

CDCL(Crypto) SAT Solvers for Cryptanalysis

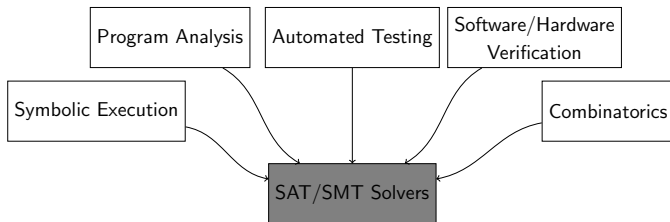
Saeed Nejadi and Vijay Ganesh

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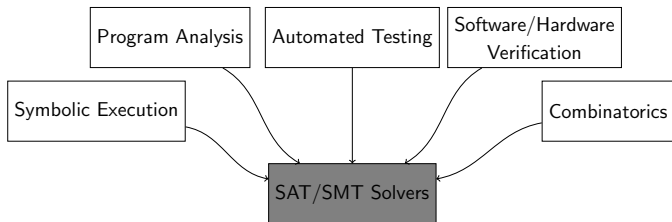
University of Waterloo

- **SAT Solvers:** Powerful general purpose search tools



Motivation

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- **Cryptanalysis:** Searching a huge search space for a secret key/value



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 - Differential cryptanalysis [Pro16]
 - RX-differentials [Ashur2017], [DW17]
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Question

Can we tailor internals of a SAT solver for a specific cryptographic problem to improve the solving time?

Outline

Boolean SAT Solvers

CDCL SAT Solvers

The CDCL(Crypto) Framework

Programmatic SAT Architecture

Case Studies

Algebraic Fault Attack

Differential Cryptanalysis

Boolean SAT Solvers

- NP-complete problem

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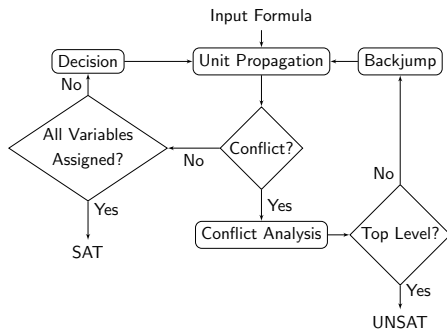
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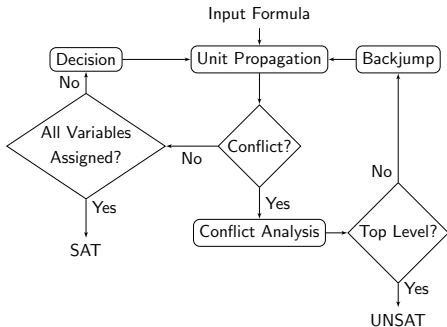
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- Example: $(x \vee y \vee \neg z) \wedge (\neg x \vee \neg y) \wedge z$

Unit propagation



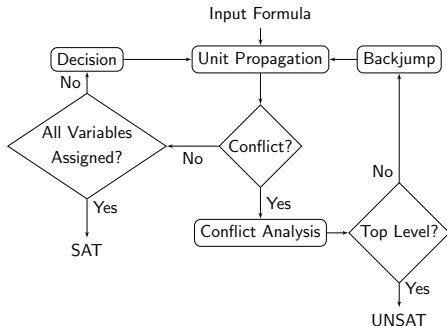
Unit propagation

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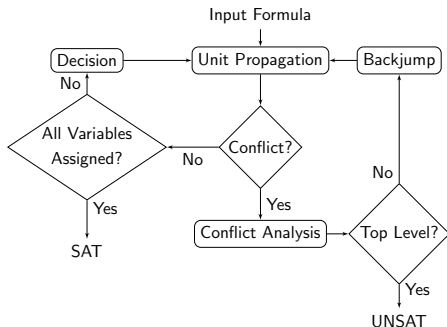
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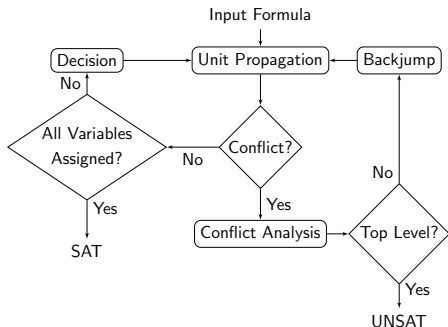
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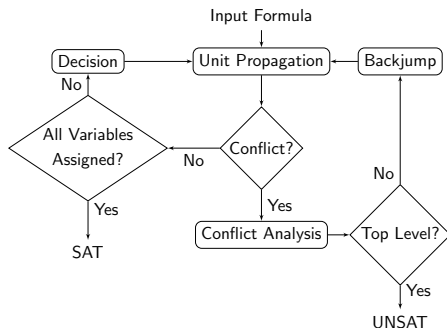
Unit propagation

