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# Security Research Between Attack and Design

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# Outline of the talk

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- ❑ Motivation
- ❑ Stochastic approach (brief reminder)
- ❑ Several variants
- ❑ Design applications
  - ❑ Interpretation of a  $\beta$ -characteristic
  - ❑ Symmetry considerations
- ❑ Conclusion

# Constructive side-channel analysis

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- ❑ A successful attack proves that there is a weakness but this information may not suffice to fix the vulnerability.
- ❑ Desirable: **constructive attacks**, which
  - ❑ point to the source of the leakage
  - ❑ (ideally) quantify the leakage
  - support target-oriented (re-)design
  - allow interaction between attack and design

# The stochastic approach

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- ❑ target: block cipher
- ❑ exploits power measurements at several time instants  $t_1 < t_2 < \dots < t_m$
- ❑ The measurement values are interpreted as values that are assumed by random variables.
- ❑ The stochastic approach combines engineers' expertise with efficient stochastic methods from multivariate statistics.

## The stochastic model (basic variant)

target algorithm: block cipher (e.g., AES; no masking)

$x \in \{0,1\}^p$  (known) part of the plaintext or ciphertext

$k \in \{0,1\}^s$  subkey **[AES: (typically)  $s = 8$  ]**

$t$  time instant

$$I_t(x,k) = h_t(x,k) + R_t$$

Random variable  
(depends on  $x$  and  $k$ )

deterministic part  
= leakage function  
(depends on  $x$  and  $k$ )

Random variable  
 $E(R_t) = 0$

quantifies the randomness of the side-channel signal at time  $t$

Noise (centered)

## The stochastic model (masking case)

$x \in \{0,1\}^p$  (known) part of the plaintext or ciphertext

$z \in M$  masking value

$k \in \{0,1\}^s$  subkey [AES: (typically)  $s = 8$ ]

$t \in \{t_1, t_2, \dots, t_m\}$  time instant

$$I_t(x, z; k) = h_t(x, z; k) + R_t$$

Random variable  
(depends on  $x, z, k$ )

quantifies the randomness of the side-channel signal at time  $t$

deterministic part  
= leakage function  
(depends on  $x, z, k$ )

Random variable  
 $E(R_t) = 0$

Noise (centered)

