

Relations Between the Generalizations of Differential and Linear Cryptanalysis

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Introduction

Block cipher :

$$egin{array}{rcl} {\mathbb F}_{{\mathcal K}}^n & : & {\mathbb F}_2^n & o & {\mathbb F}_2^n \ & & x & \mapsto & y \end{array}$$

Iterative block cipher :

$$x \longrightarrow F_{K_1} \longrightarrow F_{K_2} \longrightarrow F_{K_r} \longrightarrow F_{K_{r+1}} \longrightarrow y$$

Statistical attacks: Attacks that take advantage of a non-uniform behaviour of the cipher

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Differential Cryptanalysis [Biham Shamir 90]

Difference between plaintext and ciphertext pairs



Input difference : δ Output Difference : Δ

Differential Probability :

 $\mathbf{P}[\delta \to \Delta] = P_x[E_k(x) \oplus E_k(x \oplus \delta) = \Delta]$



Generalisations of Differential Cryptanalysis

Set of input differences : $\delta \in A$ Set of output differences : $\Delta \in B$

$$\mathbf{P}[A o B] = rac{1}{|A|} \sum_{\delta \in A} \sum_{\Delta \in B} P[\delta o \Delta]$$

Truncated Differential (TD) [Knudsen 94] :

Set A and B generally with structure : linear, affine spaces

Impossible Differential (ID) [Knudsen 99] :

Truncated differential distinguisher with probability 0

Multiple Differential [Blondeau Gérard 11] :

Sets A and B without structure

Multiple Differential using LLR or χ^2 [Blondeau Gérard Nyberg 12]



Linear Cryptanalysis [Tardy Gilbert 91] [Matsui 93]

Linear relation involving plaintext, key and ciphertext bits



Input mask : u Output mask : v Bias: $\varepsilon = 2^{-n} \# \{ x \in \mathbb{F}_2^n | u \cdot x \oplus v \cdot y = 0 \} - \frac{1}{2}$ Correlation : $\operatorname{cor}_{x}(u, v) = 2 \cdot P_{x} [u \cdot x \oplus v \cdot E_{k}(x) = 0] - 1$ $= 2 \cdot \varepsilon$



Generalizations of Linear Cryptanalysis

Set of masks $(u, v) \in U \times V \setminus \{0, 0\}$

Capacity :

$$\mathcal{C} = \sum_{u \in U \setminus \{0\}} \sum_{v \in V \setminus \{0\}} \operatorname{cor}_x^2(u, v)$$

Multiple Linear [Biryukov et al 04] : Set *U* and *V* without structure

Multidimensional Linear (ML) [Hermelin et al 08] : Set *U* and *V* are linear or affine spaces

Zero-Correlation Linear (ZC) [Bogdanov et al 10] : (Multidimensional) linear distinguisher with capacity 0



[Chabaud Vaudenay 94] :

Let $F : \mathbb{F}_2^n \to \mathbb{F}_2^m$

$$\mathbf{P}[\delta \to \Delta] = 2^{-m} \sum_{u \in \mathbb{F}_2^n} \sum_{v \in \mathbb{F}_2^m} (-1)^{u \cdot \delta \oplus v \cdot \Delta} \mathbf{cor}_x^2(u, v)$$

 Used to show that AB functions are APN (In general used for vector-Boolean functions)



Recent Links Between The Statistical Attacks



Outline

Computing Differential Probabilities using Linear Correlations Improving the Estimate of Differential Probability

Link between the TD and ML Key-Recovery Attacks Data/Time/Memory Trade-offs

Statistical Saturation (SS) Attack The SS Attack is a Truncated Differential Attack

Relation between ID and ZC Distinguishers Mathematical and Structural Relation

Conclusion



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Computation



Chabaud-Vaudenay's link:

$$\mathbf{P}[\delta \xrightarrow{F} \Delta] = 2^{-n} \sum_{u \in \mathbb{F}_2^n} \sum_{v \in \mathbb{F}_2^n} (-1)^{u \cdot \delta \oplus v \cdot \Delta} \mathbf{cor}_x^2(u, v)$$

Complexity: Computation of 2^{2n} correlations!!! \Rightarrow Impossible in practice



