

Automating the Search for Cryptanalytic Attacks

Maria Eichlseder

- FSE 2024
- Leuven · 26 Mar 2024



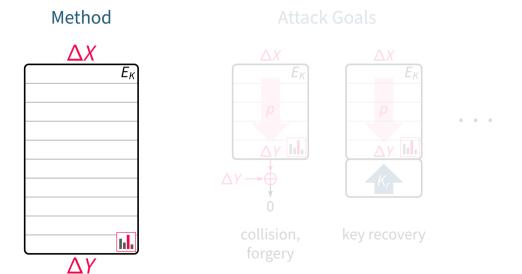
> https://iaik.tugraz.at

‡ Outline

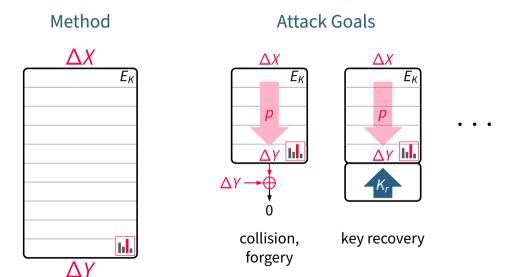
- Motivation
- Finding Distinguishers with MILP/SAT Solvers
 - Mixed-Integer Linear Programming (MILP)
 - Boolean Satisfiability and Constraint Programming (SAT/SMT, CP)
- Dedicated Algorithms
- Optimized Key Recovery Attacks
- Frameworks

Motivation

Differential Cryptanalysis [BS90]



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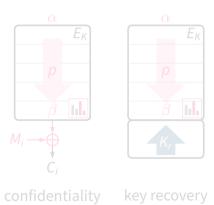


Linear Cryptanalysis [Mat93]

Method

Linear mask lpha E_K lı . Linear mask eta

Attack Goals

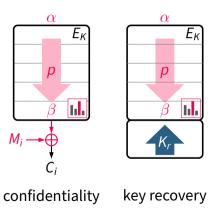


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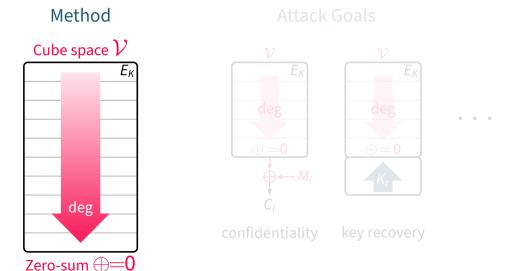
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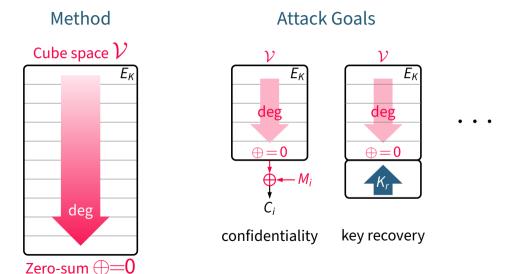
Attack Goals



Integral Cryptanalysis [Lai94; Knu94; KW02]



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How to Find Distinguishers



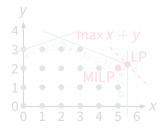
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 - SAT/SMT (Boolean SATisfiability/Sat. Modulo Theories)
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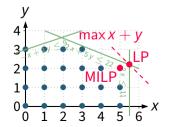
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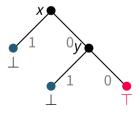




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Finding Distinguishers with MILP/SAT Solvers

Basic Approach

- Model constraints that characterize correct characteristics/solutions
 - Coarse-grained: truncated patterns (which S-boxes are active?)
 - Fine-grained: precise differences/masks

- Model cost (if applicable)
 - Express the search goal: any one / all / best / good solution(s)?