## Profiling, Step 1 (I)

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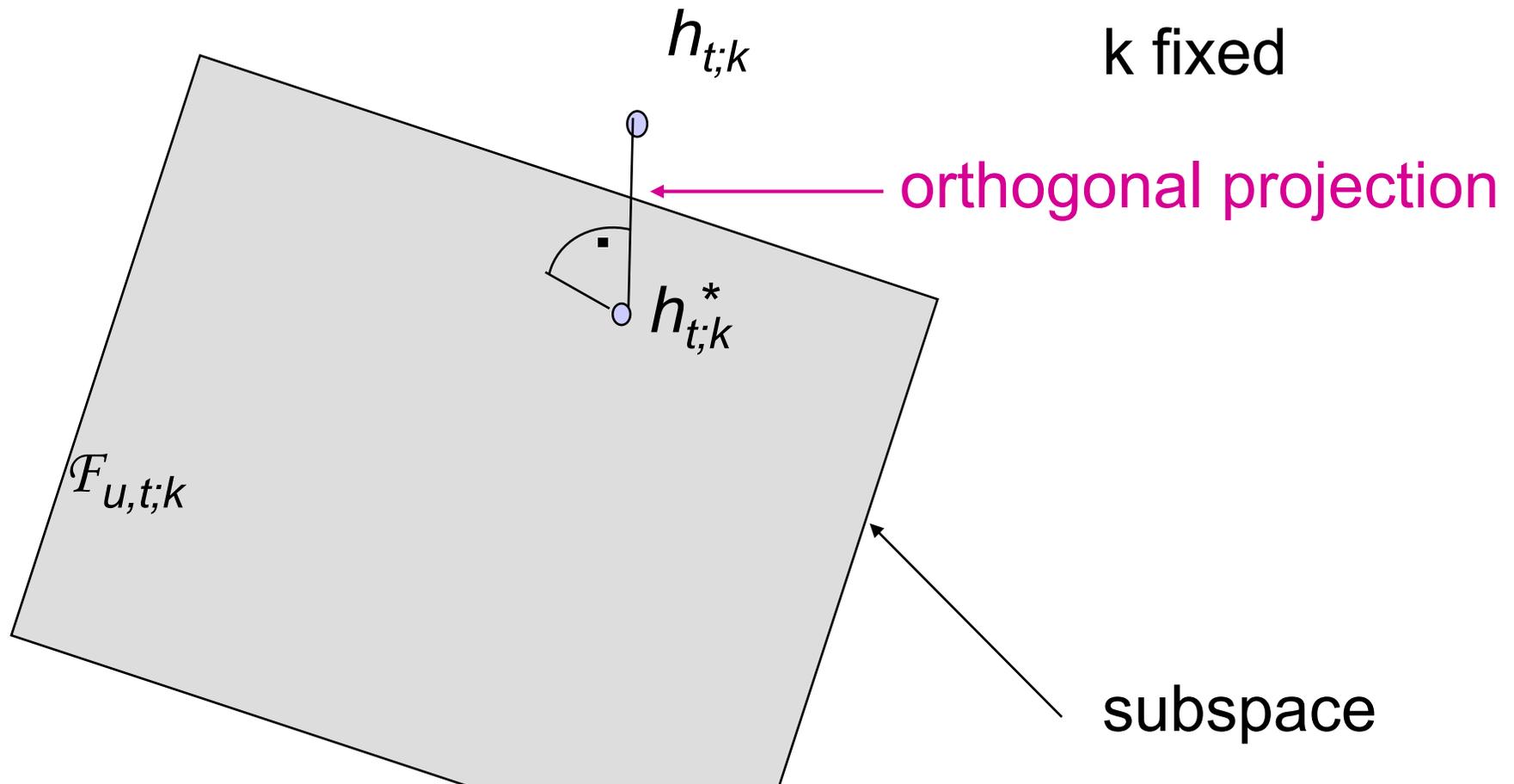
- Fix a subkey  $k \in \{0,1\}^s$ .
- The unknown function

$$h_{t;k}: \{0,1\}^p \times M \times \{k\} \rightarrow \mathbb{R}, h_{t;k}(x,z;k) := h_t(x,z;k)$$

is interpreted as an element of a high-dimensional real vector space  $\mathcal{F}$ . In particular,  $\dim(\mathcal{F}) = 2^p |M|$ .

- Goal: Approximate  $h_{t;k}$  by its image  $h_{t;k}^*$  under the orthogonal projection onto a suitably selected low-dimensional vector subspace  $\mathcal{F}_{u,t;k}$

# Geometric illustration



The image  $h_{t;k}^*$  is the best approximator of  $h_{t;k}$  in  $\mathcal{F}_{u,t;k}$

## Profiling, Step 1 (II)

$$\mathcal{F}_{u,t;k} := \{h' : \{0,1\}^p \times M \times \{k\} \rightarrow \mathbb{R} \mid \sum_{j=0}^{u-1} \beta'_{j,t;k} g_{j,t;k} \text{ with } \beta'_{j,t;k} \in \mathbb{R}\}$$

↑  
(only for masking)

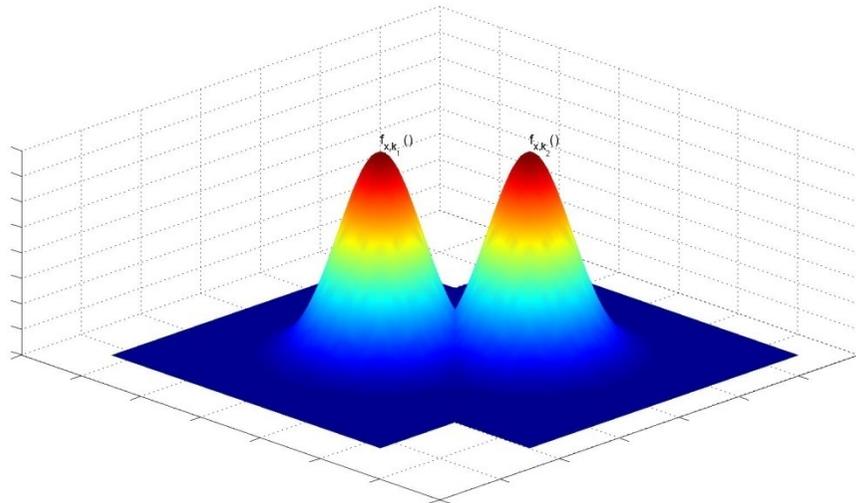
with basis functions  $g_{j,t;k} : \{0,1\}^p \times M \times \{k\} \rightarrow \mathbb{R}$

