

Automated Theorem Proving – *Foundations* –

July 2019

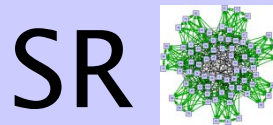
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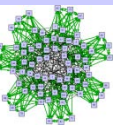
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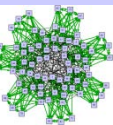
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Areas and Methods of Classic AI

- Goal (strong AI): Build artificial Human
 - Weak AI: Machine which performs a limited task like a human
- Classic areas (since late 1950s / 1960s)
 - Vision; Speech understanding, Reasoning, Robotics
 - **Reasoning** (Nilsson. *Problem Solving Methods in AI*, 1971. [3])
 - Idea: not only mathematical thinking, but thinking in general, can be expressed as logical deduction! What is proof? A finite deduction in Predicate Calculus! [2]
 - Resolution Theorem Proving (J.A. Robinson, 1965)
 - Nilsson: Problem solving methods in AI, 1971. Searching („A*, alpha-beta“), planning („blocks world“), game playing („chess“, „tic-tac-toe“)
 - Algebraic: symbolic integration heuristics (later replaced by algorithm)
 - Infrastructure: Programming languages Lisp, Prolog
 - Symbolic computation, dynamic heap storage, recursion



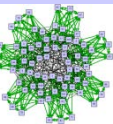
Short history of AI Reasoning with Mathematical Logic

- Math. reasoning demonstrates intelligence! (really ?)
- Proof procedures (calculi) + search heuristics (intell.?)
 - Propositional Calculus: DPLL SAT-Solving: Davis & Putnam (J.ACM 1960), Davis & Putnam & Logemann & Loveland (C.ACM 1962). No deduction here.
 - Predicate Calculus: „Resolution w. Unification“ : J.A. Robinson (J.ACM 1965)
 - Resolution is deduction, both propositional and 1st order. Idea: deduction = reasoning.
 - Largely textbook examples from 1960—1990; PROLOG came and went
- Applicability of SAT-Solving
 - First order inconsistency (Davis&Putnam 1960); hence first order proof!
 - Computer Hardware verification (switching algebra: microelectronic circuits)
 - Computer Software verification: compile program to Boolean circuit
 - Configuration: Product description (product variance), variant parts list (BoM)
- 1990s: algorithmic break-through in SAT, industrial applications
 - Intel floating point bug pushes CDCL SAT-Solving (DPLL + resolution)
 - Conflict Driven Clause Learning SAT: 100.000s clauses and variables



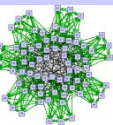
Satisfiability and Validity

- A formula F is *satisfiable* if it has a *model*. A model is a valuation $\beta: \text{Var}(F) \rightarrow \{1, 0\}$ s.th. $\beta(F)=1$. We write $\beta \models F$ or $\models_{\beta} F$. We write $\text{SAT}(F)=\text{true}$ if F is satisfiable, else $\text{SAT}(F)=\text{false}$ if F is unsatisfiable.
 - The interpretation of the logical connectives is fixed, so β only depends on the valuation of the propositional variables. There are no function or predicate symbols (and no quantifiers).
- A formula F is *valid* ($\models F$), if $\beta(F)=1$ for all β .
- F is valid iff. $\neg F$ is unsatisfiable.
 - Since $\beta(F)=1$ iff. $\beta(\neg F)=0$, and $\beta(F)=0$ iff. $\beta(\neg F)=1$.



