# Cluster Search and MILP Modeling for Differential Attacks 

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## Symmetric Cryptography



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- Stream ciphers
- Block ciphers


## Symmetric Cryptography



■ Block ciphers

## Block Cipher

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Given a key $K \in \mathbb{F}_{2}^{m}$ and a message $M \in \mathbb{F}_{2}^{N}$, a block cipher of block size n is an invertible function $E_{K}$ that encrypts the message $M$ in blocks of size $n$

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■ Used in Advance Encryption Standard (AES)

## DES

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## AES

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■ Proposed in 1998 by Daemen and Rijmen for the 1997 NIST competition

## Block Cipher Operations

Linear operations

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Linear operations<br>■ Constant additions

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## Substitution Box

A substitution box (S-box), is a non-linear operation usually represented as a look-up table:
$\begin{array}{clll}\mathbf{S}: & \mathbb{F}_{2}^{m_{1}} & \longrightarrow & \mathbb{F}_{2}^{m_{2}} \\ & \mathbf{x} & \mapsto & \mathbf{S}(\mathbf{x})\end{array}$

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AES S-box:
$\mathbb{F}_{256}$
1 If $x \neq 0$, replace $x$ by its inverse, $x=x^{-1}$ in $\mathbb{F}_{2}^{8 *}$

* Generated by the polynomial $m(X)=X^{8}+X^{4}+X^{3}+X+1$


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1 If $x \neq 0$, replace $x$ by its inverse, $x=x^{-1}$ in $\mathbb{F}_{2}^{8 *}$
2 $x=A x+b$, where $A$ is a fix $8 \times 8$ binary matrix and $b$ is a fix 8 binary vector

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Find a pair $(\Delta, \nabla) \in \mathbb{F}_{2}^{2 n}$ such that $p\left(E_{K}(x \oplus \Delta)=\nabla\right) \gg 2^{-n}$

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Find a pair $(\Delta, \nabla) \in \mathbb{F}_{2}^{2 n}$ such that $p\left(E_{K}(x \oplus \Delta)=\nabla\right) \gg 2^{-n}$
Study the propagation of input differences throughout the cipher:

$$
\nabla=E_{K}(P) \oplus E_{K}(P \oplus \Delta)
$$



## Differential Attack

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- The pair $(\Delta, \nabla)$ is referred to as a differential usually hard to find
- Differential characteristic: $\Delta=\delta_{0} \rightarrow \delta_{1} \rightarrow \cdots \rightarrow \delta_{r}=\nabla$
- Analyze the differential behavior of cipher operations


## Computing the Probability

$$
p\left(\delta_{0} \rightarrow \delta_{1} \cdots \rightarrow \delta_{r}\right)
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$\rightarrow$ On the board

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$$
\operatorname{DDT}\left(\Delta_{i}, \nabla_{0}\right)=\#\left\{\mathbf{x} \in \mathbb{F}_{2}^{n}: S(\mathbf{x}) \oplus S\left(\mathbf{x} \oplus \Delta_{i}\right)=\nabla_{o}\right\}
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| I Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 8 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 2$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 4 | 0 | 0 | 0 | 4 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 4 | 0 | 0 | 4 |  |
| $0 \times 5$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 6$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 4 | 4 | 0 | 0 | 0 |  |

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■ Follow the differential behaviour of the nonlinear operation
-> Obtain differential characteristics

## Clusters

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## Cluster

A cluster is a set of differential characteristics, for a given number of rounds, that have the same input and output difference

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A cluster is a set of differential characteristics, for a given number of rounds, that have the same input and output difference

$$
\begin{gathered}
\Delta=\delta_{0}^{j} \rightarrow \delta_{1}^{j} \rightarrow \cdots \rightarrow \delta_{r}^{j}=\nabla \\
p(\Delta \rightarrow \nabla) \approx \sum_{j} p\left(\delta_{0}^{j} \rightarrow \delta_{1}^{j} \rightarrow \cdots \rightarrow \delta_{r}^{j}\right)
\end{gathered}
$$

## Implementing Cluster Search

[Chen et all., Improved Differential Characteristic Searching Methods, 2015]

## Naive approach

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■ Finds the whole cluster (for small number of rounds)

