

$2m \frac{dx^2}{dt^2} + V\psi = E\psi$ $\psi_e = \frac{\Delta t'}{\sqrt{1-\frac{v^2}{c^2}}}$ $4\pi r^2$ $k = \frac{2\pi}{\lambda}$ $v_e = \sqrt{\frac{R M_2}{R_2}}$ $\vec{F}_m = \vec{B} I l = \frac{\mu I_1 I_2}{2\pi d} l$
 $U_{ef} = \frac{U_m}{\sqrt{2}}$ $E = h\nu$ $U = \frac{W_{AB}}{|E_{PA} - E_{PB}|} = |\varphi_A - \varphi_B|$ $X_L = \frac{U_m}{I_m} = \omega L = 2\pi f L$ $F_g = \frac{m_1 m_2}{r^2}$
 $\vec{B} = \mu \frac{NI\sqrt{2}}{l}$ $v = \frac{nh}{2\pi r m_e}$ $\varphi_E = \frac{F_e}{\rho_0} = k \frac{q}{r^2} \varphi$ $T = \frac{4 n_1 n_2}{(n_2 + n_1)^2}$ $g = \frac{m_1 m_2}{r^2}$
 $K = \frac{p^2}{2m}$ $m_0 = \frac{M_m}{N_A} = \frac{M_r \cdot 10^{-3}}{N_A}$ $m = N \cdot m_0 = \frac{Q}{v_e} \frac{M_m}{N_A}$ $E = \frac{E_c}{a} \int_{-a/L}^{+a/L} \sin(\omega t + \phi) dy$ $k_m = \frac{c}{T}$ $k = \pm \sqrt{\frac{2m}{\hbar^2} (E - V_0)}$
 $\lambda = \frac{h}{p}$ $l_t = l_0(1 + d \Delta t)$ $I = \frac{U_e}{R + \dots}$ $\omega = 2\pi f$

Tools for Cryptanalysis

宋凌

Topics

- What we have known
 - Differential/linear trails
 - Computation of the differential probability or linear correlation
- What we will study
 - Search for good differential/linear trails
 - Aid with various tools

Tools for DC/LC

- Matsui's algorithm [Mat94]
 - Branch and bound
 - Depth-first search algorithm
- **Mixed-Integer Linear Programming (MILP)[MWGP11]**
- **Boolean Satisfiability (SAT/SMT)**
- Constraint Programming (CP)
- Dedicated tools

$$\begin{aligned}
 & \frac{2m}{\hbar^2} \psi + \nabla^2 \psi = E \psi & \psi_e = \frac{4\pi r^2}{\sqrt{1-v^2/c^2}} & k = \frac{2\pi}{\lambda} & v_k = \sqrt{\frac{R M_2}{R_2}} & \vec{F}_m = \vec{B} I l = \frac{\mu I_1 I_2}{2\pi d} l \\
 & U_{ef} = \frac{U_m}{E} & E = h\nu & X_L = \frac{U_m}{I_m} = \omega L = 2\pi f L & F_g = \frac{m_1 m_2}{r^2} & \\
 & \vec{B} = \mu \frac{NI\sqrt{2}}{l} & v = \frac{nh}{2\pi r m_e} & \varphi_E = \frac{F_e}{\rho_0} = k \frac{q}{r^2} & T = \frac{4n_1 n_2}{(n_2 + n_1)^2} & g = \frac{c}{r^2} \\
 & K = \frac{p^2}{2m} & m_0 = \frac{M_m}{N_A} = \frac{M_r \cdot 10^{-3}}{N_A} & m = N \cdot m_0 = \frac{Q}{N_A} & E = \frac{E_c}{a} \int_{-a/L}^{+a/L} \sin(\omega t + \phi) dy & k_m = \frac{c}{T} & k = \pm \sqrt{\frac{2m}{\hbar^2} (E - V_0)} \\
 & \lambda = \frac{h}{p} & l_t = l_0 (1 + \alpha \Delta t) & I = \frac{U_e}{R} & & &
 \end{aligned}$$

MILP-based differential cryptanalysis

$$\begin{aligned}
 & \oint \vec{B} \cdot d\vec{l} = \mu \iint_S \vec{J} \cdot d\vec{S} & \vec{S} = \frac{1}{\mu_0} (\vec{E} \times \vec{B}) & \Delta I_B & \phi = \frac{2\pi \sin^2 \theta}{\lambda} & \oint \vec{D} \cdot d\vec{S} = Q^* \\
 & C(s) & E_k = \frac{\hbar^2 k^2}{2m} & & & \\
 & v_k = \sqrt{\frac{3kT}{m_0}} = \sqrt{\frac{3kT N_A}{M_m}} = \sqrt{\frac{3R_m T}{M_r \cdot 10^{-3}}} & E = \frac{\hbar^2 k^2}{2m} & 1 \text{ pc} = \frac{1 \text{ AU}}{r} & R = \frac{U}{I} & W_2 = U_e I t \\
 & \lambda = \frac{h\nu}{T} & F_h = S h p g & M_0 = \frac{4\pi^2 r^3}{3\pi T^2} & \vec{F}_v = \int \frac{F_n}{R} & \\
 & \left(\frac{E_t}{E} \right) = \frac{2 \cos \theta_1 \cos \theta_2}{2} & f_0 = \frac{1}{2\pi \sqrt{LC}} & \sigma = \frac{Q}{S_T} & M = F d \cos \alpha & \\
 & & & & &
 \end{aligned}$$

Mixed-Integer Linear Program (MILP)

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

$$\min -x_1 + x_2 - 2x_3 + x_4 - x_5$$

Linear objective function

subject to

$$x_1 + x_2 \leq 1$$

$$x_1 - 5x_2 + x_3 \leq 2$$

$$2x_3 + 2x_4 - 4x_5 \leq 1$$

$$x_2 - 2x_4 + x_5 \leq 0$$

$$x \in \{0, 1\}^5$$

Constraints in the form of linear inequality

Domain of variables

Mixed-Integer Linear Programming (MILP) allows some of the decision variables to be constrained to **integers** and others to be **non-integers**. **Binary** variables are common for crypto!

