

# Pluggable SAT-Solvers for SMT-Solvers

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# The SMT problem

## The SAT problem

Given a Boolean formula  $\mathcal{F}$ , is there an assignment for which  $\mathcal{F}$  evaluates to true?

## The SMT problem

SAT extended with a set of theories  $\mathcal{T}_1 \cup \mathcal{T}_2 \cdots \cup \mathcal{T}_n$

Example ( $\mathcal{EUF} \cup \mathcal{LA}(\mathbb{Z})$ ):

$$(x + 2y = 6 \vee y = 9) \wedge \neg(f(x) = f(y)) \wedge x = 2$$

## Some Useful Theories

### Theory of Linear Arithmetic ( $\mathcal{LA}$ )

$$\mathcal{F}_{\mathcal{LA}} = L_{(x=2)} \wedge L_{(x<3)}$$

### Theory of BitVectors ( $\mathcal{BV}$ )

$$\mathcal{F}_{\mathcal{BV}} = L_{(zext^{(2)}(x^{(2)})) >_u 15^{(4)}}$$

### Theory of Arrays ( $\mathcal{ARR}$ )

$$\mathcal{F}_{\mathcal{ARRULA}} = L_{(read(a,1)=2)} \wedge \neg L_{(read(write(a,3,1),i)=read(a,i))}$$

# Applications of SMT-Solvers

- Bounded Model Checking
- Equivalence Testing [GPB01]
- Property Driven Reachability Testing [CNR12]
- Scheduling [ABP<sup>+</sup>11]
- Test Case Generation [GLM12]
- Software model checking through Predicate Abstraction [FQ02]
- Program Synthesis [SGCF11]
- ...

## Pluggable SAT solvers: Motivation

- Developing a new (allround) SMT solver entails more than a new SAT solver.  $\rightarrow$  MATHSAT5  $\sim$  150kloc vs MiniSAT  $\sim$  6kloc
- Success of SAT solvers highly dependent on heuristics.
- Tuning SAT solvers requires investment of time and money.
- SAT-Solver is a deciding factor for  $BV$  &  $BV \cup ARR$  instances.
- We want to combine state-of-the-art SAT solvers & SMT solvers.
- This is NOT a straight forward bitblasting approach.

- 1 Introduction
- 2 The DPLL and DPLL( $\mathcal{T}$ ) algorithms
- 3 Architecture & Implementation
- 4 Experimental Evaluation
- 5 Demo
- 6 Conclusion & Future Work

## The DPLL algorithm

```
1: Preprocess( $\mathcal{F}$ )
2: while true do
3:    $BCP(\mathcal{F})$ 
4:   if not conflict then
5:     if all variables assigned then
6:       return SAT
7:     end if
8:      $decide()$ 
9:   else
10:     $C_{\text{conflict}} \leftarrow analyze()$ 
11:    if top level conflict found then
12:      return UNSAT
13:    end if
14:     $backtrack(C_{\text{conflict}})$ 
15:  end if
16: end while
```

# DPLL( $\mathcal{T}$ ) = DPLL + ..

- For correctness:
  - Theory consistency checks.
  - Case splitting.
- For optimization:
  - Early pruning.
  - Theory deductions.
- (Incrementality)

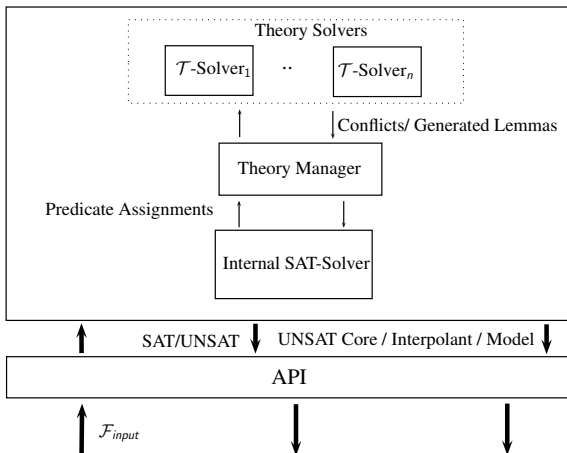
For specific details check [ST09].



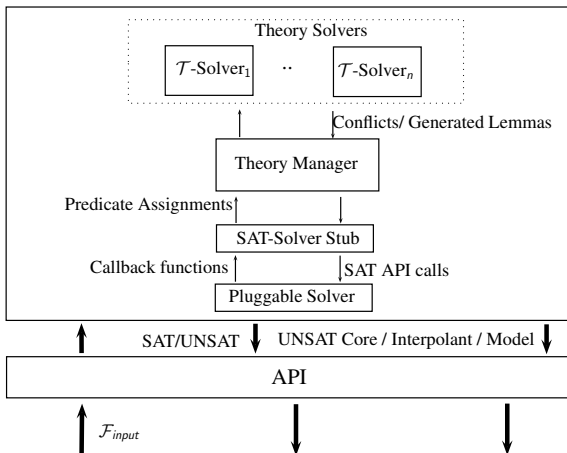
## (Simplified) DPLL( $\mathcal{T}$ ) algorithm

```
1: Preprocess( $\mathcal{F}$ )
2: while true do
3:    $BCP(\mathcal{F})$ 
4:   if not conflict and theories consistent then
5:     if all variables assigned and no case splitting needed.
6:       then
7:         return SAT
8:       end if
9:      $decide()$ 
10:  else
11:     $C_{\text{conflict}} \leftarrow analyze()$ 
12:    if top level conflict found then
13:      return UNSAT
14:    end if
15:     $backtrack(C_{\text{conflict}})$ 
16:  end if
end while
```