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- From the truncated path: selects a round with few values in the middle
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■ Finds the whole cluster (for small number of rounds)

- Uses too much memory


## MILP Modeling

MILP: Mixed-Integer Linear Programming

## MILP Modeling

ILP: Integer Linear Programming
■ Minimize or maximize an objective function

$$
\sum_{i} a_{i} X_{i}
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$\square$ Constraints $\sum b_{i} X_{i} \geq b \quad \sum c_{i} X_{i} \leq c \quad \sum d_{i} X_{i}=d$

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ILP: Integer Linear Programming

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## XOR Truth Table

| $a, b, c \in \mathbb{F}_{2}$ |  |  |
| :---: | :---: | :---: |
| $a$ | $b$ | $c=a \oplus b$ |
| 0 | 0 | 0 |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 0 |

## MILP Modeling

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Non-valid transitions:
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Non-valid transitions:

$$
\begin{gathered}
(0,0,1),(0,1,0),(1,0,0),(1,1,1) \\
a+b \geq c \quad a+c \geq b \\
a+b+c \leq 2
\end{gathered}
$$

## Step1 MILP Modeling

Working with truncated characteristics:

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Working with truncated characteristics:

## Abstracted XOR Truth Table

| $a$ | $b$ | $c=a \oplus b$ |
| :---: | :---: | :---: |
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| 0 | 1 | 1 |
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| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 0 |
| 1 | 1 | 1 |

Non-valid transitions:
$(0,0,1),(0,1,0),(1,0,0)$

## Step1 MILP Modeling

Working with truncated characteristics:

## Abstracted XOR Truth Table

| $a$ | $b$ | $c=a \oplus b$ |
| :---: | :---: | :---: |
| 0 | 0 | 0 |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 0 |
| 1 | 1 | 1 |

Non-valid transitions:
$(0,0,1),(0,1,0),(1,0,0)$ $a+b+c \neq 1 \Rightarrow$

## Step1 MILP Modeling

Working with truncated characteristics:

## Abstracted XOR Truth Table

| $a$ | $b$ | $c=a \oplus b$ |
| :---: | :---: | :---: |
| 0 | 0 | 0 |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 0 |
| 1 | 1 | 1 |

Non-valid transitions:
$(0,0,1),(0,1,0),(1,0,0)$
$a+b+c \neq 1 \Rightarrow$
$a+b \geq c \quad a+c \geq b \quad b+c \geq a$

## Step1 MILP Modeling

Working with truncated characteristics:

## Abstracted XOR Truth Table

| $a$ | $b$ | $c=a \oplus b$ |
| :---: | :---: | :---: |
| 0 | 0 | 0 |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 0 |
| 1 | 1 | 1 |

Non-valid transitions:

$$
\begin{gathered}
(0,0,1),(0,1,0),(1,0,0) \\
a+b+c \neq 1 \Rightarrow \\
a+b \geq c \quad a+c \geq b \quad b+c \geq a
\end{gathered}
$$

Step1 objective function:

## Step1 MILP Modeling

Working with truncated characteristics:

## Abstracted XOR Truth Table

| $a$ | $b$ | $c=a \oplus b$ | Non-valid transitions: |
| :---: | :---: | :---: | :---: |
| 0 | 0 | 0 | $(0,0,1),(0,1,0),(1,0,0)$ |
| 0 | 1 | 1 | $a+b+c \neq 1 \Rightarrow$ |
| 1 | 0 | 1 | $a \geq c \quad a+c \geq b \quad b+c \geq a$ |
| 1 | 1 | 0 |  |

Step1 objective function:

$$
\sum_{i, r} X_{i, r}
$$

where $i$ word position, $r$ round, $X_{i, r}=1$ if there is a non-zero value at the S-box, zero otherwise.

## Step2 MILP Modeling

How do we model a non-linear function only with linear constraints?

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How do we model a non-linear function only with linear constraints?

- H-representation of the convex-hull



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■ Product-of-Sum Representation of Boolean Functions

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■ Product-of-Sum Representation of Boolean Functions Quine-McCluskey (QM) algorithm

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How do we model a non-linear function only with linear constraints?