The basis  $g_{0,t;k}, \dots, g_{u-1,t;k}$  shall be selected under consideration of the attacked device.

The estimation of  $h^*_{t,k}$  can completely be moved to the low-dimensional subspace  $\mathcal{F}_{u,t;k}$ , which reduces the number of measurements to a small fraction.

## Profiling, Step 2

- Estimate the covariance matrix  $C$  of the noise vector  $(R_{t_1}, \dots, R_{t_m})$  (multivariate normal distribution)
- $\rightarrow$  probability densities  $\tilde{f}_{x,z;k}(\cdot)$  for  $(I_{t_1}(x,z;k), \dots, I_{t_m}(x,z;k))$ .



Example ( $m = 2$ )

## Attack phase (analogous to template attacks)

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- ❑ Perform  $N_3$  measurements on the target device
- ❑ (Maximum-likelihood principle) Decide for that subkey  $k^*$ , which maximises

$$\prod_{j=1}^{N_3} \sum_{z' \in M} \Pr(Z_j = z') \tilde{f}_{x_j, z'; k}(i_j)$$

- ❑ For non-masked implementations this formula simplifies to

$$\prod_{j=1}^{N_3} \tilde{f}_{x_j; k}(i_j)$$

# Principal Component Analysis (PCA)

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- ❑ Profiling, Step 2: The covariance matrix  $C$  may be ‘almost’ singular. This might cause large estimation errors for  $C^{-1}$ .
- ❑ Thus it is often advisable to consider only the information ‘contained’ in a subspace of  $R^m$  that is spanned by the eigenspaces of  $C$ , which belong to its largest eigenvalues: i.e., consider  $(P^t C P)$  for a suitable transformation matrix  $P$  instead of  $C$ .
- ❑ Typically, only 1-3 eigenvalues are ‘relevant’.

## Variants of the stochastic approach (I)

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Besides the “standard method” described on the previous slides there exist further variants:

❑ **Profiling with unknown masking values:**

possible but less efficient than with knowledge of the masking values since all time instants have to be handled simultaneously

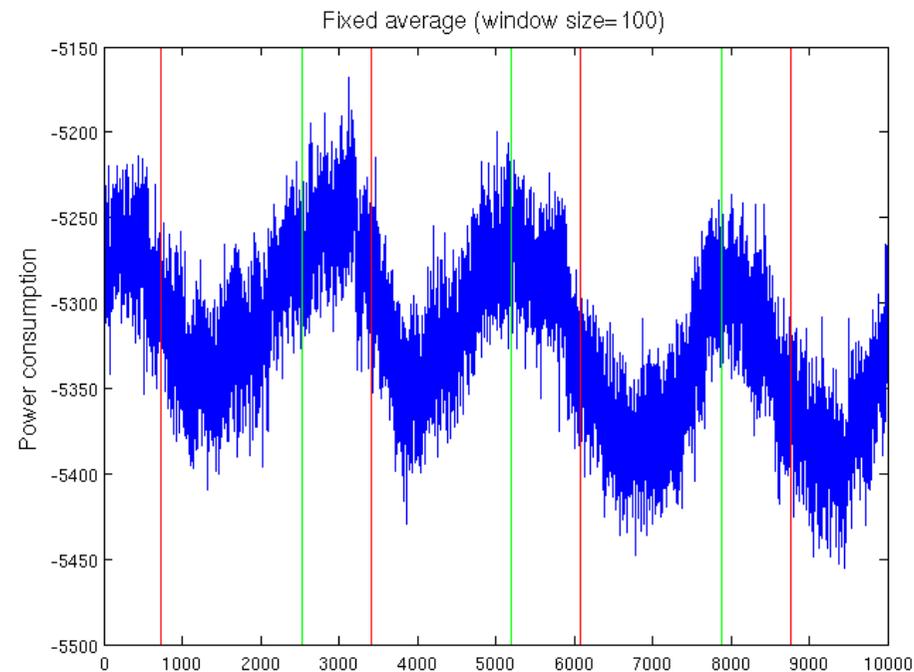
❑ **Attacking without profiling:** possible

