



Aalto University
School of Science

Links Between Truncated Differential and Multidimensional Linear Properties of Block Ciphers and Underlying Attack Complexities

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Monday 12 May 2014
EUROCRYPT, Copenhagen

Outline

Statistical Attacks

- Truncated Differential (TD) Cryptanalysis

- Multidimensional Linear (ML) Cryptanalysis

Link between ML and TD Attacks

- Mathematical Relation between ML and TD

- Complexity of TD and ML Distinguishing Attacks

Statistical Saturation Attack

- Definition

- Statistical Saturation Attack on PRESENT

Converting a ML Attack to a TD Attack

- Example on PRESENT

- Conclusion

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Differential Cryptanalysis [Biham Shamir 90]

Difference between plaintext and ciphertext pairs

Input difference : δ

Output Difference : Δ

Differential Probability :

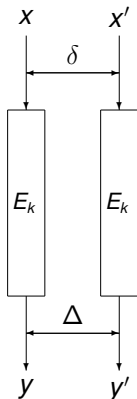
$$\mathbf{P}[\delta \rightarrow \Delta] = P_x[E_k(x) \oplus E_k(x \oplus \delta) = \Delta]$$

Truncated Differential (TD) [Knudsen 94] :

Set of input differences : $\delta \in A$

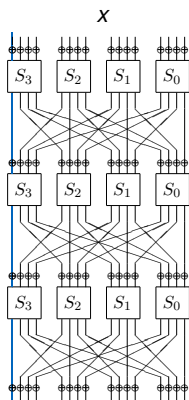
Set of output differences : $\Delta \in B$

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



Linear Cryptanalysis [Tardy Gilbert 91] [Matsui 93]

Linear relation involving plaintext, key and ciphertext bits



Input mask : u

Output mask : v

Correlation :

$$\mathbf{cor}_x(u, v) = 2 \cdot P_x [u \cdot x \oplus v \cdot E_k(x) = 0] - 1$$

Multidimensional Linear (ML) Approximation
[Hermelin et al 08] :

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

Capacity :

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

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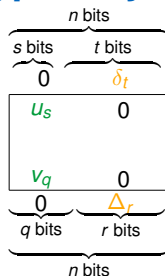
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Link between Differential and Linear Cryptanalysis

[Chabaud Vaudenay 94] :

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

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

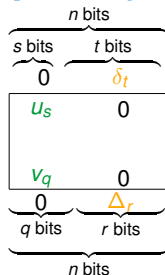


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Generalization :

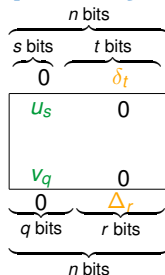
- ▶ ML : $[(u_s, 0), (v_q, 0)]_{u_s \in \mathbb{F}_2^s \setminus \{0\}, v_q \in \mathbb{F}_2^q}$ with capacity C
- ▶ TD : $[(0, \delta_t), (0, \Delta_r)]_{\delta_t \in \mathbb{F}_2^t, \Delta_r \in \mathbb{F}_2^r}$ with probability p

Link between Differential and Linear Cryptanalysis

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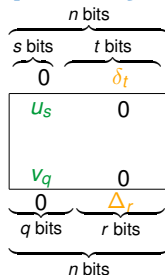
$$p = 2^{-q}(C + 1)$$

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

Data Complexity of a Distinguishing Attack

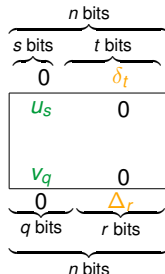
[Selçuk 07] $P_S = 50\%$ and $\varphi_a = \Phi^{-1}(1 - 2^{-a})$, with a the advantage

- ▶ Multidimensional Linear :

$$N^{ML} = \frac{2^{(s+q+1)/2}}{C} \cdot \varphi_a$$

- ▶ Truncated Differential :

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



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

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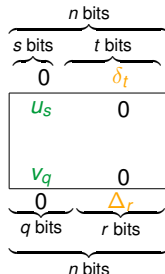
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- ▶ For $p = 2^{-q}(C + 1)$:

$$N^{TD} = \frac{2^{q+1}}{2^t \cdot C^2} \cdot \varphi_a^2$$



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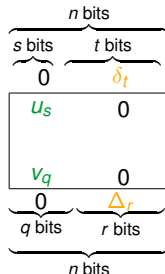
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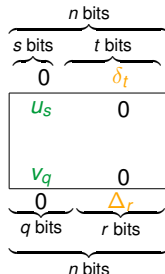