Mixed-Integer Linear Program (MILP)

Linear Programming (LP) is a method to solve optimization problems

- on the real-valued, positive decision variables $x \in \mathbb{R}^d, x \geq 0$
- with a linear objective function (min or max) $f(x) = c^{\mathsf{T}} x = \sum_{i=1}^{d} c_i x_i$
- under *J* linear constraints (s.t.) $Ax \le b$, i.e., $\sum_{i=1}^d a_{ji}x_i \le b_j$ for $1 \le j \le J$:

$$\max_{x \in \mathbb{R}^d} \left\{ c^\mathsf{T} x \mid Ax \le b \land x \ge 0 \right\}$$

Mixed-Integer Linear Programming (MILP) allows some of the decision variables to be constrained to integer values: $x \in \mathbb{Z}^i \times \mathbb{R}^{d-i}$.

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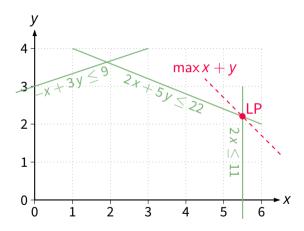
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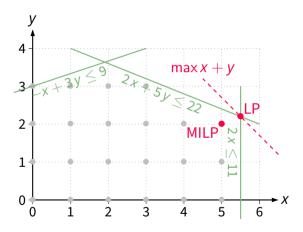
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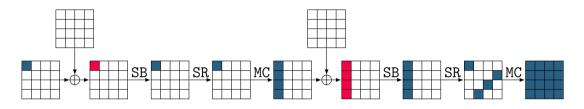
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LP vs. MILP

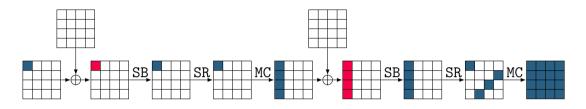


LP vs. MILP

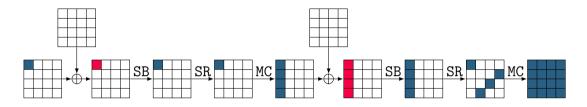




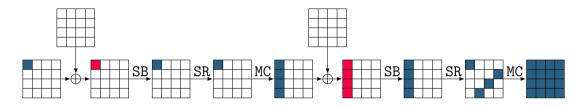
- AddRoundKey: input = output
- SubBytes: input = output, cost = sum(inputs)
- ShiftRows: variable renaming
- MixColumns: for each active column: sum(inputs) + sum(outputs) ≥ 5 (= \mathcal{B})



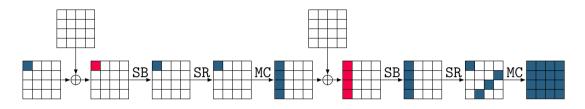
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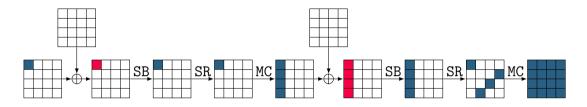
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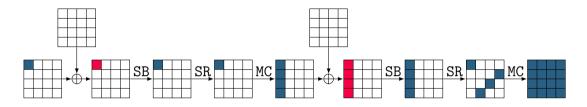
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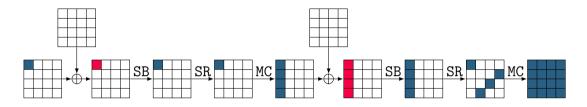
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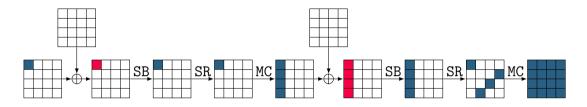
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Variables:

- $S_{r,i,j} \in \{0,1\}$: Is S-box in row i, column j in round r active?
- $M_{r,j} \in \{0,1\}$: Is MixColumns j in round r active?

Linear Program:

$$\min \sum_{r,i,j} S_{r,i,j} \qquad \qquad \text{(Min \# active S-boxes)}$$
 s.t. $\mathcal{B} \cdot M_{r,j} \leq \sum_i S_{r,i,(i+j)\%4} + \sum_i S_{r+1,i,j} \leq 8 \cdot M_{r,j} \qquad \text{(For each MixColumns)}$
$$\sum_{i,j} S_{0,i,j} \geq 1 \qquad \qquad \text{(Non-triviality)}$$

