## Links Between Differential and Linear Cryptanalysis and Boolean Functions

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## Outline

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Matsui's Algorithms
Linear Hull
Links Between Statistical Attacks
Newer Statistical Cryptanalysis
Recent Links
Multidimensional Linear and Truncated Differential
Properties
Index of Coincidence
Computing Differential Probabilities using Linear Correlations
Distinguishing Distributions
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Thanks to Céline for letting me use some of her slides.

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## Introduction

- Study of APN and PN functions is motivated by conventional differential cryptanalysis.
- Other types of attacks may (?) require stronger countermeasures.
- We will survey recent results on links between statistical attacks on block ciphers.
- The statistical models of distinguishers will be discussed.
- Some bent functions are more vulnerable than some others.


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## Brief History

- Biham-Shamir Crypto1990: Differential Cryptanalysis
- Lai, Massey, and Murphy EC1990: Markov Ciphers and Differential cryptanalysis
- K.N. EC1991: Perfect Nonlinear S-boxes


## Differential Cryptanalysis

Difference between plaintext and ciphertext pairs


Input difference $\delta$
Output Difference $\Delta$
Differential Probability:

$$
\operatorname{Pr}\left[\delta \xrightarrow{E_{k}} \Delta\right]=\operatorname{Pr}\left[E_{k}(x) \oplus E_{k}(x \oplus \delta)=\Delta\right]
$$

Markov cipher $E_{k}=f_{k} \circ g_{k}$

$$
\operatorname{Pr}\left[\delta \xrightarrow{E_{k}} \Delta\right]=\sum_{\gamma} \operatorname{Pr}\left[\delta \xrightarrow{g_{\xi}} \gamma\right] \operatorname{Pr}\left[\gamma \xrightarrow{f_{k}} \Delta\right]
$$

## Provable Security Theorem

with L. Knudsen, Crypto 1992 Rump Session, J Crypt 1995
Theorem ( $\mathcal{K N}$-Theorem) It is assumed that in a DES-like cipher with $F: \mathbb{F}_{2}^{m} \rightarrow \mathbb{F}_{2}^{n}$ the round keys are independent and uniformly random. Then the probability of an s-round differential, $s \geq 4$, is less than or equal to $2 p_{\text {max }}^{2}$.

Here

$$
p_{\max }=\max _{\beta} \max _{\alpha \neq 0} \operatorname{Pr}[\alpha \xrightarrow{F} \beta]
$$

If $F$ bijective, then the claim of Theorem holds for $s \geq 3$.
Later Aoki showed that the constant 2 can be removed.

Minimize $p_{\max } \Leftrightarrow F$ APN

## CRADIC

Cipher Resistant Against Differential Cryptanalysis
aka $\mathcal{K N}$-Cipher
6 -round Feistel cipher with round function $f: \mathbb{F}_{2}^{32} \rightarrow \mathbb{F}_{2}^{32}$ based on the power three operation in $\mathbb{F}_{2}^{33}$
No key schedule, 198-bit key
Jakobsen \& Knudsen FSE1997 break $\mathcal{K} \mathcal{N}$-Cipher

- with 512 chosen plaintexts and $2^{41}$ running time,
- or with 32 chosen plaintexts and $2^{70}$ running time
- using higher order differential cryptanalysis

Round-function based on the inverse mapping not any more resistant.
This approach was then abandonded
... but resumed again recently, see [Boura-Canteaut IEEE
Trans. IT 2013].

## Linear Cryptanalysis

- M. Matsui (EC1993 Bergen) Linear Cryptanalysis


## Linear Cryptanalysis



## Linear approximation with mask vector

 $(u, \tau, w)$ is a relation$$
u \cdot x+\tau \cdot k+w \cdot E_{k}(x)
$$

Input mask $u$
Key mask $\tau$
Output mask w
Bias:
$\varepsilon=2^{-n} \#\left\{x \in \mathbb{F}_{2}^{n} \mid u \cdot x+\tau \cdot k+w \cdot y=0\right\}-\frac{1}{2}$
Correlation: $\operatorname{cor}_{x}(u, w)=2 \varepsilon$

## Matsui's Algorithms

Matsui's Algorithm 1 is a statistical cryptanalysis method for finding one bit of the key $k$ based on the observed correlation of a linear approximation

$$
u \cdot x+w \cdot E_{k}(x)
$$

Matsui's Algorithm 2 is a statistical cryptanalysis method for finding a part of the last round key for a block cipher based on distinguishing cipher data from more random data using observed correlations of a linear approximation

$$
u \cdot x+w \cdot E_{k}^{\prime}(x)
$$

## Linear Hull

Or What is the Equivalent of Differential in Linear Cryptanalysis?