Computation



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Complexity: Computation of 2^{2n} correlations!!! \Rightarrow Impossible in practice

How to reduce the complexity:

- Using truncated output differences
 - \Rightarrow Reduce the sum over v
- Assuming δ of small weight \Rightarrow Reduce the sum over u



Truncated Output Difference

Setting:

- Affine space $\Delta_q \oplus \mathbb{F}_2^r$
- Let G be a projection of F

$$\mathbf{P}[\delta \xrightarrow{F} (\Delta_q \oplus \mathbb{F}_2^r)] = \mathbf{P}[\delta \xrightarrow{G} \Delta_q]$$





Truncated Output Difference

Setting:

- Affine space $\Delta_q \oplus \mathbb{F}_2^r$
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Truncated Output Difference

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- Let G be a projection of F

$$\mathbf{P}[\delta \xrightarrow{\mathsf{F}} (\Delta_q \oplus \mathbb{F}_2^r)] = \mathbf{P}[\delta \xrightarrow{\mathsf{G}} \Delta_q]$$



Link:

$$\mathbf{P}[\delta \stackrel{G}{\rightarrow} \Delta_q] = 2^{-q} \sum_{u \in \mathbb{F}_2^n} \sum_{\mathbf{v}_q \in \mathbb{F}_2^q} (-1)^{u \cdot \delta \oplus \mathbf{v}_q \cdot \Delta_q} \mathbf{cor}_x^2(u, \mathbf{v}_q)$$

Complexity: Computation of 2^{n+q} correlations



Assuming δ of Small Weight

Assumption: $\delta = (\delta_s, \delta_t) \in \mathbb{F}_2^s \times \mathbb{F}_2^t$ with $\delta_t = 0$

Fundamental Theorem [Nyberg 93]:

$$\sum_{u \in \mathbb{F}_2^n} (-1)^{u \cdot \delta} \mathbf{cor}_x^2(u, v_q) = \frac{2^{-t}}{\sum_{x_l \in \mathbb{F}_2^t}} \sum_{u_s \in \mathbb{F}_2^s} (-1)^{u_s \cdot \delta_s} \mathbf{cor}_{x_s}^2(u_s, v_q)$$





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Approximation:

$$\sum_{u \in \mathbb{F}_2^n} (-1)^{u \cdot \delta} \mathbf{cor}_x^2(u, v_q) \approx \frac{1}{|\mathcal{A}|} \sum_{x_t \in \mathcal{A}} \sum_{u_s \in \mathbb{F}_2^s} (-1)^{u_s \cdot \delta_s} \mathbf{cor}_{x_s}^2(u_s, v_q)$$



Method of Computation

Estimated Truncated Differential Probability:

$$\mathbf{P}[\delta \stackrel{G}{\to} \Delta_q] \approx \frac{2^{-q}}{|\mathcal{A}|} \sum_{x_t \in \mathcal{A}} \sum_{u_s \in \mathbb{F}_2^s} \sum_{v_q \in \mathbb{F}_2^q} (-1)^{u_s \cdot \delta_s \oplus v_q \cdot \Delta_q} \mathbf{cor}_{x_s}^2 (u_s, v_q)$$

Complexity: Computation of $2^{s+q}|A|$ correlations

Accuracy: Depends on the choice of s and A



PRESENT

[Bogdanov et al 08]

- 64-bit cipher
- 80-bit (128-bit) key
- 31 rounds





Setting of Experiments on PRESENT

PRESENT:

- Single-bit linear trails are dominant
- Computation of correlations using transition matrices as for instance in [Cho 10]

Setting:

Truncated differential distribution cryptanalysis
 Using LLR statistical test [Blondeau Gérard Nyberg 12]



Truncated Differential Distribution Cryptanalysis

Experiments on PRESENT :





Link Between Statistical Attacks

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Truncated Differential Distribution Cryptanalysis





Cryptanalysis:

On 19 rounds

Previously:

- Multiple differential cryptanalysis: 18 rounds
- Multidimensional linear cryptanalysis: 26 rounds



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Generalization [Blondeau Nyberg 14]:

