

Separable Statistics and Multidimensional Linear Cryptanalysis

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Briefly

- ▶ Matsui's Linear Cryptanalysis is based on the distribution of

$$x_1 \oplus \dots \oplus x_s \oplus y_1 \oplus \dots \oplus y_t,$$

where $X = x_1, \dots, x_s$ some plain-text bits and $Y = y_1, \dots, y_t$ some cipher-text bits in Algorithm1

- ▶ In Algorithm2 the bits X are inputs to the second round, and Y to the last round
- ▶ Starting point in **our work**: method for computing joint distribution of $(X, Y) = (x_1, \dots, x_s, y_1, \dots, y_t)$
- ▶ The distributions (both Matsui's and our's) are approximate
- ▶ They depend on small sets of the cipher key-bits or linear combinations
- ▶ Algorithm2-like cryptanalysis is then applied

Outline

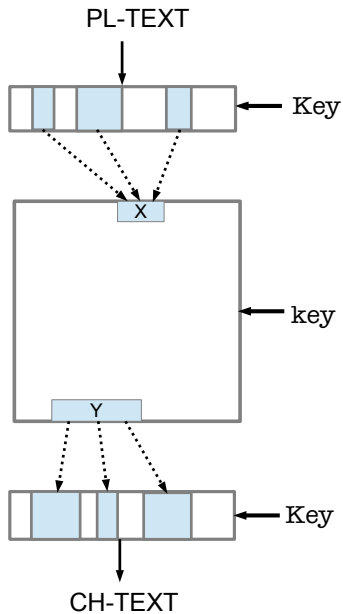
- ▶ Matsui's Algorithm² and LLR statistic
- ▶ New Statistic Construction
- ▶ Optimisation Problem and Search Algorithm
- ▶ Implementation for 16-round DES
- ▶ Multidimensional Distributions in Feistel Ciphers
- ▶ Conclusions

Outline

- ▶ Matsui's Algorithm2 and LLR statistic



Round Cipher Cryptanalysis with Algorithm2



Logarithmic Likelihood Ratio(LLR) Statistic

- ▶ To distinguish two distributions with densities $P(x)$, $Q(x)$
- ▶ By independent observations ν_1, \dots, ν_n
- ▶ Most powerful test(Neyman-Pearson lemma):
- ▶ Accept $P(x)$ if

$$\sum_{i=1}^n \ln \frac{P(\nu_i)}{Q(\nu_i)} > \text{threshold}$$

- ▶ Left hand side function is called LLR statistic

Algorithm2 Cryptanalysis with LLR statistic

- ▶ Distribution of (X, Y) depends on key-bits key
- ▶ Observation on (X, Y) depends on key-bits Key
- ▶ LLR statistic depends on $key \cup Key$
- ▶ Distinguish correct and incorrect $key \cup Key$ with LLR statistic
- ▶ by computing $2^{|key \cup Key|}$ values of LLR
- ▶ For large (X, Y) the number of the key-bits involved $|key \cup Key|$ may be too large
- ▶ Not efficient

New Statistic

- ▶ Instead of $2^{|key \cup Key|}$ computations of LLR-values
- ▶ **Our work:** $\ll 2^{|key \cup Key|}$ ($\approx 10^3$ times faster in DES)
- ▶ By using a new statistic
- ▶ Which reflects the structure of the round function
- ▶ That has a price to pay, but trade-off is positive

Outline

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- ▶ New Statistic Construction
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LLRs for Projections

- ▶ (h_1, \dots, h_m) some subvectors (projections) of (X, Y) such that
- ▶ Distribution and Observation for h_i depend on a lower number of the key-bits $key_i \cup Key_i$
- ▶ LLR_i is a LLR-statistic for h_i
- ▶ Vector (LLR_1, \dots, LLR_m) asymptotically distributed
- ▶ m -variate $\mathbf{N}(n\mu, nC)$ if $key \cup Key$ is correct
- ▶ Close to $\mathbf{N}(-n\mu, nC)$ if $key \cup Key$ is incorrect
- ▶ Mean vector μ , covariance matrix C , number of plain-texts n