## Correlation for Iterated Block Cipher

We focus on key alternating iterated block ciphers. Let ( $k_{1}, k_{2}, \ldots, k_{r}$ ) be the extended key with the round keys $k_{i}$ derived from $k$ and assume that $E_{k}$ has the following structure

$$
E_{k}(x)=g\left(\ldots g\left(g\left(g\left(x+k_{1}\right)+k_{2}\right) \ldots\right)+k_{r}\right) .
$$

Then [Daemen FSE1994]

$$
\operatorname{cor}_{x}\left(u \cdot x+w \cdot E_{k}(x)\right)=\sum_{\tau}(-1)^{\tau \cdot k} \prod_{i=1}^{r} \operatorname{cor}_{x}\left(\tau_{i} \cdot x+\tau_{i+1} \cdot g(x)\right)
$$

where $\tau=\left(\tau_{1}, \tau_{2}, \ldots, \tau_{r}\right), \tau_{1}=u$ and $\tau_{r+1}=w$.


## Estimating Correlation

- Assumption for Matsui's algorithms: magnitudes of correlations about the same for all keys.
- In general, correlation magnitude varies with the key except when there is a single dominating trail with key mask $\tau$ and trail correlation

$$
\begin{aligned}
\tilde{c}(u, \tau, w) & =\prod_{i=1}^{r} \operatorname{cor}_{x}\left(\tau_{i} \cdot x+\tau_{i+1} \cdot g(x)\right) \\
& =\operatorname{Avg}_{k} \operatorname{cor}(u \cdot x+\tau \cdot k+w \cdot y)
\end{aligned}
$$

## The Linear Hull Theorem

By Jensen's inequality

$$
\operatorname{Avg}_{k} \operatorname{cor}_{x}\left(u \cdot x+\tau \cdot k+w \cdot E_{k}(x)\right)^{2} \geq \tilde{c}(u, \tau, w)^{2}
$$

for all $\tau$, and in general a strict inequality holds. More accurately, the following theorem holds

The Linear Hull Theorem [K.N. EC1994, K.N. DAM 2001] If the round keys of a block cipher $E_{k}$ are uniformly distributed, then

$$
\operatorname{Avg}_{k} \operatorname{cor}_{x}\left(u \cdot x+w \cdot E_{k}(x)\right)^{2}=\sum_{\tau} \tilde{c}(u, \tau, w)^{2}
$$

- Squared correlations of linear hull correspond to probabilities of differentials.
- An analogue of the $\mathcal{K N}$-Theorem for linear cryptanalysis is obtained.


## More Generally: The Fundamental Theorem

$$
\begin{aligned}
& f: \mathbb{F}_{2}^{n} \times \mathbb{F}_{2}^{\ell} \rightarrow \mathbb{F}_{2}, \quad \hat{f}(u, v)=\sum_{x \in \mathbb{F}_{2}^{\prime}, z \in \mathbb{F}_{2}^{s}}(-1)^{u \cdot x+v \cdot z+f(x, z)} \\
& f_{z}(x)=f(x, z), f_{z}: \mathbb{F}_{2}^{n} \rightarrow \mathbb{F}_{2}, z \in \mathbb{F}_{2}^{\ell}
\end{aligned}
$$

Theorem [K.N. EC1994] For all $u \in \mathbb{F}_{2}^{n}$

$$
2^{\ell} \sum_{z \in \mathbb{F}_{2}^{\ell}} \widehat{f}_{z}(u)^{2}=\sum_{v \in \mathbb{F}_{2}^{\ell}} \hat{f}(u, v)^{2}, \text { or equivalently }
$$

$2^{-\ell} \sum_{z \in \mathbb{F}_{2}^{\ell}} \operatorname{cor}_{x}\left(u \cdot x+f_{z}(x)\right)^{2}=\sum_{v \in \mathbb{F}_{2}^{\ell}} \operatorname{cor}_{x, z}(u \cdot x+v \cdot z+f(x, z))^{2}$.
A. Canteaut, C. Carlet, P. Charpin, C. Fontaine. On cryptographic properties of the cosets of $\mathrm{r}(1, \mathrm{~m})$. IEEE Trans. IT 47(4), 14941513 (2001)
N. Linial, Y. Mansour and N. Nisan. Constant depth circuits, Fourier transform, and learnability. Journal of the ACM 40 (3), 607-620 (1993).

