

# On the Data Complexity of Statistical Attacks against Block Ciphers

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Some known statistical cryptanalyses:

- linear cryptanalysis [Matsui 93];
- differential cryptanalysis [Biham Shamir 91];
- higher order differential cryptanalysis [Knudsen 94];
- impossible differential cryptanalysis [Biham Biryukov Shamir 99];
- ...

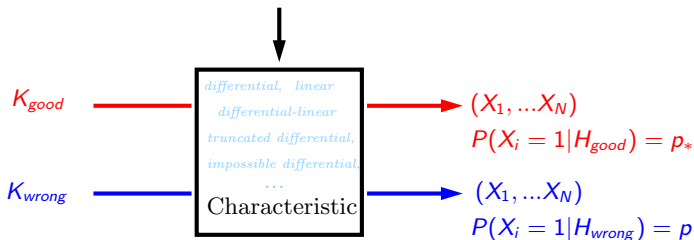
# Using a characteristic to distinguish from random

Let  $\chi$  be some characteristic on a given cipher.

- If the sub-key guess is correct :  $\chi$  occurs with probability  $p_*$ .
- If the sub-key guess is not correct :  $\chi$  occurs with probability  $p$ .

$$X_i = \begin{cases} 1 & \text{if } \chi \text{ occurs in sample } i, \\ 0 & \text{otherwise.} \end{cases}$$

$N$  samples



## Neyman-Pearson (optimal) test:

Accept a candidate  $K$  if

$$\frac{P(X_1, X_2, \dots, X_N | H_{\text{good}})}{P(X_1, X_2, \dots, X_N | H_{\text{wrong}})} > t.$$

This (likelihood) ratio only depends on  $S_N = \sum_{i=1}^N X_i$ ,  $p_*$  and  $p$  and is increasing in  $S_N$ .

Thus, the acceptance condition becomes, for some threshold  $0 < T < N$ ,

$$S_N > T$$

- $S_{N,p_*} = \sum_{i=1}^N X_i$  follows a binomial law of parameters  $(N, p_*)$ .
- $S_{N,p} = \sum_{i=1}^N X_i$  follows a binomial law of parameters  $(N, p)$ .

Two kinds of errors can be made:

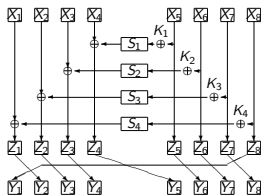
- **Non-detection error probability**  $P(S_{N,p_*} < T)$ ;
- **False alarm error probability**  $P(S_{N,p} \geq T)$ .

The **non-detection error probability** corresponds to the success probability of the cryptanalysis.

The **false alarm error probability** is the expected ratio of kept candidates and thus influences the time complexity of the cryptanalysis.

**Aim:** Finding  $N$  minimal and the corresponding  $T$  such that  $P(S_{N,p_*} < T) \leq \alpha$  and  $P(S_{N,p} \geq T) \leq \beta$  for given values of  $\alpha$  and  $\beta$ .

# Motivation



Generalized Feistel Network  
[Nyberg 96] with:

- 10 rounds;
- 4 S-boxes.

Truncated differential path:  $p_* = 1.18 \cdot 2^{-16}$  and  $p = 2^{-16}$

Differential path:  $p_* = 1.53 \cdot 2^{-27}$  and  $p = 2^{-32}$

## Question:

Which couple of parameters gives the best cryptanalysis ???



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# An algorithm for finding $N$ (1/2)

- For a fixed  $\tau = T/N$ , error probabilities decrease when  $N$  increases.
- For a fixed  $N$ , non-detection error increases with  $\tau$ .
- For a fixed  $N$ , false alarm error decreases when  $\tau$  increases.

## Problem

$N$  and  $T$  are integers  $\Rightarrow$  restrictions on  $\tau$ .

## Solution

Approximations of probabilities ( $G_{nd}$  and  $G_{fa}$ ) that:

- allow non integers values for  $N$  and  $T$ ;
- have the expected properties.