LP vs. MILP

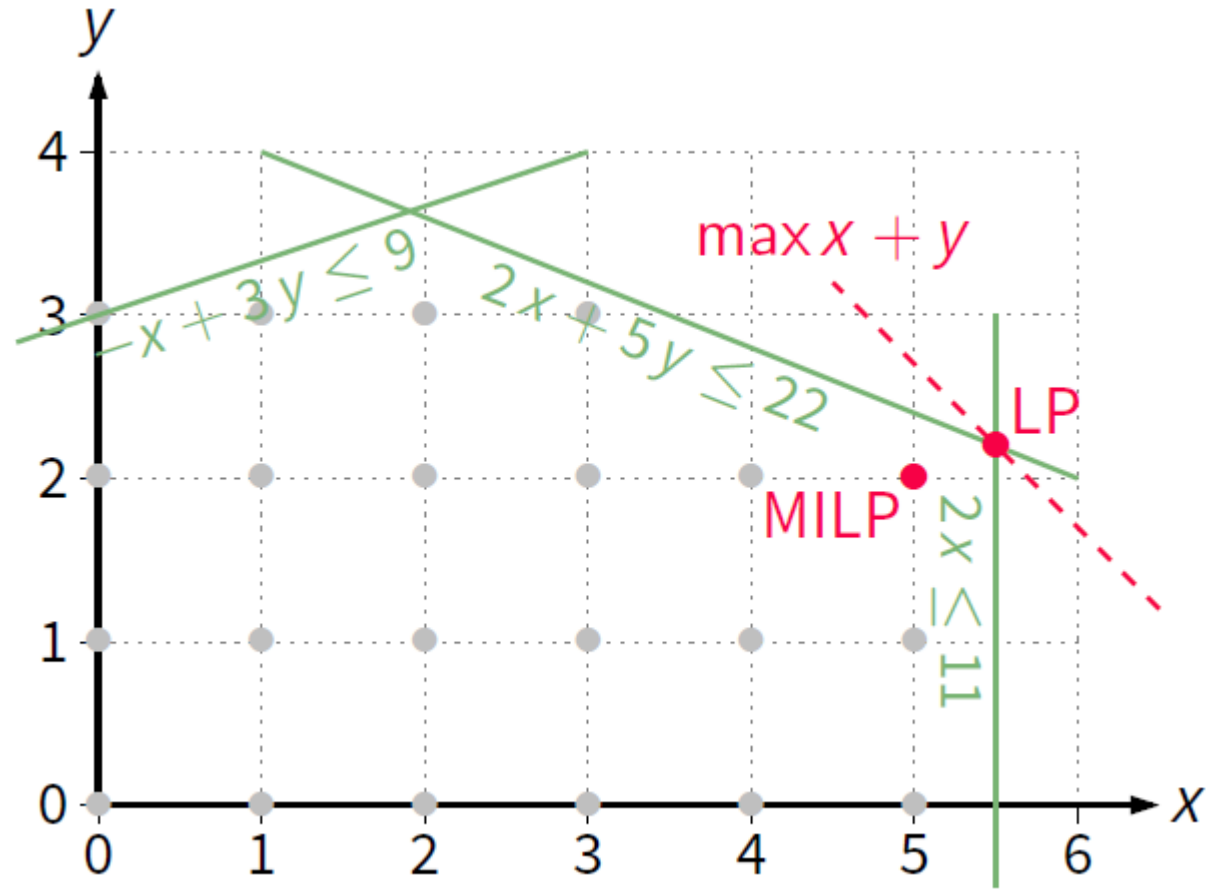
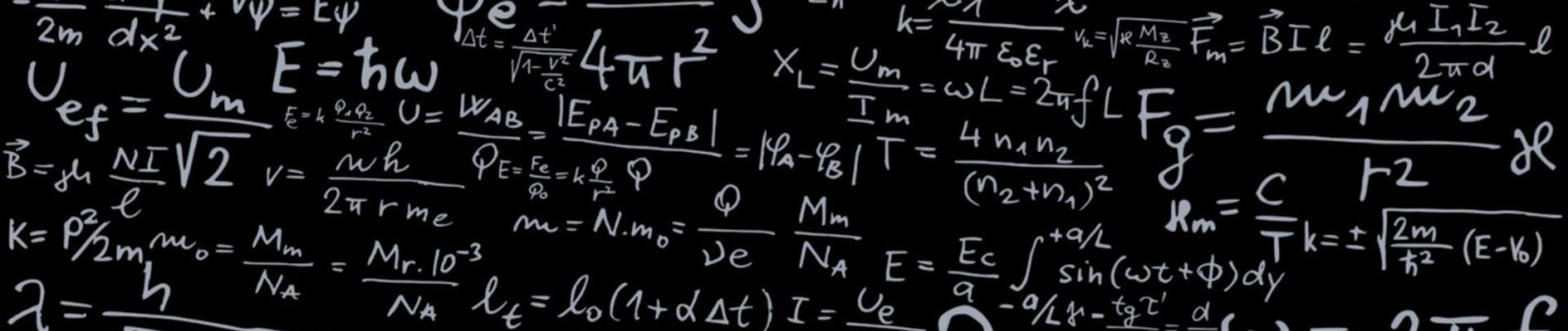


Figure from Maria's slides

MILP solvers

- Hardness of MILP solving:
 - Integer programming is NP-complete.
- Some well-known solver software:
 - CPLEX
 - **Gurobi**

These solvers can be used as stand-alone software (.lp input files) or as libraries with convenient interfaces (C/C++, python **Sagemath**, ...).



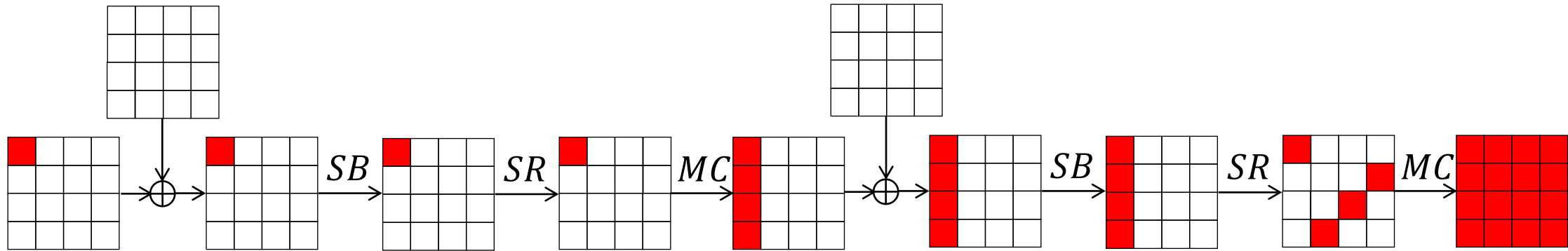
MILP-based differential cryptanalysis

- Counting the number of active S-boxes (#AS) of 4-round AES
- Search for best differential trails

Counting #AS of 4-round AES

- By the wide trail strategy, it is proved that there are at least 25 active S-boxes in 4-round AES.
- Let us prove it again with MILP.

Differential propagation of AES



Assign a binary variable $S_{r,i,j}$ for each state byte:

$S_{r,i,j} = 1$ if it is active, otherwise $S_{r,i,j} = 0$.

- **AK:** input = output
- **SB:** input = output, $\text{cost} = \sum S_{r,i,j}$
- **SR:** reorder variables
- **MC:** $w(\text{input}) + w(\text{output}) \geq 5$ (the branch number)

Modeling of AES

Variables:

- $S_{r,i,j} \in \{0,1\}$: Is the S-box in row i column j in round r active?
- $M_{r,j} \in \{0,1\}$: Is the column j of MC in round r active?
- #variables: $16 \cdot 4 + 4 \cdot 4 = 80$, # inequality: **$(1 + 8) \cdot 4 \cdot 3 + 1 = 109$**

Model:

$$\min \sum_{r,i,j} S_{r,i,j}$$

% Find the minimal #AS

$$\text{s.t.} \quad \sum_i S_{r,i,(i+j)\%4} + \sum_i S_{r+1,i,j} \geq 5 \cdot M_{r,j}$$

% For each MixColumns

$$M_{r,j} \geq S_{r,i,(i+j)\%4}, M_{r,j} \geq S_{r+1,i,j} \text{ for } i \in [0,3]$$

$$\sum_{i,j} S_{0,i,j} \geq 1$$

% At least one active byte in the input

$$S_{r,i,j}, M_{r,j} \in \{0,1\} \text{ for } r, i, j \in [0,3]$$

% Domain

Modeling of AES

Variables:

- $S_{r,i,j} \in \{0,1\}$: Is the S-box in row i column j in round r active?
- $M_{r,j} \in \{0,1\}$: Is the column j of MC in round r active?
- #variables: $16 \cdot 4 + 4 \cdot 4 = 80$, # inequality: **$2 \cdot 4 \cdot 3 + 1 = 25$**

Model:

$$\min \sum_{r,i,j} S_{r,i,j} \quad \% \text{ Find the minimal \#AS}$$

$$\text{s.t.} \quad 5 \cdot M_{r,j} \leq \sum_i S_{r,i,(i+j)\%4} + \sum_i S_{r+1,i,j} \leq 8 \cdot M_{r,j} \quad \% \text{ For each MixColumns}$$

$$\sum_{i,j} S_{0,i,j} \geq 1 \quad \% \text{ At least one active byte in the input}$$

$$S_{r,i,j}, M_{r,j} \in \{0,1\} \text{ for } r, i, j \in [0,3] \quad \% \text{ Domain}$$

A note on how we model a simple crypto problem

- Construct the model for a specific operation by hand.
- The validity can be verified.
- #variables and #inequality may vary when a different modeling is used.