## Variants of the stochastic approach (II)

- Within long measurement series the environmental conditions might change, influencing the power consumption and thereby violating the (silent) assumption of having identical conditions all the time.
- Example:  
dpa-v2 power traces

0:00 am

+24h



(time-local average power consumption)<sup>14</sup>

## Variants of the stochastic approach (III)

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- ❑ Problem: There is a drifting offset that affects the effectiveness of profiling based attacks (template attacks, stochastic approach).
- ❑ Solution: new variant of the stochastic approach that tolerates those effects (submitted).

## Newest results (dpa-v2 traces)

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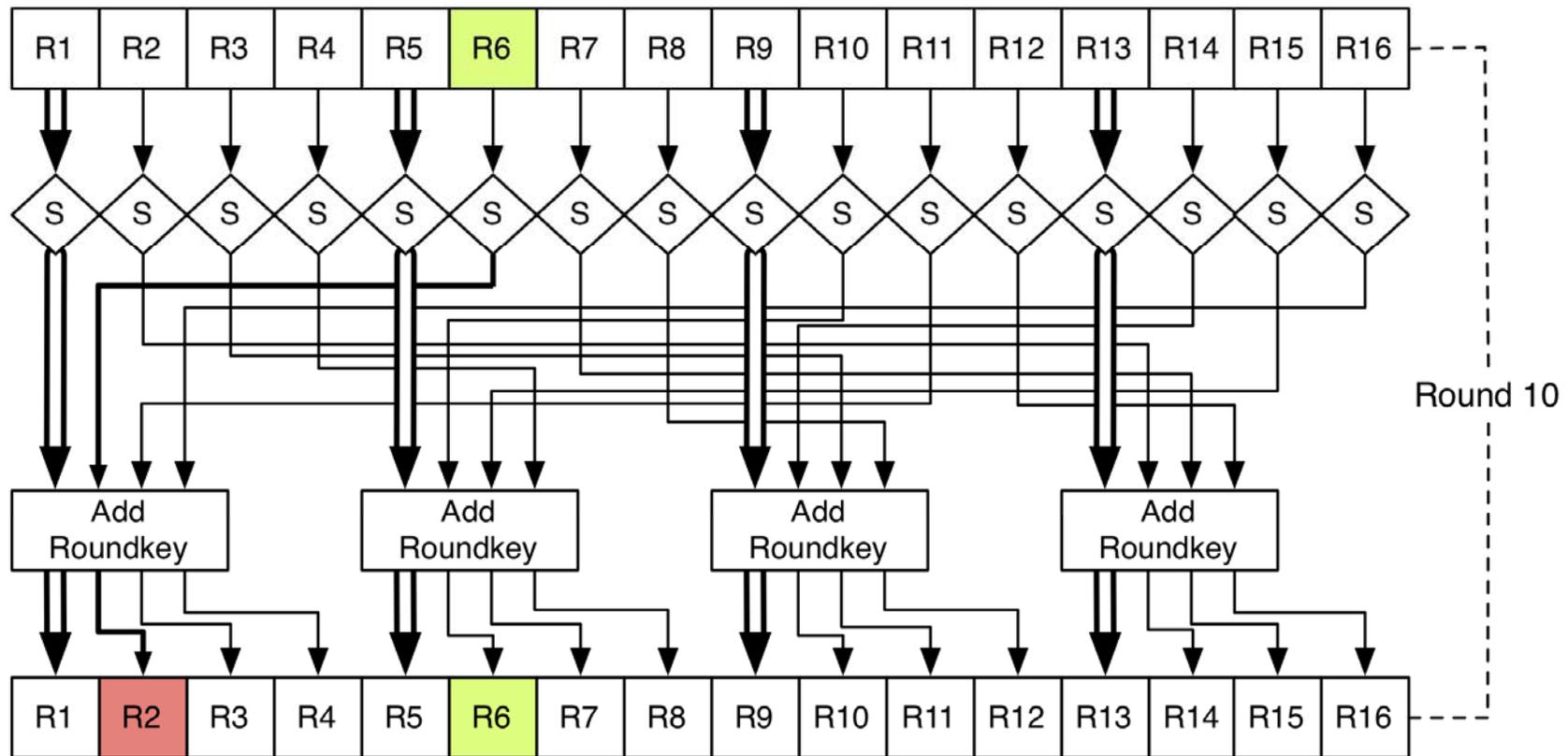
- ❑ Criterion: partial success rate  $> 80\%$ :
  - ❑ Contest winner: 5890 traces, 2<sup>nd</sup> rank: 7515 traces
  - ❑ Our new method: (checked only at the public base!):  
4117 traces
  