# An algorithm for finding $N$ (2/2)

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**Input:**  $(\alpha, \beta)$  and  $(p_*, p)$

**Output:**  $N$  and  $\tau$  the minimum number of samples and the corresponding relative threshold to reach error probabilities less than  $(\alpha, \beta)$ .

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$\tau_{\min} \leftarrow p$  and  $\tau_{\max} \leftarrow p_*$ .

**repeat**

$$\tau \leftarrow \frac{\tau_{\min} + \tau_{\max}}{2}.$$

Compute  $N_{\text{nd}}$  such that  $\forall N > N_{\text{nd}}, G_{\text{nd}}(N, \tau) \leq \alpha$ .

Compute  $N_{\text{fa}}$  such that  $\forall N > N_{\text{fa}}, G_{\text{fa}}(N, \tau) \leq \beta$ .

**if**  $N_{\text{nd}} > N_{\text{fa}}$  **then**  $\tau_{\max} = \tau$  **else**  $\tau_{\min} = \tau$

**until**  $N_{\text{nd}} = N_{\text{fa}}$ .

**return**  $N$  and  $\tau$ .

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# Number of required samples $N$ for differential and truncated-differential cryptanalyses

## Answer to the question:

In that case, truncated differential is better than differential.

$\alpha$	$\beta$	$\log(N)$ (differential)	$\log(N)$ (truncated differential)
0.5	0.001	27.35	24.31
0.5	$10^{-10}$	29.25	26.37
0.01	0.001	29.43	25.94
0.01	$10^{-10}$	30.54	27.29

Differential:  $p_* = 1.53 \cdot 2^{-27}$  and  $p = 2^{-32}$

Truncated differential:  $p_* = 1.18 \cdot 2^{-16}$  and  $p = 2^{-16}$

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# Gaussian approximation of the binomial tail

$$P[S_{N,p} \leq N\tau] \simeq \int_{-\infty}^{\tau} \frac{1}{\sqrt{2\pi Np(1-p)}} \cdot e^{-\frac{N(x-p)^2}{2p(1-p)}} dx$$

Classically used in linear cryptanalysis:

- [Matsui 93,94];
- [Gilbert 97];
- [Junod 01,03,05];
- [Selçuk 08]
- ...

But ...

... not valid everywhere. For instance, when  $N \cdot p$  is too small as in differential cryptanalysis [Selçuk 08].

# Poisson approximation of the binomial tail

$$P[S_{N,p} \leq N\tau] \simeq \sum_{k=0}^{\lfloor N\tau \rfloor} e^{-Np} \cdot \frac{(Np)^k}{k!}$$

Implicitly used in differential cryptanalysis:

- [Biham Shamir 91,93];
- [Gilbert 97];
- [Selçuk 08]
- ...

**But ...**

... not valid everywhere. For instance, when  $N \cdot p$  is too big as in linear cryptanalysis.

# A good approximation of the binomial tail

We recall the binomial tail:

$$P[S_{N,p} \leq N\tau] = \sum_{k=0}^{\lfloor N\tau \rfloor} \binom{n}{k} p^k (1-p)^{n-k}$$

Approximation found, for instance, in [Arriata, Gordon 89]:

$$P(S_{N,p_*} \leq N\tau) \underset{N \rightarrow \infty}{\sim} \frac{p_* \sqrt{1-\tau}}{(p_* - \tau) \sqrt{2\pi N\tau}} \cdot 2^{-N \cdot D(\tau || p_*)}.$$

Where the Kullback-Leibler divergence is defined by:

$$D(p || q) = p \log_2 \left( \frac{p}{q} \right) + (1-p) \log_2 \left( \frac{1-p}{1-q} \right).$$