Goal:

Model a problem with a minimal number of variables and inequality.

(The solving time is not necessarily reduced when #variables and #inequality are minimal.)

Modeling of AES and solve with Gurobi

AES4r.lp

Minimize

$S_{r0_0_0} + S_{r0_0_1} + \dots + S_{r3_3_3}$

Subject To

$S_{r0_0_0} + S_{r0_1_1} + S_{r0_2_2} + S_{r0_3_3} + S_{r1_0_0} + S_{r1_1_0} + S_{r1_2_0} + S_{r1_3_0} - 5 M_{r0_0} \geq 0$

$S_{r0_0_0} + S_{r0_1_1} + S_{r0_2_2} + S_{r0_3_3} + S_{r1_0_0} + S_{r1_1_0} + S_{r1_2_0} + S_{r1_3_0} - 8 M_{r0_0} \leq 0$

.....

$S_{r0_0_0} + S_{r0_0_1} + S_{r0_0_2} + S_{r0_0_3} + S_{r0_1_0} + S_{r0_1_1} + S_{r0_1_2} + S_{r0_1_3} + S_{r0_2_0} + S_{r0_2_1} + S_{r0_2_2} + S_{r0_2_3} + S_{r0_3_0} + S_{r0_3_1} + S_{r0_3_2} + S_{r0_3_3} \geq 1$

Binary

$S_{r0_0_0}$

.....

```
gurobi> m = read("AES4r.lp")
gurobi> m.optimize()
gurobi> m.write("filename")
```

```
Gurobi 9.1.1
```

Nodes		Current Node			Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time
0	0	1.32813	0	46	25.00000	1.32813	94.7%	-	0s
0	0	10.00000	0	52	25.00000	10.00000	60.0%	-	0s

```
Cutting planes:
Gomory: 1
MIR: 1
Zero half: 1
RLT: 2

Explored 1 nodes (102 simplex iterations) in 0.03 seconds
Thread count was 12 (of 12 available processors)

Solution count 2: 25 40

Optimal solution found (tolerance 1.00e-04)
Best objective 2.500000000000e+01, best bound 2.500000000000e+01, gap 0.0000%
gurobi>
```

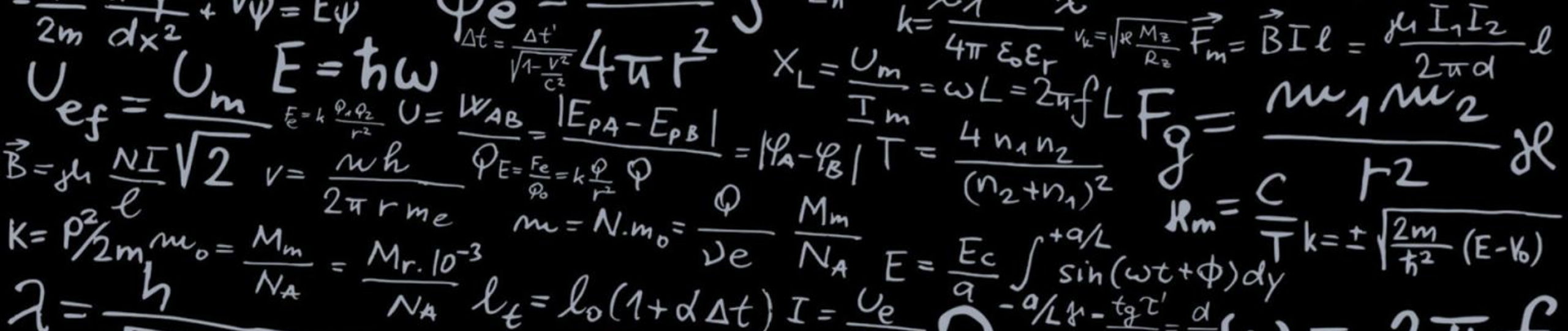
Modeling of AES and solve with Sagemath

```
#!/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 j in [0..3]:
        activecells = sum(S[r,i ,(i+j )%4] for i in [0..3]) + sum (S[r+1,i,j] for i in [0..3])
        p. add_constraint (5*M[r,j] <= activecells <= 8*M[r,j])
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))
```

<https://doc.sagemath.org/html/en/reference/numerical/sage/numerical/mip.html>

MILP Example: counting #AS of 4-round AES

```
sage: #!/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:
.....:     ^Ifor j in [0..3]:
.....:     ^I^Iactivecells = sum(S[r,i,(i+j)%4] for i in [0..3]) + sum(S[r+1,i,j] for i in [0..3])
.....:     ^I^Ip.add_constraint(5*M[r,j] <= activecells <= 8*M[r,j])
.....:     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))
.....:
25.0
25.0 {(0, 0, 0): 1.0, (0, 1, 1): 1.0, (0, 2, 2): 1.0, (0, 3, 3): 1.0, (1, 0, 0): 0.0, (1, 1, 0): 0.0, (1, 2, 0):
0.0, (1, 3, 0): 1.0, (0, 0, 1): 0.0, (0, 1, 2): 0.0, (0, 2, 3): 0.0, (0, 3, 0): 0.0, (1, 0, 1): 0.0, (1, 1, 1):
0.0, (1, 2, 1): 0.0, (1, 3, 1): 0.0, (0, 0, 2): 0.0, (0, 1, 3): 0.0, (0, 2, 0): 0.0, (0, 3, 1): 0.0, (1, 0, 2):
0.0, (1, 1, 2): 0.0, (1, 2, 2): 0.0, (1, 3, 2): 0.0, (0, 0, 3): 0.0, (0, 1, 0): 0.0, (0, 2, 1): 0.0, (0, 3, 2):
0.0, (1, 0, 3): 0.0, (1, 1, 3): 0.0, (1, 2, 3): 0.0, (1, 3, 3): 0.0, (2, 0, 0): 0.0, (2, 1, 0): 0.0, (2, 2, 0):
0.0, (2, 3, 0): 0.0, (2, 0, 1): 1.0, (2, 1, 1): 1.0, (2, 2, 1): 1.0, (2, 3, 1): 1.0, (2, 0, 2): 0.0, (2, 1, 2):
0.0, (2, 2, 2): 0.0, (2, 3, 2): 0.0, (2, 0, 3): 0.0, (2, 1, 3): 0.0, (2, 2, 3): 0.0, (2, 3, 3): 0.0, (3, 0, 0):
1.0, (3, 1, 0): 1.0, (3, 2, 0): 1.0, (3, 3, 0): 1.0, (3, 0, 1): 1.0, (3, 1, 1): 1.0, (3, 2, 1): 1.0, (3, 3, 1):
1.0, (3, 0, 2): 1.0, (3, 1, 2): 1.0, (3, 2, 2): 1.0, (3, 3, 2): 1.0, (3, 0, 3): 1.0, (3, 1, 3): 1.0, (3, 2, 3):
1.0, (3, 3, 3): 1.0, (4, 0, 0): 1.0, (4, 1, 0): 0.0, (4, 2, 0): 0.0, (4, 3, 0): 0.0, (4, 0, 1): 1.0, (4, 1, 1):
0.0, (4, 2, 1): 0.0, (4, 3, 1): 0.0, (4, 0, 2): 1.0, (4, 1, 2): 0.0, (4, 2, 2): 0.0, (4, 3, 2): 0.0, (4, 0, 3):
1.0, (4, 1, 3): 0.0, (4, 2, 3): 0.0, (4, 3, 3): 0.0}
sage: 
```