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- Logical condition techniques for 8-bit S-boxes


## Step2 MILP Modeling

How do we model a non-linear function only with linear constraints?

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■ Logical condition techniques for 8-bit S-boxes

## Minimization

■ Greedy algorithm

- MILP minimization
- Prime implicants table


## Step2 MILP Modeling

How do we model a non-linear function only with linear constraints?
■ H-representation of the convex-hull


■ Product-of-Sum Representation of Boolean Functions Quine-McCluskey (QM) algorithm
■ Logical condition techniques for 8-bit S-boxes

## Minimization

■ Greedy algorithm
■ MILP minimization

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## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]

■ Establish valid transitions: DDT

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]
■ Establish valid transitions: DDT

| : Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 8 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 2$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 4 | 0 | 0 | 0 | 4 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 4 | 0 | 0 | 4 |  |
| $0 \times 5$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 6$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 4 | 4 | 0 | 0 | 0 |  |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]
■ Establish valid transitions: *-DDT

| d: Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]
■ Establish valid transitions: *-DDT

| $\Delta:$ Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]
■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla$ : output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| i: Input <br> difference | $\nabla$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 8 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 2$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 5$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 6$ | 0 | 2 | 2 | 0 | 0 | 2 | 2 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]
■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla$ : output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| i: Input <br> difference | $\nabla$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]

■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input | $\nabla$ : output difference |  |  |  |  |  |  |  |
| difference | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 8 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 4 | 0 | 0 | 0 | 4 |
| 0x4 | 0 | 0 | 0 | 0 | 4 | 0 | 0 | 4 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 4 | 4 | 0 | 0 | 0 |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]

■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input | $\nabla$ : output difference |  |  |  |  |  |  |  |
| difference | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]

■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ H-representation of convex-hull:

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
|  | 0x0 | $0 \times 1$ | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]

■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ H-representation of convex-hull:

- Set of valid transitions DDT:


## ■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input | $\nabla$ : output difference |  |  |  |  |  |  |  |
| difference | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

## DDT Modeling with MILP

[Abdelkhalek et al., MILP Modeling for (Large) S-boxes to OptimizeProbability of Differential Characteristics, 2017]
■ Establish valid transitions: *-DDT

| : Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ H-representation of convex-hull:

- Set of valid transitions DDT:

$$
\{(0,0),(1,1),(1,2),(1,5), \ldots,(7,4)\}
$$

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
|  | 0x0 | 0x1 | $0 \times 2$ | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

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|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
|  | 0x0 | 0x1 | $0 \times 2$ | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

■ H-representation of convex-hull:

- Set of valid transitions DDT:

$$
\{(0,0),(1,1),(1,2),(1,5), \ldots,(7,4)\}
$$

- Set of valid transitions 2-DDT and 4-DDT


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|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input | $\nabla$ : output difference |  |  |  |  |  |  |  |
| difference | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

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$$
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$$

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$$
\{(0,0),(1,1),(1,2),(1,5), \ldots,(6,6)\}
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■ Establish valid transitions: *-DDT

| $\Delta:$ Input <br> difference | $\nabla:$ output difference |  |  |  |  |  |  |  |  |
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|  | $0 \times 0$ | $0 \times 1$ | $0 \times 2$ | $0 \times 3$ | $0 \times 4$ | $0 \times 5$ | $0 \times 6$ | $0 \times 7$ |  |
| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 2$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
| $0 \times 4$ | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |  |
| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 6$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
| $0 \times 7$ | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |  |

■ Use 2-DDT and 4-DDT:

| $\Delta$ : Input difference | $\nabla$ : output difference |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input | $\nabla$ : output difference |  |  |  |  |  |  |  |
| difference | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | 1 | 0 | 0 | 0 |

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- Set of valid transitions DDT:

$$
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$$

- Set of valid transitions 2-DDT and 4-DDT

$$
\begin{aligned}
& \{(0,0),(1,1),(1,2),(1,5), \ldots,(6,6)\} \\
& \{(0,0),(3,3),(3,7),(4,4), \ldots,(7,4)\}
\end{aligned}
$$