# Experimental results

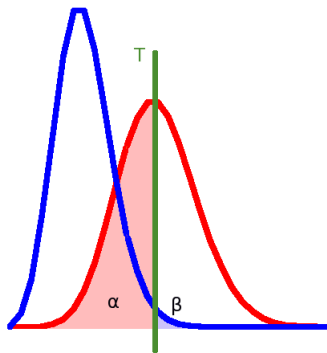
		Exact	Poisson	Gaussian	Ours
<b>Lin Crypt:</b> $p = 0.5$ $p_* = 0.5 + 2^{-10}$	$\beta$	$8.12 \cdot 10^{-5}$	$3.84 \cdot 10^{-3}$	$8.12 \cdot 10^{-5}$	$8.62 \cdot 10^{-5}$
	$\alpha$	$2.97 \cdot 10^{-2}$	$9.14 \cdot 10^{-2}$	$2.97 \cdot 10^{-2}$	$3.58 \cdot 10^{-2}$
<b>Diff Crypt:</b> $p = 2^{-27}$ $p_* = 2^{-20}$	$\beta$	$2.03 \cdot 10^{-3}$	$2.03 \cdot 10^{-3}$	$8.84 \cdot 10^{-5}$	$1.97 \cdot 10^{-3}$
	$\alpha$	$3.27 \cdot 10^{-3}$	$3.27 \cdot 10^{-3}$	$6.66 \cdot 10^{-3}$	$3.33 \cdot 10^{-3}$
<b>Trunc Diff(1):</b> $p = 2^{-4}$ $p_* = 1.01 \cdot 2^{-4}$	$\beta$	$9.29 \cdot 10^{-5}$	$1.46 \cdot 10^{-4}$	$9.23 \cdot 10^{-5}$	$9.90 \cdot 10^{-5}$
	$\alpha$	$9.80 \cdot 10^{-5}$	$1.55 \cdot 10^{-4}$	$9.89 \cdot 10^{-5}$	$1.04 \cdot 10^{-4}$
<b>Trunc Diff(2):</b> $p = 2^{-15}$ $p_* = 1.5 \cdot 2^{-15}$	$\beta$	$5.05 \cdot 10^{-5}$	$5.06 \cdot 10^{-5}$	$3.17 \cdot 10^{-5}$	$5.34 \cdot 10^{-5}$
	$\alpha$	$4.37 \cdot 10^{-4}$	$4.38 \cdot 10^{-4}$	$5.45 \cdot 10^{-4}$	$4.67 \cdot 10^{-4}$

These values are given for  $N = 2^{23}$  and  $\tau = \frac{p_* + p}{2}$ .

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# Approximation of the data complexity (1)

**Aim:** Finding a simple formula to estimate the data complexity.



Fixing  $T$  simplifies the problem.

So we take  $T = Np_*$  what implies  $\alpha \simeq 50\%$ .

## Approximation of the data complexity (2)

$$N' = -\frac{1}{D(p_* || p)} \left[ \log \left( \frac{\lambda \beta}{\sqrt{D(p_* || p)}} \right) + 0.5 \log(-\log(\lambda \beta)) \right],$$

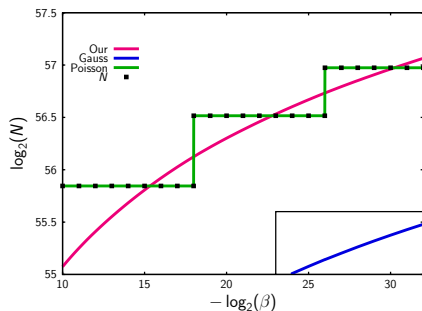
where  $\lambda = \frac{(p_* - p)\sqrt{2\pi(1-p_*)}}{(1-p)\sqrt{p_*}}$ .

$$N' \leq N_\infty \leq N' \left[ 1 + \frac{(\theta - 1) \log(\theta)}{\log(N')} \right],$$

with  $\theta = \left[ 1 + \frac{1}{2 \log(\lambda \beta)} \log \left( -\frac{\log(\lambda \beta)}{D(p_* || p)} \right) \right]^{-1}$ .