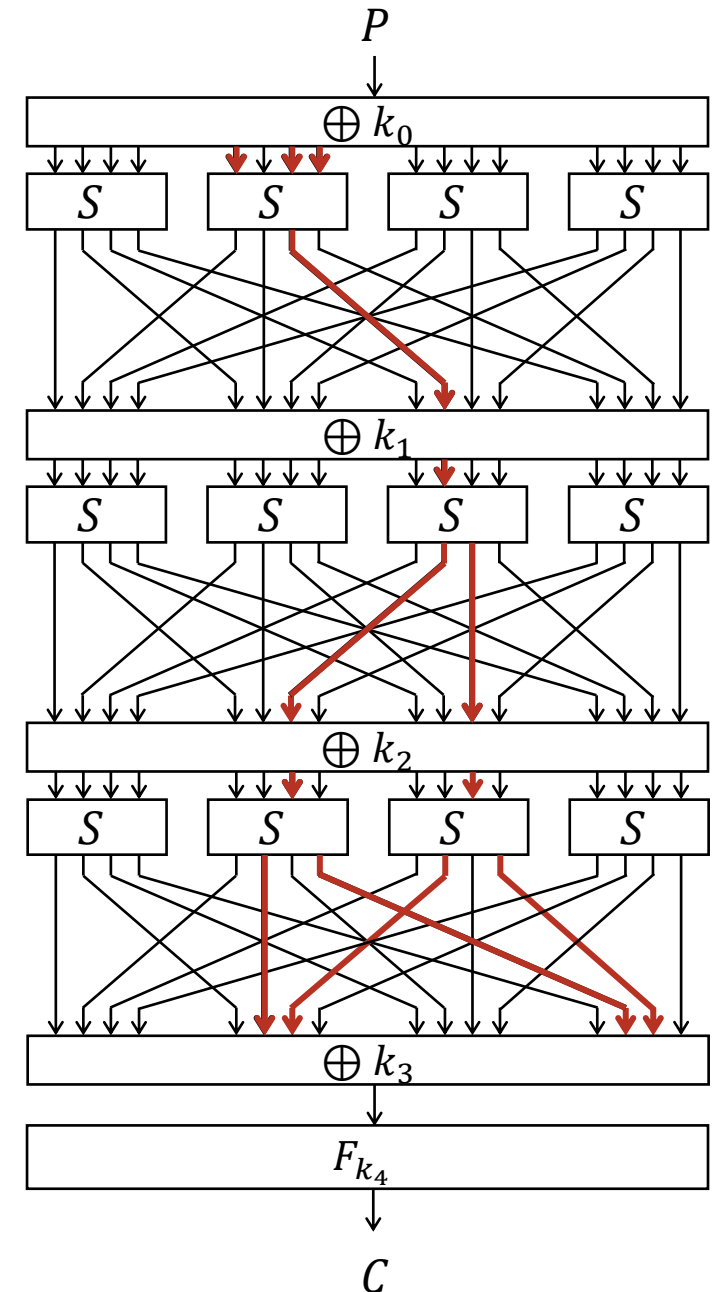
MILP-based differential cryptanalysis

- Counting the number of active S-boxes (#AS) of 4-round AES
- **Search for best differential trails**

Come back to the toy cipher

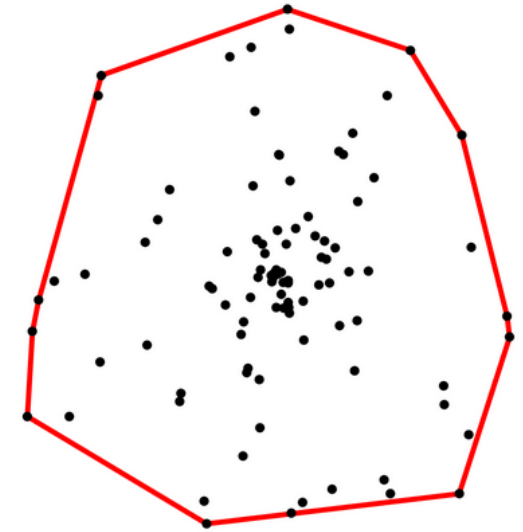
$S = \text{SBox}([14, 4, 13, 1, 2, 15, 11, 8, 3, 10, 6, 12, 5, 9, 0, 7])$

- This cipher uses a bitwise linear layer.
- Cannot treat the S-box as identity.
- How to give a bitwise model for an S-box?
 1. Convex hull computation [SHW+14]
 2. Logical computation [SHW+14, ST17]



1. Convex hull computation

- Convex hull of a set of points in \mathbb{R}^n : the smallest convex set that contains these points.
 - ✓ A convex hull can be represented by a set of linear inequalities
- Treat the set of all possible differential patterns of an S-box as a set of points in \mathbb{R}^n .
- Then we can compute the linear inequalities representation of the set of differential patterns.



1. Convex hull computation

Collect the set of all possible differentials

$$[x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

$$1. \quad 0 \rightarrow 0 \Rightarrow [0,0,0,0, 0,0,0,0]$$

$$2. \quad 1 \rightarrow 3 \Rightarrow [1,0,0,0, 1,1,0,0]$$

$$3. \quad 1 \rightarrow 7 \Rightarrow [1,0,0,0, 1,1,1,0]$$

$$4. \quad 1 \rightarrow 9 \Rightarrow [1,0,0,0, 1,0,0,1]$$

.....

```
sage: Pattern_list = [[0,0,0,0, 0,0,0,0], [0,0,0,1, 0,0,1,1], [0,0,0,1,
0,1,1,1], [0,0,0,1, 1,0,0,1], ...]
.....: A_polyhedron = Polyhedron(Pattern_list)
.....: for v in A_polyhedron.inequality_generator():
.....:     print(v)
```

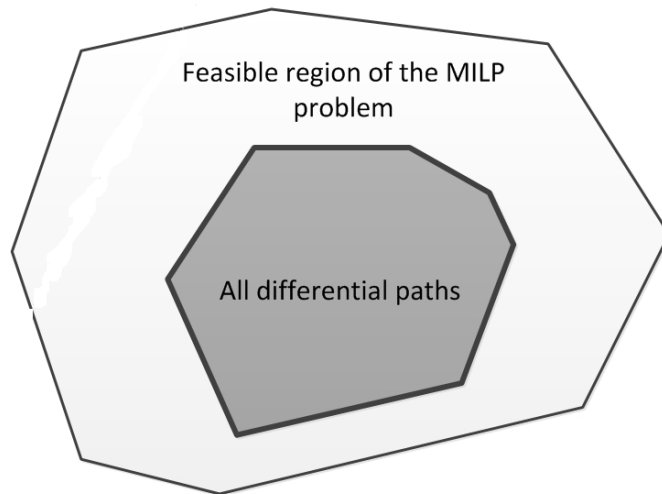
Return 410 inequalities

```
An inequality (0, -1, 0, 0, 0, 0, 0, 0) x + 1 >= 0
An inequality (-1, 0, 0, 0, 0, 0, 0, 0) x + 1 >= 0
An inequality (0, 0, -1, 0, 0, 0, 0, 0) x + 1 >= 0
An inequality (0, -1, -1, -1, -1, 0, -1, 0) x + 4 >= 0
An inequality (0, 0, 0, -1, 0, 0, 0, 0) x + 1 >= 0
An inequality (-1, -1, -1, -1, 1, 0, 1, 0) x + 3 >= 0
An inequality (0, 0, 0, 0, 0, 0, 1, 0) x + 0 >= 0
An inequality (0, 0, 0, 0, -1, 0, 0, 0) x + 1 >= 0
An inequality (-1, 1, -1, -1, -1, 0, 0, -1) x + 4 >= 0
An inequality (0, 0, 0, 0, 0, 0, 0, -1) x + 1 >= 0
An inequality (-1, 1, -1, 0, -1, 0, 1, -1) x + 3 >= 0
An inequality (-1, -1, 0, -1, 1, 0, 1, -1) x + 3 >= 0
An inequality (1, -1, -1, 0, -1, 0, 1, -1) x + 3 >= 0
An inequality (-1, -1, 1, -1, -1, 0, 1, 0) x + 3 >= 0
```

