

Wedding Boolean Solvers with Superposition: a Societal Reform

Simon Cruanes

École polytechnique and INRIA, 23 Avenue d'Italie, 75013 Paris, France
<https://who.rocq.inria.fr/Simon.Cruanes/>

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- 1 The Talented SAT and Superposition Solvers
- 2 Avatar: a Mighty Combination
- 3 Structural Induction
- 4 Conclusion

SAT solving: the Big Boolean Hammer

- **SAT**: boolean satisfiability
- the archetypical NP-complete problem

SAT solving: the Big Boolean Hammer

- **SAT**: boolean satisfiability
- the archetypical NP-complete problem
- but: good solvers exist (many breakthroughs, competition)
 - **Chaff** (2001) first CDCL solver, 2-watch literals
 - **Minisat** (2003) small, efficient, extensible free solver
 - **Lingeling**
 - **picosat**
 - ...
- gave rise to SMT solvers (alt-ergo, CVC4, Z3, yices...)
- encodings to SAT are good! (e.g. iProver, Satallax...)

Superposition: the King of Equality

In classical first-order theorem proving with \simeq , successful paradigm

- **clausal** calculus
 - literal: $s \simeq t$ or $s \not\simeq t$
 - clause: is a disjunction of literals $l_1 \vee \dots \vee l_n$
 - empty clause means \perp
- **saturation**-based reasoning
 - state: set of clauses
 - *inference rules* deduce new clauses from current ones
 - new clauses are added to the set
 - \rightarrow until fixpoint (**sat**) or \perp (**unsat**)
 - might never terminate if problem is **sat**

Superposition: Example

Let's prove $(p \wedge a \simeq b \wedge f(a) \simeq c) \Rightarrow (p \wedge \exists x f(f(b)) \simeq f(x))$.

We take RPO with $p \succ f \succ a \succ b \succ c$ as ordering.

$$\frac{\frac{a \simeq b \quad f(a) \simeq c}{f(b) \simeq c} \text{sup+} \quad \frac{p \quad \neg p \vee f(f(b)) \not\simeq f(x)}{f(f(b)) \not\simeq f(x)} \text{sup-, eq. res}}{\frac{f(c) \not\simeq f(x)}{\perp} \text{eq. res with } \{x \mapsto c\}} \text{sup-}$$

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unification and ordering are crucial

Superposition

$$\frac{C \vee s \simeq t \quad D \vee u \simeq v}{(C \vee D \vee u[t]_p \simeq v)\sigma}$$

$\sigma = \text{mgu}(u|_p, s)$, ordering conditions

Equality Factoring

$$\frac{C \vee s \simeq s' \vee t \simeq t'}{(C \vee s' \not\simeq t' \vee t \simeq t')\sigma}$$

where $\sigma = \text{mgu}(s, t)$, ordering conditions

Equality Resolution

$$\frac{C \vee s \not\simeq t}{C\sigma}$$

where $\sigma = \text{mgu}(s, t)$, ordering conditions

Superposition: why it works

Superposition is **sound** and **complete** in theory.

In practice, needs many optimizations to work:

- **redundancy** criteria (remove trivial/useless clauses)
- **simplification** rules (infer + delete)
- implementation techniques (term indexing)

Need for Split

Problem with resolution/superposition: clauses *grow*.

Typically:

$$\frac{C \vee I \quad D \vee \neg I}{C \vee D}$$

- for non-unit clauses, conclusion has $m + n - 2$ literals
- huge search space
- heavy clauses (more indexing, memory, etc.)

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Avatar: Split and Cut (1)

Often, clauses have independent **components**.

Components

- components make a partition of the clause
- no variable shared between components
- clause = boolean disjunction of its components

Example

clause $C \stackrel{\text{def}}{=} p(x) \vee q(y) \vee r(y, f(z)) \vee s$

components $\left\{ \begin{array}{l} p(x) \\ q(y) \vee r(y, f(z)) \\ s \end{array} \right.$

hence: $C = (\forall x p(x)) \vee (\forall y \forall z q(y) \vee r(y, f(z))) \vee s$

C is actually a boolean clause!