## Estimation of Correlation

Methods to catch significant trails:

- Dominant trails: By hand
- Branch and Bound algorithm
- Transition matrices


## Computing an Estimate of Correlation

$$
\begin{aligned}
& \operatorname{Avg}_{k} \operatorname{cor}_{x}\left(u \cdot x+w \cdot E_{k}(x)\right)^{2}=\sum_{\tau_{2}, \ldots, \tau_{r}} \prod_{i=1}^{r} c_{z}\left(\tau_{i} \cdot z+\tau_{i+1} \cdot g(z)\right)^{2} \\
& =\sum_{\tau_{r}} c_{z}\left(\tau_{r} \cdot z+w \cdot g(z)\right)^{2} \sum_{\tau_{r-1}} c_{z}\left(\tau_{r-1} \cdot z+\tau_{r} \cdot g(z)\right)^{2} \\
& \ldots \ldots \sum_{\tau_{3}} c_{z}\left(\tau_{3} \cdot z+\tau_{4} \cdot g(z)\right)^{2} \\
& \sum_{\tau_{2}} c_{z}\left(\tau_{2} \cdot z+\tau_{3} \cdot g(z)\right)^{2} c_{z}\left(u \cdot z+\tau_{2} \cdot g(z)\right)^{2}
\end{aligned}
$$

- This expression gives an iterative algorithm: start from the bottom line to compute for each $\tau_{3}$ the value on the last line.
- Can be made feasible by restricting to $\tau$ with low Hamming weight and keeping only the largest values from each iteration.
- Restrictions on $\tau$ will lead to a lower bound, which is still much larger than any single $\tilde{c}(u, \tau, w)^{2}$.


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## Statistical Attacks

## LINEAR CONTEXT

## DIFFERENTIAL CONTEXT

Linear Cryptanalysis [Tardy, Gilbert 92] [Matsui 93]
Differential Cryptanalysis [Biham, Shamir 90]
Differential-Linear Cryptanalysis [Langford, Hellman 94]
Truncated Differential Cryptanalysis [Knudsen 94]

Higher Order Differential cryptanalysis [Lai 94] [Knudsen 94]
Square Attack, Integral … [Daemen, Rijmen, Knudsen 97]
Statistical Saturation [Collard, Standaert 09]
Zero Correlation [Bogdanov, Rijmen 11]
Impossible Differential Cryptanalysis [Knudsen 98]
Multiple Linear Cryptanalysis
Multiple Differential Cryptanalysis [Albrecht, Leander 12]
[Biryukov, de Cannière, Quisquater 04]
Multidimensional Linear Cryptanalysis [Cho, Hermelin, Nyberg 08]

## Truncated Differential Cryptanalysis



Input difference $\delta$
Output difference $\Delta$
Set of input differences: $\delta \in C$
Set of output differences: $\Delta \in D$
Probability of truncated differential

$$
\frac{1}{|C|} \sum_{\delta \in C} \sum_{\Delta \in D} P[\delta \xrightarrow{F} \Delta]
$$

## Multidimensional Linear Cryptanalysis



Multidimensional linear approximation:
Set of masks $(u, w) \in U \times W$
Capacity: $\sum_{u \in U} \sum_{W \in W} \operatorname{cor}_{x}(u \cdot x+w \cdot y)^{2}-1$

## Recent Links

[Leander EC2011] :
Statistical Saturation $\Leftrightarrow$ Multidimensional Linear
[Bogdanov et al AsiaCrypt2012] :

$$
\text { Integral } \Leftrightarrow \text { Zero Correlation Linear }
$$

Proofs follow from the Fundamental Theorem [N 1994]
[C.Blondeau-K.N. EC2013]:
Zero Correlation Linear $\Leftrightarrow$ Impossible Differential
[C.Blondeau-K.N. EC2014]:
Multidimensional Linear $\Leftrightarrow$ Truncated Differential
Proofs follow from the Chabaud-Vaudenay Link EC1994