		Output Difference																
		0	1	2	3	4	5	6	7	8	9	A	B	C	D	E	F	
Input	0	16	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
	1	0	0	0	2	0	0	0	2	0	2	4	0	4	2	0	0	
	2	0	0	0	2	0	6	2	2	0	2	0	0	0	0	2	0	
	3	0	0	2	0	2	0	0	0	0	4	2	0	2	0	0	4	
	4	0	0	0	2	0	0	6	0	0	2	0	4	2	0	0	0	
	5	0	4	0	0	0	2	2	0	0	0	4	0	2	0	0	2	
	Differential	6	0	0	0	4	0	4	0	0	0	0	0	2	2	2	2	
		7	0	0	2	2	2	0	2	0	0	2	2	0	0	0	0	4
		8	0	0	0	0	0	0	2	2	0	0	0	4	0	4	2	2
		9	0	2	0	0	2	0	0	4	2	0	2	2	2	0	0	0
A		0	2	2	0	0	0	0	0	6	0	0	2	0	0	4	0	
B		0	0	8	0	0	2	0	2	0	0	0	0	0	2	0	2	
C		0	2	0	0	2	2	2	0	0	0	0	2	0	6	0	0	
D		0	4	0	0	0	0	0	4	2	0	2	0	2	0	2	0	
E		0	0	2	4	2	0	0	0	6	0	0	0	0	0	2	0	
F		0	2	0	0	6	0	0	0	0	4	0	2	0	0	2	0	

1. Convex hull computation

- Too many inequalities, which will make the MILP problem too difficult to be solved in practical time
 - ✓ There are redundant inequalities.
 - ✓ Can we use fewer inequalities? **Yes!**



Return 410 inequalities

```
An inequality (0, -1, 0, 0, 0, 0, 0, 0) x + 1 >= 0
An inequality (-1, 0, 0, 0, 0, 0, 0, 0) x + 1 >= 0
An inequality (0, 0, -1, 0, 0, 0, 0, 0) x + 1 >= 0
An inequality (0, -1, -1, -1, -1, 0, -1, 0) x + 4 >= 0
An inequality (0, 0, 0, -1, 0, 0, 0, 0) x + 1 >= 0
An inequality (-1, -1, -1, -1, 1, 0, 1, 0) x + 3 >= 0
An inequality (0, 0, 0, 0, 0, 0, 1, 0) x + 0 >= 0
An inequality (0, 0, 0, 0, -1, 0, 0, 0) x + 1 >= 0
An inequality (-1, 1, -1, -1, -1, 0, 0, -1) x + 4 >= 0
An inequality (0, 0, 0, 0, 0, 0, 0, -1) x + 1 >= 0
An inequality (-1, 1, -1, 0, -1, 0, 1, -1) x + 3 >= 0
An inequality (-1, -1, 0, -1, 1, 0, 1, -1) x + 3 >= 0
An inequality (1, -1, -1, 0, -1, 0, 1, -1) x + 3 >= 0
An inequality (-1, -1, 1, -1, -1, 0, 1, 0) x + 3 >= 0
```

1. Convex hull computation

Select a smaller set of inequalities via a greedy algorithm

Algorithm 1. Selecting n inequalities from the convex hull \mathcal{H} of an S-box

Input: \mathcal{H} : the set of all inequalities in the H-representation of the convex hull of an S-box; \mathcal{X} : the set of all impossible differential patterns of an S-box; n : a positive integer.

Output: \mathcal{O} : a set of n inequalities selected from \mathcal{H}

```
1  $l^* := \text{None}; \mathcal{X}^* := \mathcal{X}; \mathcal{H}^* := \mathcal{H}; \mathcal{O} := \emptyset;$   
2 for  $i \in \{0, \dots, n - 1\}$  do  
3    $l^* :=$  The inequality in  $\mathcal{H}^*$  which maximizes the number of removed  
   impossible differential patterns from  $\mathcal{X}^*$  ;  
4    $\mathcal{X}^* := \mathcal{X}^* - \{\text{removed impossible differential patterns by } l^*\};$   
5    $\mathcal{H}^* := \mathcal{H}^* - \{l^*\}; \mathcal{O} := \mathcal{O} \cup \{l^*\};$   
6 end  
7 return  $\mathcal{O}$ 
```

\mathcal{H} is the set of inequalities return by SAGE

Select the inequality which excludes the largest number of impossible differentials