Avatar: Split and Cut (3)

Idea: **box** clauses (components) into boolean literals

Boxing

- $C \mapsto \llbracket C \rrbracket$: injection into **boolean atoms**
- $\llbracket C \rrbracket$ unique modulo alpha-renaming and AC of \vee
- $\llbracket \neg C \rrbracket \equiv \neg \llbracket C \rrbracket$ (if C has 1 literal)

Avatar: Split and Cut (4)

connect FO clauses and boolean atoms: the **trail**

Trail

- $C \leftarrow \overbrace{b_1 \sqcap b_2 \sqcap \dots \sqcap b_n}^{\text{trail}}$
- means $(b_1 \sqcap b_2 \sqcap \dots \sqcap b_n) \Rightarrow C$
- Theorem: $C \leftarrow \llbracket C \rrbracket$
(proof: left to the reader)

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Usual inferences inherit trails. Example:

$$\frac{C \vee I \leftarrow \Gamma_1 \quad D \vee \neg I \leftarrow \Gamma_2}{C \vee D \leftarrow \Gamma_1 \sqcap \Gamma_2} \text{ resolution}$$

Avatar: Split and Cut (5)

Avatar keeps a set of boolean constraints $S_{\text{constraints}}$.

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Splitting inference rule

If C has components C_1, \dots, C_n (with $n \geq 2$):

$$\frac{C_1 \vee \dots \vee C_n \leftarrow \Gamma}{C_i \leftarrow \llbracket C_i \rrbracket} \text{split}(i), i \in \{1 \dots n\}$$

Also add $\Gamma \Rightarrow_b \bigsqcup_{i=1}^n \llbracket C_i \rrbracket$ to $S_{\text{constraints}}$

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Bottom inference rule

When a clause $\perp \leftarrow b_1 \sqcap \dots \sqcap b_n$ is found:

Add $\neg b_1 \sqcup \dots \sqcup \neg b_n$ to $S_{\text{constraints}}$

Avatar: the Proof Procedure

- regular superposition + SAT-solving on $S_{\text{constraints}}$
- **unsat** iff either one returns **unsat**
- pros:
 - for SAT problem, SAT-solver does all the work \rightarrow fast
 - for ground problems, *unit-superposition*
 - superposition handles smaller clauses
 - resolution divided between (FO) prover and (SAT) solver
- also, can use current SAT interpretation to filter clauses.

Voronkov claims huge performance improvements in Vampire.

Avatar: Example

Let us re-examine the same problem:

$$(p \wedge a \simeq b \wedge f(a) \simeq c) \Rightarrow (p \wedge \exists x f(f(b)) \simeq f(x)).$$

$$\begin{array}{c}
 \frac{\neg p \vee f(f(b)) \not\simeq f(x)}{\neg p \leftarrow \neg \llbracket p \rrbracket} \\
 \vdots \\
 \pi_1 \\
 \\
 \frac{p}{\perp \leftarrow \neg \llbracket p \rrbracket} \text{ sup-, eq. res} \\
 \\
 \frac{\neg p \vee f(f(b)) \not\simeq f(x)}{f(f(b)) \not\simeq f(x) \leftarrow \neg \llbracket f(f(b)) \simeq f(x) \rrbracket} \\
 \vdots \\
 \pi_2 \\
 \\
 \frac{a \simeq b \quad f(a) \simeq c}{f(b) \simeq c} \text{ sup+} \\
 \frac{f(b) \simeq c \quad f(c) \not\simeq f(x) \leftarrow \neg \llbracket f(f(b)) \simeq f(x) \rrbracket}{\perp \leftarrow \neg \llbracket f(f(b)) \simeq f(x) \rrbracket} \text{ sup- eq. res} \\
 \pi_2 \\
 \vdots
 \end{array}$$

$$S_{\text{constraints}} = \{\neg \llbracket p \rrbracket \sqcup \neg \llbracket f(f(b)) \simeq f(x) \rrbracket, \llbracket p \rrbracket, \llbracket f(f(b)) \simeq f(x) \rrbracket\} \text{ unsat}$$

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Poincaré: [l'induction est] «le raisonnement mathématique par excellence».

Herbrand universe calls for **structural induction**:

- powerful enough for data structures
- generalizes induction on naturals
- simpler than general Noetherian induction (uses subterm ordering \triangleleft)

Work inspired from [Kersani&Peltier, 2013].

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Refute the presence of a **minimal model** for a **subset** of all the clauses.

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 - e.g. for naturals, $\kappa(i) = \{0, s(j)\}$ where $j : \text{nat}$ is a fresh constant
 - e.g. it can also be $\kappa(i) = \{0, s(0), s(s(j))\}$
 - e.g. for trees, $\kappa(i) = \{E, N(j_1, t, j_2)\}$ with fresh $t : \text{term}, j_1, j_2 : \text{tree}$

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- 3 pick subset of ind. clauses, call it $S_{\min}(i)$
- 4 for every $t \in \kappa(i)$
 - (i) assert $i \simeq t \leftarrow \llbracket i \simeq t \rrbracket$
 - (ii) assume the model is minimal for $S_{\min}(i)$ and $i \simeq t$
 - (iii) seek contradictionalso add $\bigoplus_{t \in \kappa(i)} \llbracket i \simeq t \rrbracket$ to $S_{\text{constraints}}$ (where \bigoplus is “xor”)
- 5 no minimal model \Rightarrow no model \Rightarrow **unsat**

How to survive Combinatorial Explosions

Problem with previous approach:

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- smells like combinatorial explosion!
- some boolean solvers are *good* at this!

