

# Higher-Order DCA against Standard Side-Channel Countermeasures

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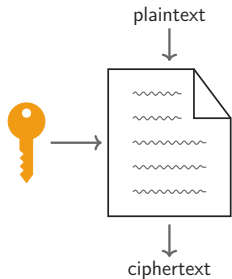
COSADE 2019, 4 April, 2019



# Overview

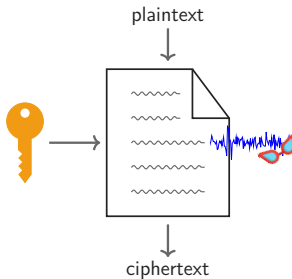
- 1 ■ White-Box Context
- 2 ■ Differential Computation Analysis
- 3 ■ Side-Channel Countermeasures
- 4 ■ Higher-Order DCA
- 5 ■ Multivariate Higher-Order DCA

# White-Box Threat Model



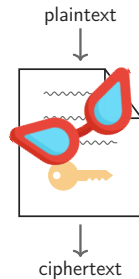
## black-box model

knowing the specification  
observing I/O behavior  
*e.g.* linear/differential cryptanalysis



## gray-box model

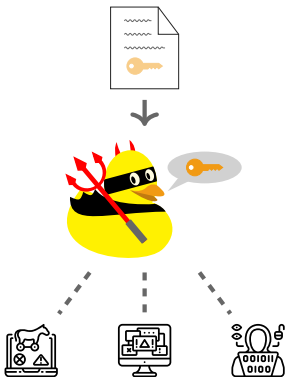
+ side-channel leakages  
(power/EM/time/...)  
*e.g.* differential power analysis



## white-box model [\[SAC02\]](#)

fully controlling the binary  
and its execution environment

# White-Box Adversary



- **Goal:** to extract a cryptographic key, ...
- **Where:** from a software impl. of the cipher
- **Who:** malwares, co-hosted applications, user themselves, ...
- **How:** (*by all kinds of means*)
  - ▶ analyze the code
  - ▶ spy on the memory
  - ▶ interfere the execution
  - ▶ ...

No provably secure white-box scheme for standard block ciphers.

# Typical Applications

## Digital Content Distribution

videos, music, games, e-books, ...



## Host Card Emulation

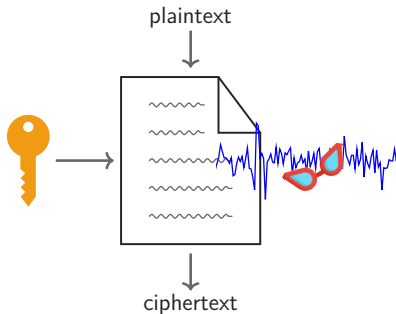
mobile payment without a secure element



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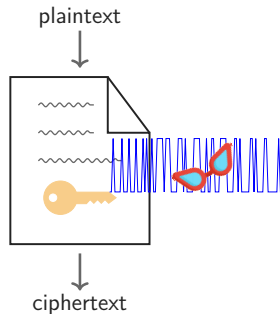
# Differential Computation Analysis [CHES16]



**gray-box model**

side-channel leakages (*noisy*)

e.g. power/EM/time/...



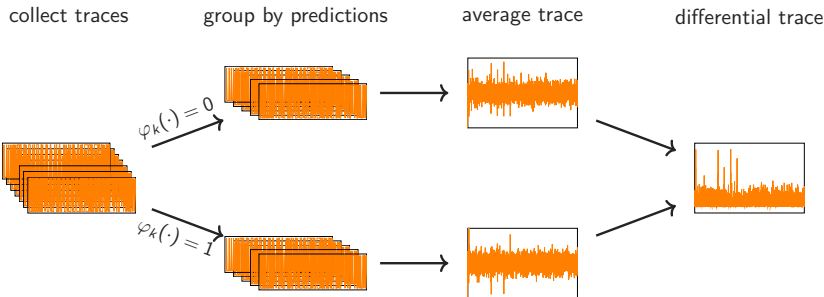
**white-box model**

computational leakage (*perfect*)

e.g. registers/accessed memory/...

# Differential Computation Analysis [CHES16]

*Differential power analysis* techniques on computational leakages



Implying strong *linear correlation* between the sensitive variables and the leaked samples in the computational traces.



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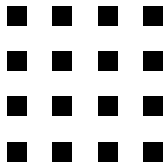
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# Masking

- Split a sensitive variable  $x$  in  $d$  shares *s.t.*

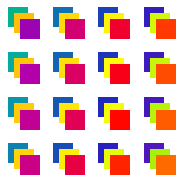
$$x = x_1 \oplus x_2 \oplus \cdots \oplus x_d$$

original states



Masking  
→

masked states



- Any combination of  $d - 1$  shares is independent with  $x$ .