1. Convex hull computation

- A set of 25 inequalities can be selected to model the DDT

$[x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$

$(-2, -2, 0, 1, 3, 4, 4, 1, 0)$ →

$(4, 1, 0, 1, 4, -2, 3, -2, 0)$

$(3, 4, 1, 3, -2, 1, -2, 1, 0)$

$(-1, 4, -1, -1, 3, -2, 4, 5, 0)$

$(2, -3, -1, -3, -2, 1, -1, -1, 8)$

$(-3, -1, -4, -2, -3, 2, -4, -1, 14)$

$(-1, -1, 1, 2, -3, -3, 3, -2, 7)$

$(1, -1, 2, -1, 1, 2, -2, 0, 2)$

$(-1, 0, 1, 0, 1, -1, -1, 0, 2)$

$(1, -2, -2, -1, -1, -2, -3, 2, 8)$

$(1, 1, 1, -2, -2, 3, 3, -2, 3)$

$(-2, -2, 2, -1, -3, -1, 3, 1, 6)$

$(2, -2, -2, 3, 1, 3, 3, 1, 0)$

$$-2x_0 - 2x_1 + 0x_2 + x_3 + 3y_0 + 4y_1 + 4y_2 + y_3 \geq 0$$

$(-1, 1, 2, 1, -1, 2, -1, 2, 1)$

$(2, 1, 2, 3, -2, -1, -1, -2, 3)$

$(-3, 2, -1, -2, -3, 1, -1, -2, 9)$

$(3, 3, -1, -1, 3, 0, 2, -1, 0)$

$(-1, -1, -1, 1, -2, -1, -2, 2, 6)$

$(-2, -1, -1, 1, -1, 2, -2, -1, 6)$

$(-1, -1, 0, -1, 1, 0, 1, -1, 3)$

$(-1, 0, 0, 1, 1, -1, -1, -1, 3)$

$(-1, 2, -1, 1, 2, 2, 0, 2, 0)$

$(2, 2, 2, -1, -1, -1, 3, 3, 0)$

$(0, 0, -1, -1, -1, -1, -1, -1, 5)$

$(3, -3, 1, -3, -1, -1, -2, 2, 7)$

2. Logical computation

Basic Idea: remove all impossible differentials for the S-box

$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

Eg. $6 \rightarrow 4 \Rightarrow [0,1,1,0, 0,0,1,0]$ ✗

$$x_0 - x_1 - x_2 + x_3 + y_0 + y_1 - y_2 + y_3 \geq -2$$

$$\text{Let } q(a) = \begin{cases} 1, & a = 0 \\ -1, & a = 1 \end{cases}$$

Then to remove $a = [0,1,1,0, 0,0,1,0]$ via

$$\sum q(a[i]) \cdot V[i] \geq 1 - H_w(a)$$

		Output Difference															
		0	1	2	3	4	5	6	7	8	9	A	B	C	D	E	F
Input	0	16	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	1	0	0	0	2	0	0	0	2	0	2	4	0	4	2	0	0
	2	0	0	0	2	0	6	2	2	0	2	0	0	0	0	2	0
	3	0	0	2	0	2	0	0	0	0	4	2	0	2	0	0	4
	4	0	0	0	2	0	0	6	0	0	2	0	4	2	0	0	0
	5	0	4	0	0	0	2	2	0	0	0	4	0	2	0	0	2
Difference	6	0	0	0	4	0	4	0	0	0	0	0	2	2	2	2	
	7	0	0	2	2	2	0	2	0	0	2	2	0	0	0	0	4
	8	0	0	0	0	0	0	2	2	0	0	0	4	0	4	2	2
	9	0	2	0	0	2	0	0	4	2	0	2	2	2	0	0	0
	A	0	2	2	0	0	0	0	0	6	0	0	2	0	0	4	0
	B	0	0	8	0	0	2	0	2	0	0	0	0	0	2	0	2
reference	C	0	2	0	0	2	2	2	0	0	0	2	0	6	0	0	
	D	0	4	0	0	0	0	4	2	0	2	0	2	0	2	0	
	E	0	0	2	4	2	0	0	0	6	0	0	0	0	0	2	0
	F	0	2	0	0	6	0	0	0	0	4	0	2	0	0	2	0

Changing any bit of a will increase the value of RHS, meaning that any other vector does not violate this inequality.

2. Logical computation

Basic Idea: remove all impossible differentials for the S-box

$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

Eg. $6 \rightarrow 4 \Rightarrow [0,1,1,0, 0,0,1,0]$ ✗

$$x_0 - x_1 - x_2 + x_3 + y_0 + y_1 - y_2 + y_3 \geq -2$$

$6 \rightarrow 6 \Rightarrow [0,1,1,0, 0,1,1,0]$ ✗

$$x_0 - x_1 - x_2 + x_3 + y_0 - y_1 - y_2 + y_3 \geq -3$$

However, these two cases can be represented as

$$[0,1,1,0, 0,*, 1,0]$$
 ✗

And removed by

$$x_0 - x_1 - x_2 + x_3 + y_0 - y_2 + y_3 \geq -2$$

		Output Difference															
		0	1	2	3	4	5	6	7	8	9	A	B	C	D	E	F
I n p u t	0	16	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	1	0	0	0	2	0	0	0	2	0	2	4	0	4	2	0	0
	2	0	0	0	2	0	6	2	2	0	2	0	0	0	0	2	0
	3	0	0	2	0	2	0	0	0	0	4	2	0	2	0	0	4
	4	0	0	0	2	0	0	6	0	0	2	0	4	2	0	0	0
D i f f e r e n c e	5	0	4	0	0	0	2	2	0	0	4	0	2	0	0	2	
	6	0	0	0	4	0	4	0	0	0	0	0	0	2	2	2	2
	7	0	0	2	2	2	0	2	0	0	2	2	0	0	0	0	4
	8	0	0	0	0	0	0	2	2	0	0	0	4	0	4	2	2
	9	0	2	0	0	2	0	0	4	2	0	2	2	2	0	0	0
	A	0	2	2	0	0	0	0	0	6	0	0	2	0	0	4	0
	B	0	0	8	0	0	2	0	2	0	0	0	0	0	2	0	2
C	0	2	0	0	2	2	2	0	0	0	0	2	0	6	0	0	
D	0	4	0	0	0	0	0	4	2	0	2	0	2	0	2	0	
E	0	0	2	4	2	0	0	0	6	0	0	0	0	0	2	0	
F	0	2	0	0	6	0	0	0	0	4	0	2	0	0	2	0	

Full model

Variables:

- $S_{r,i} \in \{0,1\}$: Is the S-box i in round r active?
- $x_{r,i,j} \in \{0,1\}$: Is the j -th bit of S-box i in round r active?
- #variables: $16 \cdot (R+1) + 4 \cdot R$, # inequality: **$2 \cdot 4 \cdot R + 4 \cdot R \cdot T$, $T = 25$**

$$\min \sum_{r,i} S_{r,i}$$

% Find the minimal #AS

s.t. $S_{r,i} \leq \sum_j x_{r,i,j} \leq 4 \cdot S_{r,i}$
25 inequalities for input-output patterns

% For each S-box

$$\sum_i S_{0,i} \geq 1$$

% At least one active S-box in the input

$$S_{r,i}, x_{r,i,j} \in \{0,1\} \text{ for } r \in [0,R), i, j \in [0,3]$$

% Domain

Find a minimal set [ST17]

☹️ The greedy algorithm does not guarantee a minimal set to be returned.

😊 Minimize the number of inequalities via MILP

Introduce variables z_i to denote whether inequality i is selected or not.

Impossible patterns which should be removed

$$\text{minimize } \sum_{i=1}^N z_i.$$

$$z_2 + z_8 + z_N \geq 1,$$

$$z_2 + z_3 + z_7 \geq 1,$$

⋮

$$z_1 + z_3 + z_4 + z_9 \geq 1,$$

as the constraint for R_0 ,

as the constraint for R_1 ,

⋮

as the constraint for $R_{|\mathcal{R}|-1}$.

	Patterns in \mathcal{R}											
	R_0	R_1	R_2	R_3	R_4	R_5	R_6	R_7	R_8	R_9	⋯	$R_{ \mathcal{R} -1}$
Inequality 1	0	0	0	1	1	0	1	0	0	0	⋯	1
Inequality 2	1	1	0	1	0	0	0	0	0	1	⋯	0
Inequality 3	0	1	1	0	0	1	0	0	1	0	⋯	1
Inequality 4	0	0	0	0	0	0	1	0	0	0	⋯	1
Inequality 5	0	0	0	1	0	0	0	0	1	0	⋯	0
Inequality 6	0	0	1	0	0	0	0	0	0	0	⋯	0
Inequality 7	0	1	0	0	1	1	0	0	0	0	⋯	0
Inequality 8	1	0	0	1	0	0	0	1	0	0	⋯	0
Inequality 9	0	0	1	0	0	0	0	1	0	0	⋯	1
⋮												
Inequality N	1	0	0	1	0	0	1	0	1	1	⋯	0

Choose from a set with N inequalities

What else MILP can do

- Search for other distinguishers
 - Impossible differential
 - Zero-correlation trail
 - Division trail
 - Boomerang distinguishers
 - Demirci-Selcuk MitM attack
 - Cube attack
 - ...

Limitation

- Not so useful for complex bitwise descriptions and characteristics
 - Language of linear inequalities is not so natural for crypto
 - Too many integer variables lead to bad solver performance

SAT problem

- The Boolean satisfiability problem (SAT) considers the satisfiability of a given Boolean formula.
- It was shown that the problem is NP-complete. However, modern SAT solvers based on backtracking search can solve problems of practical interest with millions of variables.
- Conjunctive Normal Form (CNF, 合取范式)

$$\underbrace{(a)}_{\text{literal}} \wedge \underbrace{(\bar{b})}_{\text{literal}} \wedge \underbrace{(\bar{a} \vee b \vee c)}_{\text{literal literal literal}}_{\text{clause}}$$

