

# Side Channel Analysis enhancement: A proposition for measurements resynchronisation

Nicolas Debande, Youssef Souissi, Maxime Nassar  
Thanh-Ha Le, Sylvain Guilley, Jean-Luc Danger

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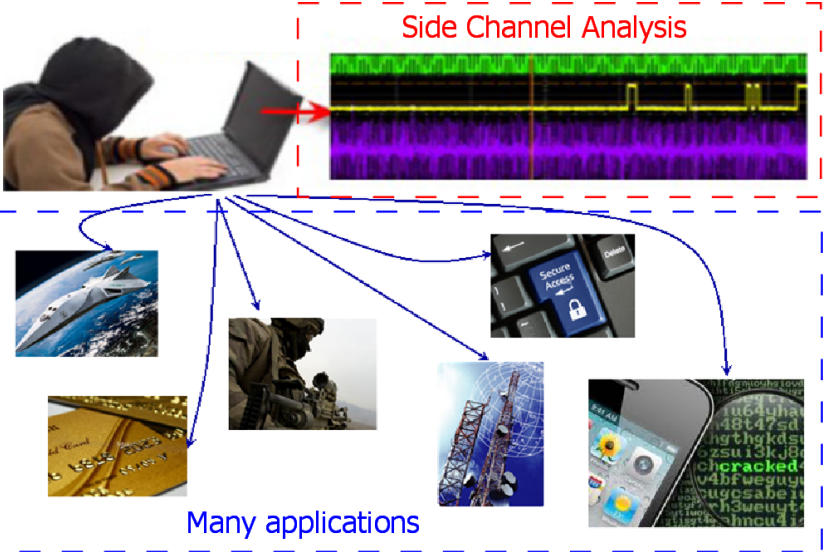


# Summary

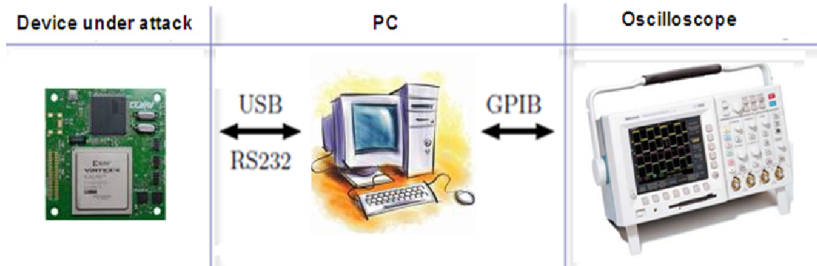
- 1 Introduction
- 2 Curves Registration
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# Embedded Device Context

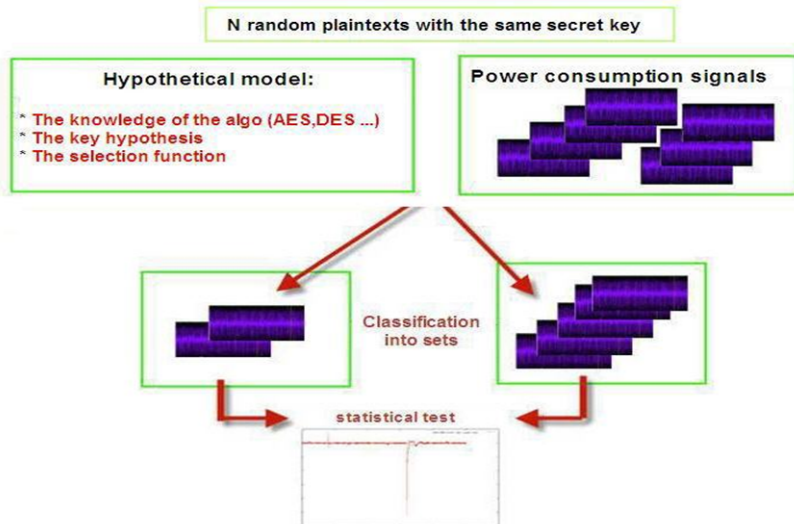


# Embedded Device Context



Side Channel Analysis tools

# Side Channel Analysis



- 1 Introduction
- 2 Curves Registration
  - Importance of Curves Synchronization
  - The Problem
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# Importance of Curves Synchronization