• ML: $[(u_s, 0), (v_q, 0)]_{u_s \in \mathbb{F}_2^s \setminus \{0\}, v_q \in \mathbb{F}_2^q}$

• TD :
$$[(0, \delta_t), (0, \Delta_r)]_{\delta_t \in \mathbb{F}_2^t, \Delta_r \in \mathbb{F}_2^r}$$

 $\begin{array}{c}
 n \text{ bits} \\
 \overline{ \begin{array}{c} s \text{ bits} \\ 0 \end{array}} & \overline{ \begin{array}{c} t \text{ bits} \\ \delta_t \end{array}} \\
 U_s & 0 \\
 U_s & 0 \\
 V_q & 0 \\
 \underline{ \begin{array}{c} 0 \\ V_q \end{array}} & \underline{ \begin{array}{c} \Delta_r \\ r \text{ bits} \end{array}} \\
 n \text{ bits} \end{array}}$

with capacity C

with probability p



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Let
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• TD :
$$[(0, \delta_t), (0, \Delta_r)]_{\delta_t \in \mathbb{F}_2^t, \Delta_r \in \mathbb{F}_2^r}$$

with capacity C with probability p

 $p=2^{-q}(C+1)$



Link Between Statistical Attacks

n bits

t bits

0

s bits

Us

Va

0 *q* bits

[Chabaud Vaudenay 94] :

Let
$$F : \mathbb{F}_2^n \to \mathbb{F}_2^m$$

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Generalization [Blondeau Nyberg 14]:

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• TD :
$$[(0, \delta_t), (0, \Delta_r)]_{\delta_t \in \mathbb{F}_2^t, \Delta_r \in \mathbb{F}_2^r}$$



with capacity C with probability p

$p = 2^{-q}(C+1)$

- TD is a chosen plaintext (CP) attack
- ML is a known plaintext (KP) attack



[Selçuk 07] $P_S = 50\%$ and $\varphi_a = \Phi^{-1}(1 - 2^{-a})$, with *a* the advantage

Multidimensional Linear :



Truncated Differential :

$$\mathcal{N}^{TD} = rac{2^{-q+1}}{M \cdot (p-2^{-q})^2} \cdot arphi_a^2,$$



where *M* is the size of a structure (usually $M = 2^t$)



[Selçuk 07] $P_S = 50\%$ and $\varphi_a = \Phi^{-1}(1 - 2^{-a})$, with *a* the advantage

Multidimensional Linear :

$$N^{ML} = rac{2^{(s+q+1)/2}}{C} \cdot \varphi_{a}$$

Truncated Differential :

$$N^{TD} = rac{2^{-q+1}}{M \cdot (p-2^{-q})^2} \cdot arphi_a^2,$$



where *M* is the size of a structure (usually $M = 2^t$)

► For
$$p = 2^{-q}(C+1)$$
:
 $N^{TD} = \frac{2^{q+1}}{2^t \cdot C^2} \cdot \varphi_a^2$



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[Selçuk 07] $P_S = 50\%$ and $\varphi_a = \Phi^{-1}(1 - 2^{-a})$, with *a* the advantage

$$N^{ML} = \frac{2^{(s+q+1)/2}}{C} \cdot \varphi_a$$
For $p = 2^{-q}(C+1)$:
$$N^{TD} = \frac{2^{q+1}}{2^t \cdot C^2} \cdot \varphi_a^2$$

$$N^{TD} = \frac{1}{2^n} \cdot (N^{ML})^2$$



Link Between Statistical Attacks

and the Second

[Selçuk 07] $P_S = 50\%$ and $\varphi_a = \Phi^{-1}(1 - 2^{-a})$, with *a* the advantage

$$N^{ML} = \frac{2^{(s+q+1)/2}}{C} \cdot \varphi_a$$
For $p = 2^{-q}(C+1)$:
$$N^{TD} = \frac{2^{q+1}}{2^t \cdot C^2} \cdot \varphi_a^2$$