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MILP - Example application: AES - Code in sagemath

```
#!/usr/bin/env sage
rounds = range(4)
p = MixedIntegerLinearProgram(maximization=False)
S = p.new_variable(name='sbox', binary=True)
M = p.new_variable(name='mcol', binary=True)
for r in rounds:
    for i in [0..3]:
        activecells = sum(S[r,i,(i+j)\%4] \text{ for } i \text{ in } [0..3]) \setminus
                     + sum(S[r+1,i,j] for i in [0..3])
        p.add_constraint(5*M[r,j] <= activecells <= 8*M[r,j])</pre>
p.add_constraint(sum(S[0,i,j] for i in [0..3] for j in [0..3]) >= 1)
p.set_objective(sum(S[r,i,j] for r in rounds for i in [0..3] \
                                                for j in [0..3]))
p.solve()
print(p.get_objective_value(), p.get_values(S))
```

MILP - Advanced models

- Modeling more complex relations of "allowed transitions" accurately
 - Activity patterns for linear layers: XOR, near-MDS matrices [ABD+23], ...
 - DDT/LAT for bitwise S-box models [SHW+14b; SHW+14a], ARX [FWG+16]
- Need to translate vertex representation into half-space representation

XOR:
$$\begin{cases} \mathsf{inputs}\,\Box, \Box \to \mathsf{output}\,\Box \\ \mathsf{inputs}\,\Box, \blacksquare \to \mathsf{output}\,\blacksquare \\ \mathsf{inputs}\,\blacksquare, \Box \to \mathsf{output}\,\blacksquare \\ \mathsf{inputs}\,\blacksquare, \blacksquare \to \mathsf{output}\,\Box \mathsf{or}\,\blacksquare \end{cases} \Rightarrow \begin{cases} I_1 + I_2 \ge O \\ I_1 + O \ge I_2 \\ I_2 + O \ge I_1 \end{cases}$$

For large tables, this becomes very heavy (e.g., 8-bit S-boxes [AST+17; SW23])

SAT/SMT/CP – Different Levels of Convenience

1 SAT (Satisfiability) Solvers: Find valid solution or prove unsatisfiability of CNF

$$\bigwedge_i \bigvee_j \ell_{i,j}$$
 with literals $\ell_{i,j} \in \{\mathsf{v}_{i,j}, \neg \mathsf{v}_{i,j}\}$

Example solvers: MiniSAT, lingeling, and a myriad others

- 2 SMT (Sat. Modulo Theories) Solvers: Accept a more general grammar including bitvector operations such as integer addition.

 Example solvers: STP ("Simple Theorem Prover"), ...
- 3 CP (Constraint Programming) Solvers: Accept an even more general grammar (depends on solver).

Example solvers/frameworks: MiniZinc, Z3, Choco, ...

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SAT/SMT/CP for Finding Distinguishers [MP13; Köl14]

- Solves a constraint satisfaction problem, may not be optimal
 - "Emulate" optimization: "is there a solution better than $X, X + 1, X + 2, \dots$?"
- Useful to find valid solutions under some constraints
 - Finding characteristics that follow a given truncated pattern
 - Finding solutions for other crypto problems (preimage, ...)
- On the solution of the solutio
 - Not so good for modelling a cost sum or optimization [EME22]
 - Not perfectly parallelizable

Dedicated Tools for Hash Functions: Examples

SHA-1: HashClash
https://github.com/cr-marcstevens/hashclash

SHA-2: nldtool
https://github.com/iaikkrypto/nldtool

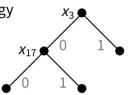
■ SHA-3: KeccakTools

○ https://github.com/KeccakTeam/KeccakTools

Dedicated Guess-and-Determine Search

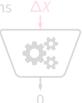
Guess-and-Determine Search is a general search strategy

- Traverse search tree to find a valid solution
- SAT solvers use it on CNF level
- This is an example on small (differential) circuits



• nldtool: Automated search for characteristics and solutions

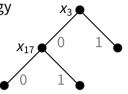
- Hash collision search
- Application example: SHA-2 [MNS11; MNS13; DEM15]



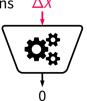
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Guess-and-Determine Search Algorithm

while there are undetermined bits do

Decision (Guessing)