Definition

A *quantified boolean formula* (or *QBF*) is defined as $Q_1x_1 Q_2x_2 \dots Q_nx_n F$ where F is a boolean formula, $\{x_1, \dots, x_n\}$ is the set of boolean variables in F , and every $Q_i \in \{\exists, \forall\}$.

- complexity: PSPACE-complete (expand **and/or** tree)
- but: benefits from advances in SAT
- SAT is QBF with \exists only

Example

$\forall a \exists b \forall c ((a \sqcup b) \sqcap (c \sqcup \neg b))$ is a false QBF.

Problem

Induction → second-order
clauses → first-order

Clause Contexts

Problem

Induction \rightarrow second-order
clauses \rightarrow first-order

Solution

Use **clause contexts**

- a clause with a *hole* \diamond
- noted $C[\diamond]$
- can be *applied* to a term: $C[t] \stackrel{\text{def}}{=} [\diamond \mapsto t]C[\diamond]$

Example

- $\neg p[\diamond]$ to prove $\forall n p(n)$
- $n + s(\diamond) \not\approx s(n + \diamond)$ to prove $\forall n \forall m n + s(m) \simeq s(n + m)$

Keep S_{input} and $S_{\text{min}}()$ Separate

S_{input} clauses deducible from input

- non-inductive clauses are the theory
- inductive clauses \rightarrow find new contexts (heuristic)

$S_{\text{min}}()$ clauses deducible from induction hypothesis only

- induction hypothesis
- minimality assumptions
- saturate with inference rules

\rightarrow do not mix them!

Keep S_{input} and $S_{\text{min}}()$ Separate (cont'd)

- Use a special *marker*, `input`, to annotate clauses from S_{input}
- remember, trails are inherited in inferences
- Redundancy criterion blocks interactions between S_{input} and $S_{\text{min}}()$

$$\frac{C \leftarrow \text{input}, \llbracket D[\diamond] \in S_{\text{min}}(i) \rrbracket, \Gamma}{\top}$$

Example

- $(\neg p(n) \leftarrow \text{input}) \in S_{\text{input}}$ (provable)
- $(\neg p(n) \leftarrow \llbracket \neg p[\diamond] \in S_{\text{min}}(n) \rrbracket) \in S_{\text{min}}()$ (ind. hypothesis)

Express Induction Hypothesis

For inductive constant i , set of all contexts is $S_{\text{cand}}(i)$

- $C[i] \leftarrow \overbrace{[[C[\diamond] \in S_{\text{min}}(i)]]}^{\text{in the subset?}} \sqcap \overbrace{[[\text{init}(C[\diamond], i)]]}^{\text{provable from } S_{\text{input}}?}$
- boolean valuation of atoms $[[C[\diamond] \in S_{\text{min}}(i)]]$ determine subset $S_{\text{min}}(i)$
- $[[\text{init}(C[\diamond], i)]]$ added to QBF when $C[i]$ subsumed by some clause

- model minimal for $S_{\min}(i) \neq \emptyset$
 $\Rightarrow \exists C[\diamond] \in S_{\min}(i)$ with $\begin{cases} C[i] \text{ true} \\ C[\diamond] \text{ false on a smaller term } j \end{cases}$

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 - “winning” witness $C[j]$ annotated with $\llbracket \text{minimal}(C[\diamond], i, j) \rrbracket$
 - for each $C[\diamond]$ and $j \triangleleft t \in \kappa_{\downarrow}(i)$ (subterm of inductive type):
 $\neg C[j] \leftarrow \llbracket \text{minimal}(C[\diamond], i, j) \rrbracket \cap \llbracket C[\diamond] \in S_{\min}(i) \rrbracket \cap \llbracket i = t \rrbracket$
(then reduced to CNF)
- means: “ $C[\diamond]$ is the context for which $S_{\min}(i)$ is minimal”
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- in QBF, disjunction that forces the *choice* of $C[\diamond]$ in $S_{\min}(i)$

