Automated Reasoning: the big picture

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Outline

The field of automated reasoning: a brief overview Theorem proving as a search problem A taxonomy of theorem-proving strategies

The field of automated reasoning: a brief overview

Theorem proving as a search problem

A taxonomy of theorem-proving strategies

A central problem in automated reasoning

S: set of *assumptions* properties of the object of study (e.g., system, circuit, program, data type, communication protocol, mathematical structure)

 φ : *conjecture* a property to be verified

Problem: does φ follow from S?

$$S \models^{?} \varphi$$

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Automated reasoning and knowledge representation

Knowledge representation:

finding formalisms for ${\cal S}$ and φ to represent desired aspects of the analyzed systems

Automated reasoning:

studying and implementing reasoning techniques to solve the entailment problem (S $\models^? \varphi$)

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Automated reasoning in first order logic

Representation formalism: first order logic (FOL)

Motivation: FOL provers applied successfully to, e.g.,

software and hardware verification, e.g.,

- cryptographic protocols
- message-passing systems
- software specifications
- theorem proving support to model checking

proving non-trivial mathematical theorems in, e.g.,

- Boolean algebras
- theories of rings, groups and quasigroups
- many-valued logic

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Automated reasoning: building proofs or models

$$S \models^{?} \varphi$$

Theorem proving:

finding a *proof* of φ from S and answer affirmatively

Model building:

finding a *model* of $S \cup \{\neg \varphi\}$, that is a *counter-example* for $S \models \varphi$, and answer negatively

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Theorem proving: deduction or induction

 $S \models \varphi$:

 φ is true in all models (systems, worlds ...) where ${\it S}$ is true

Deductive theorem proving:

$$S \models \varphi$$

Inductive theorem proving:

$$S \models \varphi \sigma$$

for all ground substitutions σ

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Automated reasoning problems are very hard

In first order logic

- **Deductive theorem proving** is only semi-decidable
- Inductive theorem proving is not even semi-decidable
- **Model building** is not even semi-decidable

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Automatic and interactive theorem proving

Automatic theorem proving:

the machine alone is expected to find a proof

Interactive theorem proving:

a proof is born out of the interaction between human and machine

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Automatic deductive theorem proving

Automatic theorem proving:

deductive theorem proving

Interactive theorem proving: induction, model generation and reasoning in higher-order logics

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Refutational theorem proving

- Direct proof:
 deriving φ from S without making use of φ itself
- Proof by way of contradiction or by refutation: showing that S ∪ {¬φ} generates a contradiction (⊥), S ∪ {¬φ} is inconsistent, hence S ⊨ φ

Too difficult to find a proof ignoring the conjecture: theorem-proving methods work *refutationally*.

Decidable instances of reasoning problems

Decidable instances of reasoning problems do exist

Decidability may stem from imposing restrictions on

- 1. the logic
- 2. the form of admissible formulae
- 3. the theory presented by the assumptions

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Examples of decidable instances

- 1. propositional logic: the SAT problem
- 2. Bernays-Schönfinkel class:

$$\exists x_1, \ldots x_n, \forall y_1, \ldots y_m, P[x_1, \ldots x_n, y_1, \ldots y_m]$$

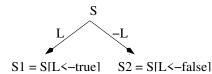
where P is quantifier-free and function-free

3. *Presburger arithmetic* or theories of data structures, such as *lists* or *arrays*

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SAT: Davis-Putnam-Logemann-Loveland procedure

Case analysis or splitting + unit propagation:



- **Unit clause rule:** if *L* is a clause, only one branch
- **Pure literal rule:** if *L* is pure (only one sign), only one branch
- Control: depth-first search (DFS) with backtracking + refinements

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SAT: Boolean Ring simplification

Let + be exclusive or and juxtaposition be and:

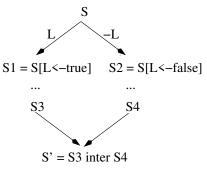
$$\begin{array}{ll} \mathbf{x}\mathbf{x} = \mathbf{x} & \mathbf{x}\mathbf{0} = \mathbf{0} & \mathbf{x}\mathbf{1} = \mathbf{1} \\ \mathbf{x} + \mathbf{x} = \mathbf{0} & \mathbf{x} + \mathbf{0} = \mathbf{x} & -x = x \\ xy = yx & (xy)z = x(yz) \\ x + y = y + x & (x + y) + z = x + (y + z) & \mathbf{x}(\mathbf{y} + \mathbf{z}) = \mathbf{x}\mathbf{y} + \mathbf{x}\mathbf{z} \end{array}$$

 $x \lor y$ is xy + x + y and $\neg x$ is x + 1

- Simplification by equations in bold face as rewrite rules
- **Unique normal form**: 0, 1 or a Boolean polynomial
- Distributivity: exponential growth of the normal form
- ► A solution: DPLL + BR representation + BR simplification

SAT: Stålmarck's method

- Same framework as DPLL (sort of)
- Dilemma rule:



Control: DFS with iterative deepening (DFID) to control how deep to go in the dilemma's branches

Back from SAT to FOL theorem proving

Semi-decidability:

No procedure is guaranteed to halt and

- return a positive answer and a proof whenever S ∪ {¬φ} is inconsistent
- return a negative answer and a model whenever S ∪ {¬φ} is consistent

The best one can have is a semi-decision procedure

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Semi-decision procedures

A semi-decision procedure is guaranteed to halt and return a positive answer and a proof whenever $S \cup \{\neg\varphi\}$ is inconsistent.