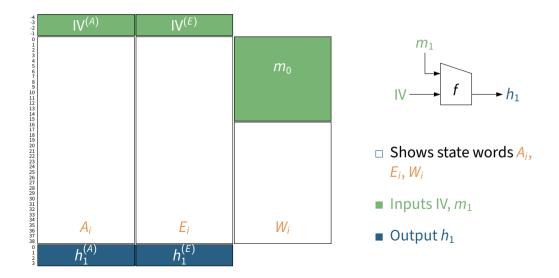
- 1 Pick an undetermined bit
- 2 Constrain this bit

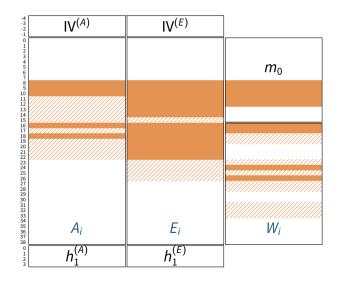
Deduction (Propagating)

- 3 Propagate the new information to other variables and equations
- 4 if no inconsistency is detected, goto step 1

Correction (Backtracking)

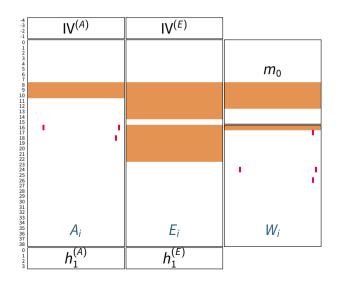
- **5 if** possible, apply a different constraint to this bit, goto step 3
- 6 else undo guesses until this critical bit can be resolved



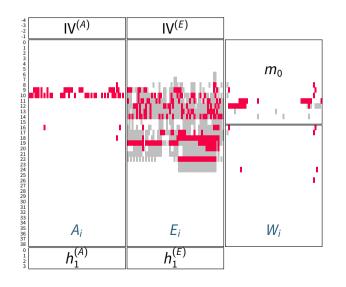


Starting point:

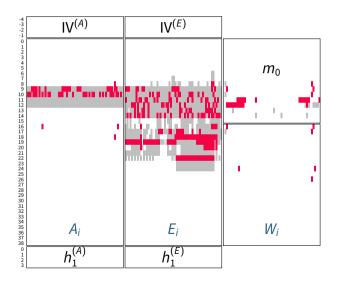
- "Local Collision" with few active message words
- Active words with differences [?]
- No differences [-] (cancellation required)
- □ No differences [-]



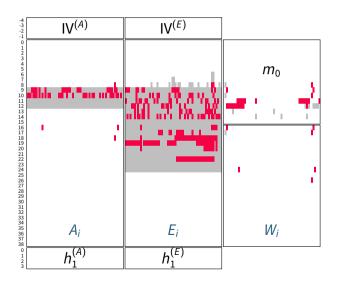
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- Fixed inactive bits [0,1]



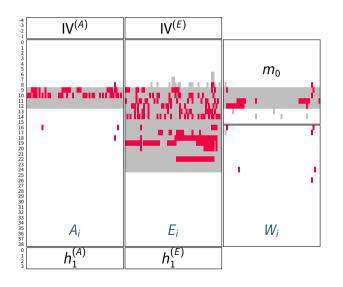
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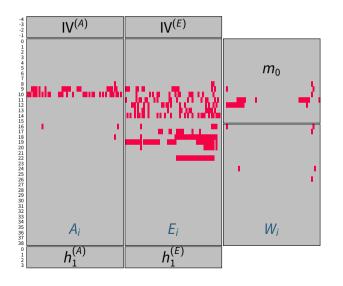
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κ.

Optimized Key Recovery Attacks

The Need for Tools

- Key recovery has long been ignored
- Fewer choices to make for the attacker

...but ...