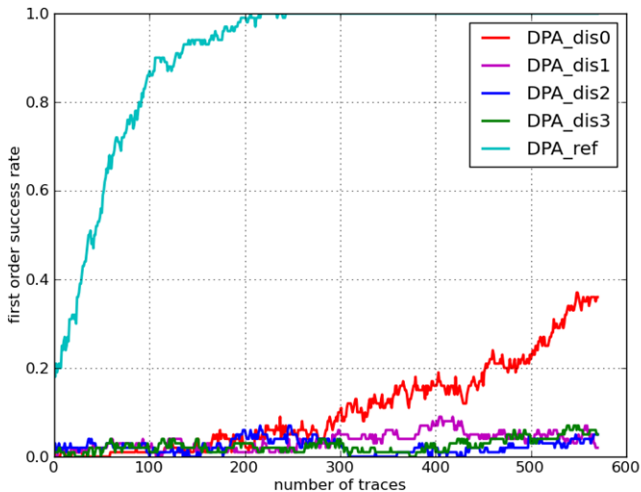


Figure: Desynchronization effect on DPA efficiency



# Importance of Curves Synchronization

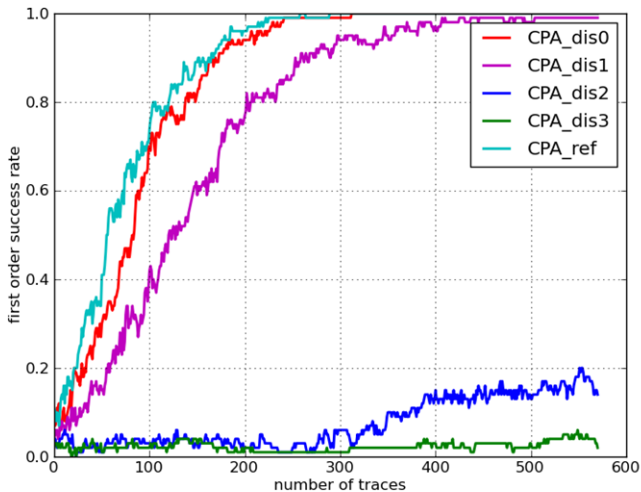
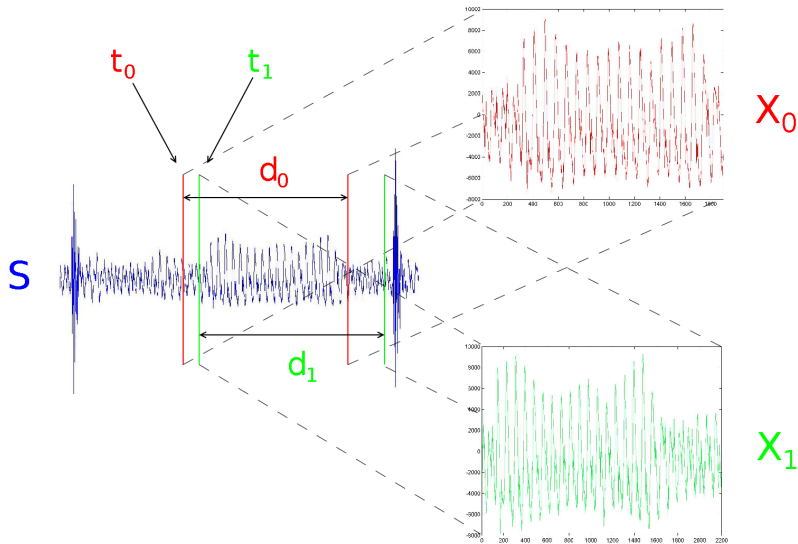


Figure: Desynchronization effect on CPA efficiency

# The Problem



# The Problem

$(t_0, d_0)$  and  $(t_1, d_1)$  are the **temporal basis** of  $X_0$  and  $X_1$

There exists a warping function  $W(t)$  such as:

$$X_0(t) = X_1(W(t)) \quad (1)$$

Let's define  $W(t) = a + b \cdot t$ , where  $a$  is the shift and  $b$  is the zoom

In our example:  $a = t_1 - t_0$ ,  $b = \frac{d_1}{d_0}$

# The Problem

In the Side Channel Analysis context, the evaluator has  $N$  curves  $X_i$  from  $N$  different sources  $S_i$ ,  $i \in \llbracket 1, N \rrbracket$ .

The problem is to find  $W_i(t) = a_i + b_i \cdot t$  such as  $(t_{\text{ref}}, d_{\text{ref}})$  is the temporal basis of  $X_i(W_i(t))$ , for all  $i \in \llbracket 1, N \rrbracket$ .

Some already known methods:

Cross-Correlation	$O(n \log(n))$	→ Only for shifted traces
POC <sup>2</sup>	$O(n \log(n))$	→ Only for shifted traces
DTW <sup>3</sup>	$O(n)$	→ Works with shifted and stretched traces

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<sup>2</sup>N. Homma *et al.*, CHES 2006

<sup>3</sup>J. G. J. van Woudenberg *et al.*, CT-RSA 2011

- 1 Introduction
- 2 Curves Registration
- 3 Resynchronization by Moment**
  - James method
  - Our Proposition
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## Moment Based Synchronization <sup>4</sup>

Temporal mean:

$$\mu_{X_i}^{(1)} = \int t.X_i(t) dt \quad (2)$$

$\mu_{X_i}^{(1)}$  allows to find  $a_i$ .