- ❑ Criterion: global success rate  $> 80\%$ 
  - ❑ Contest winner: 7061 traces, 2<sup>nd</sup> rank: 10666 traces
  - ❑ Our new method (checked only at the public base!):  
6705 traces
  
- ❑ Note: The best results were gained with a 93-dimensional subspace.

## Note

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- ❑ For design purposes only profiling step 1 is relevant.

All experiments were performed on the  
**SASEBO FPGA evaluation board**



## AES implementation on an FPGA

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Target: key byte  $k_{(2)} \in \{0,1\}^8$  in Round 10

$R_{(x)}$  value of byte register  $x$  after Round 10

9-dimensional vector subspace:

$$g_{0,t;k(2)} ((R_{(2)}, R_{(6)}), k_{(2)}) = 1$$

$$g_{j,t;k(2)} ((R_{(2)}, R_{(6)}), k_{(2)}) = 2((R_{(6)} \oplus S^{-1}(R_{(2)} \oplus k_{(2)}))_j - 0.5) \\ \text{for } 1 \leq j \leq 8$$

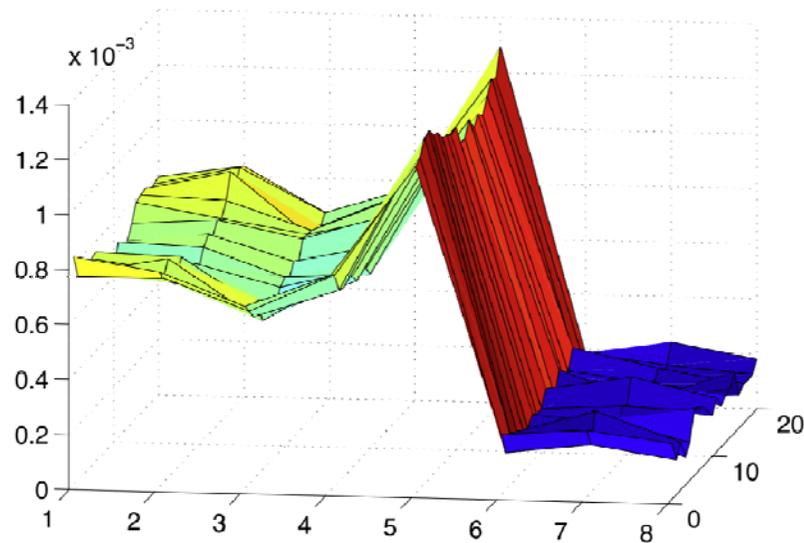
NOTE: Factor '2' and summand ' - 0.5 ' imply that the basis is orthonormal.