Semantic Entailment

- Formula F entails formula G under a valuation β ($F \models_{\beta} G$) (resp. G follows from F), if $\beta(G)=1$ whenever $\beta(F)=1$.
(β must completely assign all variables in F and G)
- If $F \models_{\beta} G$ for *all* complete valuations β , then G follows from F , and we write $F \models G$.
- Proposition: $F \models G$ iff $F \wedge \neg G \models \perp$
 - Let $F \models G$. Then $\beta(\neg G)=0$ whenever $\beta(F)=1$.
 - Let $F \wedge \neg G \models \perp$. Hence if $\beta(F)=1$ then $\beta(\neg G)=0$, and so $\beta(G)=1$.
- Corollary: $F \models G$ iff $\text{SAT}(F \wedge \neg G) = \text{false}$ (and $\models \neg F \vee G$)
 - There is no counterexample where $\beta(F)=1$ and $\beta(G)=0$
- → Semantic entailment can be proved by SAT-Solving



Propositional Algebra: Equivalences

We write $F \equiv G$ if $\models (F \leftrightarrow G)$.

- $\beta(F) = \beta(G)$ for all valuations β .
- Attn: \equiv is a meta-symbol, not a propositional connective!
- \equiv denotes an equivalence relation

➤ Distributivity:

$$F \vee (G \wedge H) \equiv (F \vee G) \wedge (F \vee H)$$

$$F \wedge (G \vee H) \equiv (F \wedge G) \vee (F \wedge H)$$

➤ Absorption:

$$F \vee (F \wedge G) \equiv F \wedge (F \vee G) \equiv F$$

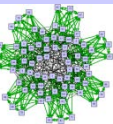
➤ DeMorgan's Laws:

$$\neg(F \wedge G) \equiv \neg F \vee \neg G$$

$$\neg(F \vee G) \equiv \neg F \wedge \neg G$$

➤ Commutativity and Associativity of \wedge , \vee .

➤ ... and many more ...



Decision Procedures and SAT-Solving

- Bad news: SAT is NP-complete (Cook 1972)
 - No known algorithm decides SAT (true/false) in polynomial time
- Good News: SAT(F) is decidable!
 - Truth Tables → guaranteed exponential, toy problems only
 - Disjunctive normal form (DNF) → easily exp., small problems ok.
 - Tableaus → similar to DNF, easily exp., small problems ok.
 - Boolean Polynomials → little use.
 - Binary Decision Diagrams (ROBDD) → model checking use, 100s variables ok, $O(1)$ SAT-solving, easy model counting, canonical form
 - Propositional Resolution → too many deductions, theoretical importance
 - Davis-Putnam-Logemann-Loveland (DPLL) → toy problems
 - DPLL based CDCL SAT-Solving → practically efficient for science and industry, 100,000+ variables, method of choice, very robust, much research.



Method Example: Truth Table and DNF

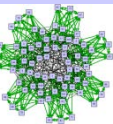
$$F = ((x \wedge \neg(y \rightarrow z)) \oplus x)$$

| x | y | z | $y \rightarrow z$ | $\neg(y \rightarrow z)$ | $x \wedge \neg(y \rightarrow z)$ | $(x \wedge \neg(y \rightarrow z)) \oplus x$ |
|---|---|---|-------------------|-------------------------|----------------------------------|---|
| 0 | 0 | 0 | 1 | 0 | 0 | 0 |
| 0 | 0 | 1 | 1 | 0 | 0 | 0 |
| 0 | 1 | 0 | 0 | 1 | 0 | 0 |
| 0 | 1 | 1 | 1 | 0 | 0 | 0 |
| 1 | 0 | 0 | 1 | 0 | 0 | 1 |
| 1 | 0 | 1 | 1 | 0 | 0 | 1 |
| 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 1 | 1 | 1 | 1 | 0 | 0 | 1 |

F is satisfiable, but not valid. Exponential number of rows.

DNF: enumerate all points =1

$$\text{DNF}(F) = (x \wedge \neg y \wedge \neg z) \vee (x \wedge \neg y \wedge z) \vee (x \wedge y \wedge z) = (x \wedge \neg y) \vee (x \wedge y \wedge z)$$



Application Example: Mercedes Configuration.