LLR for Two Normal Distributions

- ▶ LLR statistic S to distinguish two normal distributions $\mathbf{N}(n\mu, nC)$ and $\mathbf{N}(-n\mu, nC)$
- ▶ S degenerates to linear:
- ▶ $S(\text{key} \cup \text{Key}, \nu) = \sum_{i=1}^m S_i(\text{key}_i \cup \text{Key}_i, \nu_i)$,
- ▶ where $S_i = \omega_i \text{LLR}_i$ weighted LLR statistic for h_i
- ▶ ν observation on (X, Y) and ν_i observation on h_i
- ▶ S is **separable**
- ▶ For polynomial distributions the theory of separable statistics was developed by Ivchenko, Medvedev,.. in 1970-s

Distribution

- ▶ S distributed 1-variate $\mathbf{N}(u, u)$ if $key \cup Key$ correct
- ▶ Close to $\mathbf{N}(-u, u)$ if incorrect
- ▶ for an explicit positive u

Cryptanalysis

- ▶ Find $key \cup Key$ s.t.

$$S(key \cup Key, \nu) > threshold$$

- ▶ without brute forcing $key \cup Key$
- ▶ Can be done as
- ▶ $S(key \cup Key, \nu) = \sum_{i=1}^m S_i(key_i \cup Key_i, \nu_i)$
- ▶ and $|key_i \cup Key_i|$ is much smaller than $|key \cup Key|$
- ▶ $|key \cup Key| = 54$ and $|key_i \cup Key_i| \approx 20$ in DES
- ▶ By solving efficiently an optimisation problem with a Search Algorithm

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- ▶ Optimisation Problem and Search Algorithm
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Optimisation Problem Example

S_1	0.1	0.2	0.3	0.1
$x_1 \oplus x_3$	0	0	1	1
x_2	0	1	0	1

S_2	0.5	0.1
$x_1 \oplus x_2$	0	1

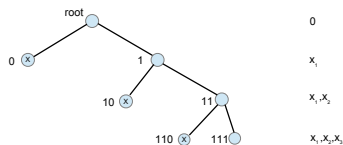
S_3	0.4	0.5	0.7	0.1
x_1	0	0	1	1
$x_2 \oplus x_3$	0	1	0	1

find binary x_1, x_2, x_3 s.t.

$$S(x_1, x_2, x_3) = S_1(x_1 \oplus x_3, x_2) + S_2(x_1 \oplus x_2) + S_3(x_1, x_2 \oplus x_3) > 1.3$$

Threshold is 1.3, solution 111

Search Tree



- ▶ One walks over a search tree and checks if the inequality $S_1(x_1 \oplus x_3, x_2) + S_2(x_1 \oplus x_2) + S_3(x_1, x_2 \oplus x_3) > 1.3$
- ▶ feasible under current fixation
- ▶ Cut if not feasible. Continue if feasible
- ▶ One is to check 6 linear inequalities. Brute force takes 8
- ▶ Same way one solves

$$S(\text{key} \cup \text{Key}, \nu) = \sum_{i=1}^m S_i(\text{key}_i \cup \text{Key}_i, \nu_i) > \text{threshold}$$

Success Probability & Number of $(key \cup Key)$ -candidates

- ▶ Search tree output is $(key \cup Key)$ -candidates for the final brute force
- ▶ The distribution of $S(key \cup Key, \nu)$ is known
- ▶ So one can compute success probability and
- ▶ The number of wrong solutions, that is $(key \cup Key)$ -candidates

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- ▶ Implementation for 16-round DES
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Two 14-bit vectors

- ▶ $\text{DES}_K(X_0, X_1) = (X_{17}, X_{16})$
- ▶ Matsui's best linear approximation

$$X_2\{24, 18, 7\} \oplus X_{15}\{15\} \oplus X_{16}\{24, 18, 7, 29\}$$

- ▶ We use two 14-bit vectors

$$X_2[24, 18, 7, 29], X_{15}[16, 15, \dots, 11], X_{16}[24, 18, 7, 29]$$
$$X_1[24, 18, 7, 29], X_2[16, 15, \dots, 11], X_{15}[24, 18, 7, 29]$$