□ Let

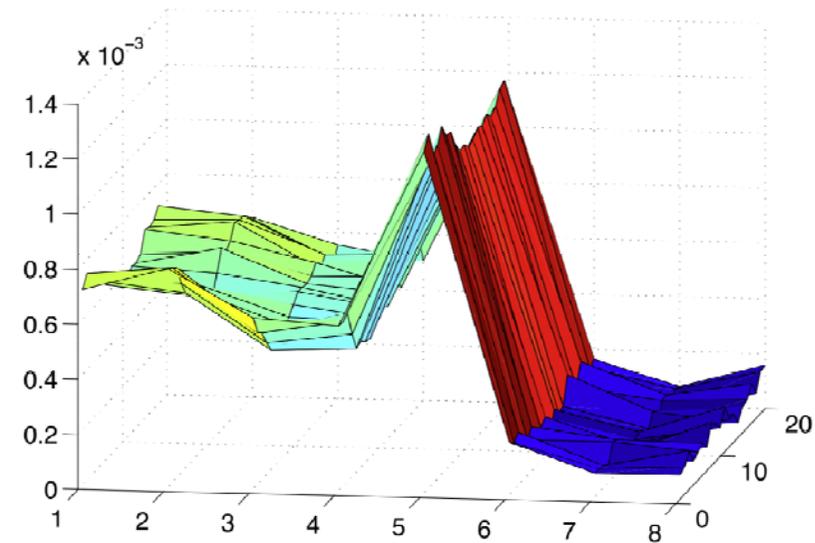
$$h_{t;k}^*(x, k) = \sum_{j=0}^{u-1} \beta_{j,t;k}^* g_{j,t;k}(x, k)$$

- A large coefficient  $|\beta_{j,t;k}^*|$  for some  $j > 0$  means that the ‘direction’ of the basis vector  $g_{j,t;k}$  has considerable impact on the subkey-dependent part of the leakage  $h_{t;k}$ .

# $\beta$ -characteristics of an S-Box design (FPGA, TBL)



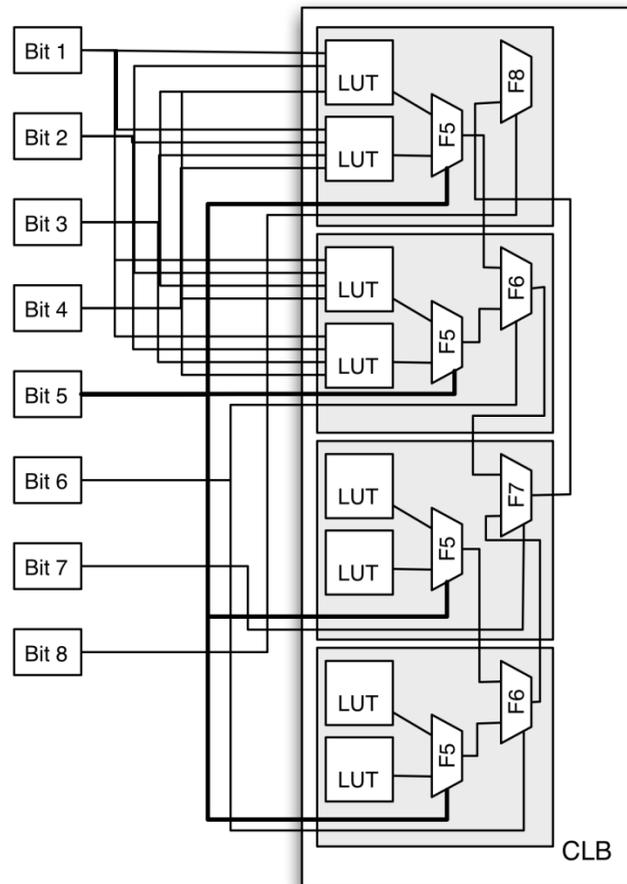
**AES TBL,  $k_{(1)} = 19$ :**  
 **$|\beta_1|, \dots, |\beta_8|$  for  $t_1, \dots, t_{20}$**



**AES TBL,  $k_{(1)} = 209$ :**  
 **$|\beta_1|, \dots, |\beta_8|$  for  $t_1, \dots, t_{20}$**

plots for two different keys and several time instants

## Brief analysis of the implementation (FPT 2010)



- ❑ Part of the SBox implementation after the synthesis process and the Place & Route process
- ❑ The first layer of the multiplexer network is switched by the 5<sup>th</sup> bit
- ❑ Different propagation delays caused by LUT to the multiplexer produces data-dependent glitches.
- ❑ This implies bit-specific higher power consumption.

(The design was synthesized for the Virtex-II pro family.)

## Remark

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- Generally speaking, higher-dimensional subspaces  $\mathcal{F}_{u,t;k}$  may provide more precise leakage models (work in progress).
- Another important question remains: Is the choice of the basis vectors appropriate?