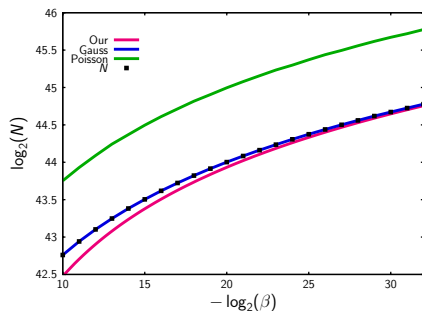
This is a good approximation of  $N$  when  $\beta$  tends to 0.

# Experimental results (1)



Differential cryptanalysis of DES

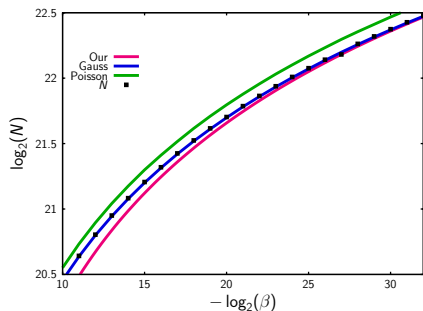
$$p_* = 1.87 \cdot 2^{-56}, p = 2^{-64}$$



Linear cryptanalysis of DES

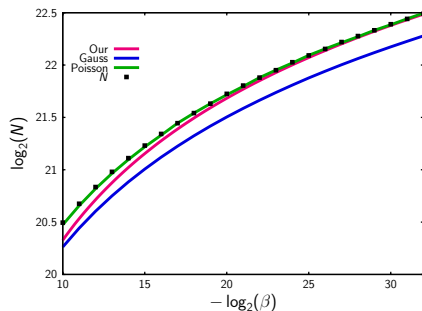
$$p_* = 0.5 + 1.19 \cdot 2^{-21}, p = 0.5$$

# Experimental results (2)



Truncated differential (1)

$$p_* = 1.01 \cdot 2^{-4}, p = 2^{-4}$$



Truncated differential (2)

$$p_* = 1.5 \cdot 2^{-15}, p = 2^{-15}$$

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## Simplified formula for the data complexity

Recall that:

$$N' = -\frac{1}{D(p_*||p)} \left[ \log \left( \frac{\lambda\beta}{\sqrt{D(p_*||p)}} \right) + 0.5 \log(-\log(\lambda\beta)) \right],$$

Using Taylor series,  $\log(2\sqrt{\pi D(p_*||p)})$  is a good estimate of  $\log(\lambda)$ .

$$N'' = -\frac{\log(2\sqrt{\pi}\beta)}{D(p_*||p)}.$$

So comparing the data complexity of two statistical cryptanalyses boils down to comparing the Kullback Leibler divergences of those cryptanalyses.



# Behavior of the data complexity for some statistical attacks

Attack	Parameters		Classical results	$\frac{1}{D(p_*    p)}$
Linear	$p = 0.5$	$p_* - p \ll p$	$\frac{1}{(p_* - p)^2}$	$\frac{1}{2(p_* - p)^2}$
Differential	$p_* \ll 1$	$p_* \gg p$	$\frac{1}{p_*}$	$\frac{1}{p_* \log_2(p_*/p) - p_*}$
Differential-linear	$p = 0.5$	$p_* - p \ll p$	$\frac{1}{(p_* - p)^2}$	$\frac{1}{2(p_* - p)^2}$
Truncated differential	$p_* \ll 1$	$p_* - p \ll p$	unknown	$\frac{2p}{(p_* - p)^2}$
Impossible differential	$p_* = 0$	$p \ll 1$	implicitly : $\frac{1}{p}$	$\frac{1}{p}$
k-th order differential	$p_* = 1$	$p \ll 1$	1	$-\frac{1}{\log_2 p}$

We were interested in the **Data Complexity** of statistical attacks.

This work provides:

- an algorithm to accurately compute the **DC**;
- an asymptotic formula of the **DC**;
- the asymptotic behavior of the **DC**.

Perspectives:

- Key ranking.
- Using an approximation that catches the lattice behavior of the considered random variables.
- Generalizing this work to other distributions than Bernoulli.