

# Behind the Scene of Side Channel Attacks

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ANSSI

# Agenda

- 1 Side Channel Attacks (SCA)**
  - a. Background
  - b. Contributions
- 2 Linear Regression Attack (LRA)**
  - a. LRA Basics
  - b. Experimental Results
- 3 Template Attack (TA)**
  - a. Template Attack Basics
  - b. Experimental Results
- 4 Conclusion**



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## 1 Side Channel Attacks (SCA)

a. Background

b. Contributions

## 2 Linear Regression Attack (LRA)

a. LRA Basics

b. Experimental Results

## 3 Template Attack (TA)

a. Template Attack Basics

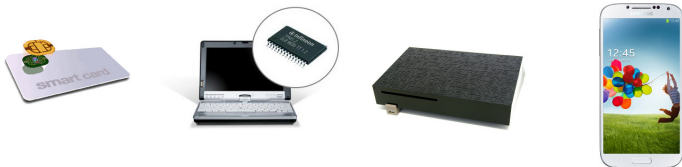
b. Experimental Results

## 4 Conclusion



# Context

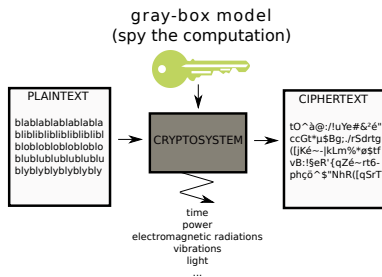
- Since the 90's, increasing use of **Embedded Systems**
  - ▶ 7G smartcards sold in 2012 (SIM, banking, pay-TV, ID, ...)



- Embedded Systems integrating **Cryptography** are susceptible to **Side Channel Cryptanalysis**

# Side Channel Cryptanalysis [Kocher et al - Crypto99]

- A CMOS device leaks info. about its state during a computation through **side-channels** e.g.: time, power consumption, EM radiations, ...
- **SCA** exploit these **physical leakages** to guess a **secret**



# Generic SCA Flow

1. Collect  $N$  side channel traces w. known inputs  
 $t_1 \rightarrow Enc(p_1, k), \dots, t_N \rightarrow Enc(p_N, k)$
2. Choose sensitive variable depend. on input & secret  
e.g. AES Sbox output  $\rightarrow v_i^{\hat{k}} = S(p_i \oplus \hat{k})$
3. Choose a Leakage Model  
e.g. Hamming Weight (H)
4. Compute predictions for each key hypothesis  
 $\hat{k} = 0 \rightarrow H(v_1^{\hat{k}=0}), \dots, H(v_N^{\hat{k}=0})$   
...  
 $\hat{k} = 255 \rightarrow H(v_1^{\hat{k}=255}), \dots, H(v_N^{\hat{k}=255})$
5. Use a distinguisher to discriminate the correct key by comparing the  $N$  traces and the predictions



# SCA flow and Leakage Model: 3 cases

## 1. Select a priori a Leakage Model

- ▶ Hamming Weight, Hamming Distance
- ▶ Used in classical SCA (DPA, CPA, MIA, ...)

## 2. Select a priori a space of Leakage Models

- ▶ Attack will *guess* the correct model in selected space
- ▶ Used in Linear Regression Attack (LRA)

## 3. Infer a Leakage Model through profiling before attack

- ▶ A preliminary step is performed on an *open copy* of the device to build a leakage model for each key value
- ▶ Used in Template Attack (TA)



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# Microelectronics & Side Channel Cryptanalysis

## ■ Moore's law:

- ▶ Nb. of transistors on ICs doubles approx. every two years
- ▶ CMOS processes decrease  
1995 → CMOS process 350nm / 2013 → CMOS process 22nm

## ■ Consequence:

- ▶ **intra-chip variability** increases  
⇒ bits leak differently ones from others
- ▶ **inter-chip variability** increases  
⇒ two identical ICs behave differently

⇒ New **challenges** for **Side Channel Attacks** ?!



