$$

$$\frac{s \simeq t \ u \neq v \lor R}{\sigma(s) \neq \sigma(t), \dots} \quad \text{if } u[p \leftarrow \sigma(t)] \neq v \lor R \quad \text{if } u|_p = \sigma(s) = \sigma(s$$

Rewriting

Idea: Replace terms by semantically equal but >-smaller terms

 $\begin{array}{c} (\text{Rewriting of negative literals})\\ \underline{s \simeq t \quad u \not\simeq v \lor R}\\ \hline s \simeq t \quad u[p \leftarrow \sigma(t)] \not\simeq v \lor R \end{array} \quad \begin{array}{c} \text{if } u|_p \ = \ \sigma(s) \ \text{and} \ \sigma(s) \ > \\ \sigma(t) \end{array}$

(Rewriting of positive literals)

 $s \simeq t \quad u \simeq v \lor R$ $s \simeq t \quad u[p \leftarrow \sigma(t)] \simeq v \lor R$

 $\begin{array}{ll} \text{if } u|_p = \sigma(s) \text{ and } \sigma(s) > \\ \sigma(t) \text{ and } u \simeq v \text{ is not maximal in } u \simeq v \lor R \text{ or } u < v \text{ or } \\ p \neq \lambda \end{array}$

Reminder: Rewriting in Action



- Maximally simplify g with respect to all unit equations in P
- Maximally simplify all clauses inserted into U with respect to all unit equations in P
- Check for all clauses in *P* if they can be simplified with *g*