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$$N^{TD} = \frac{1}{2^n} \cdot (N^{ML})^2$$

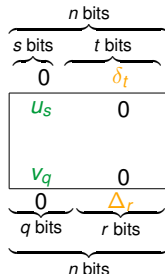


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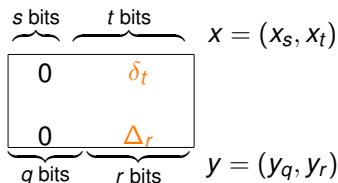
▶ For $p = 2^{-q}(C + 1)$:

$$N^{TD} = \frac{2^{q+1}}{2^t \cdot C^2} \cdot \varphi_a^2$$

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

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

Truncated Differential Distinguisher



M : size of a structure

S : number of structures

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

$D = 0$

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

Truncated Differential Distinguisher

Time Complexity : Verifying all pairs

$$Time \approx S \cdot M^2 / 2$$

Memory Complexity : Storing all ciphertexts inside a structure

$$Memory \approx M$$

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Multidimensional Linear Distinguisher

$$\begin{array}{c} \underbrace{\hspace{2cm}}_{s \text{ bits}} \quad \underbrace{\hspace{2cm}}_{t \text{ bits}} \quad x = (x_s, x_t) \\ \begin{array}{|c|c|} \hline u_s & 0 \\ \hline v_q & 0 \\ \hline \end{array} \\ \underbrace{\hspace{1cm}}_{q \text{ bits}} \quad \underbrace{\hspace{1cm}}_{r \text{ bits}} \quad y = (y_q, y_r) \end{array}$$

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 $(x_s^i, y_q^j)_{i,j}$

$$Memory \approx 2^{s+q}$$

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Complexities of TD and ML Attacks

- ▶ ML distinguisher :

$$\text{Data} = N^{ML}$$

$$\text{Time} \approx N^{ML}$$

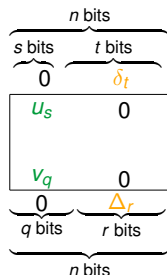
$$\text{Memory} \approx 2^{s+q}$$

- ▶ TD distinguisher :

$$\text{Data} = N^{TD} = S \cdot 2^t < N^{ML}$$

$$\text{Time} \approx N^{TD} \cdot 2^{t-1}$$

$$\text{Memory} \approx 2^t$$



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 : (y_1, b_1) (y_2, b_2) (y_1, b_3) (y_3, b_4)
1 pair with equal y_i
- ▶ Previous algorithm : 6 comparisons

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Improved Version :

- ▶ Count the occurrences of each y_i :

