Efficient SAT in the Alt-Ergo SMT solver

after the work of the Alt-Ergo team

July 2021

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1) Motivations

Motivation = software correctness:

- tests are widely used but can't certify that there is no bug
- proofs of program (deductive verification) with proof assistant don't scale
- → automatic provers

Automatic provers :

- what kind of anwser?
- counterexamples ?
- specialized vs generalist provers ?
- how much time to get the answer?
- •

Here we will focus on the algorithmic efficiency.

2) SAT and SMT solvers

SAT: satisfiability of propositional/boolean formula

$$A \lor B \Rightarrow C \Leftrightarrow (A \land B) \lor (B \land C)$$

Theory:

- 2-SAT (SAT with clauses of size 2 in CNF) is solvable in P time
- 3-SAT (SAT with clauses of size 3 in CNF) NP-complete
- validity of F (tautology) = unsatisfiability of ¬ F (can't find interpretation such that true)

SMT: each literal is a predicate expressed in a specific theory and possibly with quantification, equality, ...

$$(x > 4) \land (f(x) < 2.0) \land (\forall i : \mathbb{Z}.f(i) = 0)$$
Solveur SAT

Solveur de théories

Moteur d'instanciation

Conjonctive normal form (CNF):

 boolean formulas are composed of the ∧ of clauses and clauses are the ∨ of literals and literals are a variable or its negation.

$$\rightarrow \neg (A \rightarrow B) \lor (C \rightarrow A) \leadsto (A \lor \neg C) \land (\neg B \lor \neg C \lor A)$$

Negative normal form (NNF):

- boolean formulas are composed with \land , \lor and negation applied to one variable only.
- \rightarrow A \land (B $\lor \neg C$) = (A \land B) \lor (A $\land \neg C$)
- → not unique!

Unit clauses and Boolean Constrain Propagation (BCP):

- assignment makes every literal in the clause unsatisfied but leaves a single literal undecided: last one has to be true for the clause to be true
- \rightarrow $(\neg A \lor \neg B \lor C) \land (\neg C \lor D)$ with A and B assigned leads to a constrain propagated to C and then propagated to D

SAT History:

- 1960 David Putman
- 1962 David Logemann Loveland (DPLL)
- ⇒ chronological backtracking without learning
 - 1967 Tseitin algo to put in CNF (conjonctive normal form) (linear in the number of clauses instead of exponential) → (... ∨ ... ∨ ...) ∧ (... ∨ ... ∨ ...) ∧ ...
- → faster boolean constrain propagation (BCP)
 - 1996 conflict driven clause learning (CDCL) ⇒ non-chronological backjumping + learning
 - 2001 optimized BCP

SMT Decision procedure = combination of solvers for differents theories :

- 1980: Shostak algorithm: deciding combination of theories (original paper is wrong, no proof of termination, corrected in 2003) conjonction of equalities with uninterpreted symbols
- → Alt-Ergo is one of the last to use it
 - 1979 : Nelsen-Oppen algorithm

Difficulties with SMT solvers:

- combination of decision procedures
- deal with quantification efficiently
- polymorphism
- AC (associative, commutative) symbols

Shostak theories: canonizer returns normal form and solver take an equality and returns a substitution Examples :

- linear integers arithmetic
- pairs, records
- fragment of bitvectors

Algorithm : take each equality u = v :

- u' = canon(u), v' = canon(v)
- $\sigma = \text{solve}(u', v')$
- ullet apply σ to representants of equivalence class
- how to combine canon / solve / sigma such that the algorithm is complete and terminating?