## Chabaud-Vaudenay Link

[Chabaud-Vaudenay EC1994]

$$
F: \mathbb{F}_{2}^{n} \rightarrow \mathbb{F}_{2}^{m}
$$

Link between differential and linear cryptanalysis

$$
\operatorname{Pr}[\delta \xrightarrow{F} \Delta]=2^{-m} \sum_{u \in \mathbb{F}_{2}^{n}} \sum_{w \in \mathbb{F}_{2}^{m}}(-1)^{u \cdot \delta+w \cdot \Delta} \operatorname{cor}(u \cdot x+w \cdot F(x))^{2}
$$

- Used for theory (almost bent $\Rightarrow$ APN)
- Not really used for cryptanalysis


## Splitting the Spaces



Focus on the left side:
multidimensional linear context

- all non-zero input and output masks
truncated differential context
- zero input and output differences

Omit the right side:
multidimensional linear context

- zero input and output masks
truncated differential context
- all input and output differences


## Zero Correlation Linear



Zero Correlation


## Zero Correlation Linear :

$$
\operatorname{cor}_{x}\left(\left(a_{s}, 0\right),\left(b_{q}, 0\right)\right)=0
$$

$$
\text { for all }\left(a_{s}, b_{q}\right) \in \mathbb{F}_{2}^{s} \times \mathbb{F}_{2}^{q} \backslash\{(0,0)\}
$$

## Impossible Differential

## Truncated Differential:

$\sum_{\delta_{t} \in \mathbb{F}_{2}^{t} \Delta_{r} \in \mathbb{F}_{2}^{r}} \sum_{\operatorname{Pr}}\left[\left(0, \delta_{t}\right) \rightarrow\left(0, \Delta_{r}\right)\right]=2^{t-q}$
If $\mathrm{t}=\mathrm{q}$ and $\delta_{t} \neq 0$

$\mathbb{F}_{2}^{t}$
mpossible Differential:
$\operatorname{Pr}\left[\left(0, \delta_{t}\right) \rightarrow\left(0, \Delta_{r}\right)\right]=0$ for all $\left(\delta_{t}, \Delta_{r}\right) \in \mathbb{F}_{2}^{t} \times \mathbb{F}_{2}^{r} \backslash\{(0,0)\}$

## Zero Correlation Linear and Impossible Differential



Impossible


If $t=q$
Zero Correlation Linear Distinguisher
is equivalent to
Impossible Differential Distinguisher

## Multidimensional Linear and Truncated Differential



Multidimensional Linear Distinguisher
is equivalent to
Truncated Differential Distinguisher

## Statistical Saturation Attack

For fixed $x_{s} \in \mathbb{F}_{2}^{s}$ denote by $C\left(x_{s}\right)$ the capacity of the distribution of $y_{q}$.

Chosen plaintext sampling for evaluation of the uniformity of the distribution of $y_{q}$, for a fixed $x_{s}$.

## Focus on Distributions

Distribution of values $\left(x_{s}, y_{q}\right) \in \mathbb{F}_{2}^{s} \times \mathbb{F}_{2}^{q}$

- Multidimensional Linear has

$$
\operatorname{Pr}\left(x_{s}, y_{q}\right)=2^{-(s+q)} \sum_{u_{s}, w_{q}}(-1)^{u_{s} \cdot x_{s}+w_{q} \cdot y_{q}} \operatorname{cor}\left(\left(u_{s}, 0\right) \cdot x+\left(w_{q}, 0\right) \cdot y\right)
$$

- Truncated Differential probability

$$
P=2^{-t} \sum_{\delta_{t} \in \mathbb{F}_{2}^{t} \Delta_{r} \in \mathbb{F}_{2}^{r}} \sum_{\operatorname{Pr}}\left[\left(0, \delta_{t}\right) \rightarrow\left(0, \Delta_{r}\right)\right]
$$

These are just different approaches to sampling of the cipher data and measuring the nonuniformity of the same distribution of $\left(x_{s}, y_{q}\right) \in \mathbb{F}_{2}^{s} \times \mathbb{F}_{2}^{q}$.