Standard deviation:

$$\mu_{X_i}^{(2)} = \int (t - \mu_{X_i}^{(1)})^2 . X_i(t) dt \quad (3)$$

$\mu_{X_i}^{(2)}$  allows to find  $b_i$ .

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<sup>4</sup>G. James, Annals of Applied Statistics 2007

## Methodology

For each curve  $X_i$ :

- Compute  $\tilde{X}_i = S(X_i)$ , where  $S()$  combines a smoothing and a weighting function  
→ This step aims to emphasize characteristic patterns of the curve
- Compute  $\mu_{\tilde{X}_i}^{(1)}$  and  $\mu_{\tilde{X}_i}^{(2)}$
- Compute  $b_i = \sqrt{\frac{\mu_{\tilde{X}_i}^{(2)}}{\mu_{\text{ref}}^{(2)}}}$  and  $a_i = \mu_{\tilde{X}_i}^{(1)} - b_i \cdot \mu_{\text{ref}}^{(1)}$

## Some difficulties . . .

- Each support  $S_i$  is different
  - ↪ Implies an error margin on  $\mu_{\tilde{X}_i}^{(1)}$  and  $\mu_{\tilde{X}_i}^{(2)}$
  - ↪ The weighting phase is important!
- $S_i$  is mainly periodic implies that there is no suitable weighting function



## Assumptions

The curves  $X_i$  are mainly periodic and its period  $T_i$  are known by the evaluator.

↔ The evaluator can compute  $b_i = \frac{T_i}{T_{\text{ref}}}$

**Idea:** to work with averaged period  $\mathcal{T}_i$  on each curve.

We redefine  $\mu^{(1)}$ :

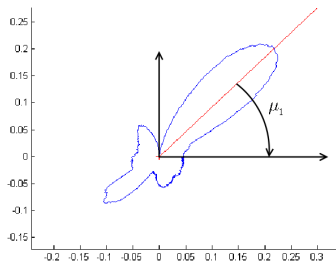
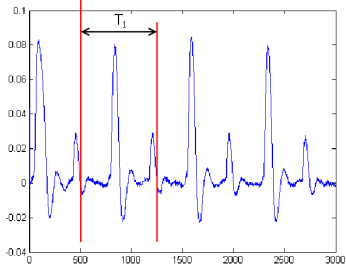
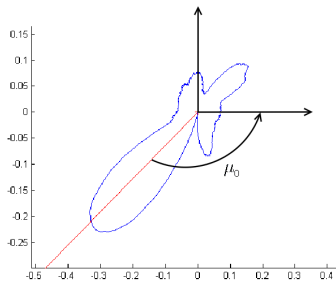
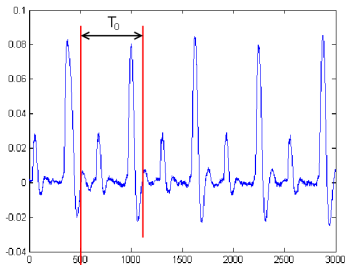
$$\mu_{\mathcal{T}_i}^{(1)} = \sum_{t=1}^{T_i} \mathcal{T}_i(t) e^{\frac{-2\pi i}{T_i} t} dt \quad (4)$$

## Methodology

For each curve  $X_i$ :

- Compute an averaged period  $\mathcal{T}_i$
- Compute  $\tilde{\mathcal{T}}_i = S(\mathcal{T}_i)$ , where  $S()$  combines a smoothing and a weighting function
- Compute  $\mu_{\tilde{\mathcal{T}}_i}^{(1)}$
- Compute  $a_i = \frac{\arg\left(\mu_{\tilde{\mathcal{T}}_i}^{(1)}\right) \cdot \mathcal{T}_i}{2\pi}$

# Resynchronization by Moment



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# Experimental results

## Case of shift only

	Set	$X_i$	RM( $X_i$ )	AOC( $X_i$ )
Averaged	1	0.1390	0.9937	1.000
	2	0.0444	0.9239	1.000
	3	0.0244	0.8807	0.9919
Noised	1	0.1357	0.8960	0.9240
	2	0.0404	0.8486	0.9239
	3	0.0316	0.8181	0.9176

Table: Difference of efficiency between AOC (cross-correlation) and RM (Resynchronization by Moment)

Metric used:  $\rho = \frac{A(\sum X_i)}{\sum A(X_i)}$ , where  $A(X)$  returns the amplitude of  $X$

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## Advantages of RM

- Lower computing complexity ( $O(n)$ )
- Works with stretched curves

## Perspectives

- Experimental results obtained without smoothing and weighting phase  
↳ Possible enhancement by working on this step
- Experimental works on shifted and stretched traces
- Synchronization of curves acquired with power varying  
↳ Use of a circular standard deviation  $\mu_{\tilde{T}_i}^{(2)}$

# Thank you for your attention

`nicolas.debande@morpho.com`  
`nicolas.debande@telecom-paristech.fr`