Example: E-Class

- approx. 1.500 codes (options, countries, ..)
- approx. 3.000 rules in product description
- rule based BOM with approx. 35.000 parts



$$B(P09) = 297 + 540 + 543;$$

$$B(610) = (512 / 527 / 528) + 608 + -978;$$

$$Z(P09) = (M271 + -M013 / M272) + 830 / \\ Z04 + -(M273 + Z27)$$

$$Z(682) = 623 / 830 / 513L$$

$$B(450) = L + 965 + 670 + 837 + -P34 + \\ ((M271 / M651) + (953 / 955) + (100A / 200A) + (334 / 335) + (301 / 336 / 337) / \\ / M642 + (2XXL / 557L / 571L) + (953 / 955) + (100A / 200A) + (334 / 335) + (301 / 337));$$



Application Example: Mercedes Configuration.

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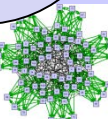
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P09: Sonnenschutzpaket
297: Rollos Fond-Türen links+rechts
540: Rollo elektr. Heckfenster
543: Sonnenbl. li+re m. Make-up-Sp.
610: Nachtsichtsystem
512: Comand APS mit DVD-Wechsl.
527: Comand DVD APS mit NAVI
528: Comand DVD+ ohne NAVI
608: Autom. Fernlichtschaltung
978: Spezial-Fahrgestell
682: Feuerlöscher
830: Zusatzteile China
623: Golf-Staaten-Ausführung
513L: Belgien
M271: R4-Ottomotor
M272: V6-Ottomotor
M273: V8-Ottomotor
M013: Motor leistungsreduziert
Z04: B4 Panzerung
Z27: Bodenpanzerung
450: TAXI-International



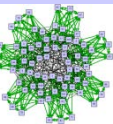
Application Example: *Checking Configuration Options*

- Car order is a list L of options (option codes)
 - Codes in L are *true*, all others are *false* (for this car)
- Compile all configuration rules into a single formula \mathbf{C}
 - Solutions $\beta(\mathbf{C})=\text{true}$ are constructible orders
 - Restriction $\mathbf{C}|_{o=\text{true}}$ sets $o=\text{true}$ in \mathbf{C} : all cars with option o
- Checking the configuration rule-set
 - Is option o available? $\rightarrow \text{SAT}(\mathbf{C}|_{o=\text{true}}) = \text{true}?$
 - Is option o forced? $\rightarrow \text{SAT}(\mathbf{C}|_{o=\text{false}}) = \text{false}?$
 - Is option o available in country c ? $\rightarrow \text{SAT}(\mathbf{C}|_{c=\text{true},o=\text{true}}) = \text{true}?$
 - Is option o forced in country c ? $\rightarrow \text{SAT}(\mathbf{C}|_{c=\text{true},o=\text{false}}) = \text{false}?$
- Configuration step
 - Is option o available after selecting options o_1, \dots, o_n ?
 $\rightarrow \text{SAT}(\mathbf{C}|_{o_1=\text{true}, \dots, o_n=\text{true}, o=\text{true}}) = \text{true}?$



Conjunctive Normal Form (CNF)

- Resolution and DPLL / CDCL SAT-Solving require CNF
- CNF is a conjunction (\wedge) of *clauses* ($\ell_1 \vee \dots \vee \ell_n$)
 - The ℓ_i are *literals* (positive or negative variables)
 - Example: $F = (x \vee \neg y) \wedge (x \vee \neg z) \wedge (z \vee \neg y) \wedge (z \vee \neg x)$
 - Simplified set notation: $F = \{\{x, \neg y\}, \{x, \neg z\}, \{z, \neg y\}, \{z, \neg x\}\}$
 - CNF lists a set of (simple) *constraints*, which must be **simultaneously** satisfied for $F=1$.
 - CNF enumerates conditions for $F \neq 0$.
 - CNF enumerates the negations of F 's roots.
- CNF conversion (theoretically)
 - Push negations \neg inside \rightarrow negation normal form NNF
 - Apply distributivity law: $F \vee (G \wedge H) \equiv (F \vee G) \wedge (F \vee H)$