Simplification

Compute normal form of clause C with respect to P:

```
while C is not in normal form:
   for all literals | in C:
      for all terms t in 1:
         for all subterms s of t:
            for all unit clauses l=r in P:
                sigma = match(l,s)
                if sigma and other conditions:
                    replace s by sigma(r)
                else:
                   sigma = match(r,s)
                    if sigma and other conditions:
                       replace s by sigma(1)
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Simplification

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```

... and |P| is growing to $\approx 10^5$ in 300 seconds!

A *term index* is a data structure supporting one or more of the following operations:

- Given a term t, find all terms s from some set S such that
 - s matches t (our current use case)
 - *t* is the subterm to be rewritten
 - Every s is a potential left hand side of a unit clause
 - t matches s
 - s and t can be unified
- Indexing can be...
 - Perfect all retrieved terms s are in the retrieval relation with the query term t
 - Non-perfect index returns a (hopefully small) superset of candidates

Top Symbol Hashing

- Assume $t \equiv \mathbf{f}(t_1, \ldots, t_n)$
- Observation:

►
$$\sigma(t) = s$$
 implies $s \equiv \mathbf{f}(s_1, \dots, s_n)$
► $t = \sigma(s)$ implies $s \equiv \mathbf{f}(s_1, \dots, s_n)$ or $s \equiv x \in V$

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• ... because
$$\sigma$$
 never affects f

If we organize *P* by top symbol of potentially maximal sides of unit clauses, we can avoid most matching attempts!

Fingerprint/Discrimination Tree Indexing

- ► Core idea: Iterated top symbol hashing
 - Traverse tern left-to-right
 - Each function symbol potentially restricts candidates for matching/unification
- ► Index data structure: Trie
 - Convert terms to symbol strings: $add(s(X), Y) \Longrightarrow add s X Y'$
 - Organize strings in a trie
 - Leaf nodes carry original terms and associated date (e.g. originating clause)
- Retrieval:
 - Convert query term into symbols string
 - Follow all compatible branches
 - If a leaf is reached, try candidates stored there
 - ► For *perfect discrimination trees*, only terms compatible with the retrieval relation will be found
 - ► For fingerprint indexing and non-perfect discrimination tree indexing: Candidates must be checked

Impressionist Art



- Non-perfect discrimination tree
- Clause set P from final state of 6th Lusk/Overbeck problem
 - Unit-equational example
 - ► A ring with x * x * x = x is commutative
 - ▶ Historically considered hard
 - Now: 0.2 seconds on this computer

Indexing - Summary

- ► Term indexing can convert the time needed to find inference partners from roughly O(|P|) to O(log(|P|))
- Unification indices speed up paramodulation
 - ▶ E.g. Fingerprint Indexing (Choice for E)
 - E.g. Discrimination Tree Indexing (but ugly unification)
- ► Find-Matching indices speed up forward rewriting
 - E.g. Discrimination Tree Indexing (Choice for E)
 - E.h. Fingerprint Indexing
- ► Find-Matched indices speed up backwards-rewriting
 - E.g. Fingerprint Indexing (Choice for E)
- Subsumption indices speed up subsumption
 - Index clauses, not terms
 - E.g. Feature Vector Indexing

Search Control

- ► Heuristics are crucial for first-order theorem provers
 - Practical experience is clear
 - Proof search happens in an infinite search space
 - Proofs are rare
- Three major choice points
 - Choice of the term ordering
 - Choice of the literal selection strategy
 - Choice of the next given clause

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Basic Clause Selection Heuristics

- ► Basic idea: Clauses ordered by heuristic evaluation
 - Heuristic assigns a numerical value to a clause
 - ▶ Clauses with smaller (better) evaluations are processed first
- ► Example: Evaluation by symbol counting
 - ▶ $|{f(X) \neq a, P(a) \neq $true, g(Y) = f(a)}| = 10$
 - ▶ Motivation: Small clauses are general, \Box has 0 symbols
 - Best-first search
- Example: FIFO evaluation
 - Clause evaluation based on generation time (always prefer older clauses)
 - Motivation: Simulate breadth-first search, find shortest proofs
- ► Combine best-first/breadth-first search
 - E.g. pick 4 out of every 5 clauses according to size, the last according to age

Clause Selection Heuristics in E



- Many symbol-counting variants
 - E.g. Assign different weights to symbol classes (predicates, functions, variables)
 - E.g. Goal directed: lower weight for symbols occurring in original conjecture
 - E.g. ordering-aware/calculus-aware: higher weight for symbols in inference terms
- Arbitrary combinations of base evaluation functions
 - E.g. 5 priority queues ordered by different evaluation functions, weighted round-robin selection

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E can simulate nearly all other approaches to clause selection!

Folklore on Clause Selection/Evaluation

- ► FIFO is obviously fair, but awful *Everybody*
- ► Preferring small clauses is good *Everybody*
- Interleaving best-first (small) and breadth-first (FIFO) is better