$$N^{TD} = \frac{1}{2^n} \cdot (N^{ML})^2$$

 $N^{TD} \leq N^{ML}$ with equality when using the full codebook



Link Between Statistical Attacks

and the Second

Truncated Differential Distinguisher



M : size of a structure S : number of structures

$$N^{TD} = S \cdot M$$

for *S* values of $x_s \in \mathbb{F}_2^s$ do Create a table *T* of size *M* for *M* values of $x_t \in \mathbb{F}_2^t$ do $(y_q, y_r) = E((x_s, x_t))$ $T[x_t] = y_q$ for all pairs (x_t, x_t') do

if $(T[x_t] \oplus T[x'_t]) == 0$ then D+=1 For S structures

For all elements in a structure

Store the partial ciphertexts

Count the number of pairs which have no difference on the q bits



D=0

Truncated Differential Distinguisher

Time Complexity : Verifying all pairs $\label{eq:Time} Time \approx S \cdot M^2/2$

Memory Complexity : Storing all ciphertexts inside a structure Memory $\approx M$

 $\begin{array}{l} D=0\\ \text{for }S \text{ values of } x_s\in \mathbb{F}_2^s \text{ do}\\ \text{ Create a table }T \text{ of size }M\\ \text{ for }M \text{ values of } x_t\in \mathbb{F}_2^t \text{ do}\\ (y_q,y_r)=E((x_s,x_t))\\ T[x_t]=y_q\\ \text{ for all pairs }(x_t,x_t') \text{ do}\\ \text{ if }(T[x_t]\oplus T[x_t'])==0 \text{ then}\\ D+=1 \end{array}$

For S structures

For all elements in a structure

Store the partial ciphertexts

Count the number of pairs which have no difference on the q bits



Multidimensional Linear Distinguisher



Set a counter *D* to 0 Create a table *T* of size 2^{q+s} for N^{ML} plaintexts do $(y_q, y_r) = E((x_s, x_t))$ $T[(x_s, y_q)] += 1$ for all (x_s, y_q) do $D += (T[(x_s, y_q)] - N/2^{q+s})^2$

For N^{ML} plaintexts Count the number of occurrences of each pair (x_s, y_q) Compute the statistic



Multidimensional Linear Distinguisher

Time Complexity : Reading all messages Time $\approx N^{ML}$

Memory Complexity : Storing the number of occurrences of $(x_s^i, y_a^j)_{i,i}$ *Memory* $\approx 2^{s+q}$

Set a counter D to 0 Create a table T of size 2^{q+s} for N^{ML} plaintexts do $(y_a, y_r) = E((x_s, x_t))$ $T[(x_s, y_a)] += 1$ all (x_s, y_q) do $D += (T[(x_s, y_q)] - N/2^{q+s})^2$ for all (x_s, y_q) do

For *N^{ML}* plaintexts Count the number of occurrences of each pair (x_s, y_q) Compute the statistic
Complexities of TD and ML Attacks

ML distinguisher :

$$Data = N^{ML}$$

Time $pprox N^{ML}$
Memory $pprox 2^{s+q}$

► TD distinguisher : $Data = N^{TD} = S \cdot 2^{t} < N^{ML}$ $Time \approx N^{TD} \cdot 2^{t-1}$

Memory $\approx 2^t$



Question : Can we decrease the time complexity of a TD attack?



TD with Less Time Complexity

 Dominant part: Verifying the output difference for each pair of ciphertexts

Example :

- ▶ 4 ciphertexts : (y₁, b₁) (y₂, b₂) (y₁, b₃) (y₃, b₄)
 1 pair with equal y_i
- Previous algorithm : 6 comparisons



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 1 pair with equal y_i
- Previous algorithm : 6 comparisons

Improved Version :

Count the occurrences of each y_i:

and compute $D = \sum_i T[y_i](T[y_i] - 1)/2 = 1$



TD with Less Time Complexity

D = 0for *S* values of $x_s \in \mathbb{F}_2^s$ do Create a table *T* of size 2^{*q*} for *M* values of $x_t \in \mathbb{F}_2^t$ do $(y_q, y_r) = E((x_s, x_t))$ $T[y_q] += 1$ for all $y_q \in \mathbb{F}_2^q$ do $D += T[y_q](T[y_q] - 1)/2$