Some SMT solvers commonly used in Why3:

• 2002 : CVC

• 2003 : SMT-lib

2006: CVC3, Alt-Ergo (Inria, then 2013 OcamlPro)

• 2007 : Z3

• 2012 : CVC4

2017 : Alt-Ergo with floats

• 2021 : CVC5

Alt-Ergo:

- generalist (= not specialized for some specific theory)
- motivations: Why3 (first-order + polymorphism), its input syntax is common with the one used in Why3

History of Alt-Ergo:

- 10/2006 : naive SAT, linear arithmetic
- 02/2007 : polymorphism
- 07/2008 : backjumping
- 07/2009 : AC symbols
- 05/2010 : tableaux method, non-linear arithmetic
- 04/2011 : enum types, graphical interface
- 12/2011 : record types
- 01/2013 : model productions
- 09/2013 : OcamlPro
- 12/2014 : plugin architecture
- 2015-2017 : optimizations
- •
- 2018 : Albin Coquereau thesis

3) Albin Coquereau thesis (2018)

2018 : Albin Coquereau thesis : Chapter 3 :

- study of a naive coupling of SAT-CDCL with Alt-Ergo
- Ocaml vs C++ => garbage collector responsible for cache misses
- new SAT is efficient but the new SAT-CDCL/SMT coupling less efficient than the previous SAT-Tableaux/SMT coupling in historical Alt-Ergo

Chapter 4:

efficient CDCL(T) in Alt-Ergo

Chapter 5:

- lib-smt2 syntax support
- syntax extension to polymorphism

Chapter 6:

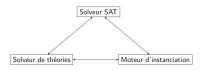
• SMT 2018 competition results

4) Efficient SAT in Alt-Ergo SMT

Four algorithms:

- CDCL
- Tableaux (historique)
- Tableaux assisted by CDCL
- → CDCL + pertinence calculus Tableaux-method

From this:



to this:



```
(\neg A \lor B)
\land (\neg C \lor D)
\land (\neg E \lor \neg F)
\land (F \lor \neg E \lor \neg B)
\land (E \lor G)
\land (E \lor \neg G \lor \neg B)
```