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| $0 \times 0$ | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |  |
| $0 \times 1$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
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| $0 \times 3$ | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |  |
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| $0 \times 5$ | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |  |
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| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x2 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x3 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x4 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x5 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x6 | 0 | 1 | 1 | 0 | 0 | 1 | 1 | 0 |
| 0x7 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| $\Delta$ : Input | $\nabla$ : output difference |  |  |  |  |  |  |  |
| difference | 0x0 | 0x1 | 0x2 | 0x3 | 0x4 | 0x5 | 0x6 | 0x7 |
| 0x0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x2 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 | 1 |
| 0x4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 | 1 |
| 0x5 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0x7 | 0 | 0 | 0 | 1 | , | 0 | 0 | 0 |

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$$
\begin{aligned}
& \{(0,0),(1,1),(1,2),(1,5), \ldots,(6,6)\} \\
& \{(0,0),(3,3),(3,7),(4,4), \ldots,(7,4)\}
\end{aligned}
$$

$$
\sum_{i=0}^{7} a_{i}^{j} x_{i}+a^{j} \geq 0
$$

## MILP Minimization of the H -representation

[Sasaki and Todo, New Algorithm for Modeling S-box in MILP Based Differential and Division Trail Search, 2017]

■ Assign a variable to each inequality (Value 0 or 1)

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- 1 if it is used
- 0 otherwise
- Relate each inequality to the points that satisfy it

■ Minimize the number of inequalities constrained to: all points must be included

■ Step2 objective: Maximize the probability of the transitions throughout 2-DDT and 4-DDT

## Lightweight Cryptography

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Need for encryption and authentication on constrained devices

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- Small hardware footprint


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■ Low-energy consumption

## Warp

[Banik et all., WARP : Revisiting GFN for Lightweight 128-bit Block Cipher, 2020]

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■ 128-bit Generalize Feistel cipher

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$F=A R K \circ S$

- 128-bit Generalize Feistel cipher
- 128-bit key size


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## Warp

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$F=A R K \circ S$

- 128-bit Generalize Feistel cipher
- 128-bit key size
- Linear key schedule
- 41 round function iterations


## Clusters for Warp

| Rounds | S-boxes | n_sol | Step2 -log(prob) | Cluster size | Cluster prob |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 10 | 17 | 2 | 34 | 4 | 32 |
| 11 | 22 | 2 | 44 | 4 | 42 |
| 12 | 28 | 4 | 56 | 16 | 53 |
| 13 | 34 | 2 | 68 | 512 | 59 |

## Clusters on Warp

| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 7, | d, | a, | $d$ | 0 , | 0, | 0, | 0, | 0 , | a, | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0, | a, | d, | 0 , | 0 , | 0, | $d$ | 0, | 0, |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | d, | 7 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0. | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | a, | $d$, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | $a$, | a, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | a, | 0 , | 0, | 0 , | 0 , | 0, | 0, | 0 , | 0, |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0, |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0, | 0 , | 0 , | 0, | 0 , | 0 , | a, | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0, | 0 , | 0 , |
| $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0, | $a$, | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0. | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | a, | 0 |
| 0 , | 0 , | 0 , | 0, | a, | 0 , | a, | a, | 0 , | 0, | 0 , | 0, | 0, | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0, | 0, | 0 , | a, | 0 , | 0, | 0, | 0 , | a, |
| 0 , | $a$, | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5. | 0 , | 0 , | 0 , | a, | 0 , | 0 , | a, | 0, | a, | 0 , | 0 , | 0 , | 0 , | $a$, | 0 , | 0, | 0 , | 0 , | 0 |
| 0. | 0, | 0 , | 0. | 7 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | a, | a, | 0 , | 0 , | a, | 0, | 0 , | d, | 0 , | 0 , | 5, | 0 , | 0 , | a, | 0, | $a$, | 0. |