## Leakage model $\mathcal{B}$

The basis vectors

$$g_{0,t;k(2)}((R_{(2)}, R_{(6)}), k_{(2)}) = 1$$

$$g_{j,t;k(2)}((R_{(2)}, R_{(6)}), k_{(2)}) = 2((R_{(6)} \oplus S^{-1}(R_{(2)} \oplus k_{(2)}))_j - 0.5) \\ \text{for } 1 \leq j \leq 8$$

depend on  $((R_{(2)}, R_{(6)}), k_{(2)})$  only through

$$\Phi((R_{(2)}, R_{(6)}), k_{(2)}) := R_{(6)} \oplus S^{-1}(R_{(2)} \oplus k_{(2)}) \\ \text{(symmetry assumption } \mathcal{B})$$

The same symmetry property holds for

$$h^*_{t,k(2)}((R_{(2)}, R_{(6)}), k_{(2)}) \text{ and } \tilde{h}^*_{t,k(2)}((R_{(2)}, R_{(6)}), k_{(2)})$$

## Alternate leakage model $\mathcal{A}$

The basis vectors

alternate 9-dimensional vector subspace:

$$g'_{0,t;k(2)}((R_{(2)}, R_{(6)}), k_{(2)}) = 1$$

$$g'_{j,t;k(2)}((R_{(2)}, R_{(6)}), k_{(2)}) = 2 \left( (S^{-1}(R_{(2)} \oplus k_{(2)}))_j - 0.5 \right) \\ \text{for } 1 \leq j \leq 8$$

depend on  $((R_{(2)}, R_{(6)}), k_{(2)})$  only through

$$\Phi'((R_{(2)}, R_{(6)}), k_{(2)}) := S^{-1}(R_{(2)} \oplus k_{(2)}) \\ \text{(alternate symmetry assumption } \mathcal{A})$$

## Symmetries

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If symmetry assumption  $\mathcal{B}$  is true then  $\beta_{j,t;k'}^* = \beta_{j,t;k''}^*$  for all  $k', k''$ .

Analogously, if symmetry assumption  $\mathcal{A}$  is true then  $\beta'_{j,t;k'}^* = \beta'_{j,t;k''}^*$  for all  $k', k''$ .

Note: In case of a (perfect) symmetry it suffices to estimate  $h_{t,k}^*$  for any single key  $k$ .

## Symmetry metric

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The ratio

$$\frac{2\sqrt{\sum_{j>0}(\tilde{\beta}_{j,t;k'} - \tilde{\beta}_{j,t;k''})^2}}{\sqrt{\sum_{j>0}\tilde{\beta}_{j,t;k'}^2} + \sqrt{\sum_{j>0}\tilde{\beta}_{j,t;k''}^2}}$$

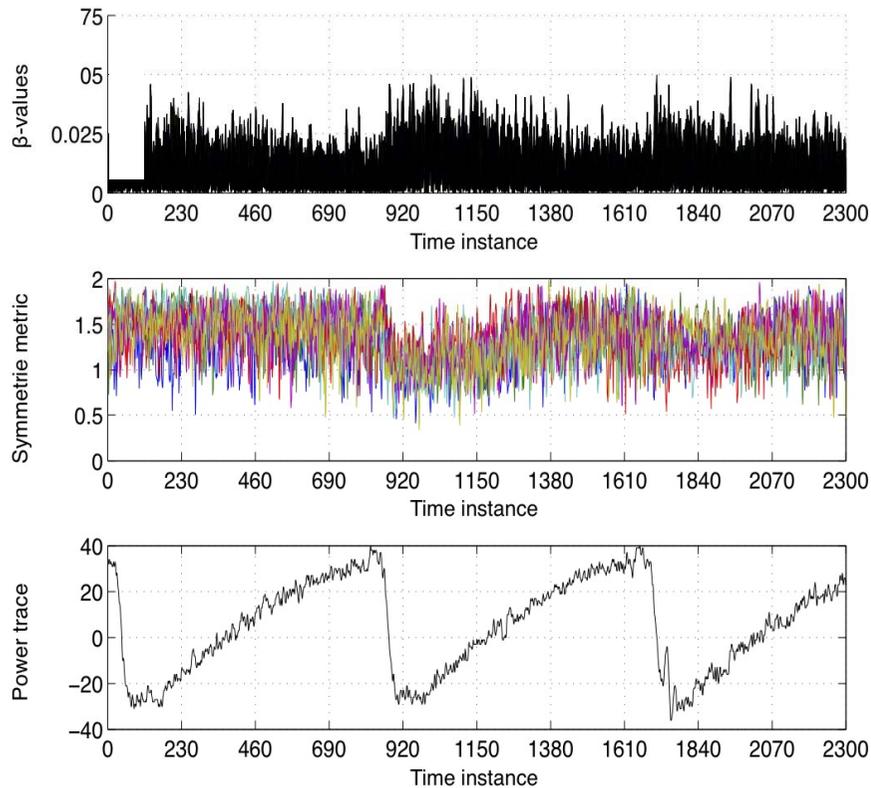
provides a symmetry metric.