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$\downarrow_{|V|}$$

$$\downarrow_{|V|}$$

$$\downarrow_{|V|}$$

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$\downarrow_{|V|}$$

$$A$$

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

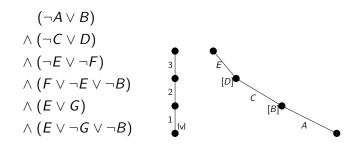
$$\land (E \lor \neg G \lor \neg B)$$

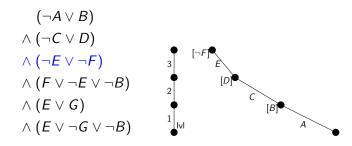
$$\downarrow_{|V|}$$

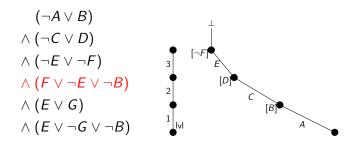
$$\downarrow_{|V|}$$

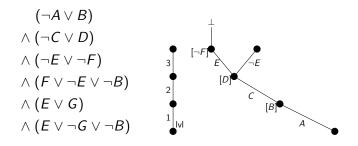
$$\downarrow_{|V|}$$

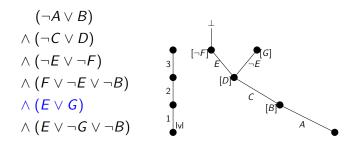
$$\downarrow_{|V|}$$

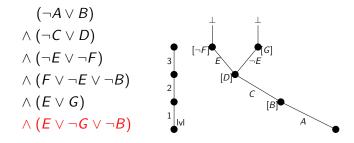


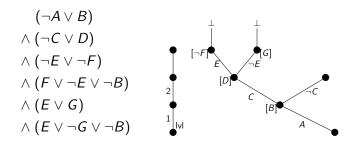


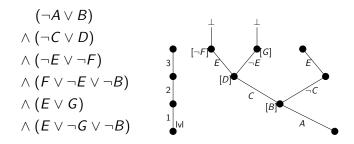


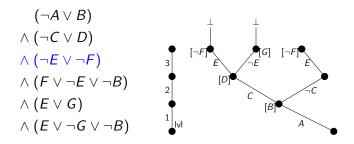


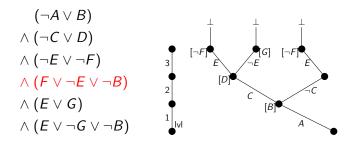


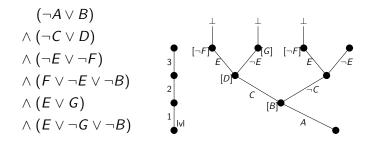


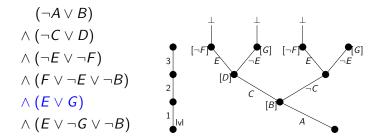


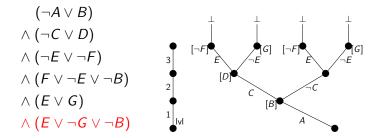


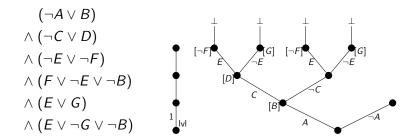


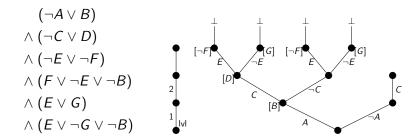


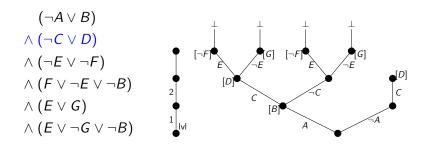


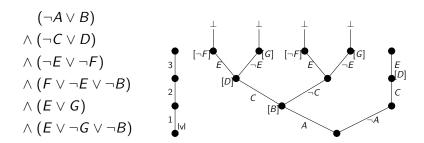


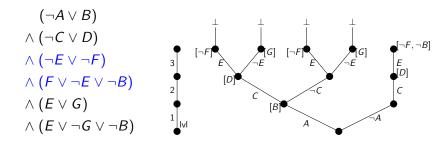


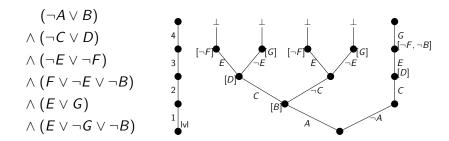


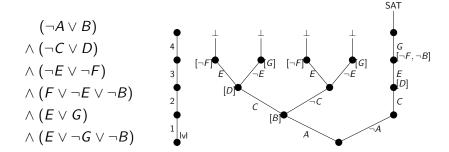












```
Input: Γ: CNF, Δ: Boolean Model
   Output: Satisfiability status

    IvI ← 0

2 while true do
        (\Delta, Conflict) \leftarrow BCP(\Gamma, \Delta)
        if Conflict ≠ Ø then
              if lvl = 0 then
                   return UNSAT
             else
                   |v| \leftarrow |v| - 1
                  \Delta \leftarrow backtrack(\Gamma, \Delta)
        else if all variables are assigned in \Delta then
10
             return SAT
11
12
        else
              L \leftarrow choose(\Gamma, \Delta)
13
             |v| \leftarrow |v| + 1
             \Delta \leftarrow L :: \Delta
```

- when conflict, backtracks to the last non-BCP decision
- → doesn't learn anything but that there is a conflict
- ightarrow backtrack only one level

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$\downarrow \qquad \qquad \downarrow \qquad \downarrow \qquad \qquad \downarrow \qquad \downarrow$$

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$\downarrow_{[\neg F]}$$

$$\downarrow_{[B]}$$

$$\downarrow_{[B]}$$

Logical resolution :

$$(P \lor Q) \land (\neg P \lor R) = (Q \lor R)$$

Adding a new learned clause = the logical resolution with the conflict :

$$(\neg A \land B) \lor (\neg E \land \neg F) \lor (F \land \neg E \land \neg B) = (\neg A \land B) \lor (\neg E \land \neg B) = (\neg A \land \neg E)$$