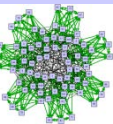


Example: CNF from function table

$$F = ((x \wedge \neg(y \rightarrow z)) \oplus x)$$

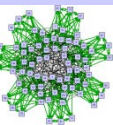
| x | y | z | $y \rightarrow z$ | $\neg(y \rightarrow z)$ | $x \wedge \neg(y \rightarrow z)$ | $(x \wedge \neg(y \rightarrow z)) \oplus x$ |
|---|---|---|-------------------|-------------------------|----------------------------------|---|
| 0 | 0 | 0 | 1 | 0 | 0 | 0 |
| 0 | 0 | 1 | 1 | 0 | 0 | 0 |
| 0 | 1 | 0 | 0 | 1 | 0 | 0 |
| 0 | 1 | 1 | 1 | 0 | 0 | 0 |
| 1 | 0 | 0 | 1 | 0 | 0 | 1 |
| 1 | 0 | 1 | 1 | 0 | 0 | 1 |
| 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 1 | 1 | 1 | 1 | 0 | 0 | 1 |

CNF: negate all roots: $\neg(\neg x) \wedge \neg(x \wedge y \wedge \neg z) \equiv x \wedge (\neg x \vee \neg y \vee z)$
 $\equiv x \wedge (\neg y \vee z)$



Complexity of CNF-Transformation

- Application of Distributivity Law may double the formula size
 - $F \vee (G \wedge H) \equiv (F \vee G) \wedge (F \vee H)$
- Worst case: $\text{CNF}(F)$ exponentially larger than F
 - Very serious impediment for applications
- Example: $(x_{11} \wedge x_{12}) \vee \dots \vee (x_{n1} \wedge x_{n2})$.
 - CNF contains 2^n clauses with n variables each (\rightarrow Induction)
- Solution in 2 steps
 1. Tseitin transformation
 2. Plaisted-Greenbaum Transformation
- Transformations maintain satisfiability, not equivalence!



CNF: Tseitin-Transformation

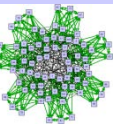
➤ Key idea:

- introduce auxiliary variables to represent (sub-)formulae
- introduce additional constraints to link variables to formulae

➤ General concept, we use it only for CNF

- In $F \vee (G \wedge H)$, represent $(G \wedge H)$ by new x
- Add constraint $x \leftrightarrow (G \wedge H)$.
- Hence $F \vee (G \wedge H)$ is replaced by $(F \vee x) \wedge (x \leftrightarrow (G \wedge H))$
- Now convert $x \leftrightarrow (G \wedge H)$ to $x \rightarrow (G \wedge H)$ and $(G \wedge H) \rightarrow x$, giving clauses $(\neg x \vee G)$, $(\neg x \vee H)$, $(\neg G \vee \neg H \vee x)$ in addition to $(F \vee x)$.
- This is the „full“ Tseitin-Transformation
- Only satisfiability is preserved: $F \vee (G \wedge H) \cong (F \vee x) \wedge (x \leftrightarrow (G \wedge H))$
- Model count is preserved

➤ Bad news?: Now G and H are doubled instead of F!?



CNF: Tseitin-Transformation

➤ Good news!

- Think recursively: May assume G and H are already represented by variables g and h .
- Then we get clauses $(\neg x \vee g)$, $(\neg x \vee h)$, $(\neg g \vee \neg h \vee x)$ in addition to $(F \vee x)$ where only 2 variables are doubled
- and only one set of 3 clauses and 3 variables each for g and h .

➤ Plaisted & Greenbaum Transform: $(F \vee x) \wedge (x \rightarrow (G \wedge H))$

- David A. Plaisted, Steven Greenbaum: A Structure Preserving Clause Form Transform. *J. Symbolic Computation* Vol 2(3), 1986.
- Often also misnamed „Tseitin-Transform“
- Drops constraint $((G \wedge H) \rightarrow x)$
- Model count may (but need not) double with every aux. Variable
- Theorem: $F \vee (G \wedge H) \cong (F \vee x) \wedge (x \rightarrow (G \wedge H))$
 - Intuition: if G and H are satisfied, we don't care about the Tseitin variable



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