Product of sums

- EXample solvers:
 - MiniSAT, CryptoMiniSAT, Plingeling,

Model DDT of the S-box

Basic Idea: remove all impossible differentials for the S-box

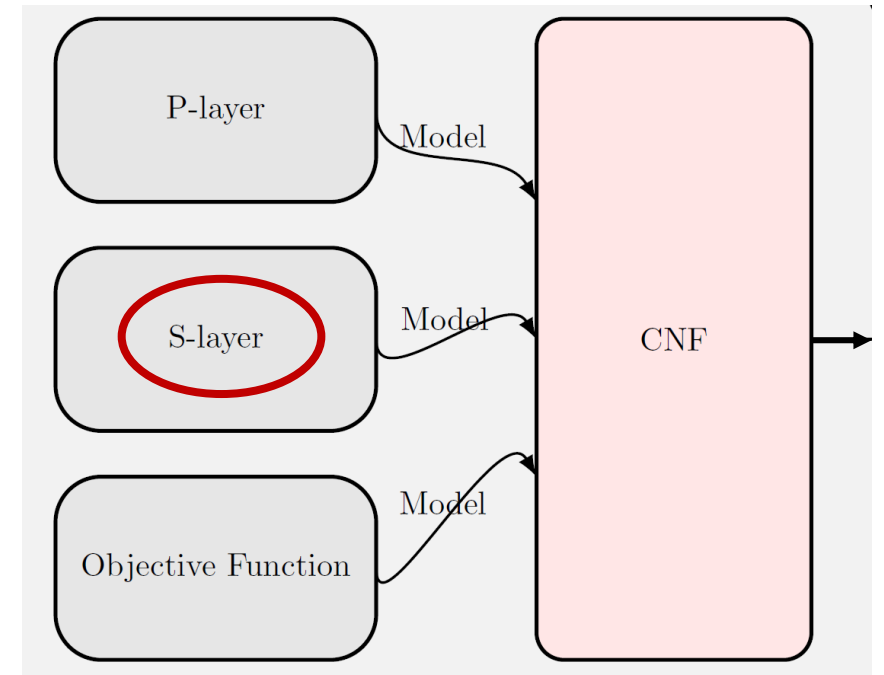
$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

Eg. $6 \rightarrow 4 \Rightarrow [0, 1, 1, 0, 0, 0, 1, 0]$ \boxtimes

$$x_0 \vee \bar{x}_1 \vee \bar{x}_2 \vee x_3 \vee y_0 \vee y_1 \vee \bar{y}_2 \vee y_3 \quad (= \text{true})$$

That is, remove $a = [0, 1, 1, 0, 0, 0, 1, 0]$ via

$$\bigvee (V[i] \oplus a[i]) \quad (= \text{true})$$



Changing any bit of a will make the clause true, meaning that any other vector does not violate this clause.

Model DDT of the S-box

Basic Idea: remove all impossible differentials for the S-box

$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

Eg. $6 \rightarrow 4 \Rightarrow [0, 1, 1, 0, 0, 0, 1, 0]$ \boxtimes

$$x_0 \vee \bar{x}_1 \vee \bar{x}_2 \vee x_3 \vee y_0 \vee y_1 \vee \bar{y}_2 \vee y_3$$

$6 \rightarrow 6 \Rightarrow [0, 1, 1, 0, 0, 1, 1, 0]$ \boxtimes

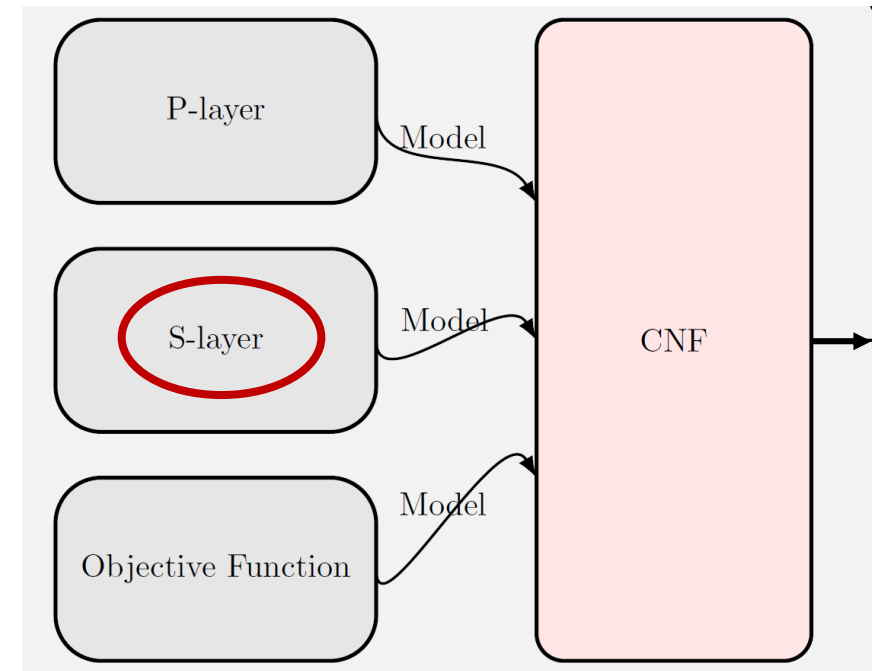
$$x_0 \vee \bar{x}_1 \vee \bar{x}_2 \vee x_3 \vee y_0 \vee \bar{y}_1 \vee \bar{y}_2 \vee y_3$$

However, these two cases can be represented as

$$[0, 1, 1, 0, 0, *, 1, 0] \boxtimes$$

And removed by

$$x_0 \vee \bar{x}_1 \vee \bar{x}_2 \vee x_3 \vee y_0 \vee \bar{y}_2 \vee y_3$$



Model DDT of the S-box

Take the activeness into account:

$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3, S_{r,i}]$$

$$\text{Eg. } 6 \rightarrow 4 \Rightarrow [0, 1, 1, 0, 0, 0, 1, 0, 0] \boxtimes$$

$$[0, 1, 1, 0, 0, 0, 1, 0, 1] \boxtimes$$

$$6 \rightarrow 6 \Rightarrow [0, 1, 1, 0, 0, 1, 1, 0, 0] \boxtimes$$

$$[0, 1, 1, 0, 0, 1, 1, 0, 1] \boxtimes$$

Simplify the product of sums

- Quine-McCluskey (QM) algorithm & Espresso algorithm.
- **Software: Logic Friday.**

Convert integer constraints into CNF

- $\sum_{r,i} S_{r,i} \leq w$, i.e., set the number of active S-boxes to $\leq w$.
- Employ the cardinality constraint.
- Cost $2wn + n - 3w - 1$ clauses.
 - when $n = 16, w = 4$ it requires 131 clauses

Cardinality constraint:

$$\sum_{\xi=0}^{\mu-1} p_{\xi} \leq w, w \geq 1.$$

$$\left\{ \begin{array}{l} \overline{p_0} \vee u_{0,0} = 1 \\ \overline{u_{0,j}} = 1 \\ \overline{p_i} \vee u_{i,0} = 1 \\ \overline{u_{i-1,0}} \vee u_{i,0} = 1 \\ \overline{p_i} \vee \overline{u_{i-1,j-1}} \vee u_{i,j} = 1 \\ \overline{u_{i-1,j}} \vee u_{i,j} = 1 \\ \overline{p_i} \vee \overline{u_{i-1,w-1}} = 1 \\ \overline{p_{\mu-1}} \vee \overline{u_{\mu-2,w-1}} = 1 \end{array} \right. .$$

$$u_{i,j} \quad (0 \leq i \leq \mu - 2, 0 \leq j \leq w - 1)$$

$$\begin{aligned}
 & \frac{2m}{\hbar^2} \psi = E\psi & \psi_e = \frac{4\pi r^2}{\Delta t} & k = \frac{2\pi}{\lambda} & v_k = \sqrt{\frac{R M_z}{R_z}} & \vec{F}_m = \vec{B} I l = \frac{\mu I_1 I_2}{2\pi d} l \\
 & U_{ef} = \frac{U_m}{E} & E = h\nu & X_L = \frac{U_m}{I_m} = \omega L = 2\pi f L & F_g = \frac{m_1 m_2}{r^2} & \\
 & \vec{B} = \mu \frac{NI\sqrt{2}}{l} & v = \frac{nh}{2\pi r m_e} & \varphi_E = \frac{E_e}{\varphi_0} = k \frac{Q}{r^2} & T = \frac{4n_1 n_2}{(n_2 + n_1)^2} & g = \frac{c}{T} k = \pm \sqrt{\frac{2m}{\hbar^2} (E - V_0)} \\
 & K = \frac{p^2}{2m} & m_0 = \frac{M_m}{N_A} = \frac{M_r \cdot 10^{-3}}{N_A} & m = N \cdot m_0 = \frac{Q}{v_e} \frac{M_m}{N_A} & E = \frac{E_c}{a} \int_{-a/L}^{+a/L} \sin(\omega t + \phi) dy & \\
 & \lambda = \frac{h}{p} & l_t = l_0(1 + \alpha \Delta t) & I = \frac{U_e}{R} & &
 \end{aligned}$$