## Clusters on Warp

| 0 , | 0 , | 0 , | 0, | 0 , | 0, | 7. | d, | $a$, | $d$ | 0 , | 0, | 0, | 0, | 0 , | a, | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | a, | d, | 0 , | 0 , | 0, | d, | 0 , | 0, |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | a, | $a$, | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | d, | 7 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | a, | d, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | a, | $a$, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | a, | 0, | 0 , | 0 , | 0 , | 0. | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0, | 0, | 0 , | 0 , | 0, | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | $a$, | 0 , | 0, | 0, | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0, | 0, | a, | 0 , |
| 0 , | 0 , | 0 , | 0 , | a, | 0 , | a, | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0, | 0 , | a, |
| 0 , | a, | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5, | 0. | 0 , | 0 , | a, | 0 , | 0 , | a, | 0 , | a, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0, | 0, | 0 , |
| 0. | 0 , | 0 , | 0. | 7, | 0, | a, | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0, | 0 , | a, | a, | 0 , | 0 , | $a_{1}$ | 0, | 0 , | d, | 0 , | 0 , | 5, | 0 , | 0 , | a, | 0, | a, | 0. |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 7 , | d, | a, | $d$ | 0 , | 0 , | 0, | 0, | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | d, | 0 , | 0 , | 0 , | d, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | a, | 0. | 0, | 0 , | 0, | 0, | 0 , | 0, | 0, | d, | 7 , | 0 , | 0 , | 0 , | 0, | 0 , | a, | 0 , | 0. | 0 , | 0, |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | a, | d, | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | a, | a, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0, |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0. | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| a, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | a, | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 5, | 0 , |
| 0 , | 0 , | 0 , | 0 , | a, | 0, | a, | 5. | 0 , | 0 , | 0 , | 0, | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0. | 0 , | a, |
| 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5. | 0. | 0 , | 0 , | a, | 0 , | 0, | $a_{1}$ | 0 , | a, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 7 , | 0, | a, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | $a$, | a, | 0 , | 0 , | $a_{1}$ | 0 , | 0 , | d, | 0 , | 0 , | 5, | 0 , | 0 , | a, | 0, | a, | 0 , |

## Clusters on Warp

|  |  |  |  |  |  |  |  |  |  |  | 0, |  |  |  |  |  |  |  |  |  |  |  |  | a, | d, | 0, | 0, | 0, | d, | 0 , | 0, |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | a, | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | d, | 7. | 0 , | 0, | 0 , | 0 , | 0 , | $a$, | 0, | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | $a_{1}$ | d, | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0. | a, | $a$, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0, | 0 , | 0 , |
| 0. | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0. | 0 , | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0, | 0 , | 0, | 0 , | 0. | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | $a$, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | $a$, | 0, | 0. | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0, | 0, | 0 , | 0 , |
| a, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | a, | 0 , | 0, | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0, | 0, | 0, | 0 , | 0, | 0, | a, | 0 , |
