

# On the Data Complexity of Statistical Attacks against Block Ciphers

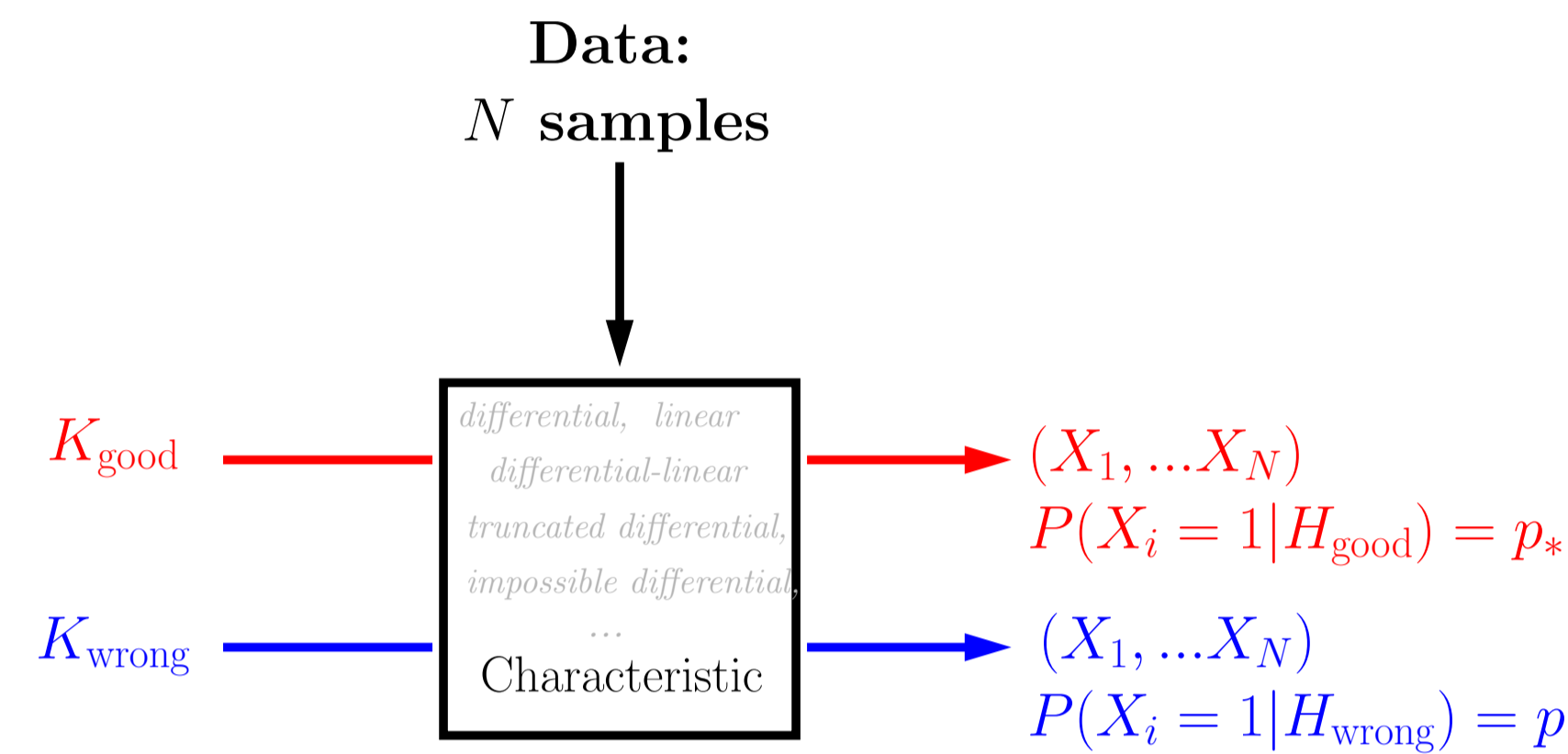
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## Issue & Notation

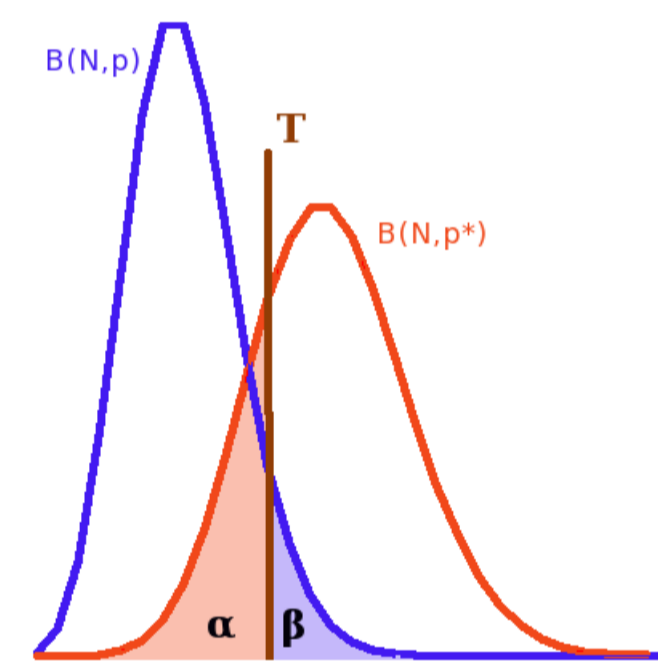
**Context:** statistical attacks based on a distinguisher.