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$\land (\neg A \lor \neg E)$$

$$\land (\neg A)$$

$$(\neg A \lor B)$$

$$\land (\neg C \lor D)$$

$$\land (\neg E \lor \neg F)$$

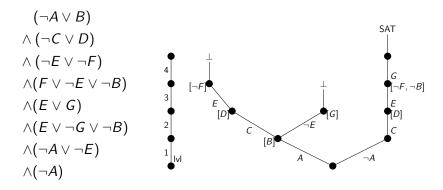
$$\land (F \lor \neg E \lor \neg B)$$

$$\land (E \lor G)$$

$$\land (E \lor \neg G \lor \neg B)$$

$$\land (\neg A \lor \neg E)$$

$$\land (\neg A)$$



```
\label{eq:while true do} \begin{aligned} & (\Delta, Conflict) \leftarrow BCP(\Gamma, \Delta) \\ & (\Delta, Conflict) \leftarrow \emptyset \text{ then} \\ & \text{if } Konflict \neq \emptyset \text{ then} \\ & \text{if } M = 0 \text{ then} \\ & \text{return UNSAT} \\ & \text{else} \\ & \left\{ L \lor C, bj.JM \right\} \leftarrow resolve(\Gamma, \Delta, Conflict) \\ & \Gamma \leftarrow \Gamma \cup \left\{ L \lor C \right\} \\ & \left\{ (\Delta, M) \leftarrow backjump(\Delta, bj.M) \right\} \\ & \Delta \leftarrow L :: \Delta \\ & \text{else if all variables are assigned in } \Delta \text{ then} \\ & \Gamma \text{ return SAT} \\ & \text{else} \\ & L \leftarrow choose(\Gamma, \Delta) \\ & M \leftarrow M + 1 \\ & \Delta \leftarrow L :: \Delta \end{aligned}
```

- when conflict, backtracks to the latest guess that affects a literal in the learned clause
- → mistake → clause learning → don't do the same mistake
- → clause learning feeds the BCP mechanism

From SAT CDCL:

```
\label{eq:while true do} \begin{cases} (\Delta, Conflict) \leftarrow BCP(\Gamma, \Delta) \\ (\Delta, Conflict) \neq \emptyset \text{ then} \\ \text{if } Konflict \neq \emptyset \text{ then} \\ \text{if } W = 0 \text{ then} \\ \text{ir terum UNSAT} \\ \text{else} \\ \left\{ (L \lor C, bj. | \mathcal{M}) \leftarrow resolve(\Gamma, \Delta, Conflict) \\ \Gamma \leftarrow \Gamma \cup (L \lor C) \\ (\Delta, \mathcal{M}) \leftarrow backjump(\Delta, bj. | \mathcal{M}) \\ \Delta \leftarrow L :: \Delta \\ \text{else if all variables are assigned in } \Delta \text{ then} \\ \text{i return SAT} \\ \text{else} \\ L \leftarrow choose(\Gamma, \Delta) \\ \mathcal{M} \leftarrow \mathcal{M} + 1 \\ \Delta \leftarrow L :: \Delta \end{cases}
```

To SMT CDCL(Theory):

```
while true do (\Delta, Conflict) \leftarrow BCP(\Gamma, \Delta) if Conflict \neq \emptyset then else (T, Conflict \neq \emptyset) if then if N = 0 then | f(N) = N = N if N = 0 then | f(N) = N = N if N = 0 then | f(N) = N = N if N = N = N then | f(N) = N = N if N = N = N then | f(N) = N = N if N = N = N then | f(N) = N = N if N = N = N then | f(N) = N = N if N = N = N then | f(N) = N = N if N = N = N then | f(N) = N then | f(N) = N = N then | f(N) = N = N then | f(N) = N then |
```

Advantage of the CDCL method :

fast

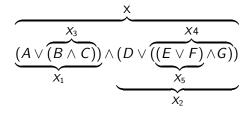
Disadvantage of the CDCL method:

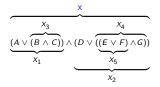
- needs CNF
- → returns the full boolean model
- → don't take in account the original shape of the formula
- → too much instanciations in SMT theories