| 0 , | 0 , | 0 , | 0, | a, | 0 , | a, | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0, | a, | 0, | 0 , | 0, | 0 , | a, |
| 0 , | $a$, | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5, | 0 , | 0 , | 0 , | a, | 0 , | 0, | a, | 0 , | a, | 0 , | 0 , | 0 , | 0, | a, | 0 , | 0 , | 0, | 0 , | 0 , |
| 0, | 0 , | 0, | 0. | 7, | 0, | a, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | a, | a, | 0 , | 0, | a, | 0, | 0 , | d, | 0, | 0, | 5, | 0 , | 0, | a, | 0, | a, | 0. |
| 0 , | 0 , | 0, | 0, | 0 , | 0 , | 7 , | d, | $a$, | d, | 0 , | 0 , | 0, | 0 , | 0 , | $a$, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 | $a$, | d, | 0 , | 0 , | 0 , | d, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | $a$, | $a$, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | $d$, | 7, | 0 , | 0, | 0 , | 0, | 0 , | a, | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | d, | 0, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | a, | $a$, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | a, | 0, | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0, | 0. | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0, | 0, | 0 , | 0, | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | $a$, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | $a$, | 0, | 0, | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , |
| a, | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0, | 0 , | 5 , | 0 , |
| 0 , | 0 , | 0 , | 0. | $a$, | 0, | a, | 5, | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | a, | 0 , | 0, | 0, | 0 , | a, |
| 0 , | a, | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 5, | 0 , | 0 , | 0 , | a, | 0 , | 0 , | $a_{1}$ | 0, | a, | 0 , | 0, | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0, | 0, | 7, | 0, | a, | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0. | 0 , | $a$, | a, | 0 , | 0, | $a_{1}$ | 0 , | 0, | d, | 0, | 0, | 5, | 0, | 0, | a, | 0, | a, | 0 , |
| 0 , | 0 , | 0 , | 0. | 0. | 0 , | 7, | d, | a, | d, | 0 , | 0 , | 0, | 0. | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0, | a, | d, | 0 | 0. | 0 , | d, | 0 , | 0 , |
| 0 , | 0 , | 0, | 0. | 0. | 0, | 0, | 0 , | 0 , | 0. | a, | a, | 0, | 0, | 0, | 0 , | 0, | 0 , | 0, | 0 , | d, | 7. | 0, | 0, | 0 , | 0. | 0, | $a$, | 0, | 0. | 0 , | 0, |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | $a$, | d, | 0, | 0 , | 0 , | 0, | 0 , | 0, | 0. | 0, | 0, | 0, | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0. | 0. | 0 , | 0, | 0 , | 0 , | 0. | 0, | 0 , | 0, | 0. | $a$, | $a$, | 0 , | 0 , | 0. | 0 , | 0 , | 0. | 0 , | 0. | 0. | 0. | 0, | 0. | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | a, | 0 , | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0. | 0 , | 0, | a, | 0, | 0, | 0. | 0 , | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0. | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | a, | 0 , | 0 , | 0, | 0 , | 0. | 0 , | 0, | 0, | 0. | 0 , | 0 , | 0 , | 0 , |
| a, | 0 , | 0 , | 0. | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0. | 0, | 0, | 0, | 0, | 0, | d, | 0 , |
| 0 , | 0 , | 0 , | 0, | a, | 0 , | $a$, | d, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0. | 0 , | 0. | 0 , | 0, | a, | 0. | 0 , | 0, | 0 , | a, |
| 0 , | a, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5, | 0 , | 0 , | 0 , | $a$, | 0 , | 0 , | a, | 0 , | a, | 0 , | 0, | 0 , | 0, | $a$, | 0, | 0 , | 0. | 0 , | 0 , |
| 0 , | 0 , | 0, | 0, | 7. | 0 , | a, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0 , | a, | a, | 0 , | 0, | a, | 0 , | 0 , | d, | 0. | 0 , | 5, | 0, | 0, | a, | 0, | a, | 0 , |