Two hypotheses:

- $H_{\text{good}}$ : “correct subkey guess”
- $H_{\text{wrong}}$ : “non correct subkey guess”



- $S_{N,p_*} = \sum_{i=1}^N X_i$  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:

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

**Aim:** Finding the minimal number  $N$  of samples required to reach some given error probabilities.

## An algorithm for finding $N$

**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  $(\alpha, \beta)$ .

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

**repeat**

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

Compute  $N_{\text{nd}}$  such that  $\forall N > N_{\text{nd}}, P(S_{N,p_*} < \tau N) \leq \alpha$ .

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

**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  $\tau$ .

## Related work

[Junod 01,03,05]: Extensive study of linear cryptanalysis (Gaussian approximation).

[Baignères, Junod and Vaudenay 04]: General study of distinguishers. In particular the exponential behavior of binomial tail is exhibited.

[Selçuk 08]: Key ranking study for linear and differential cryptanalysis.

“It can reasonably be said that the normal approximation for the binomial counters may not be accurate when  $Np_*(1-p_*) < 4$ ”

## Some approximations

Approximations of the binomial law :

- Linear cryptanalysis  $\rightarrow$  Gaussian approximation is valid.
- Differential cryptanalysis  $\rightarrow$  Poisson approximation is valid.
- Truncated-differential cryptanalysis  $\rightarrow$  ???

Parameters	Exact probabilities	Poisson	Gaussian	Kullback
<b>L:</b> $N = 2^{23}, \tau = 0.5 + 2^{-10.58}$ $p_* = 0.5 + 2^{-10}, p = 0.5$	$\beta = 8.12 \cdot 10^{-5}$ $\alpha = 2.97 \cdot 10^{-2}$	$\beta = 3.84 \cdot 10^{-3}$ $\alpha = 9.14 \cdot 10^{-2}$	$\beta = 8.12 \cdot 10^{-5}$ $\alpha = 2.97 \cdot 10^{-2}$	$\beta = 8.62 \cdot 10^{-5}$ $\alpha = 3.58 \cdot 10^{-2}$
<b>D:</b> $N = 2^{23}, \tau = 2^{-23}$ $p_* = 2^{-20}, p = 2^{-27}$	$\beta = 2.03 \cdot 10^{-3}$ $\alpha = 3.27 \cdot 10^{-3}$	$\beta = 2.03 \cdot 10^{-3}$ $\alpha = 3.27 \cdot 10^{-3}$	$\beta = 8.84 \cdot 10^{-5}$ $\alpha = 6.66 \cdot 10^{-3}$	$\beta = 1.97 \cdot 10^{-3}$ $\alpha = 3.33 \cdot 10^{-3}$
<b>TD(1):</b> $N = 2^{23}, \tau = 1.005 \cdot 2^{-4}$ $p_* = 1.01 \cdot 2^{-4}, p = 2^{-4}$	$\beta = 9.29 \cdot 10^{-5}$ $\alpha = 9.80 \cdot 10^{-5}$	$\beta = 1.46 \cdot 10^{-4}$ $\alpha = 1.55 \cdot 10^{-4}$	$\beta = 9.23 \cdot 10^{-5}$ $\alpha = 9.89 \cdot 10^{-5}$	$\beta = 9.90 \cdot 10^{-5}$ $\alpha = 1.04 \cdot 10^{-4}$
<b>TD(2):</b> $N = 2^{23}, \tau = 1.25 \cdot 2^{-15}$ $p_* = 1.5 \cdot 2^{-15}, p = 2^{-15}$	$\beta = 5.05 \cdot 10^{-5}$ $\alpha = 4.37 \cdot 10^{-4}$	$\beta = 5.06 \cdot 10^{-5}$ $\alpha = 4.38 \cdot 10^{-4}$	$\beta = 3.17 \cdot 10^{-5}$ $\alpha = 5.45 \cdot 10^{-4}$	$\beta = 5.34 \cdot 10^{-5}$ $\alpha = 4.67 \cdot 10^{-4}$

Error probabilities obtained with different binomial approximations compared to real ones for:

**L:** linear cryptanalysis, **D:** differential cryptanalysis and **TD(1)-(2):** truncated differential cryptanalysis.

## Approximating binomial tails

**Aim :** Finding a valid approximation for all parameters  $p_*$  and  $p$ .  
Approximation from [Arriata, Gordon, 89].

$$P(S_{N,p} \geq \tau N) \underset{N \rightarrow \infty}{\sim} \frac{(1-p)\sqrt{\tau}}{(\tau-p)\sqrt{2\pi N(1-\tau)}} 2^{-ND(\tau||p)},$$

The Kullback-Leibler divergence:

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

## Rather good approximations

Approximation for  $N$  in the case where  $\tau = p_*$  (implying  $\alpha \approx 0.5$ ).