- ▶ Considered independent as they incorporate different bits
- ▶ Computing their distributions took a few seconds

Projections

- ▶ 28 projections

$$X_2[24, 18, 7, 29], X_{15}[i, j], X_{16}[24, 18, 7, 29]$$

$$X_1[24, 18, 7, 29], X_2[i, j], X_{15}[24, 18, 7, 29]$$

- ▶ For each projection LLR depends on (≤ 21) key-bits
- ▶ 54 key-bits overall
- ▶ Two separable statistics for two independent bunches of the projections
- ▶ Search Algorithm combines (≤ 21)-bit values to find 54-bit candidates
- ▶ Those candidates are brute forced

One Particular Projection

- ▶ projection h_1 :

$$X_2[24, 18, 7, 29], X_{15}[16, 15], X_{16}[24, 18, 7, 29]$$

- ▶ $key_1 \cup Key_1$ incorporates 20 unknowns

$$\begin{aligned} & X_{63}, X_{61}, X_{60}, X_{53}, X_{46}, X_{42}, X_{39}, X_{36}, X_{31}, \\ & X_{30}, X_{27}, X_{26}, X_{25}, X_{22}, X_{21}, X_{12}, X_{10}, X_7, X_5, \\ & X_{57} + X_{51} + X_{50} + X_{19} + X_{18} + X_{15} + X_{14} \end{aligned}$$

x_i key-bits of 56-bit DES key

- ▶ 2^{20} values of $S_1 = \omega_1 LLR_1$
- ▶ Similar for other 27 projections

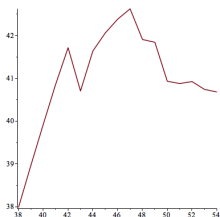
Key-variables Order for the Search Tree

- ▶ One needs $key \cup Key$ ordered to run a tree search
- ▶ x_2 appears in 14(maximal number) of $key_i \cup Key_i$, etc

$x_2, x_{19}, x_{60}, x_{34}, x_{10}, x_{17}, x_{59}, x_{36}, x_{42}, x_{27}, x_{25},$
 $x_{52}, x_{11}, x_{33}, x_{51}, x_9, x_{23}, x_{28}, x_5, x_{55}, x_{46}, x_{22},$
 $x_{62}, x_{15}, x_{37}, x_{47}, x_7, x_{54}, x_{39}, x_{31}, x_{29}, x_{20}, x_{61},$
 $x_{63}, x_{30}, x_{38}, x_{26}, x_{50}, x_1, x_{57}, x_{18}, x_{14}, x_{35}, x_{44},$
 $x_3, x_{21}, x_{41}, x_{13}, x_4, x_{45}, x_{53}, x_6, x_{12}, x_{43}$

Search Tree Algorithm Run

- ▶ We fix a desirable success rate 0.83
- ▶ solve equation $n = |\text{keys to brute force}|$ in n
- ▶ got $n = 2^{41.8}$



- ▶ The number of tree nodes is shown, \log_2 scale
- ▶ $|(\text{key} \cup \text{Key})\text{-candidates}| = 2^{39.8}$, $|\text{keys to brute force}| = 2^{41.8}$
- ▶ Number of nodes is $2^{45.5} \ll 2^{54}$. Constructing the nodes is faster (in bit operations) than final brute force
- ▶ Improves Matsui's result on DES ($n = 2^{43}$, 0.85)

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- ▶ Multidimensional Distributions in Feistel Ciphers
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r-Round DES

- ▶ $\text{DES}_K(X) = Y$, where X random, \mathcal{E} any event
- ▶ We want to compute $\Pr(\mathcal{E})$ in r -round DES. Let's formalise
- ▶ X_0, X_1, \dots, X_{r+1} random independently generated 32-bit blocks. Event \mathcal{C} defines DES:

$$X_{i-1} \oplus X_{i+1} = F_i(X_i, K_i), \quad i = 1, \dots, r$$

- ▶ K_1, \dots, K_r fixed round keys. We need

$$\Pr(\mathcal{E}|\mathcal{C}) = \frac{\Pr(\mathcal{E}\mathcal{C})}{\Pr(\mathcal{C})} = 2^{32r} \Pr(\mathcal{E}\mathcal{C})$$