For S structures

For all elements in a structure

Count the number of occurrences of the partial ciphertexts

Compute the statistic

$$egin{aligned} \mathsf{Data} &= \mathsf{N}^{\mathsf{TD}} = \mathcal{S} \cdot \mathcal{M} < \mathsf{N}^{\mathsf{ML}} \ && \mathsf{Time} &\approx \mathsf{max}(\mathsf{N}^{\mathsf{TD}}, \mathcal{S} \cdot 2^q) \ && \mathsf{Memory} &pprox 2^q \end{aligned}$$



KP ML and CP TD Attacks : An Example on PRESENT [Cho 10] :

- ML distinguisher on 24 rounds
- KP ML attack on 26 rounds (inversion of the first and last round)

First round : (In Cho's ML characteristic)



► KP ML ⇒ Guess 16-key bits



KP ML and CP TD Attacks : An Example on PRESENT [Cho 10] :

- ML distinguisher on 24 rounds
- KP ML attack on 26 rounds (inversion of the first and last round)
- First round : (In Cho's ML characteristic)



- KP ML \Rightarrow Guess 16-key bits
- Using the link between TD and ML
 - CP TD \Rightarrow Guess 4, 8, 12, 16-key bits

Example of CP TD Attack on 24 Rounds of PRESENT

Data Complexity (Data) :



- The *Data* of a KP ML is proportional to $\varphi_a = \Phi^{-1}(1 2^{-a})$
- The Data of a CP TD is proportional to φ_a^2
- Depending of the size of the fixation, the data complexity of a CP ML attack can be smaller than for a KP ML attack



Example of CP TD Attack on 24 Rounds of PRESENT

Fixing 4 bits :

Model	а	Data	Memory	Time ₁	Time ₂
CP TD	10	2 ^{54.75}	2 ²⁹	2 ^{54.75}	2 ⁷⁰
KP ML	5	2 ^{57.14}	2 ³²	2 ^{57.14}	2 ⁷⁵

*Time*₁: Complexity of the distillation phase *Time*₂: Complexity of the search phase

 Data, time and memory complexities of the CP TD attack are smaller than those of a KP ML attack



Example of CP TD Attack on 26 Rounds of PRESENT



- A CP TD attack on 26 rounds of PRESENT with less memory than the KP ML attack
- The previous differential-type attack was on 19 rounds



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Statistical Saturation (SS) Attack [Collard Standaert 09]

Idea :

- "Dual" of the saturation attack
- Takes advantage of several plaintexts with some fixed bits while the others vary randomly
- We observe the diffusion of the fixed bits during the encryption process

Application on PRESENT [Bogdanov et al 08] :

- Distinguisher on 20 / 21 rounds
- Key-recovery on 24 rounds



Statistical Saturation Distinguisher

 $\begin{array}{l} D=0\\ \text{for }S \text{ values of } x_s \in \mathbb{F}_2^s \text{ do}\\ \text{Create a table }T \text{ of size } 2^q\\ \text{for }M \text{ values of } x_t \in \mathbb{F}_2^t \text{ do}\\ (y_q, y_r) = E((x_s, x_t))\\ T[y_q]+=1\\ \text{for all } y_q \in \mathbb{F}_2^q \text{ do}\\ D+=T[y_q](T[y_q]-1)/2 \end{array}$

For S structures

For all elements in a structure

Count the number of occurrences of the partial ciphertexts

Compute the statistic

This distinguisher is the same as the improved truncated differential distinguisher



Link [Leander 11] :

For a fixed $x_s \in \mathbb{F}_2^s$, we denote by $C(x_s)$ the capacity of the distribution of y_q :

$$C = 2^{-s} \sum_{x_s \in \mathbb{F}_2^s} C(x_s)$$



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SS attacks link mathematically with ML attacks



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SS is a chosen plaintext (CP) attack ML is a known plaintext (KP) attack



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SS attacks link mathematically with ML attacks

SS is a chosen plaintext (CP) attack ML is a known plaintext (KP) attack

SS attacks link algorithmically with TD attacks



On the SS Attack on PRESENT [Collard Standaert 09]