## Clusters on Warp

|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | a, | d, | 0, | 0, | 0, | d, | 0, | 0 , |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | $a$, | a, | 0 , | 0. | 0 , | 0 , | 0 , | 0, | 0 , | 0, | d, | 7, | 0 , | 0 , | 0 , | 0 , | 0, | $a$, | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0 , | $a_{1}$ | $d$, | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0. | a, | $a$, | 0 , | 0, | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0, | 0, | 0. | 0, | 0 , | a, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0. | 0 , | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0. | 0 , | 0 , | 0 , | a, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | $a$, | 0 , | 0. | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , |
| a, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0, | 0, | 0, | 0, | 0, | 0, | 0, | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0, | 0 , | a, | 0 , |
| 0 , | 0 , | 0 , | 0, | a, | 0 , | a, | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | a, | 0 , | 0, | 0 , | 0 , | a, |
| 0 , | a, | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5, | 0 , | 0 , | 0 , | a, | 0 , | 0 , | a, | 0, | $a$, | 0 , | 0 , | 0 , | 0, | a, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0, | 0 , | 0, | 0. | 7, | 0, | a, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | a, | a, | 0, | 0, | a, | 0, | 0 , | d, | 0, | 0 , | 5, | 0 , | 0 , | a, | 0 , | a, | 0, |
| 0 , | 0 , | 0, | 0, | 0 , | 0 , | 7 , | d, | $a$, | d, | 0 , | 0 , | 0, | 0 , | 0 , | $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | a, | d, | 0 , | 0 , | 0 , | d, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | $a$, | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | $d$, | 7, | 0 , | 0, | 0 , | 0, | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | d, | 0 , | 0, | 0. | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | $a$, | 0 , | 0, | 0, | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0, | 0 , | $a$, | 0 , | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0, | 0. | 0 , | 0, | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0, | 0, | 0 , | 0 , | 0, | 0, | 0, | 0 , | 0 , | a, | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | a, | 0. | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , |
| a, | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | a, | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 5, | 0 , |
| 0 , | 0 , | 0 , | 0. | $a$, | 0, | a, | 5, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0, | 0 , | 0 , | a, |
| 0 , | a, | 0, | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5, | 0 , | 0 , | 0 , | a, | 0 , | 0, | $a_{1}$ | 0, | $a$, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0, | 0, | 7, | 0, | a, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0. | 0, | $a$, | a, | 0, | 0, | $a_{1}$ | 0, | 0, | d, | 0 , | 0 , | 5, | 0 , | 0, | a, | 0 , | a, | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 7, | d, | a, | d, | 0 , | 0 , | 0 , | 0, | 0 , | $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0, | a, | d, | 0, | 0 , | 0 , | d, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | a, | 0 , | 0, | 0. | 0 , | 0, | 0 , | 0, | 0 , | $d$, | 7. | 0 , | 0, | 0 , | 0. | 0, | $a$, | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | d, | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | $a$, | $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | a, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | a, | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0. | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | a, | 0 , | 0, | 0 , | 0 , | 0, | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , |
| a, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0 , | a, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0. | 0, | 0 , | 0 , | 0 , | 0 , | d, | 0 , |
| 0 , | 0 , | 0 , | 0, | $a$, | 0 , | $a$, | d, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0, | $a$, | 0. | 0 , | 0, | 0 , | a, |
| 0 , | a, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5 , | 0 , | 0 , | 0 , | $a$, | 0 , | 0, | a, | 0 , | $a$, | 0 , | 0. | 0 , | 0 , | $a$, | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0, | 0, | 7. | 0, | a, | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0, | a, | a, | 0, | 0, | a, | 0, | 0, | d, | 0, | 0 , | 5. | 0, | 0 , | a, | 0 , | a, | 0, |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 7. | d, | $a$, | d, | 0 , | 0, | 0, | 0 , | 0 , | a, | 0, | 0, | 0, | 0 , | 0 , | 0, | 0 , | 0 , | $a$, | d, | 0 , | 0 , | 0, | d, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | $a$, | a, | 0 , | 0. | 0 , | 0 , | 0 , | 0, | 0, | 0, | d, | 7, | 0 , | 0 , | 0 , | 0 , | 0, | $a$, | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | $a$, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0. | 0 , | 0 , | $a$, | d, | 0, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0. | 0 , | 0 , | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0 , | 0. | a, | $a$, | 0 , | 0, | 0 , | 0. | 0, | 0 , | 0 , | 0, | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
| 0 , | 0 , | 0, | 0. | 0. | 0, | 0, | 0, | 0 , | 0 , | 0 , | 0, | 0, | 0. | 0, | 0 , | 0, | 0, | 0, | 0, | 0. | 0, | 0 , | a, | 0 , | 0, | 0, | 0, | 0, | 0, | 0 , | 0 , |
| 0 , | 0 , | 0 , | 0, | 0 , | 0 , | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0, | 0 , | 0, | 0. | 0, | 0 , | 0 , | a, | 0, | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , |
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| 0 , | a, | 0 , | 0. | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 0 , | 5, | 0. | 0 , | 0 , | a, | 0 , | 0 , | a, | 0, | a, | 0 , | 0 , | 0 , | 0, | a, | 0 , | 0, | 0 , | 0 , | 0 , |
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[Beierle et all., The SKINNY Family of Block Ciphers and its Low-Latency Variant MANTIS, 2016]

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$$
\left(\begin{array}{llll}
1 & 0 & 1 & 1 \\
1 & 0 & 0 & 0 \\
0 & 1 & 1 & 0 \\
1 & 0 & 1 & 0
\end{array}\right)
$$

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| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 11 | 52 | 2 | 32 | 2 | 16 | 1 | 10 | 3 |
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- Problem


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- Problem $\rightarrow$ Many solutions don't have a Step2 solution

■ Increase the objective of Step1 $\rightarrow$ Too many solutions
■ Proposed solution: Add extra variables to control instantiation probability

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# Cluster Search and MILP Modeling for Differential Attacks 

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October 27th, 2022