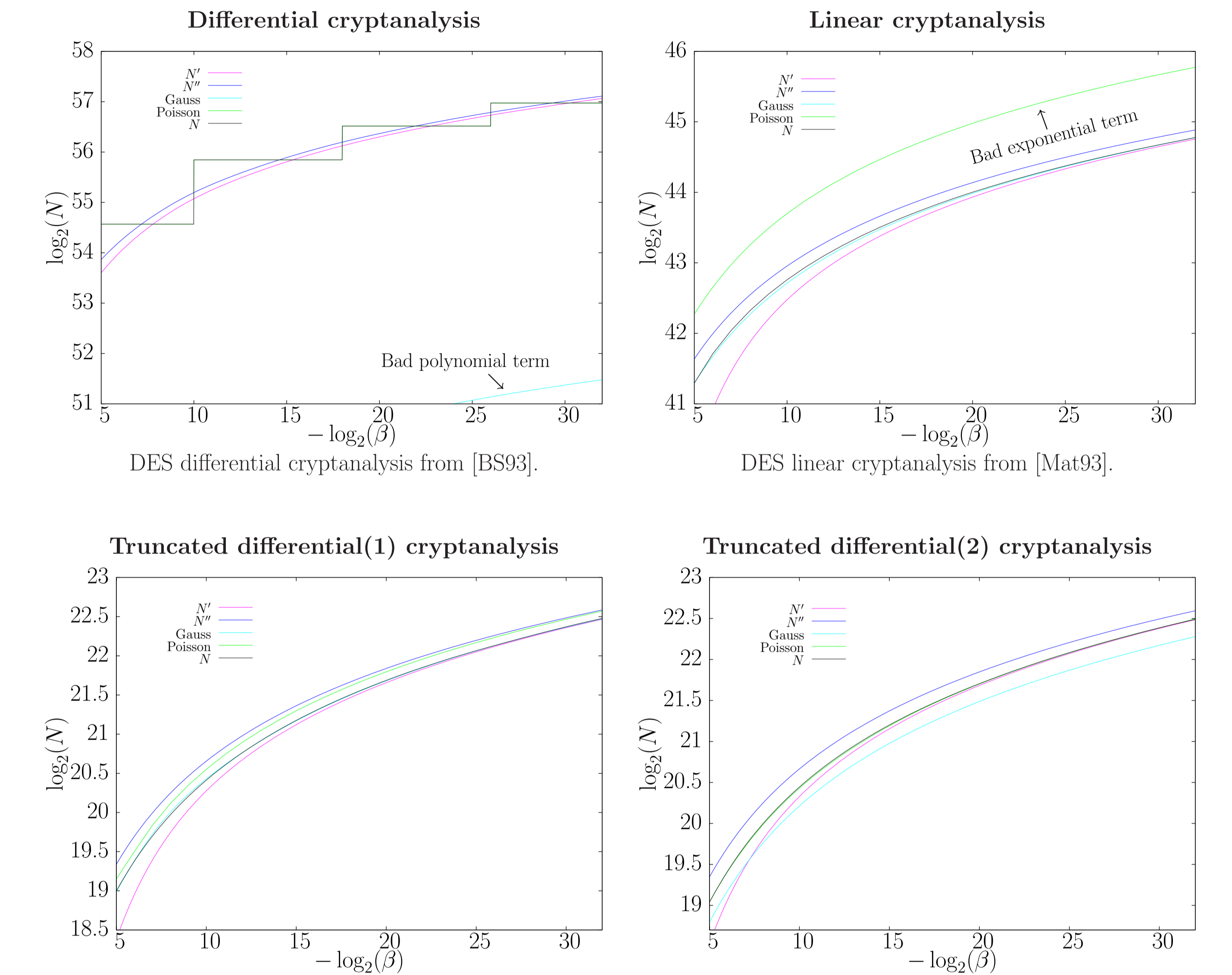
$$N' = -\frac{1}{D(p_*||p)} \left[ \log_2 \left( \frac{\lambda\beta}{\sqrt{D(p_*||p)}} \right) + 0.5 \log_2 (-\log_2(\lambda\beta)) \right],$$

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

This formula can be simplified with a small loss of precision to give the estimate  $N''$ .

$$N'' = -\frac{\log_2(2\sqrt{\pi}\beta)}{D(p_*||p)}.$$

## Experimental results



## Asymptotic data complexity

The estimate  $N''$  shows that for fixed parameters  $\alpha$  and  $\beta$ , the required number of requests depends essentially on the Kullback-Leibler divergence.

Attack	Parameters	Classical results	$\frac{1}{D(p_*  p)}$
Linear	$p = 0.5$ $p_* = 0.5 + \epsilon$	$\frac{1}{(p_* - p)^2}$	$\frac{1}{2(p_* - p)^2}$
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	$\frac{1}{\log_2 p}$