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- $\neg z, \quad \neg z \rightarrow x, \quad x \rightarrow \neg y$



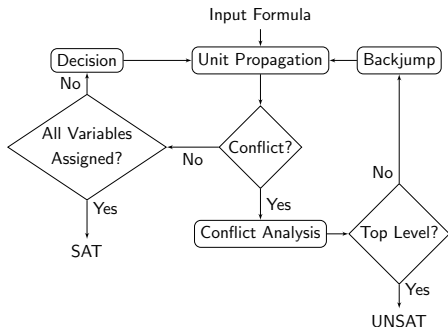
Decision/Branching heuristics

- Pick an unassigned variable and set it to False/True



Conflict analysis

- Find the root cause of conflict
- Encode it as a clause and add it back to the formula



The CDCL(Crypto) Framework

Lost in Translation

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- “Better” encoding vs. “Better” Propagation

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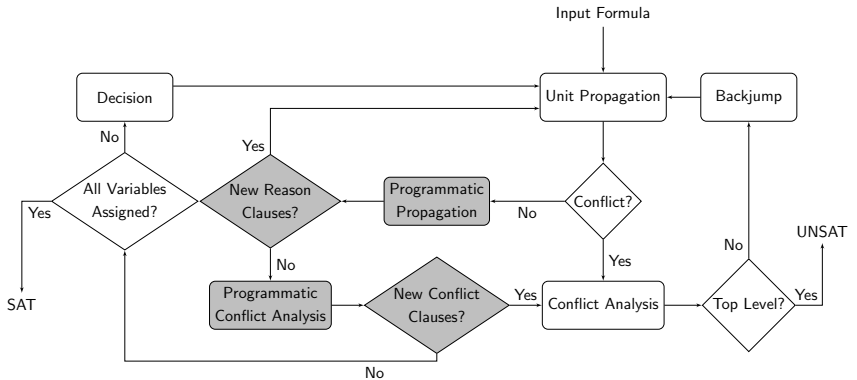
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- Conflict analysis callback
 - Called after *propagation* is done
 - Checks if partial assignment cannot be extended to a full solution

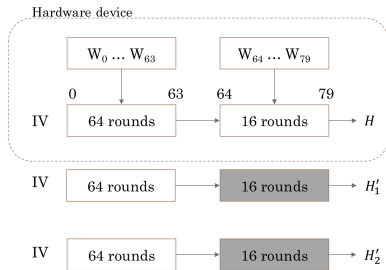
Programmatic SAT



Case Studies

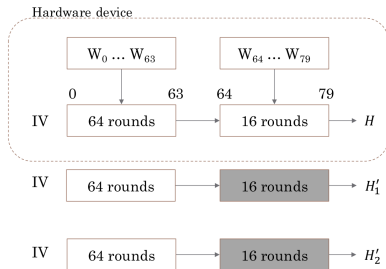
Hardware Fault Injection

- SHA-1 hash function



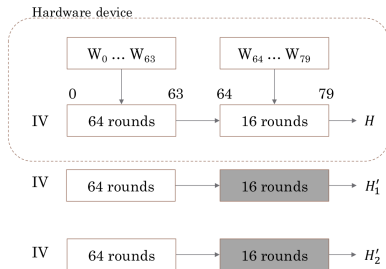
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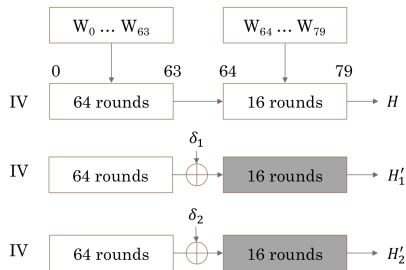
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- Using heat, EM, laser, ...