- Optimizations involve choices and tradeoffs
- Precise evaluation is tedious
- Optimal" distinguisher doesn't guarantee optimal attack

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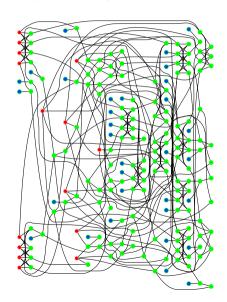
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- Guess-and-determine attacks:
 - Autoguess [HE22a] a
- Differential cryptanalysis:

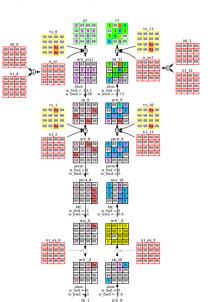
Integral cryptanalysis



ahttps://github.com/hadipourh/autoguess

- Guess-and-determine attacks:
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 - KYRYDI | BDD+24 | (7)
- Integral cryptanalysis



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bhttps://extgit.iaik.tugraz.at/castle/tool/keyrecoverytool

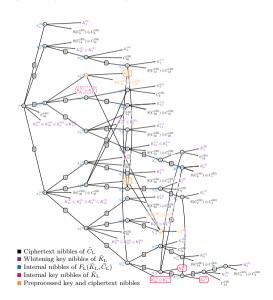
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chttps://gitlab.inria.fr/capsule/kyrydi

- Guess-and-determine attacks:
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 - Graph-based [HE22b] 🖸 d
 - AutoPSy [HSE23] (7)



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ehttps://github.com/hadipourh/zero

- % Preferably use a joint model for distinguisher and key recovery
 - ➤ Only works for satisfiability-based distinguishers
- **Complexity formulas** are often complicated
 - Mix of polynomial/exponential terms; simplified assumptions
- Multi-step processes lead to heavy models
- Very different types of key schedules
- Many different optimizations and strategies

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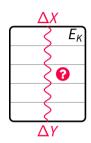
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Impossibility-based Distinguishers

Some distinguishers are based on the **non-existence** of a valid characteristic:

- Differential > Impossible differentials
- Linear > Zero-correlation linear approximations
- Integral > Division/monomial trail; ZC-based integrals

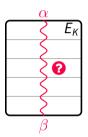


However, models for full attacks need **solution-based** distinguisher models (or a quantified language like QSAT)

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- Differential > Impossible differentials
- Linear > Zero-correlation linear approximations
- Integral > Division/monomial trail; ZC-based integrals

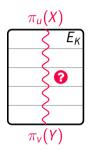


However, models for full attacks need solution-based distinguisher models (or a quantified language like QSAT)

Impossibility-based Distinguishers

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Two Ways of Modelling Impossibility

Unsatisfiability-based:

- $\Delta X = \Delta X = \Delta X$
- > First specify distinguisher, then check
- Precise, but potentially slow

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Example: Finding Full ID/ZC/Integral Attacks [HSE23]

■ Impossible-differential (ID) attacks

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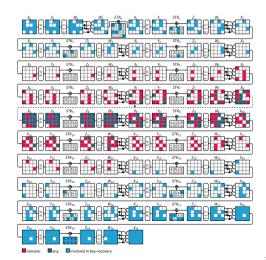


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- ZC-based integral attacks*



Frameworks

What exactly is a "Framework"?

Judging from paper titles, we have a plethora of frameworks, but ...

- Generality & Applicability
- Reuseability & Extensibility
- Maintainability & Verifiability

Frameworks: Examples

- CryptoSMT [Köl14; AK18] https://github.com/kste/cryptosmt
 - Differential/linear trails, clustering, key/preimage recovery, ...;
 based on SMT (STP, Boolector, CryptoMiniSat)
- CASCADA [RR22] https://github.com/ranea/CASCADA
 - Differential/linear trails, impossible differentials/zero-correlation, ...;
 based on SMT
- ☐ CLAASP [BGG+23] ↑ https://github.com/Crypto-TII/claasp
 - All of the above, neural tests; supports many solvers

Frameworks: Challenges

- Cipher representation
 - Based on building blocks? As a DAG? Software/hardware code?

- Efficiency vs. precision
- Simplicity vs. dedicated optimizations
- Meta-challenges: Conflicting incentives in academia

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Conclusion

- Tools and solvers can help find attacks and derive bounds
- Q Very active area with many open challenges
 - More efficient and precise models
 - Application to other design paradigms and attack techniques
 - Modeling full attacks (not just the distinguisher)
 - Frameworks and reuseability

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