

Solving Non-Linear Random Sparse Equations over Finite Fields

Thorsten Schilling

UiB

May 6, 2009



Algebraic Cryptanalysis

- Express cipher as system of equations
- Well known examples: AES, DES, Trivium etc.
- Obtaining solution to systems might break cipher



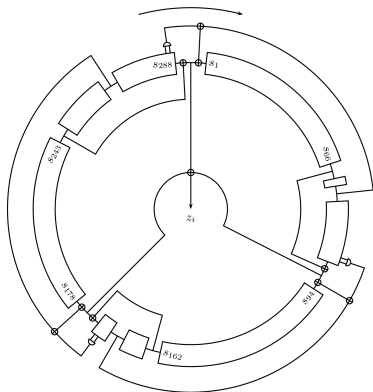
l -sparse Equation System

$$f_1(X_1) = 0, f_2(X_2) = 0, \dots, f_m(X_m) = 0$$

- $X_i \subseteq X$
- $|X_i| \leq l$



Example Trivium



- 80 bits IV, 80 bits key
- 288 bits internal state
- Output bit is linear combination of state bits
- Overall very "simple" design



Example Trivium (2)

Expressed as equation system

$$a_i = c_{i-66} + c_{i-111} + c_{i-110}c_{i-109} + a_{i-69}$$

$$b_i = a_{i-66} + a_{i-93} + a_{i-92}a_{i-91} + b_{i-78}$$

$$c_i = b_{i-69} + b_{i-84} + b_{i-83}b_{i-82} + c_{i-87}$$

Output bit z_i

$$z_i = c_{i-66} + c_{i-111} + a_{i-66} + a_{i-93} + b_{i-69} + b_{i-84}$$



Example Trivium (2)

Expressed as equation system

$$a_i = c_{i-66} + c_{i-111} + c_{i-110}c_{i-109} + a_{i-69}$$

$$b_i = a_{i-66} + a_{i-93} + a_{i-92}a_{i-91} + b_{i-78}$$

$$c_i = b_{i-69} + b_{i-84} + b_{i-83}b_{i-82} + c_{i-87}$$

Output bit z_i

$$z_i = c_{i-66} + c_{i-111} + a_{i-66} + a_{i-93} + b_{i-69} + b_{i-84}$$

→ For state recovery solve system:

- 6-sparse
- 951 variables
- 663 quadric equations, 288 linear



Solving Strategies

- Linearization
- Gröbner Basis Algorithms



Solving Strategies

- Linearization
- Gröbner Basis Algorithms
- SAT-Solving
 - ▶ Is $\phi = (x_i \vee x_j \vee \dots \vee x_k) \wedge \dots \wedge (x_u \vee x_v \vee \dots \vee x_w)$ SAT?
 - ▶ Theoretical worst case bounds [Iwama,04]:

sparsity	3	4	5	6
worst case	1.324^n	1.474^n	1.569^n	1.637^n



Solving Strategies

- Linearization
- Gröbner Basis Algorithms
- SAT-Solving
 - ▶ Is $\phi = (x_i \vee x_j \vee \dots \vee x_k) \wedge \dots \wedge (x_u \vee x_v \vee \dots \vee x_w)$ SAT?
 - ▶ Theoretical worst case bounds [Iwama,04]:

sparsity	3	4	5	6
worst case	1.324^n	1.474^n	1.569^n	1.637^n

- Gluing & Agreeing [Raddum, Semaev]
 - ▶ Expected Running Times:

sparsity	3	4	5	6
Gluing[Semaev,WCC'07]	1.262^n	1.355^n	1.425^n	1.479^n
Agr.-Gluing[Semaev,ACCT'08]	1.113^n	1.205^n	1.276^n	1.334^n



SAT-solving - DPLL/Davis-Putnam-Logemann-Loveland

General structure of the algorithm (input ϕ in CNF):

- Extend a partial guess
- Propagate information
- Clause Resolution
- Backtrack if a conflict was detected



Past Major Enhancements

- Algorithmic [Silvia, Sakallah 1996]
 - ▶ Non-chronological backtracking
 - ▶ Conflict clauses
- Technical
 - ▶ Watched literals [Moskewicz et. al. 2001]
- Instance specific
 - ▶ Guessing heuristics



A New Approach

- Generalization of a backtracking solving method for sparse non-linear equation systems over finite fields
 - ▶ CNF-instances are specialization
- Exploit sparsity of equations



Equation as Symbol

- $f(X_i) = 0$ as pair of sets $S_i = (X_i, V_i)$
- X_i variables in which the equation is defined
- V_i its satisfying vectors/assignments

Random equation over \mathbb{F}_2 expected $|V_i| = 2^{|X_i|-1}$



Equation as Symbol

- $f(X_i) = 0$ as pair of sets $S_i = (X_i, V_i)$
- X_i variables in which the equation is defined
- V_i its satisfying vectors/assignments