Algebraic Fault Analysis

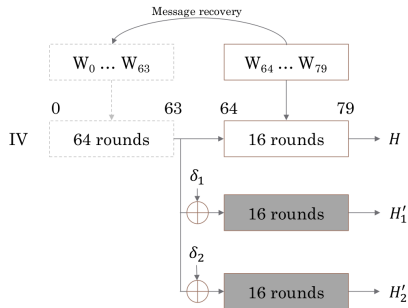
- $H = f_{0..79}(IV, W_{0..79})$
- This equation system will be encoded into CNF.
- Fault model: Constraints on δ_i .



$$H'_i = f_{64..79}(f_{0..63}(IV, W_{0..63}) \oplus \delta_i, W_{64..79})$$

Algebraic Fault Analysis

- Abstract away the common parts
- Verification of the solution will be needed

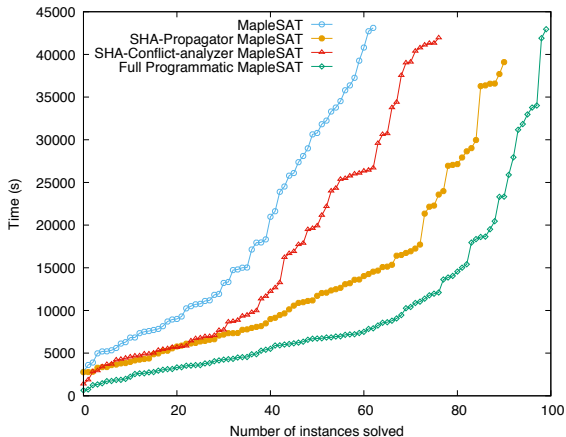


Algebraic Fault Analysis - Programmatic Approach

- Base SAT solver: MapleSAT
- Programmatic conflict analyzer
 - Embedding the verification loop
 - As soon as message word variables are set, they are ready to be verified
 - Early embedded check vs. Straightforward check after solving completely
- Programmatic propagator
 - Improving the propagation flow of multi-operand additions
 - Generating *reason clauses* in each column addition when output bits are missed

Algebraic Fault Analysis - Results

- Recovering SHA-256 message bits
- 14.3x speed-up on average
- 17 fewer faults were needed compared to the previous works



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- Gathering statistical information about the differences (finding bias/non-randomness).
- Differential Path: A trace of differentials over smaller steps in the function
- Collision: a differential path with final difference equal to zero.

- Guess-and-determine solvers:
 - Very similar search approach to SAT solvers
 - Dedicated propagators for differential propagation rules
 - Dedicated branching heuristics
 - State-of-the-art results on SHA-256 collision (31 steps)

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- Example rule: $r = IF(x, y, z)$, $r \leftarrow r \oplus x$, $x \leftarrow x \oplus r$.
- Full representation yields a blow up in size of the encoding
- An opportunity for Programmatic propagation

Differential Cryptanalysis - Results

- Collision on round-reduced SHA-256
- Modified the starting differential path of [Pro16]
- Base solver: MapleSAT
- Programmatic Propagator: Implemented a subset of differential propagation rules
- Programmatic conflict analyzer: Detects impossible differentials

- Found collision for 25 rounds of SHA-256 using MapleSAT(Crypto) in ~ 3.5 hours

Conclusions

- A framework on top of CDCL SAT solvers to implement cryptographic reasonings.
- Showcased the power of the framework in two cryptanalysis tasks.
- A bridge between two ends of a spectrum:
 - Performance of dedicated cryptanalysis tools
 - Flexibility and search power of SAT solvers
- Beating state-of-the-art in some cases but still a long way to match state-of-the-art in other cases

Thanks!
Questions?



Glenn De Witte.

Automatic sat-solver based search tools for cryptanalysis.

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- Here the T -solver can return the clause $\neg A$ (i.e. $x^2 \geq 0$).