## The Mathematical Link

The capacity $C$ of the multidimensional linear distribution is defined as

$$
C=\sum_{\left(u_{s}, w_{q}\right) \neq 0} \operatorname{cor}\left(\left(u_{s}, 0\right) \cdot x+\left(w_{q}, 0\right) \cdot y\right)^{2} .
$$

We obtain the link [BN 2014]

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

or

$$
P=2^{s} \sum_{x_{s}, y_{q}} \operatorname{Pr}\left(x_{s}, y_{q}\right)^{2} .
$$

## Coincidences of $\left(x_{s}, y_{q}\right)$

## Index of Coincidence

When solving the period of the key of a Vigènere cipher we count coincidences in letters to evaluate the nonuniformity of the distribution of the alphabet.[Friedman 1922]

Index of Coincidence is a method of ciphertext only differential cryptanalysis, but the idea generalizes to plaintext-ciphertext pairs:

$$
\operatorname{Pr}\left[\left(0, \delta_{t}\right) \rightarrow\left(0, \Delta_{r}\right)\right]=\operatorname{Pr}\left(x_{s} \leftrightarrow x_{s}^{\prime}, y_{q}=y_{q}^{\prime}\right) \leftrightarrow \sum \operatorname{Pr}\left(x_{s}, y_{q}\right)^{2}
$$

So we can evaluate the $\chi^{2}$ statistic of the distribution of $\left(x_{s}, y_{q}\right)$ using truncated differential frequences.
Differential (collision) approach is used in distribution context.

## Efficient Online Entropy Estimator (Röck 2011)

■ Number of comparisons before finding last occurrence of current element:

$$
\begin{aligned}
& \ell_{t}^{r}= \begin{cases}r & \text { if } x_{t} \neq x_{t-j}, 1 \leq j \leq r, \\
\min \left\{0 \leq j \leq r-1: x_{t-1-t}=x_{t}\right\} & \text { otherwise. }\end{cases}
\end{aligned}
$$

- Estimator: $\hat{H}_{\mathrm{pv}}^{r}\left(\mathbf{x}_{[t-r, t]}\right)=\frac{1}{\ln (2)} \sum_{j=1}^{\ell_{t}^{f}} \frac{1}{j}$

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## Using Correlations to Compute Differential Probabilitites

- For some ciphers like PRESENT
- it is easier to estimate linear correlations than differential probabilities
- Single-bit linear trails are dominant
- Computation of correlations using transition matrices as for instance in [Cho CT-RSA2010]
- Use the Chabaud-Vaudenay Link to compute differential probabilities using linear correlations [C.Blondeau-K.N. EC2013]
- Use the linear property of the cipher to mount a differential type of attack [C.Blondeau-K.N. EC2014]


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## Distinguishing Distributions

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## Distinguishing Test

- Distinguishing probability distributions over a large set of values of size $M$
- Uniform distribution
- Non-uniform distribution $p$ with known capacity

$$
C(p)=M \sum_{\eta=1}^{M}\left(p(\eta)-\frac{1}{M}\right)^{2}
$$

- Problem. Determine the data complexity estimates of the $\chi^{2}$ distinguisher.
- Solution. Use statistic

$$
T=N M \sum_{\eta=1}^{M}\left(q(\eta)-\frac{1}{M}\right)^{2},
$$

where $q$ is the distribution obtained from the data of amount $N$.

- Need to determine the probability distribution of $T$ in both cases.


## $\chi^{2}$ Distributions of $T$

- If $q$ is drawn from uniform distribution, then

$$
T=T_{0}=\sum_{\eta=1}^{M} \frac{(N q(\eta)-N / M)^{2}}{N / M} \sim \chi_{M-1}^{2} .
$$

- If $q$ is drawn from nonuniform distribution $p$, then

$$
T=T_{1}=\sum_{\eta=1}^{M} \frac{(N q(\eta)-N / M)^{2}}{N / M} \sim \chi_{M-1}^{2}(\delta),
$$

where

$$
\delta=\sum_{\eta=1}^{M} \frac{(N p(\eta)-N / M)^{2}}{N / M}=N C(p) .
$$

- Denote $C(p)=C$.