Example:

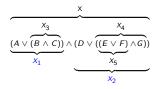
$$\varphi = (A \vee (B \vee \varphi_1)) \wedge (\neg A \vee (\neg B \vee \neg \varphi_1))$$

assignment $\{A \to true; B \to false\}$ is a model for φ but CDCL won't find it

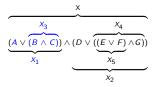




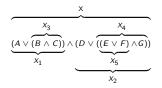


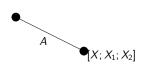


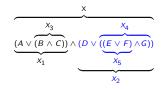


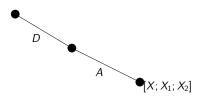


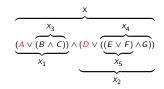


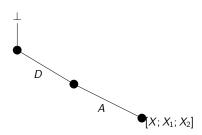


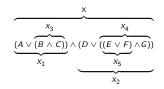


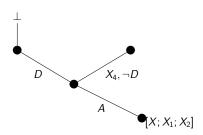


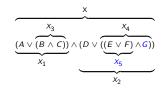


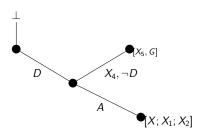


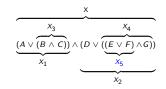


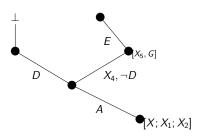


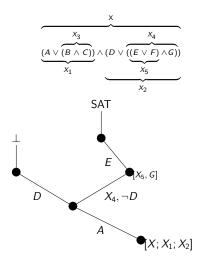












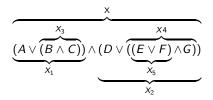
Modèle booléen renvoyé par le solveur SAT : $\{A; \neg D; E; G\}$ X_i n'ont de l'influence que sur la partie SAT

```
Input: Φ : Set of formulas in Negatif Normal Form
   Output: Satisfiability status, explanation
1 Function solve (Φ)
       (\Phi, Error) \leftarrow propagate(\Phi)
                                         // Boolean constraint propagation
       if Error then
          reason \leftarrow explain \ conflict()
          return (UNSAT.reason)
6
      else
          if \exists A \lor B \in \Phi then
              (\Phi, Error) \leftarrow (assume(\Phi, [\{A\}])
                                                                 // decide on A
              if Error then
                 reason \leftarrow explain \ conflict()
                 return (UNSAT, reason)
12
13
                  (status, reason) \leftarrow solve(\Phi)
14
                 if status \neq UNSAT then
15
                     return (status.reason)
16
                 else
17
                     if A \in reason then
                         (\Phi, Error) \leftarrow (assume(\Phi, [\{\neg A\}; \{B\}]))
18
                         // backtrack and propagate \neg A and B
19
                         if Error then
                             reason \leftarrow explain \ conflict()
21
                             return (UNSAT.reason)
                             return solve(\Phi)
24
                     else
25
                         return (UNSAT.reason)
                                                           // backjump further
             return (SAT.0)
27
                                                                  // \Phi is empty
```

- "proxy" variables
- when conflict, backtracks is changing literal (we were taking negation with the DPLL/CDCL)
- truth value of a variable \leadsto truth value of a disjonction
- negative normal form

Reduction of the literals sent to the SMT solver components (decision procedure + instanciation) with the Tableau-method :