Attack on r + 4 rounds with $M = 2^{32}$



[Collard Standaert 09] Data increases linearly [Leander 11] Estimate of the capacity C [Our work] Data is $N = \frac{2^{q+1}}{M \cdot C^2} \cdot \varphi_a^2$



On the SS Attack on PRESENT [Collard Standaert 09]

Attack on r + 4 rounds with $M = 2^{32}$



- The attack has been verified experimentally [Kerckhof et al 11]
- Our estimate match with the experiments (N around 2⁵¹ for 19 rounds)



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Mathematical and Structural Relation

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Mathematical Relation between ID and ZC

[Blondeau Nyberg 2013]

- $\succ \mathsf{TD} : [(0, \Delta_t), (0, \Gamma_r)]_{\Delta_t \in \mathbb{F}_2^t \setminus \{0\}, \ \Gamma_r \in \mathbb{F}_2^r}$
- $\blacktriangleright \mathsf{ML} : [(U_s, 0), (V_q, 0)]_{U_s \in \mathbb{F}_2^s \setminus \{0\}, V_q \in \mathbb{F}_2^q}$

with probability *p* with capacity *C*

$$\frac{2^t - 1}{2^t} \cdot p = 2^{-q} \cdot (C + 1) - 2^{-t}$$



If t = q: ZC and ID distinguishers are mathematically equivalent



Mathematical Relation between ID and ZC

[Blondeau Nyberg 2013]



If t = q: ZC and ID distinguishers are mathematically equivalent Observation :

Independent of the cipher and its structure

However: $(2^t - 1)(2^{n-t} - 1) \approx 2^n$ IDs are involved

In practice, the considered spaces are smaller



ID and ZC Distinguishers

Number of Rounds of the Distinguisher:

Ciphers	ID	ZC
LBlock / TWINE	14	14
MARS	11	11
SMS4	11	11
Skipjack	24	17
Skipjack (only rule A)	16	16
Four-Cell	18	12



ID and ZC Distinguishers

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Example of Patterns (for LBlock) :

- Impossible differential :
 - $(0000000,00 \triangle 00000) \nrightarrow (0 \Gamma 000000,00000000)$
- ► Zero correlation approximation : (000 U0000, 0000000) → (00000000, 0 V000000)



Example of Constructions





- F-layer
- X-layer
- P-layer





















Rules to find ZC and ID distinguishers

Differential Context :

Linear Context :



Rules to find ZC and ID distinguishers

Differential Context :

Linear Context :

⊕ and • "play orthogonal roles"



Mirror Round Function



- \mathcal{M} is the matrix representation of the mirror round function
- In general $\mathcal{M}^T \neq \mathcal{R}$
- Used to find ZC distinguishers [Soleimany Nyberg 2013]



Example of ID distinguisher







Example of ZC distinguisher









Matrix Method

Impossible Differential Context :

- Truncated input difference Δ
- Truncated output difference F
- ► If there is an inconsistency between $\mathcal{R}^m \cdot \Delta$ and $\mathcal{R}^{-\ell} \cdot \Gamma$, we have an ID distinguisher on $m + \ell$ rounds

Zero-Correlation Context :

- Truncated input mask U
- Truncated output mask V
- ▶ If there is an inconsistency between $\mathcal{M}^m \cdot U$ and $\mathcal{M}^{-\ell} \cdot V$, we have a ZC distinguisher on $m + \ell$ rounds



Equivalence between ID and ZC distinguishers

If it exists a linear relation between \mathcal{M} and \mathcal{R} or \mathcal{R}^{-1} , the existence of an ID distinguisher involving M differentials is equivalent to the existence of a ZC distinguisher involving Mlinear masks.