	y_1	y_2	y_3
$T[y_i]$	2	1	1

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

TD with Less Time Complexity

$D = 0$

for 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

$$\text{Data} = N^{TD} = S \cdot M < N^{ML}$$

$$\text{Time} \approx \max(N^{TD}, S \cdot 2^q)$$

$$\text{Memory} \approx 2^q$$

TD with Less Time Complexity

$D = 0$

```
for  $S$  values of  $x_s \in \mathbb{F}_2^s$  do  
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Remark :

This distinguisher is the same as the statistical saturation (SS) distinguisher

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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

Link between SS, TD and ML distinguishers

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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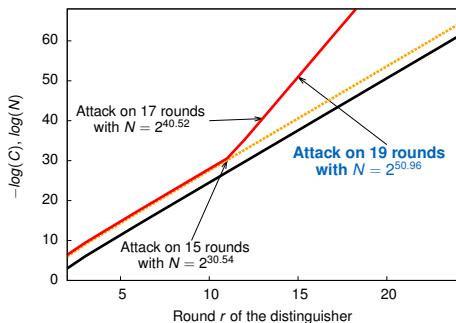
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On the SS Attack on PRESENT [Collard Standaert 09]

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



[Collard Standaert 09]
Data increases linearly

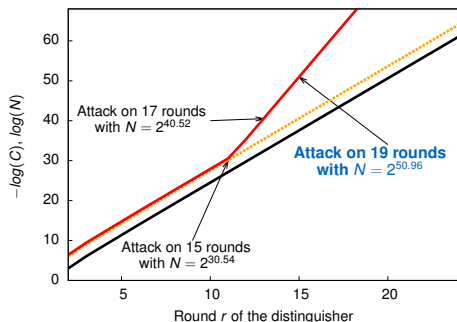
[Leander 11]
Estimate of the capacity C

[Our work]

$$\text{Data is } N = \frac{2^{q+1}}{M \cdot C^2} \cdot \varphi_a^2$$

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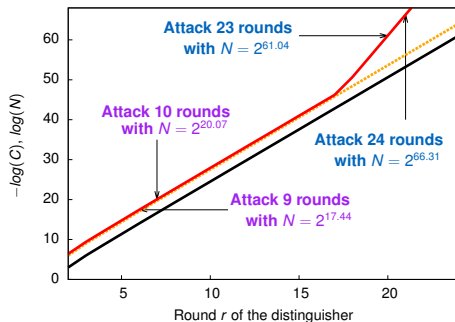
— [Our work]

$$\text{Data is } N = \frac{2^{q+1}}{M \cdot C^2} \cdot \varphi_a^2$$

- ▶ The attack has been verified experimentally [Kerckhof et al 11]
- ▶ Our estimate match with the experiments (N around 2^{51} for 19 rounds)

On the SS Attack on PRESENT [Collard Standaert 09]

Attack on $r + 3$ rounds with $M = 2^{48}$



— Estimate of the capacity C

— Data is proportional to $\frac{1}{C}$

— Data is $N = \frac{2^{q+1}}{M \cdot C^2} \cdot \varphi_a^2$

► In this model, one can only perform an attack 23 rounds

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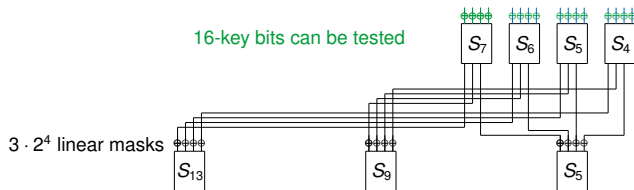
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KP ML and CP TD Attacks : An Example on PRESENT

[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)



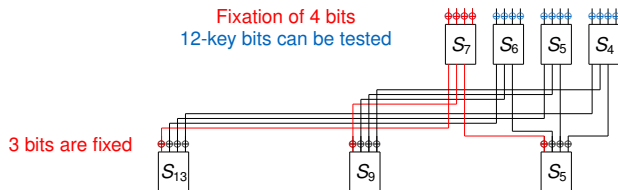
- ▶ KP ML \Rightarrow Guess 16-key bits

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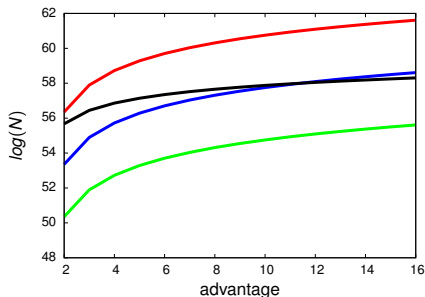
First round : (In Cho's ML characteristic)



- ▶ 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*) :



- KP ML
- CP TD fixing 4 bits
- CP TD fixing 8 bits
- CP TD fixing 12 bits

- ▶ 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 :

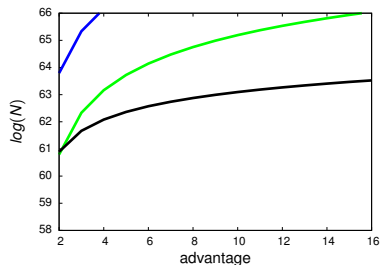
Model	a	Data	Memory	$Time_1$	$Time_2$
CP TD	10	$2^{54.75}$	2^{29}	$2^{54.75}$	2^{70}
KP ML	5	$2^{57.14}$	2^{32}	$2^{57.14}$	2^{75}

$Time_1$: Complexity of the distillation phase

$Time_2$: Complexity of the search phase

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

Example of CP TD Attack on 26 Rounds of PRESENT



— KP ML
— CP TD fixing 4 bits
— CP TD fixing 8 bits

Model	a	Data	Memory	$Time_1$	$Time_2$
CP TD	4	$2^{63.16}$	2^{29}	$2^{63.16}$	2^{76}
KP ML	4	$2^{62.08}$	2^{32}	$2^{62.08}$	2^{76}

- ▶ 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

Conclusion

In this work :

- ▶ We analyze the complexities of some statistical attacks and their relation
- ▶ We show that the SS attack is a TD attack
- ▶ We illustrate that a KP ML attack can be converted to a CP TD attack with smaller complexities

Thank You