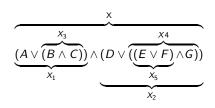
- keep the original formula
- graph search: "pertinent" literals to keep are the ones assigned to true, during the graph search



$$A, G = \top$$

$$B, D = \bot$$

$$E, F, C = -$$

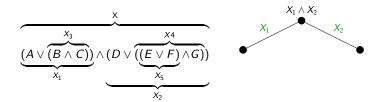




$$A, G = \top$$

$$B, D = \bot$$

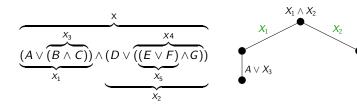
$$E, F, C = -$$



$$A, G = \top$$

$$B, D = \bot$$

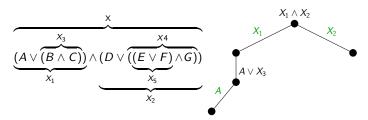
$$E, F, C = -$$



$$A, G = \top$$

$$B, D = \bot$$

$$E, F, C = -$$

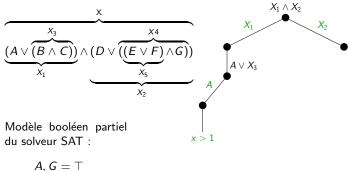


Modèle booléen partiel du solveur SAT :

$$A, G = \top$$

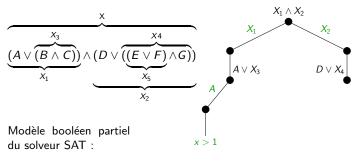
$$B, D = \bot$$

$$E, F, C = -$$



 $B, D = \bot$

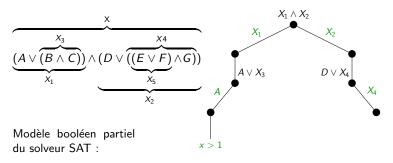
E,F,C = -



$$A, G = \top$$

$$B, D = \bot$$

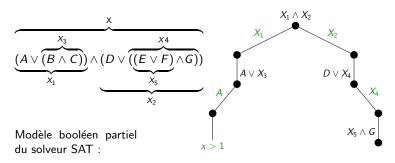
$$E, F, C = -$$



$$A, G = \top$$

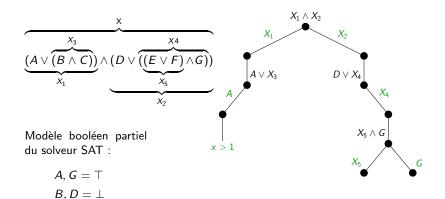
$$B, D = \bot$$

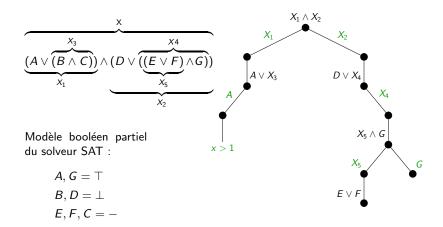
$$E, F, C = -$$

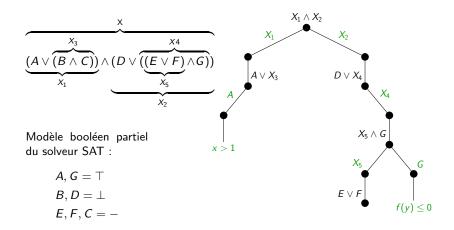


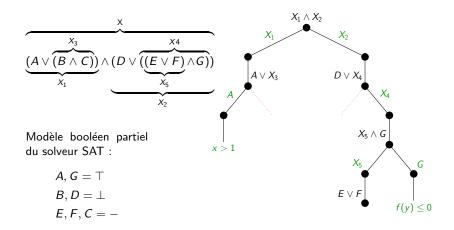
$$A, G = \top$$
 $B, D = \bot$
 $E, F, C = -$