# Our Contributions

- Propose a study on the **practicality** of:
  - ▶ **Linear Regression Attack (LRA)**
  - ▶ **Template Attack (TA)**
- Perform experiments on **3 different microcontrollers**:
  - ▶ Device A - CMOS process 350nm - AES128 enc.  
51600 points per trace - highest SNR<sup>1</sup>: 0.3
  - ▶ Device B - CMOS process 130nm - AES128 enc.  
16800 points per trace - highest SNR<sup>1</sup>: 0.6
  - ▶ Device C - CMOS process 90nm - AES128 enc.  
12800 points per trace - highest SNR<sup>1</sup>: 0.09
- Use of **3 copies** of each device for cross-tests

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<sup>1</sup>SNR: Signal-to-Noise Ratio



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# Linear Regression Attack [Doget et al - Cosade11]

- **Leakage function  $\mathcal{L}$** : models the handling of sens. var.  $v$   
handling of  $v \rightarrow \mathcal{L}(v) + \mathfrak{B}$ , with  $\mathfrak{B}$  a gaussian noise
- In LRA,  $\mathcal{L}$  assumed **unknown** and viewed as a **multivariate polynomial** in the bit-coordinates  $v_i$  of  $v$  w. coefs. in  $\mathbb{R}$   

$$\mathcal{L}(v) = \underbrace{\epsilon_0 v_0 + \epsilon_1 v_1 + \dots}_{\text{linear part}} + \underbrace{\epsilon_{0,1} v_0 v_1 + \epsilon_{0,2} v_0 v_2 + \dots}_{\text{quadratic part}} + \underbrace{\dots}_{\text{etc}}$$
- In LRA, **guessing  $\mathcal{L}$**  is hence **equivalent** to **solve**  
a **polynomial interpolation** in a noisy context  
 $\Rightarrow$  use of linear regression techniques



## LRA Issues

- Previous works reporting experiments on LRA consider **side channel traces** composed of one **unique point**
- In practice, side channel traces are never composed of one **unique point**, but rather several **thousands**
- Classical strategy consists in applying SCA on **each time sample** and to keep the key candidate **maximizing the score** over all time samples
- In our experiments, such a strategy did not work for LRA



## Our Solution

- From our experiments, we observed that **correct key  $k$  is ranked first** at time samples where:
  - ▶ the distance  $score(k) - \mathbb{E}[score(k)]$  is large
  - ▶  $Var[score(k)]$  is small
  
- We hence deduced a **normalization step**:
  - ▶ center the scores
  - ▶ divide by their variance
  - ▶  $normalized\_score(k) = \frac{score(k) - \mathbb{E}[score(k)]}{Var[score(k)]}$



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## LRA vs. Normalized LRA (device A - 350nm)