## Normal Approximations of Distributions of $T$

- If $q$ is drawn from uniform distribution, then

$$
T=T_{0} \sim \mathcal{N}(M, 2 M)
$$

- If $q$ is drawn from nonuniform distribution with capacity $C$, then

$$
T=T_{1} \sim \mathcal{N}(M+N C, 2(M+2 N C))
$$

- Data complexity

$$
N \geq \frac{\sqrt{M}}{C} \phi
$$

For typical error probabilities, we take $\phi=4$.

## Experiment on a Large Distribution



## Zero-Correlation Distribution

- With full code book of data the distribution of $\left(x_{s}, y_{q}\right)$ should be exactly uniform
- We must do sampling without replacement
- Using hypergeometric distribution, with data size $N$ and distribution size $M+1=2^{s+q}$, we get

$$
\operatorname{Exp}(T)=M \frac{2^{n}-N}{2^{n}-1} \quad \text { and } \quad \operatorname{Var}(T)=2 M\left(\frac{2^{n}-N}{2^{n}-1}\right)^{2}
$$

- Using normal approximation, we get data-complexity

$$
N \approx 2^{n-\frac{s+q}{2}} \phi
$$

Data-sampling without replacement would be more correct also for ordinary linear cryptanalysis, in particular, when close to full code book.

## Sampling Without Replacement



## Zero-Correlations on LBlock (Small Variant)



## Capacities of Bent Functions

- Capacities in some special multidimensional linear setting for certain vectorial Boolean functions were determined in [M.Hermelin-K.N. BFCA2008, M.Hermelin-K.N. CCDS2012].
- Capacity of multidimensional linear approximation of bent function $f: \mathbb{F}_{2}^{n} \rightarrow \mathbb{F}_{2}^{m}$

$$
\begin{aligned}
C & =\sum_{\left(a_{s}, b_{q}\right) \neq 0} \operatorname{cor}\left(a_{s} \cdot x, b_{q} \cdot f(x)\right)^{2} \\
& =2^{s}\left(2^{q}-1\right) 2^{-n}
\end{aligned}
$$

where $0 \leq s \leq n$ and $0<q \leq m$.

- A bent function can be distinguished from a random function using data size

$$
N=2^{n-\frac{s+q}{2}} \phi
$$

## Statistical Saturation Distinguisher of Maiorana-McFarland

- Consider Maiorana-McFarland function $f=\left(f_{1}, \ldots f_{m}\right)$

$$
f_{i}\left(x_{s}, x_{t}\right)=A^{i}\left(x_{s}\right) \cdot x_{t}+g_{i}\left(x_{s}\right)
$$

where $s=t=q=m=n / 2$ [K.N. EC1991].

- For fixed $x_{s} \neq 0, f\left(x_{s}, x_{t}\right)$ is a linear function, and for $x_{s}=0$ it is constant.
- The capacity of the multidimensional distribution of this bent function is equal to $2^{-s}\left(2^{s}-1\right)$ and the multidimensional linear attack has data complexity $N=2^{s} \phi$
- $C\left(x_{s}\right)=0$, for $x_{s} \neq 0$, and $C(0)=2^{s}-1$.
- Pick random $x_{s}$. It takes a few data to verify if $f\left(x_{s}, x_{t}\right)$ is constant. If it is not constant, the distribution of $f\left(x_{s}, x_{t}\right)$ is uniform as the function is bijective. It takes about

$$
N=2^{s+2-\frac{s}{2}}=2^{\frac{s}{2}} \phi
$$

data to distinguish it from random.

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## Conclusions

- Since the invention of linear and differential cryptanalysis researchers have examined their relationships and discovered analogies between their properties.
- Linear hull vs. differential.
- We extended the Chabaud-Vaudenay link to truncated differentials and multidimensional linear approximations.
- Differential attacks can be seen as extensions of linear cryptanalysis.
- Distribution of cipher data values and $\chi^{2}$ statistic offer a sufficiently general setting to handle both differential and linear statistical cryptanalysis.
- Chosen plaintext data sampling can be used for linear cryptanalysis and, vice versa, known plaintext data sampling for differential cryptanalysis.
- Chosen plaintext attack on the vectorial Maiorana-McFarland bent function.