Random equation over \mathbb{F}_2 expected $|V_i| = 2^{|X_i|-1}$

Example:

$$f_e(x_1, x_2, x_3) = x_1 x_2 \oplus x_3 = 0$$

becomes

S_e	x_1	x_2	x_3
a_0	0	0	0
a_1	0	1	0
a_2	1	0	0
a_3	1	1	1



Gluing

Create new symbol $S_1 \circ S_2 = (X_1 \cup X_2, U)$ from two symbols S_1, S_2 where

$$U = \{(a_1, b, a_2) \mid (a_1, b) \in V_1 \text{ and } (b, a_2) \in V_2\}$$



Gluing

Create new symbol $S_1 \circ S_2 = (X_1 \cup X_2, U)$ from two symbols S_1, S_2 where

$$U = \{(a_1, b, a_2) \mid (a_1, b) \in V_1 \text{ and } (b, a_2) \in V_2\}$$

Example:

$$\begin{array}{c|c|c|c} S_0 & 1 & 2 & 3 \\ \hline a_0 & 0 & 0 & 0 \\ a_1 & 0 & 1 & 0 \\ a_2 & 1 & 0 & 0 \\ a_3 & 1 & 1 & 1 \end{array} \circ \begin{array}{c|c|c} S_1 & 3 & 4 & 5 \\ \hline b_0 & 1 & 0 & 0 \\ b_1 & 1 & 0 & 1 \\ b_2 & 1 & 1 & 1 \end{array} = \begin{array}{c|c|c|c|c} S_0 \circ S_1 & 1 & 2 & 3 & 4 & 5 \\ \hline c_1 & 1 & 1 & 1 & 0 & 0 \\ c_2 & 1 & 1 & 1 & 0 & 1 \\ c_3 & 1 & 1 & 1 & 1 & 1 \end{array}$$



Agreeing

Delete from symbols S_1, S_2 vectors whose projections do not match in their common variables $X_1 \cap X_2$. If symbol gets "empty" \rightarrow contradiction.



Agreeing

Delete from symbols S_1, S_2 vectors whose projections do not match in their common variables $X_1 \cap X_2$. If symbol gets "empty" \rightarrow contradiction.

Example:

$$\begin{array}{c|c|c|c} S_0 & 1 & 2 & 3 \\ \hline a_0 & 0 & 0 & 0 \\ a_1 & 0 & 1 & 0 \\ a_2 & 1 & 0 & 0 \\ a_3 & 1 & 1 & 1 \end{array}, \quad \begin{array}{c|c|c|c} S_1 & 3 & 4 & 5 \\ \hline b_0 & 1 & 0 & 0 \\ b_1 & 1 & 0 & 1 \\ b_2 & 1 & 1 & 1 \end{array} \quad \text{become} \quad \begin{array}{c|c|c|c} S_0 & 1 & 2 & 3 \\ \hline a_3 & 1 & 1 & 1 \end{array}, \quad \begin{array}{c|c|c|c} S_1 & 3 & 4 & 5 \\ \hline b_0 & 1 & 0 & 0 \\ b_1 & 1 & 0 & 1 \\ b_2 & 1 & 1 & 1 \end{array}$$



Gluing-Agreeing Algorithm

Glue intermediate symbol with another symbol, then agree the intermediate equation system.



Agreeing2

Aim: Reduce number of steps for Agreeing

Addresses of common projections stored as tuples

$$\begin{array}{c|c|c|c} S_0 & 1 & 2 & 3 \\ \hline a_0 & 0 & 0 & 0 \\ a_1 & 0 & 1 & 0 \\ a_2 & 1 & 0 & 0 \\ a_3 & 1 & 1 & 1 \end{array} , \begin{array}{c|c|c} S_1 & 3 & 4 & 5 \\ \hline b_0 & 0 & 0 & 0 \\ b_1 & 1 & 0 & 0 \\ b_2 & 1 & 0 & 1 \\ b_3 & 1 & 1 & 1 \end{array} \rightarrow \{a_0, a_1, a_2; b_0\}, \{a_3; b_1, b_2, b_3\}$$



Mark groups of common projections

Example: a_0, a_1, a_3 marked $\Rightarrow b_1, b_2, b_3$ marked

$$\{a_0, a_1, a_2; b_0\}, \{a_3; b_1, b_2, b_3\}$$

S_0	1	2	3	,	S_1	3	4	5
a_0	0	0	0		b_0	0	0	0
a_1	0	1	0		b_1	1	0	0
a_2	1	0	0		b_2	1	0	1
a_3	1	1	1		b_3	1	1	1

$$\{\cancel{a_0}, \cancel{a_1}, a_2; b_0\}, \{\cancel{a_3}; \cancel{b_1}, \cancel{b_2}, \cancel{b_3}\}$$

S_0	1	2	3	,	S_1	3	4	5
a_2	1	0	0		b_0	0	0	0



Agreeing2

Advantages

- Information propagation through tuples
- Asymptotically faster
- Overhead for reading and writing decreases



Continuous Implicit Agreeing

- Tree search through possible Gluings
- Works only on tuple markings
- Keep obtained knowledge
 - ▶ Persistent marking
- Flexible in the Gluing order



General Algorithm

- 1 Pick symbol
- 2 Mark all yet unmarked assignments, except one
 - ▶ Guess assignment
- 3 Run Agreeing2
- 4 Backtrack on contradiction



Guessing Heuristics

- Which symbol keeps search tree narrow
- Gluing complexity roughly $2^{\max_i |X(i)| - i} \rightarrow$ sort and keep $|X(i)| - i = |X_0 \cup X_1 \cup \dots \cup X_i| - i$ low
- Better: choose symbol with smallest $|V_i|$ (only unmarked assignments)
- Other heuristics possible, e.g. maximum $|X_i|/|V_i|$ etc.



Example Result

$$n = m = 150, l = 5$$

MiniSAT		dynglue	
Decisions	6844	Guesses	1549
Conflicts	5046	Contradictions	280
Propagations	120239	Tuple Propagations	541128
Time	0.15s	Time	0.08s



Open Questions

- Conflict handling
- Learning
 - ▶ Dynamic
 - ▶ Static
- Heuristics
 - ▶ Random systems
 - ▶ Equation systems from ciphers