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QBF is needed to *enumerate* the characteristic function for $S_{\min}(i)$
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Formula

$$F \stackrel{\text{def}}{=} \exists_{a \in S_{\text{atoms}}} a \\ \forall_{C[\diamond] \in S_{\text{cand}}(i)} \llbracket C[\diamond] \in S_{\min}(i) \rrbracket \\ \exists_{t \in \kappa(i)} \llbracket i = t \rrbracket \\ \exists_{C[\diamond] \in S_{\text{cand}}(i)} \llbracket \text{init}(C[\diamond], i) \rrbracket \\ \exists_{j \downarrow t \in \kappa_{\downarrow}(i), C[\diamond] \in S_{\text{cand}}(i)} \llbracket \text{minimal}(C[\diamond], i, j) \rrbracket \\ \left(\prod_{x \in S_{\text{constraints}}} x \right) \sqcap \left(\text{empty} \sqcup \bigsqcup_{t \in \kappa(i)} \left\{ \begin{array}{l} \llbracket i = t \rrbracket \sqcap \\ \text{minimal}(t) \end{array} \right\} \right)$$

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$$\exists_{j \downarrow t \in \kappa_{\downarrow}(i), C[\diamond] \in S_{\text{cand}}(i)} \llbracket \text{minimal}(C[\diamond], i, j) \rrbracket$$
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$$\text{empty} \stackrel{\text{def}}{=} \prod_{C[\diamond] \in S_{\text{cand}}(i)} \neg \llbracket C[\diamond] \in S_{\min}(i) \rrbracket$$

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$$\text{minimal}(t) \stackrel{\text{def}}{=} \prod_{j \triangleleft t} \bigsqcup_{C[\diamond] \in S_{\text{cand}}(i)} \left(\llbracket C[\diamond] \in S_{\min}(i) \rrbracket \cap \llbracket \text{minimal}(C[\diamond], i, j) \rrbracket \right)$$

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Questions?

Example

n ^o	clause	constraint	source
1	$p(0, a)$		axiom
2	$\neg p(x, y) \vee p(s(x), f(y))$		axiom
3	$\neg p(n, x) \leftarrow \text{input}$		axiom
4	$n \simeq 0 \leftarrow \llbracket n \simeq 0 \rrbracket$	$\left\{ \begin{array}{l} \llbracket n \simeq 0 \rrbracket \sqcup \\ \llbracket n \simeq s(n') \rrbracket \end{array} \right.$	split
5	$n \simeq s(n') \leftarrow \llbracket n \simeq s(n') \rrbracket$		split
6	$\perp \leftarrow \text{input} \sqcap \llbracket n \simeq 0 \rrbracket$	$\neg \llbracket n \simeq 0 \rrbracket$	sup (1,4)
7	$\neg p(s(n'), x) \leftarrow \left\{ \begin{array}{l} \llbracket n \simeq s(n') \rrbracket \sqcap \\ \llbracket \text{init}(C[\diamond], n) \rrbracket \sqcap \\ \llbracket C[\diamond] \in S_{\min}(n) \rrbracket \end{array} \right.$		hypothesis
8	$p(n', b) \leftarrow \left\{ \begin{array}{l} \llbracket n \simeq s(n') \rrbracket \sqcap \\ \llbracket \text{minimal}(C[\diamond], n, n') \rrbracket \sqcap \\ \llbracket C[\diamond] \in S_{\min}(n) \rrbracket \end{array} \right.$		hypothesis

Example (cont'd)

n ^o	clause	constraint
9	$p(s(j), f(b)) \leftarrow \begin{array}{l} \llbracket \mathbf{n} \simeq s(\mathbf{n}') \rrbracket \sqcap \\ \llbracket \text{minimal}(C[\diamond], \mathbf{n}, \mathbf{n}') \rrbracket \sqcap \\ \llbracket C[\diamond] \in S_{\min}(\mathbf{n}) \rrbracket \end{array}$	
10	$\perp \leftarrow \begin{cases} \llbracket \mathbf{n} \simeq s(\mathbf{n}') \rrbracket \sqcap \\ \llbracket \text{minimal}(C[\diamond], \mathbf{n}, \mathbf{n}') \rrbracket \sqcap \\ \llbracket \text{init}(C[\diamond], \mathbf{n}) \rrbracket \sqcap \\ \llbracket C[\diamond] \in S_{\min}(\mathbf{n}) \rrbracket \end{cases}$	$\begin{array}{l} \neg \llbracket \mathbf{n} \simeq s(\mathbf{n}') \rrbracket \sqcup \\ \neg \llbracket \text{minimal}(C[\diamond], \mathbf{n}, \mathbf{n}') \rrbracket \sqcup \\ \neg \llbracket \text{init}(C[\diamond], \mathbf{n}) \rrbracket \sqcup \\ \neg \llbracket C[\diamond] \in S_{\min}(\mathbf{n}) \rrbracket \end{array}$