# Shuffling

- **Time shuffling:** randomize the order of computations

iteration in *normal* order



iteration in *randomized* order



# Shuffling

- **Time shuffling:** randomize the order of computations

iteration in *normal* order

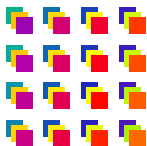


iteration in *randomized* order



- But not enough: traces can be *memory* aligned
- **Memory shuffling:** randomize the memory locations of shares

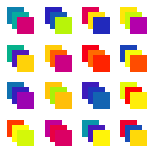
masked states



memory shuffling



memory shuffled states



# Masking and Shuffling: Security

- No external random source
- Security requirements (informally) for PRNG (seeded by the input plaintext):
  - ▶ *Pseudorandomness*: unpredictable outputs
  - ▶ *Obscurity*: hiding design
  - ▶ *Obfuscation*: preventing reverse-engineering
- Masking is good enough to prevent DCA.
- However, still vulnerable to *linear decoding analysis* (LDA)  
[\[ia.cr/2018/098; AC18\]](#)
- Necessary to introduce *noise*

What about masking + shuffling?

# This Work

- We quantify the security brought by masking and shuffling for a passive adversary by introducing
  - ▶ the higher-order variant of DCA attack
  - ▶ and an optimized multivariate version
- We analyze both attacks and verify our results by simulations
- We showcase the masking and shuffling orders that should be taken in practice

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# DCA: a Formal Description

- $N \times t$  matrix  $(v_{i,j})_{i,j}$ :  $N$  computational traces of  $t$  time slots
- $\varphi_k(x)$ : key dependent predictions
- $C$ : correlation measurement

$$\gamma_k = \max_{1 \leq j \leq t} C \left( (v_{i,j})_i, (\varphi_k(x_i))_i \right)$$

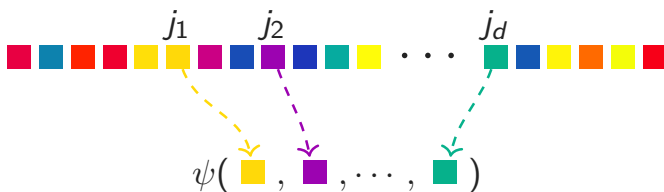
- Success probability:

$$p_{\text{succ}} = \Pr(\operatorname{argmax}_{k \in \mathcal{K}} \gamma_k = k^*) .$$



# Introducing *Higher-Order* DCA

- Trace **pre-processing**: a  $d$ -th order traces contains  $q = \binom{t}{d}$  points:



- Perform DCA attacks on the higher-order traces

# Higher-Order DCA against Masking

If only using *masking*:

- $\exists$  fixed  $j_1^* < \dots < j_d^*$  s.t.  $\varphi_{k^*}(x) = v_{j_1^*} \oplus \dots \oplus v_{j_d^*}$  for all traces
- Hence, the natural *combination function* is

$$\psi(v_{j_1}, \dots, v_{j_d}) = v_{j_1} \oplus \dots \oplus v_{j_d}$$

- Correlation measurement

$$C_k = \#\text{traces s.t. } \varphi_k(x) = v_{j_1} \oplus \dots \oplus v_{j_d}$$

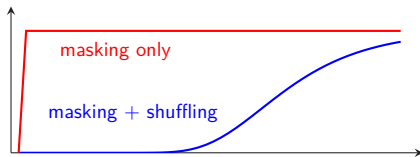
- Even for small  $N$ ,

$$\gamma_k = \max_j C_k \quad \text{satisfys} \quad \begin{cases} = N & \text{if } k = k^* \\ < N & \text{if } k \neq k^* \end{cases}$$

# HO-DCA against Masking and Shuffling

If using both *masking* and *shuffling*:

- $\nexists$  fixed  $j_1^* < \dots < j_d^*$  s.t.  $\varphi_{k^*}(x) = v_{j_1^*} \oplus \dots \oplus v_{j_d^*}$  for all traces
- More traces are required to be successful:



- Limitation: each sample in the higher-order traces is considered *independently*

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# Multivariate Higher-Order DCA

- The multivariate attack optimizes the analysis by exploiting joint information of the higher-order samples on the secrets
- Our proposal is based on a maximum likelihood distinguisher

$$\gamma_k = \Pr \left( K = k \mid (\mathbf{V}_i)_i = (\mathbf{v}_i)_i \wedge (X_i)_i = (x_i)_i \right)$$

- We show that

$$\gamma_k \propto \prod_{i=1}^N C_k(\mathbf{v}_i, x_i)$$

where *the counter*

$$C_k(\mathbf{v}, x) := \#d\text{-tuples} \quad s.t. \quad v_{j_1} \oplus \cdots \oplus v_{j_d} = \varphi_k(x) \quad \text{in one trace.}$$