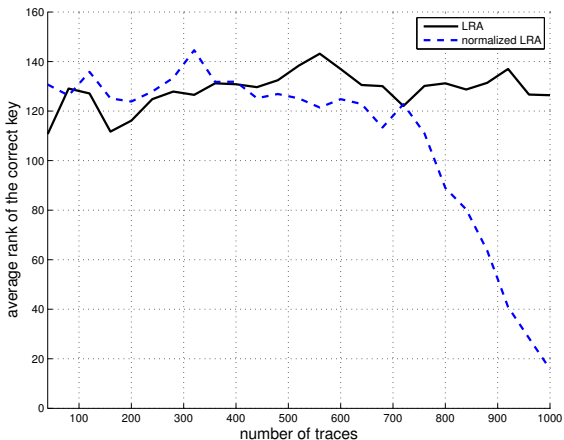


Figure: Correct key rank evolution vs. nb. of traces





# LRA vs. Normalized LRA (device B - 130nm)

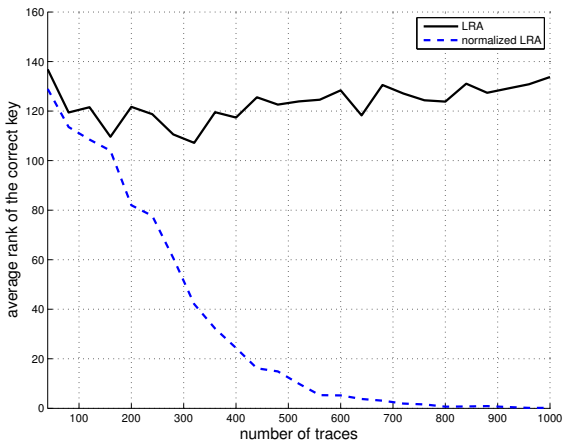


Figure: Correct key rank evolution vs. nb. of traces



# LRA vs. Normalized LRA (device C - 90nm)

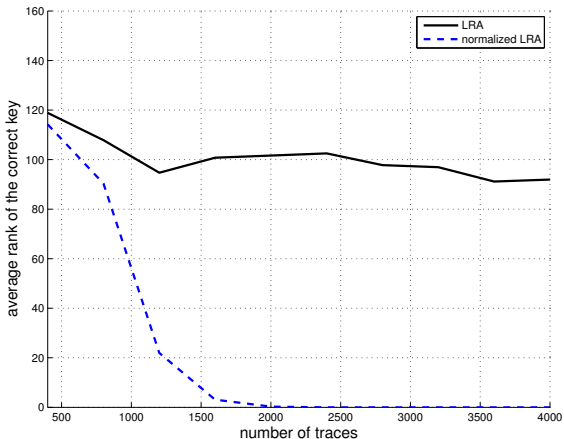


Figure: Correct key rank evolution vs. nb. of traces



# Normalized LRA vs. CPA (device C - 90nm)

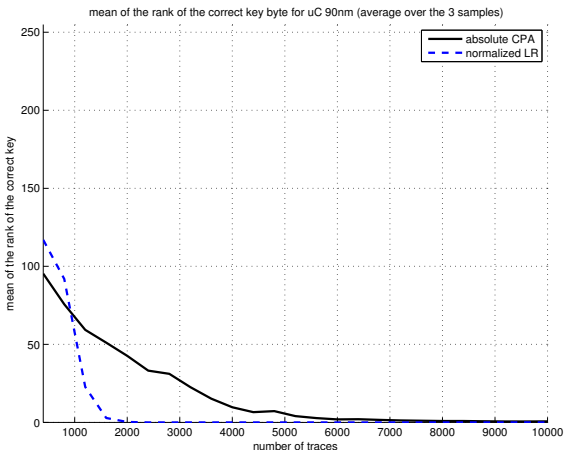


Figure: Correct key rank evolution vs. nb. of traces



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# Template Attack (TA) [Chari et al - CHES02]

## 1. Profiling Phase (performed on an open device copy)

1.1 Collect  $M$  side channel traces w. known inputs & keys

$$t_1 \rightarrow Enc(p_1, k_1), \dots, t_N \rightarrow Enc(p_M, k_M)$$

1.2 Choose **sensitive variable** depend. on input & secret

$$\text{e.g. AES Sbox output} \rightarrow v_i^{\hat{k}} = S(p_i \oplus \hat{k})$$

1.3 Compute the **pdf** of the leakage for each key value

$$pdf_{\hat{k}=0}, \dots, pdf_{\hat{k}=255}$$

## 2. Attack Phase (performed on a device copy set at an unknown secret)

2.1 Collect  $N$  side channel traces w. diff. inputs

$$t_1 \rightarrow Enc(p_1, k), \dots, t_N \rightarrow Enc(p_N, k)$$

2.2 Use a **maximum likelihood test** to discriminate the correct key by comparing the  $N$  traces and the pdfs

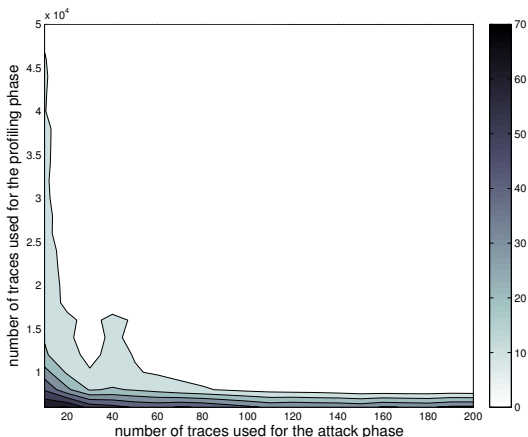


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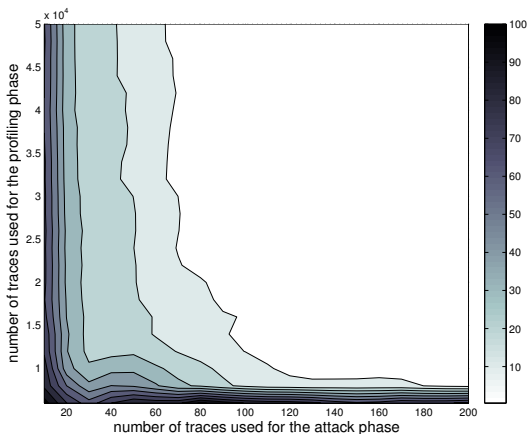
# TA on device A (350nm) - copy 1 → copy 1