This metric has several interesting properties.

In particular, it is invariant under all orthonormal bases of  $\mathcal{F}_{u,t;k}$  with first vector  $g_{0,t;k}=1$ .

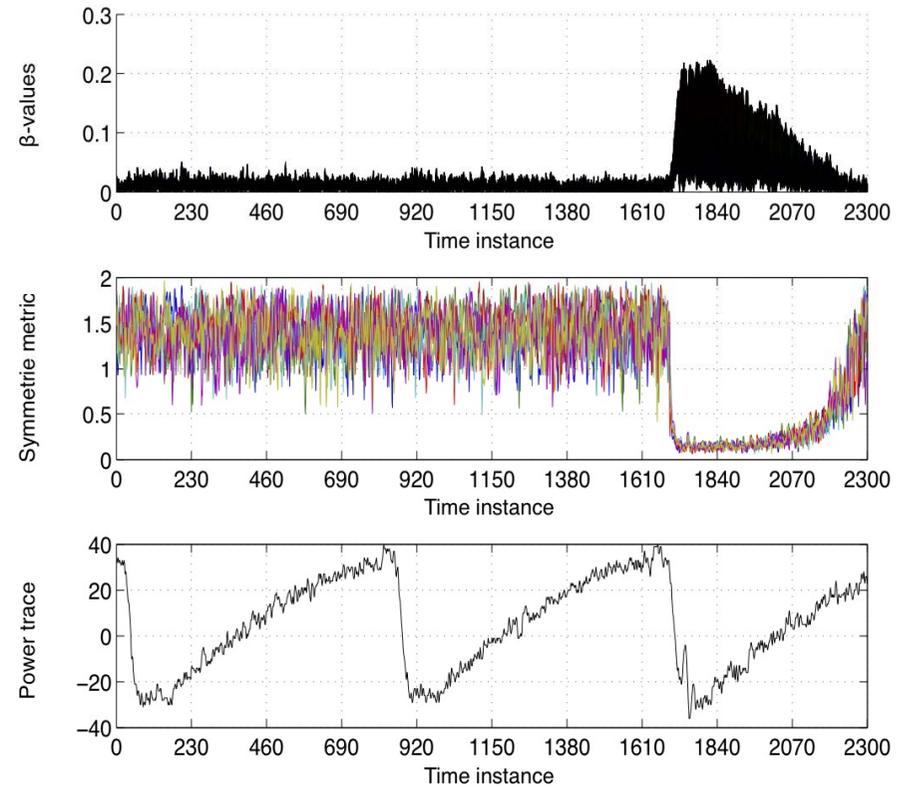
The data-independent coefficient  $\tilde{\beta}_{0,t;k}$  is neglected.

# Empirical Results (DSD 2011, to appear)



Round 10

leakage model  $\mathcal{A}$



Round 10

leakage model  $\mathcal{B}$

## Conclusion

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- ❑ The stochastic approach is useful for many applications.
- ❑ It is a powerful attack method but it can also serve as a tool to support the design of secure implementations (constructive side-channel analysis).
- ❑ In particular, the  $\beta$ -characteristic allows to draw conclusions on specific features of an implementation, and leakage models can be verified or falsified.