# Analysis of Multivariate HO-DCA

- **Goal:** to compute the success rate

$$\Pr(\forall k^\times \neq k^*, \gamma_{k^*} > \gamma_{k^\times}) = \Pr(\gamma_{k^*} > \gamma_{k^\times})^{|\mathcal{K}|-1}$$

- **Assumption:** each shuffled trace consists of  $d$  shares + uniform variables elsewhere

- We define the *zero-counter* event

$$\mathcal{Z}_k = \{\exists \text{ a trace s.t. } C_k(\mathbf{v}, x) = 0\}$$

- By the law of total probability

$$\Pr(\gamma_{k^*} > \gamma_{k^\times}) = \Pr(\gamma_{k^*} > \gamma_{k^\times} | \mathcal{Z}_{k^\times}) + \Pr(\gamma_{k^*} > \gamma_{k^\times} | \neg \mathcal{Z}_{k^\times})$$

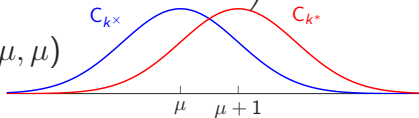
$$\triangleright \mathcal{Z}_{k^\times} \text{ happens} \implies \gamma_{k^*} > \gamma_{k^\times} = 0$$

# $\mathcal{Z}_{k^\times}$ does not Happen

- It is easy to show that

$$\Pr(\gamma_{k^*} > \gamma_{k^\times} | \neg \mathcal{Z}_{k^\times}) = \Pr\left(\frac{1}{N} \sum_{i=1}^N (\log C_{k^*} - \log C_{k^\times}) > 0 | \neg \mathcal{Z}_{k^\times}\right)$$

- Approximately,  $C_{k^*} \sim \mathcal{N}(\mu + 1, \mu)$  and  $C_{k^\times} \sim \mathcal{N}(\mu, \mu)$



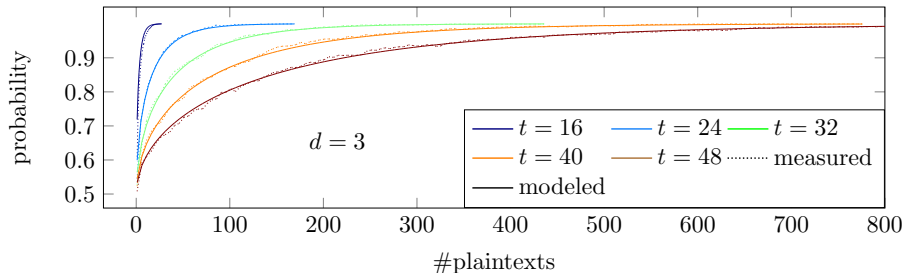
- Thanks to central limit theorem and Taylor expansion

$$p_{\text{succ}} = \Theta\left(\text{erf}\left(\frac{1}{2}\sqrt{\frac{N}{\binom{t}{d}}}\right)\right)$$

- Implying the trace complexity  $N = \mathcal{O}\left(\binom{t}{d}\right)$

# Experimental Verification

- The analysis involves approximations, e.g.:
  - ideal assumption on the traces
  - Gaussian approximations of the counters
  - Taylor expansion truncation, etc
- The accuracy is verified by simulations.





# Attacking Complexity

- Trace complexity:  $N = \mathcal{O}\left(\binom{t}{d}\right)$ .
- Computation complexity:  $\mathcal{O}\left(|\mathcal{K}| \cdot N \cdot \binom{t}{d}\right) = \mathcal{O}\left(|\mathcal{K}| \cdot \binom{t}{d}^2\right)$ .
- A 7-th order masking will bring approximately 85-bit security.

**Table:**  $d$ -th order attacks to achieve 90% success probability, where  $|\mathcal{K}| = 256$ .

$d$	$\log_2 N$	$\log_2 \text{time}$	$d$	$\log_2 N$	$\log_2 \text{time}$	$d$	$\log_2 N$	$\log_2 \text{time}$
3	10.6	32.7	5	21.0	53.5	7	31.6	74.6
4	15.8	43.1	6	26.3	64.1	8	36.9	85.3

# Conclusion

- DCA is an adaption of DPA attack
- It is natural to adapt classical DPA countermeasures
- We propose to higher-order DCA attacks to analyze the effectiveness
- We give close formulae for their success rates and we verify them by simulations
- The security level of this approach is quantified:
  - ▶ trace complexity:  $N = \mathcal{O}\left(\binom{t}{d}\right)$
  - ▶ computation complexity:  $\mathcal{O}\left(|\mathcal{K}| \cdot \binom{t}{d}^2\right)$
- Attackers are forced to perform active attack / reverse engineering

Thank You !