MILP vs. SAT

$$\begin{aligned}
 & \oint \vec{B} \cdot d\vec{l} = \mu \iint_S \vec{J} \cdot d\vec{S} & \vec{S} = \frac{1}{\mu_0} (\vec{E} \times \vec{B}) & \Delta I_B & \phi = \frac{2\pi \sin^2 \theta}{\lambda} & \oint \vec{D} \cdot d\vec{S} = Q^* \\
 & C(s) & v_k = \sqrt{\frac{3kT}{m_0}} = \sqrt{\frac{3kT N_A}{M_m}} = \sqrt{\frac{3R_m T}{M_r \cdot 10^{-3}}} & E = \frac{\hbar^2 k^2}{2m} & 1 \text{ pc} = \frac{1 \text{ AU}}{r} & R = \frac{U}{I} & W_z = U_e I t \\
 & \lambda = \frac{h\nu}{T} & F_h = S h p g & f_0 = \frac{1}{2\pi \sqrt{LC}} & \sigma = \frac{Q}{S_T} & M = F d \cos \alpha & \\
 & \left(\frac{E_t}{E} \right) = \frac{2 \cos \theta_1 \cos \theta_2}{1 + \cos \theta_1 \cos \theta_2} & & & & &
 \end{aligned}$$

MILP vs. SAT on modeling DDT

Basic Idea: remove all impossible differentials for the S-box

$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

Eg. $6 \rightarrow 4 \Rightarrow [0, 1, 1, 0, 0, 0, 1, 0]$ \boxtimes

$$x_0 - x_1 - x_2 + x_3 + y_0 + y_1 - y_2 + y_3 \geq -2$$

Changing any bit of a will **increase the value of RHS**, meaning that any other vector does not violate this inequality.

$$V = [x_0, x_1, x_2, x_3, y_0, y_1, y_2, y_3]$$

Eg. $6 \rightarrow 4 \Rightarrow [0, 1, 1, 0, 0, 0, 1, 0]$ \boxtimes

$$x_0 \vee \bar{x}_1 \vee \bar{x}_2 \vee x_3 \vee y_0 \vee y_1 \vee \bar{y}_2 \vee y_3 = 1$$

Changing any bit of a will **make the clause true**, meaning that any other vector does not violate this clause.

Obtain minimized MILP models for DDT via the minimized product-of-sums representation.

$$\begin{aligned}
 & \frac{1}{2m} \frac{d^2x}{dt^2} + V\psi = E\psi & \psi_e = \frac{4\pi r^2}{\sqrt{1-\frac{v^2}{c^2}}} & k = \frac{2\pi}{\lambda} & v_k = \sqrt{\frac{R M_z}{R_z}} & \vec{F}_m = \vec{B} I l = \frac{\mu I_1 I_2}{2\pi d} l \\
 & U_{ef} = \frac{U_m}{\sqrt{2}} & E = h\nu & X_L = \frac{U_m}{I_m} = \omega L = 2\pi f L & F_g = \frac{m_1 m_2}{r^2} & \\
 & \vec{B} = \mu \frac{NI\sqrt{2}}{l} & v = \frac{nh}{2\pi r m_e} & \varphi_E = \frac{E_e}{\varphi_0} = k \frac{Q}{r^2} & T = \frac{4n_1 n_2}{(n_2 + n_1)^2} & g = \frac{c}{r^2} \\
 & K = \frac{p^2}{2m} & m_0 = \frac{M_m}{N_A} = \frac{M_r \cdot 10^{-3}}{N_A} & m = N \cdot m_0 = \frac{Q}{v_e} \frac{M_m}{N_A} & E = \frac{E_c}{a} \int_{-a/L}^{+a/L} \sin(\omega t + \phi) dy & R_m = \frac{c}{T} & k = \pm \sqrt{\frac{2m}{\hbar^2} (E - V_0)} \\
 & \lambda = \frac{h}{p} & l_t = l_0(1 + d\Delta t) & I = \frac{U_e}{R} & & &
 \end{aligned}$$

赛题解读

$$\begin{aligned}
 & \oint \vec{B} \cdot d\vec{l} = \mu \int_S \vec{J} \cdot d\vec{S} & \vec{S} = \frac{1}{\mu_0} (\vec{E} \times \vec{B}) & \Delta I_B & \phi = \frac{2\pi \sin^2 \theta}{\lambda} & \oint \vec{D} \cdot d\vec{S} = Q^* \\
 & C(s) & v_k = \sqrt{\frac{3kT}{m_0}} = \sqrt{\frac{3kT N_A}{M_m}} = \sqrt{\frac{3R_m T}{M_r \cdot 10^{-3}}} & E = \frac{\hbar^2 k^2}{2m} & 1 \text{ pc} = \frac{1 \text{ AU}}{r} & R = \frac{U}{I} & W_z = U_e I t \\
 & \lambda = \frac{h\nu}{T} & F_h = S h \rho g & f_0 = \frac{1}{2\pi \sqrt{LC}} & \sigma = \frac{Q}{S_T} & M = F d \cos \alpha & \\
 & \left(\frac{E_t}{E_i} \right) = \frac{2 \cos \theta_1 \cos \theta_2}{1 + \cos \theta_1 \cos \theta_2} & & & & &
 \end{aligned}$$

题目二

\mathbb{Z}_2^n 上非空子集线性不等式完全刻画问题

例如, $n=3$, $A=\{(000),(101),(011),(110)\}$ 。我们可以构造一组线性

不等式组 L :

$$\begin{cases} x_1 + x_2 \geq x_3 \\ x_1 + x_3 \geq x_2 \\ x_2 + x_3 \geq x_1 \\ x_1 + x_2 + x_3 \leq 2 \end{cases},$$

其由 4 个不等式组成。容易验证, 上述线性不等式组 L 关于 (x_3, x_2, x_1)

的解集恰好为 A 。

8个小题

$$Score_{i,j} = \delta_j c_j / l_{i,j}, \quad c_j = \min\{l_{1,j}, l_{2,j}, \dots, l_{m,j}, r_j\},$$

表 1 每道小题的参数设置

小题序号	n	元素个数
1	6	29
2	8	97
3	10	317
4	12	2017
5	14	6361
6	16	32386
7	20	491144
8	24	8115092

表 2 每道小题权重分值 δ_j 和不等式个数参考值 r_j 的取值

小题序号 j	δ_j	r_j
1	100	8
2	200	16
3	200	30
4	200	180
5	400	800
6	600	2300
7	800	36000
8	1000	576000

要求：完整刻画，使用的不等式尽可能少。

举例：第一小题

- $n = 6$, 29个元素
- 采用两种方法
 - Convex hull computation + greedy algorithm
 - Logic Friday

- The number of inequalities is 14. Minimized.

q_1_n=6.txt

```
00 09 0B 0D 0F 12 13 16
17 19 1A 1D 1E 24 25 26
27 29 2B 2C 2E 32 33 34
35 39 3A 3C 3F
```

Summary

- MILP-based cryptanalysis
- SAT-based cryptanalysis
- 应用于密码数学挑战赛题目二

- **Questions?**

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