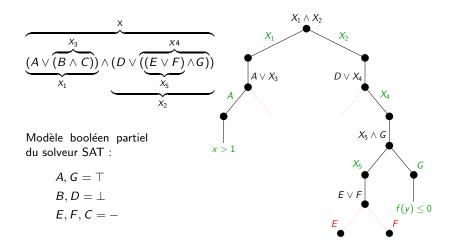
E, F, C = -

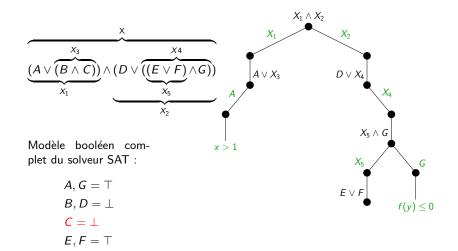


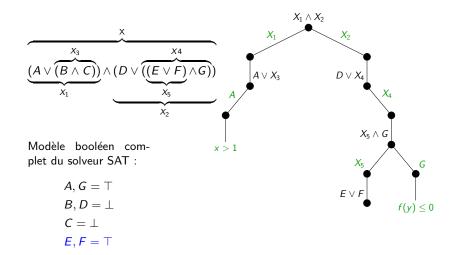


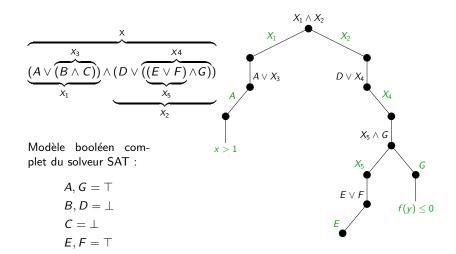


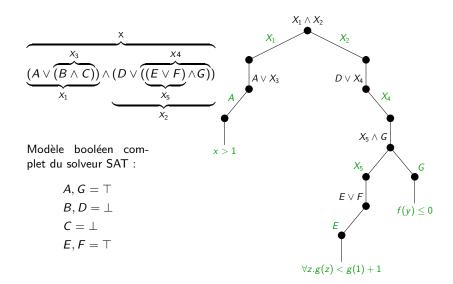




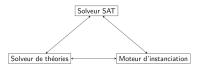




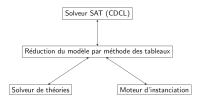




From this:



to this:



$$(A \vee (B \vee C)) \wedge (D \vee ((E \vee F) \wedge G))$$

- CDCL \rightsquigarrow complete boolean model $\{A, \neg B, \neg C, \neg D, E, F, G\}$
- CDCL-tableaux \rightsquigarrow reduced boolean model $\{A, E, G\}$ sent to the combinator of theories or instanciation engine

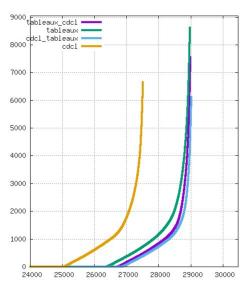
CDCL(Tableaux(T))

Résultats

	# buts	cdcl	tableaux	cdcl + tableaux
BWARE-DAB	860	98.7% (258s)	100% (417s)	100%(47s)
BWARE-RCS3	2256	98.7% (742s)	98.9% (685s)	99.0%(725s)
BWARE-p4	9341	98.4% (2097s)	99.3% (2279s)	99.4(790s)
BWARE-p9	371	64.7% (1104s)	67.9% (342s)	72.2%(492s)
EACSL	959	75.6% (64s)	93.3% (258s)	92.3%(293s)
SPARK	16773	80.5% (1769s)	83.6% (2757s)	84.0%(2298s)
WHY3	2003	38.9% (616s)	72.0% (1876s)	69.8% (1471s)
Total	32563	84.5% (6652s)	89.0% (8617s)	89.1%(6119s)

- ► Solution efficace (-29% en temps)
- Permet à Alt-Ergo d'être performant sur des problèmes fortement booléens
- ▶ Tout en restant performant sur des problèmes quantifiés

Results:



 $\label{eq:figure 4.19-Graphique du temps de résolution des solveurs SAT d'Alt-Ergo en fonction du nombre de buts résolus sur des fichiers issus de la preuve de programme.$

5) Conclusion

SMT solvers:

- hard to be efficient and generalist
- heuristics working on some problems but not on the others
- ullet result of \sim 40 years of research and experiments

Alt-Ergo has now four core solvers :

- cdcl performance on boolean problems
- tableaux to reduce boolean problems
- tableaux-cdcl not good enough
- cdcl-tableaux ⇒ best performance
- → default since Alt-Ergo 2.3.0 (February 2019)