- ▶ infeasible as \mathcal{C} depends on all key-bits

Relax \mathcal{C}

- ▶ One chooses a larger event \mathcal{C}_α (that is \mathcal{C} implies \mathcal{C}_α)

$$X_{i-1}[\alpha_i] \oplus X_{i+1}[\alpha_i] = F_i(X_i, K_i)[\alpha_i], \quad i = 1, \dots, r$$

- ▶ where $\alpha = (\alpha_1, \dots, \alpha_r)$. Then

$$\Pr(\mathcal{C}_\alpha) = 2^{-\sum_{i=1}^r |\alpha_i|}$$

- ▶ Let's accept

$$\Pr(\mathcal{E}|\mathcal{C}) \approx \Pr(\mathcal{E}|\mathcal{C}_\alpha) = \frac{\Pr(\mathcal{E}\mathcal{C}_\alpha)}{\Pr(\mathcal{C}_\alpha)} = 2^{\sum_{i=1}^r |\alpha_i|} \Pr(\mathcal{E}\mathcal{C}_\alpha)$$

- ▶ \mathcal{C}_α depends on a lower number of the key-bits. Now feasible and may be computed exactly

Regular Trails

- ▶ To compute the distribution of

$$Z = X_0[\alpha_1], X_1[\alpha_2 \cup \beta_1], X_r[\alpha_{r-1} \cup \beta_r], X_{r+1}[\alpha_r]$$

- ▶ One chooses event \mathcal{C}_α , where $\alpha = (\alpha_1, \dots, \alpha_r)$, and the trail

$$X_i[\beta_i], F_i[\alpha_i], \quad i = 1, \dots, r$$

- ▶ The trail is called regular if

$$\gamma_i \cap (\alpha_{i-1} \cup \alpha_{i+1}) \subseteq \beta_i \subseteq \gamma_i, \quad i = 1, \dots, r$$

where $X_i[\gamma_i]$ input bits relevant to $F_i[\alpha_i]$

- ▶ For a regular trail $\Pr(Z = A | \mathcal{C}_\alpha)$ is computed with a convolution-type formula, only depends on α_i

Convolution Formula

- ▶ $Z = X_0[\alpha_1], X_1[\alpha_2 \cup \beta_1], X_r[\alpha_{r-1} \cup \beta_r], X_{r+1}[\alpha_r]$
- ▶ Then $\Pr(Z = A_0, A_1, A_r, A_{r+1} | \mathcal{C}_\alpha) =$

$$\frac{2^{\sum_{i=2}^{r-1} |\alpha_i|}}{2^{\sum_{i=1}^r |(\alpha_{i-1} \cup \alpha_{i+1}) \setminus \beta_i|}} \sum_{A_2, \dots, A_{r-1}} \prod_{i=1}^r \mathbf{q}_i(A_i[\beta_i], (A_{i-1} \oplus A_{i+1})[\alpha_i], k_i)$$

- ▶ probability distribution on round sub-vectors

$$\mathbf{q}_i(b, a, k) = \Pr(X_i[\beta_i] = b, F_i[\alpha_i] = a | K_i[\delta_i] = k)$$

- ▶ $K_i[\delta_i]$ key-bits relevant to $F_i[\alpha_i]$
- ▶ May be computed iteratively by splitting encryption into two parts. A few seconds for 14-round DES

Outline

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- ▶ Conclusions

Conclusions

- ▶ Method of computing joint distribution of encryption internal bites X, Y (for Feistel ciphers) is found
 - ▶ Conventional LLR statistic is inefficient for large X, Y . New statistic reflects round function structure
 - ▶ We computed its distribution and able to predict success probability and the size of the final brute force
 - ▶ Efficient Search Algorithm to find key-candidates which fall into critical region is presented
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- ▶ Got an improvement over Matsui's results in DES (at least in bit operations)
 - ▶ Predicted correctly success probability(8-round DES) and the number of final key-candidates(16-round DES)
 - ▶ Search Algorithm is 10^3 times faster than brute forcing all key-bits which affect the statistic