Given $\ensuremath{\mathcal{Q}}$ a permutation matrix, the relation is

• Feistel-type ($\mathcal{R} = \mathcal{P} \cdot \mathcal{F}$) :

$$\mathcal{R} = \mathcal{Q} \cdot \mathcal{M} \cdot \mathcal{Q}^{-1}$$
 or $\mathcal{R} = \mathcal{Q} \cdot \mathcal{M}^{-1} \cdot \mathcal{Q}^{-1}$

• Skipjack-type ($\mathcal{R} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{F}$) :

 $\mathcal{R} = \mathcal{Q} \cdot \mathcal{M} \cdot \mathcal{Q}^{-1} \text{ or } \mathcal{F} \cdot \mathcal{P} \cdot \mathcal{X} = \mathcal{Q} \cdot \mathcal{M}^{-1} \cdot \mathcal{Q}^{-1}$

• EGFN-type (
$$\mathcal{R} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{F}$$
) :

$$\mathcal{R} = \mathcal{Q} \cdot \mathcal{M} \cdot \mathcal{Q}^{-1}$$
 or $\mathcal{R} = \mathcal{Q} \cdot \mathcal{M}^{-1} \cdot \mathcal{Q}^{-1}$ or $\mathcal{F} \cdot \mathcal{P} \cdot \mathcal{X} = \mathcal{Q} \cdot \mathcal{M}^{-1} \cdot \mathcal{Q}^{-1}$


Illustration of the Proof for a Type-I Feistel



Inverse function





Illustration of the Proof for a Type-I Feistel



 $(\alpha, \beta, \delta, \gamma) \rightarrow (\alpha, \delta, \gamma, \beta)$



Illustration of the Proof for a Type-I Feistel





Round function



 $\mathcal{R}=\mathcal{P}\cdot\mathcal{X}\cdot\mathcal{F}$



Round function

 $\mathcal{R}=\mathcal{P}\cdot\mathcal{X}\cdot\mathcal{F}$

Inverse function

$$\begin{split} \mathcal{R}^{-1} &= \mathcal{F}^{-1} \cdot \mathcal{X}^{-1} \cdot \mathcal{P}^{-1} \\ &= \mathcal{P}^{-1} \cdot \mathcal{F}_*^{-1} \cdot \mathcal{X}_*^{-1} \end{split}$$



$$\mathcal{X}_*^{-1} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{P}^{-1}$$



Round function



 $\mathcal{R}=\mathcal{P}\cdot\mathcal{X}\cdot\mathcal{F}$

Exchange the order of the operations



Inverse function







$$\mathcal{X}_*^{-1} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{P}^{-1}$$

Round function



 $\mathcal{R}=\mathcal{P}\cdot\mathcal{X}\cdot\mathcal{F}$

Exchange the order of the operations



Inverse function



$$\begin{split} \mathcal{R}^{-1} &= \mathcal{F}^{-1} \cdot \mathcal{X}^{-1} \cdot \mathcal{P}^{-1} \\ &= \mathcal{P}^{-1} \cdot \mathcal{F}_*^{-1} \cdot \mathcal{X}_*^{-1} \end{split}$$

Equivalent formulation



 $\mathcal{F}_*^{-1} = \mathcal{P} \cdot \mathcal{F} \cdot \mathcal{P}^{-1}$

$$\mathcal{X}_*^{-1} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{P}^{-1}$$

Bound function



 $\mathcal{R} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{F}$

Exchange the order of the operations



 $\mathcal{X}_*^{-1} \cdot \mathcal{P}^{-1} \cdot \mathcal{F}_*^{-1}$

Inverse function



$$\mathcal{F}_*^{-1} = \mathcal{P} \cdot \mathcal{F} \cdot \mathcal{P}^{-1}$$

$$\mathcal{X}_*^{-1} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{P}^{-1}$$

 $\mathcal{R}^{-1} = \mathcal{F}^{-1} \cdot \mathcal{X}^{-1} \cdot \mathcal{P}^{-1}$ $\mathcal{P}^{-1} \cdot \mathcal{F}_*^{-1} \cdot \mathcal{X}_*^{-1}$

Equivalent formulation Permutation of the branches



 $\mathcal{P}^{-1} \cdot (\mathcal{P} \cdot \mathcal{X}^{-1}_* \cdot \mathcal{P}^{-1}) \cdot \mathcal{F}^{-1}_*$