 - "The optimal pick-given ratio is 5" Otter
- ► Processing all initial clauses early is good Waldmeister
- Preferring clauses with orientable equation is good DISCOUNT
- Goal-direction is good E

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- Processing all initial clauses early is good Waldmeister
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- ▶ Goal-direction is good E

Can we confirm or refute these claims?
- ► Prover: E 1.9.1-pre
- ▶ 14 different heuristics
 - 13 selected to test folklore claims (interleave 1 or 2 evaluations)
 - Plus modern evolved heuristic (interleaves 5 evaluations)
- ► TPTP release 6.3.0
 - Only (assumed) provable first-order problems
 - 13774 problems: 7082 FOF and 6692 CNF
- Compute environment
 - StarExec cluster: single threaded run on Xeon E5-2609 (2.4 GHz)
 - ▶ 300 second time limit, no memory limit (≥64 GB/core physical)







Meet the Heuristics

Heuristic	Rank	Successes		Successes within 1s	
		total	unique	absolute	of column 3
FIFO	14	4930 (35.8%)	17	3941	79.9%
SC12	13	4972 (36.1%)	5	4155	83.6%
SC11	9	5340 (38.8%)	0	4285	80.2%
SC21	10	5326 (38.7%)	17	4194	78.7%
RW212	11	5254 (38.1%)	13	5764	79.8%
2SC11/FIFO	7	7220 (52.4%)	24	5846	79.7%
5SC11/FIFO	5	7331 (53.2%)	3	5781	78.3%
10SC11/FIFO	3	7385 (53.6%)	1	5656	77.6%
15SC11/FIFO	6	7287 (52.9%)	6	5006	82.5%
GD	12	4998 (36.3%)	12	5856	78.4%
5GD/FIFO	4	7379 (53.6%)	62	4213	80.2%
SC11-PI	8	6071 (44.1%)	13	4313	86.3%
10SC11/FIFO-PI	2	7467 (54.2%)	31	5934	80.4%
Evolved	1	8423 (61.2%)	593	6406	76.1%

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 - In general, only modest advantage for symbol counting (36% FIFO vs. 39% for best SC)
 - Exception: UEQ (32% vs. 63%)

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 - In general, only modest advantage for symbol counting (36% FIFO vs. 39% for best SC)
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- Interleaving best-first/breadth-first is better confirmed
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- Goal-direction is good partially confirmed
 - GD on its own performs similar to SC
 - GD shines in combination with FIFO

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- Evolving Evolved paid off
 - Significantly better than best "naive" heuristic
 - ▶ 10× more unique solutions than second-best

Bonus Material

Version control (git and beyond)

- Value of version control should be self-evident
- But: Have the system print a version number
 - Automatically record the version number with experiments
 - Loudly complain about bug reports without version number (and command line and input file)
- Automatic input format
 - ▶ E supports TPTP-3 (TFA/FOF/CNF), TPTP-2 and LOP
 - Originally only selectable by option (e.g. --tptp3-in)
 - Now automatic default based on first token
 - Automatic format is a big win for convenience

90% of the Iceberg is Under Water



Alternative Term Implementations

- ► Shared terms (e.g. in E)
 - Structure as for normal terms
 - Only one copy of structurally identical terms (f(a, a) has two nodes)
 - Can safe 80% to 99.9% of term nodes
 - Can be used to cache values/results
 - Can use garbage collection (even in C) saves a lot of time and headaches!
- ► Flat terms (e.g. in Waldmeister, Twee)
 - Term is flat string of symbol encoding
 - Very space-efficient
 - Very fast traversal for matching
 - ► Harder to manipulate

Implementing Term Orderings

- Lexicographic Path Ordering (LPO)
 - Based on function symbol precedence
 - Top symbols and symbol bags decide
 - Lexicographic decomposition for identical top symbols
 - LPO has theoretical advantages
 - ► Can orient equations towards sides with more variable occurrences
 - Can orient distributively the right way with $\times > +$: $X \times (Y + Z) \rightarrow (X \times Y) + (X \times Z)$
- Knuth-Bendix-Ordering (KBO)
 - Based on symbols weights and precedence
 - Weight, top symbol, decomposition, variable condition
 - KBO has practical advantages
 - Orients towards syntactically small terms (keeps terms smaller)
 - More efficient to evaluate
- For efficient algorithms: Löchner's Things to Know When Implementing [KL]PO [5, 6]

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- Building a basic saturating prover is not hard
 - Binary resolution+factoring
 - Unification
 - ▶ (Subsumption)
- Building a superposition prover is harder
 - Term orderings
 - Rewriting/Simplification
- Being competitive is harder still
 - Indexing
 - Search heuristics
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The knowledge needed to write a decent prover is now mostly publicly available!

Conclusion II

- Maintenance and evolution are hardest!
 - Users want new features
 - Conference/competition deadline always loom
 - Result: Code clutter
 - Unexpected dependencies and side effects
- ▶ ... so keep it clean!
 - Have a plan
 - Code clean and general building blocks
 - ▶ Take the time to refactor/cleanup (yeah, right!)
 - ... or throw it away and start over
- ▶ Remember: 90% of the iceberg...

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Most Important: Have Fun!

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