## DPLL( $\mathcal{T}$ ) Architectural Overview



## DPLL( $\mathcal{T}$ ) + Pluggable Solver Architectural Overview



## Pluggable SAT solvers: A quick overview

- 3rd Party SAT solvers can be plugged in MATHSAT5 by:
  - Implementing a worker interface.
  - Invoking required callback functions during search.
- The worker interface allows MATHSAT5 to:
  - Specify the problem for the SAT solver to solve.
  - Communicate deduced values.
- Callbacks allow the SAT solver to:
  - Communicate found (partial) models to MATHSAT5
  - Invoke  $\mathcal{T}$ -consistency checks.

## Pluggable SAT solvers: Requirements

- Must be able to act as an enumerator.
- Should support, solving under assumptions.
- Able to create new variables, add new clauses during search.
- Support variable freezing and reintroduction of eliminated variables.
- In order to support popping, must be able to delete all clauses containing certain variables.

## Worker Interface Functions

```
void solve(std::vector<int>& assump ,  
          std::vector<int>& c_assump );  
  
bool add_clause(std::vector<int>& clause ,  
               bool permanent ,  
               bool during_callback );  
  
void set_frozen(int var , bool b);  
  
int new_var(bool polarity , bool dvar);  
  
void enqueue_assignment(int assignment );  
  
void remove_clauses_containing(int v);
```

## Callback functions

```
TCODE no_conflict_ater_bcp(std::vector<int>& conf);  
  
TCODE model_found(std::vector<int>& conflict);  
  
void inform_hook_of_assignment(int assignment);  
  
void inform_hook_of_new_level();  
  
void inform_hook_of_backtrack(int level);  
  
void ask_hook_for_t_reason(int assignment,  
                           std::vector<int>& r);
```

## Pluggable SAT solvers: Two Case Studies

- Extending Minisat (& Cleaneling).
- Extending Fiver.



## Case Study A: Extending Minisat

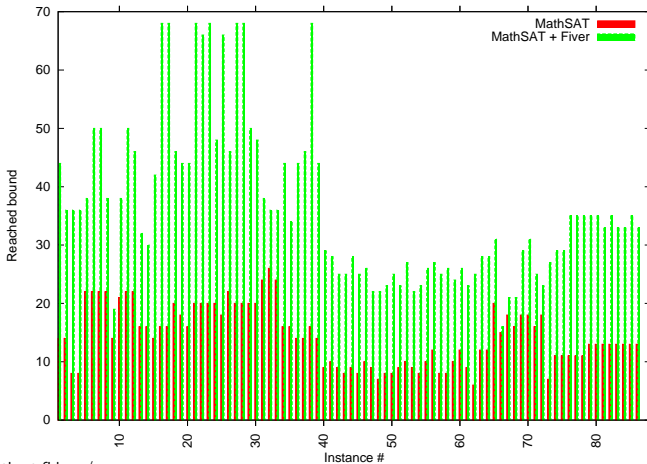
- The internal addClause method should be changed such that:
  - Clauses are added at the correct level.
  - For conflicts, jump back to the level, the conflict was introduced.
- The analyze method must take into account that assignments can be from deductions, asking the reason if necessary.
- After each round of BCP, in search which does not result in, theory propagation should be called -until fixpoint-.

- Once a complete model has been found a complete theory check should be called.
- Changing cleaneling is pretty similar!
- Example implementations for pluggable versions of Minisat & Cleaneling are available @ <http://mathsat.fbk.eu>.
- The changes required for each solver are around 180 lines of code.
- Does not support variable elimination.

## Case Study B: Extending Fiver

- Done completely at Intel. -minus some help in debugging-
- Supports preprocessing!

## Analysis of pluggable solver performance on Intel $\mathcal{BV}$ instances



## Analysis of pluggable solver performance on $BV \cup ARR$ instances

Benchmark Family	Size	MATHSAT5 <sub>MINISAT</sub>			
		#Solved	RT (sec)	#TO	#MO
brummayerbiere2	22	<b>15</b>	<b>1831</b>	<b>5</b>	<b>2</b>
brummayerbiere	293	184	17044	97	12
calc2	36	<b>36</b>	<b>4183</b>	<b>0</b>	<b>0</b>
stp	40	<b>29</b>	<b>1765</b>	<b>3</b>	<b>8</b>

Benchmark Family	Size	MATHSAT5			
		#Solved	RT (sec)	#TO	#MO
brummayerbiere2	22	15	2218	5	2
brummayerbiere	293	<b>229</b>	<b>25698</b>	<b>64</b>	<b>0</b>
calc2	36	30	7855	6	0
stp	40	26	2659	6	8

# DEMO



## Conclusion

We have presented a framework with which SAT-Solvers can be plugged in MATHSAT5 and used transparently. Next we have demonstrated the utility of such an approach on different instances.

## Future Work

- Provide support for proof logging, needed for other MathSAT functionalities such as Interpolation.
- Experiment with different type enumerators such as look-ahead Solvers.
- Experiment with pluggable Theory Solvers.



Questions?  
<http://mathsat.fbk.eu>

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