Bound function



 $\mathcal{R} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{F}$

Exchange the order of the operations



Inverse function

$$\mathcal{F}_*^{-1} = \mathcal{P} \cdot \mathcal{F} \cdot \mathcal{P}^{-1}$$

$$\mathcal{X}_*^{-1} = \mathcal{P} \cdot \mathcal{X} \cdot \mathcal{P}^{-1}$$

 $\mathcal{R}^{-1} = \mathcal{F}^{-1} \cdot \mathcal{X}^{-1} \cdot \mathcal{P}^{-1}$ $= \mathcal{P}^{-1} \cdot \mathcal{F}_{*}^{-1} \cdot \mathcal{X}_{*}^{-1}$

Equivalent formulation Permutation of the branches



 $\mathcal{X}_*^{-1} \cdot \mathcal{P}^{-1} \cdot \mathcal{F}_*^{-1}$





 $\mathcal{P}^{-1} \cdot (\mathcal{P} \cdot \mathcal{X}_*^{-1} \cdot \mathcal{P}^{-1}) \cdot \mathcal{F}_*^{-1}$

The inverse function is "equivalent" to the mirror function



Example of Equivalence

Round Function of the Twine Block Cipher:



 ${\mathcal P}$ defined from

 $\pi = \{5, 0, 1, 4, 7, 12, 3, 8, 13, 6, 9, 2, 15, 10, 11, 14\}$

We have $\mathcal{M} = \mathcal{Q} \cdot \mathcal{R} \cdot \mathcal{Q}^{-1}$ for \mathcal{Q} defined from

 $\gamma = \{ \texttt{16}, \texttt{15}, \texttt{12}, \texttt{11}, \texttt{14}, \texttt{13}, \texttt{10}, \texttt{9}, \texttt{8}, \texttt{7}, \texttt{4}, \texttt{3}, \texttt{6}, \texttt{5}, \texttt{2}, \texttt{1} \}$



Example of Inequivalence

Some of the Feistels of [Suzaki et al 2010]

For instance
$$\mathcal{R} = \mathcal{P} \cdot \mathcal{F}$$
 with $\mathcal{F} =$

and \mathcal{P} is defined from

 $\pi = \{1, 2, 9, 4, 11, 6, 7, 8, 5, 12, 13, 10, 3, 0\}$

- The original Skipjack (ID: 24 rounds, ZC: 17 rounds)
 - Rule-B followed by Rule-A is equivalent to





Outline

Computing Differential Probabilities using Linear Correlations Improving the Estimate of Differential Probability

Link between the TD and ML Key-Recovery Attacks Data/Time/Memory Trade-offs

Statistical Saturation (SS) Attack The SS Attack is a Truncated Differential Attack

Relation between ID and ZC Distinguishers Mathematical and Structural Relation

Conclusion



Differential-Linear Cryptanalysis

[Langford and Hellman 94] [Biham et al 02]

 $\blacktriangleright E = E_0 \circ E_1$

- A truncated differential on E₀
- A linear approximation on E₁

In [Blondeau Leander Nyberg 14]:

- We analyze the model.
- We generalize it to the case of multiple linear approximations and multiple input differences.
- We show that a differential-linear attack is a truncated differential attack.



Integral distinguishers and ZC distinguishers

The link of [Bogdanov et al 12]:

Let $F : \mathbb{F}_2^{\alpha} \times \mathbb{F}_2^{\beta} \to \mathbb{F}_2^{\gamma} \times \mathbb{F}_2^{\delta}$ a cipher with $H(x, y) = (H_1(x, y), H_2(x, y))$. If the input and output linear masks *u* and *v* are

If the input and output linear masks u and v are independent, the approximation $\langle v, H(x) \rangle \oplus \langle u, x \rangle$ has correlation zero for any $u = (u_1, 0)$, and any $v = (v_1, 0) \neq 0$ (ZC distinguisher) if and only if the function $H_1(\lambda, y)$ is balanced for any λ (ZC integral distinguisher).

A ZC distinguisher with independent masks on *r* rounds \Rightarrow an integral distinguisher on $r' \ge r$ rounds

An integral distinguisher on *r* rounds with balanced output set \Rightarrow a ZC distinguisher on *r* rounds with independent masks



Conclusion

- Some strong relations between statistical attacks have been identified in the last 3 years
- Nevertheless some questions remain regarding the links with some other statistical attacks
- Based on these relations we wonder if we can simplify the security analysis of a symmetric cryptographic primitive