However, if $S \cup \{\neg \varphi\}$ is consistent, the procedure is not guaranteed to halt.

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Search for proofs

Intuition of the source of semi-decidability:

- Proofs are *finite*, if they exist, but
- There is an *infinite search space* of consequences where to look for a contradiction

A machine can explore only a *finite* part of this *infinite* space

Challenge: to find a proof using as little resources as possible

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Theorem-proving strategies

 Inference system: set of inference rules defines the search space of all possible inferences

Search plan:

controls the application of the inference rules guides the search for a proof

Inference system + search plan = *theorem-proving strategy*

Since we are looking for a proof: *Proof system* + search plan = *proof procedure*

From non-determinism to determinism

- Inference system: non-deterministic set of inference rules
- Search plan:

determines the unique derivation, e.g.,

$$S_0 \vdash S_1 \vdash \ldots S_i \vdash S_{i+1} \vdash \ldots$$

that the strategy computes from $S_0 = S \cup \{\neg \varphi\}$

- A TP strategy or proof procedure is deterministic
- S_i : state, e.g.: a set of clauses; a set of clauses and a tableau

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Soundness and adequacy

- ▶ Soundness: if $S_i \vdash S_{i+1}$ then $S_i \models S_{i+1}$
- Adequacy: if $S_i \vdash S_{i+1}$ then $S_{i+1} \models S_i$

Adequacy was also called monotonicity:

 $S_i \vdash S_{i+1}$ implies $Th(S_i) \subseteq Th(S_{i+1})$

where $Th(S) = \{\psi \mid S \models \psi\}$

Refutational completeness and fairness

Refutational completeness: if $S_0 = S \cup \{\neg \varphi\}$ is inconsistent, inference system generates at least a derivation $S_0 \vdash S_1 \vdash \ldots S_i \vdash S_{i+1} \vdash \ldots$ such that S_k contains \perp for some k

Fairness:

search plan considers eventually all inferences that may be necessary to generate such an S_k

Uniform fairness:

search plan considers eventually all irredundant expansion inferences

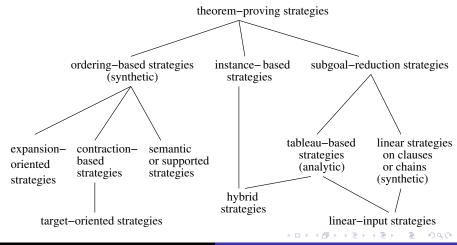
Formal definitions: e.g., with well-founded proof orderings

Refutational completeness

If the inference system (proof system) is *refutationally complete* and the search plan is *fair*, then the strategy (proof procedure) is *refutationally complete*:

if $S_0 = S \cup \{\neg \varphi\}$ is inconsistent, the unique *derivation* $S_0 \vdash S_1 \vdash \ldots S_i \vdash S_{i+1} \vdash \ldots$ computed by the strategy is such that S_k contains \perp for some k

A taxonomy of theorem-proving strategies



Ordering-based strategies I

Expansion inference rule:

$$\begin{array}{cccc} A_1 & \dots & A_n \\ \hline B_1 & \dots & B_m \end{array}$$

where m > 1e.g., resolution and paramodulation

Contraction inference rule:

$$\begin{array}{cccc} A_1 & \dots & A_n \\ \hline B_1 & \dots & B_m \end{array}$$

where $m \ge 0$ e.g., subsumption and equational simplification

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Ordering-based strategies II

Expansion inferences:

$$rac{S_i}{S_{i+1}}$$
 $S_i \subset S_{i+1}$

Contraction inferences:

$$\frac{S_i}{S_{i+1}} \ S_i \not\subseteq S_{i+1} \quad S_{i+1} \prec S_i$$

where \prec is a well-founded ordering

Database of clauses: indexing techniques

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Subgoal-reduction based strategies I

- Model elimination (ME)
- Linear resolution
- Matings
- Connections or matrices

All eventually understood in the context of *clausal normalform tableaux* e.g., *ME-tableaux*

Subgoal-reduction based strategies II

Free-variable tableaux

Clausal normalform tableaux:

- Extension: extend branch with fresh copy of clause
- Closure: close branch with unifiable complementary literals + apply mgu
- (Strong) link condition: extend only if (adjacent) complementary literals unify

Rigid variables

Ordering-based and subgoal-reduction strategies I

	Ordering-based	Subgoal-reduction	
Data	set of objects	one goal-object at a time	
Proof attempts built	many implicitly	one at a time	
Backtracking	no	yes	
Contraction	yes	no	

Ordering-based and subgoal-reduction strategies II

	Ordering-based	Subgoal-reduction
Visited search space	all generated clauses	all tried tableaux
Active search space	all kept clauses	current tableau
Generated proof	ancestor-graph of \Box	closed tableau

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Instance-based strategies

- Forerunner: Gilmore's multiplication method (1960)
- Recent methods:
 - Generate ground instances of clauses in set to be refuted (e.g., by *hyperlinking*)
 - Apply a SAT solver and iterate
- More recent methods:
 - SAT solver as model generator
 - Generate ground instances to eliminate models

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Hybrid strategies

Combine tableaux and instance generation, e.g.:

- Give up on instantiating rigid variables in the tableau
- Backtracking no longer needed
- Add instance generation, e.g., by hyperlinking

Intuitively, the information lost by not instantiating the tableau is generated as instances of clauses in the given set.

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