**Figure:** Correct key rank evolution vs. nb. of traces for the profiling phase (y-axis) and the attack phase (x-axis)



## TA on device A (350nm) - copy 1 $\rightarrow$ copy 2

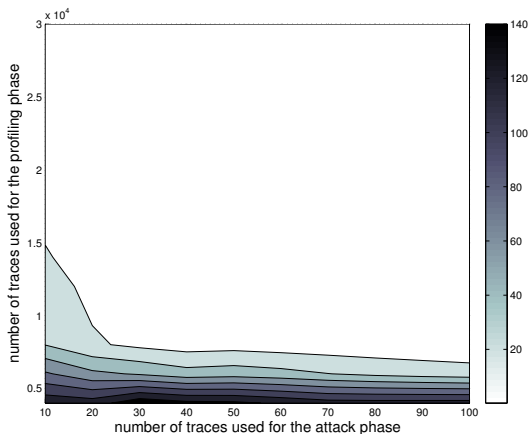


**Figure:** Correct key rank evolution vs. nb. of traces for the profiling phase (y-axis) and the attack phase (x-axis)





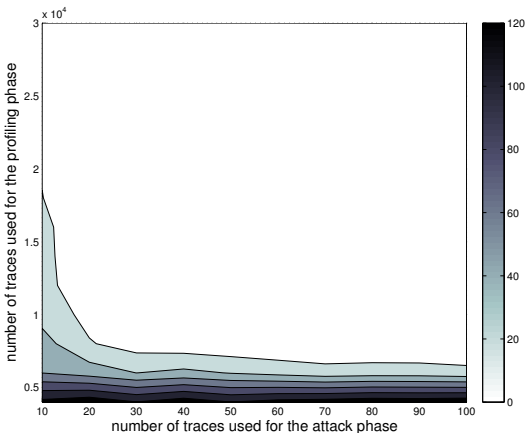
# TA on device B (130nm) - copy 1 $\rightarrow$ copy 1



**Figure:** Correct key rank evolution vs. nb. of traces for the profiling phase (y-axis) and the attack phase (x-axis)



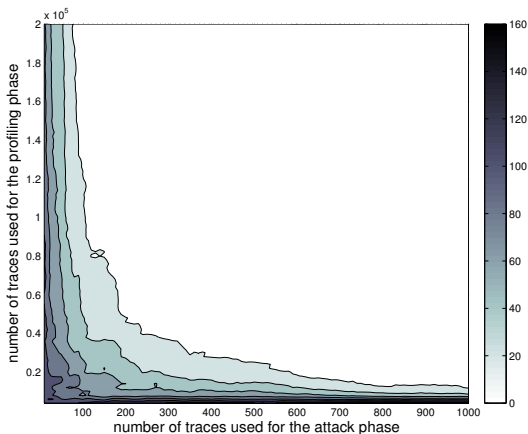
## TA on device B (130nm) - copy 1 $\rightarrow$ copy 2



**Figure:** Correct key rank evolution vs. nb. of traces for the profiling phase (y-axis) and the attack phase (x-axis)



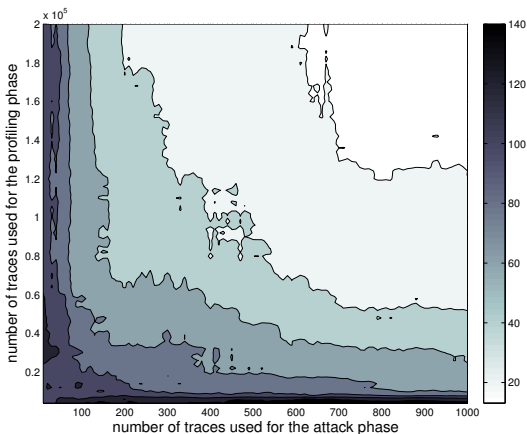
# TA on device C (90nm) - copy 1 → copy 1



**Figure:** Correct key rank evolution vs. nb. of traces for the profiling phase (y-axis) and the attack phase (x-axis)



## TA on device C (90nm) - copy 1 → copy 3



**Figure:** Correct key rank evolution vs. nb. of traces for the profiling phase (y-axis) and the attack phase (x-axis)



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## Our Results

1. Improvement to apply LRA in a practical setup
2. Experiments show that LRA is more effective than classical SCA as CMOS process tends to nanometer scale
3. Experiments show that TA work well:
  - ▶ even if both phases are performed on diff. device copies
  - ▶ TA effectiveness outperforms unprofiled SCA
4. Partition method allowing to implement efficiently all SCA
  - ▶ algo. complexity does not depend from nb. of traces (not described in this presentation)



## Some Numbers

uC	CPA	LRA	TA <sup>2</sup>	TA <sup>3</sup>
350nm	250	1000	80	10
130nm	350	800	100	10
90nm	7500	2000	700	100

**Table:** average nb. of traces to retrieve the correct key

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<sup>2</sup>Template Attack inter-chip

<sup>3</sup>Template Attack intra-chip



Talk Finished !

Thanks for your attention !

Questions ?

full version: <http://eprint.iacr.org/2013/794>

