

Une approche spectrale pour l'analyse de corrélation sur la consommation

BRANDON DRAVIE
Centre de Recherche en Automatique de Nancy (CRAN)

Supervisor : GILLES MILLERIOUX
Co-supervisor : PHILIPPE GUILLOT

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Introduction

Side Channel Attack

When running an electronic device, it emits **signals** due to **power consumption** or **waves** due to **electromagnetic radiations**.

Observe and exploit these emanations can help to extract informations within the device at a precise running time.

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A **cryptographic device** (smart card, FPGA device...) performs operations on **secret data** that are hold inside the circuit of the devices:

- running the cryptographic algorithm \implies physical emanations or leakage due to functions that operate on secret data (ex. non-linear function)
- exploiting physical leakage \implies physical attacks that reveal the secret data

This kind of attack is called **Side Channel Attack** or SCA for short.

Contents

- 1 Correlation Power Analysis (CPA)
- 2 Fundamentals (Fourier Transform of Functions on binary word)
- 3 Modelling the attack
 - Physical principles
 - Performing the attack
 - Experimental Results

1 Correlation Power Analysis (CPA)

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Correlation Power Analysis (CPA)

Side Channel Attack

- Simple Power Analysis (SPA) [Koc96]
- Differential Power Analysis [PJB99]
- **Correlation Power Analysis (CPA)** [BCO04]

Correlation Power Analysis (CPA)

Side Channel Attack

- Simple Power Analysis (SPA) [Koc96]
- Differential Power Analysis [PJB99]
- **Correlation Power Analysis (CPA)** [BCO04]

Test Bench:

- smart card (containing the cryptographic algorithm to attack)
- oscilloscope (to perform measurement when the smart card is working)
- Personal Computer (to drive the smart card and the oscilloscope).

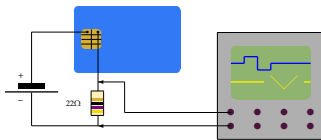


Figure : Test bench.

Correlation Power Analysis

A CPA establishes correlation between :

- **real power consumption values** obtained by measuring the power consumption when running the smart card
- and
- **hypothetical power consumption values** obtained from a **leakage model**.

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Function on binary words

$\varphi : \{0,1\}^n \mapsto \mathbb{R}$: Function on binary words. $\Phi = \{\varphi : \{0,1\}^n \rightarrow \mathbb{R}\}$

Scalar product

$\varphi, \psi \in \Phi$:

$$\langle \varphi, \psi \rangle = \sum_{x \in \{0,1\}^n} \varphi(x)\psi(x).$$

Norm

$$\|\varphi\| = \sqrt{\langle \varphi, \varphi \rangle} = \sqrt{\sum_{x \in \{0,1\}^n} \varphi(x)^2}$$

The norm of $\varphi \in \Phi$ is called energy of φ .

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Fourier Transform

$$\begin{aligned} \widehat{\varphi}(u) : \{0,1\}^n &\mapsto \mathbb{R} \\ u &\mapsto \widehat{\varphi}(u) = \frac{1}{\sqrt{2^n}} \sum_{x \in \{0,1\}^n} \varphi(x)(-1)^{u \cdot x} \end{aligned}$$

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Isometry of Fourier Transform

$$\langle \varphi, \psi \rangle = \langle \widehat{\varphi}, \widehat{\psi} \rangle.$$

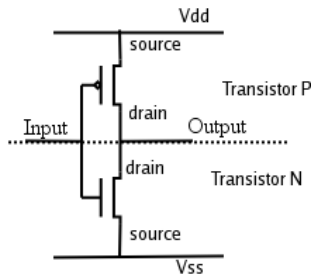
Energy conservation law

$$\|\varphi\| = \|\widehat{\varphi}\|$$

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Physical principles

Circuit CMOS (Complementary Metal-Oxide Semiconductor):



Hypothesis

The power consumption of a CMOS circuit is proportional to the number of logic gates that switch.

Target of the attack

The target of the CPA attack is an **S-Box** function implemented as **Look Up Table (LUT)** in the circuitry of the device.

S-Boxes performed essentially **non-linear operations** of the encryption algorithm.

S-Box

$$f : \{0,1\}^n \mapsto \{0,1\}^m$$

it computes a quantity (within the cryptographic device)

$$y = f(x + k^*)$$

where x is a **known value (by the adversary)** and k^* is an **unknown value (by the adversary)**

\implies aim of the CPA attack on the S-Box: recover the value k^* .

Example $n = m = 4, f : \{0,1\}^4 \mapsto \{0,1\}^4$

0 1 9 14 13 11 7 6 15 2 12 5 10 4 3 8

Leakage model and estimation of the power consumption φ

Leakage model (Hamming weight)

$$\varphi(x) = \text{WH}(f(x + k^*)) + \varepsilon(x) + C$$

$$\varphi(x) = \sum_{i=1}^m f_i(x + k^*) + \varepsilon(x) + C \quad (1)$$

- $\sum_{i=1}^m f_i(x + k^*)$ is the **Hamming weight** of f
- $\varepsilon(x)$ is a random noise that depends on x
- C is a constant related to the power consumption of the smart card that does not depend on the value x .

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The adversary needs to **guess** the value of k^* and then **assume** a value k that is not **equal** to k^* .

For this value k (with $g(x + k) = \sum_{i=1}^m f_i(x + k) :=$ power consumption model):

$$\varepsilon_k(x) = \varphi(x) - g(x + k) - C. \quad (2)$$

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likelihood value of k^*

The value of k^* is the value of k for which the energy of the noise ε_k is minimal:

$$k = k^* \Leftrightarrow \|\varepsilon_k(x)\| \text{ minimal for } k$$

Steps of the CPA attack

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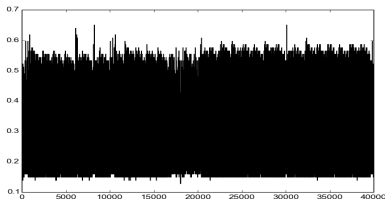
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- 2 measure and record power consumption values denoted by φ :
 - for all value of x is associated a power consumption curve called a **trace**
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 - each trace admits a number NBSAMPLES of samples where each sample corresponds to a power consumption at a given time t of the execution time of the algorithm
- 3 analyse the traces by applying Fourier transform to find the probable value of k^*

Estimation of the likelihood value of k^*

$$\text{Energy conservation law} \implies \|\varepsilon_k\| = \|\hat{\varepsilon}_k\| = \sqrt{\sum_{u \in \{0,1\}^n} \hat{\varepsilon}_k(u)^2} = E(k)$$

$$\text{find } k^* \implies \text{find } \arg \min_k \sum_{u \in \{0,1\}^n} \hat{\varepsilon}_k(u)^2 = \arg \min_k E(k)^2$$

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$$\text{Equation (2)} \implies \widehat{\varepsilon}_k(u) = \widehat{\varphi}_t(u) - (-1)^{u \cdot k} \widehat{g}(u) - C\sqrt{2^n} \delta_0(u). \quad (3)$$

The constant value C can be discarded by considering the functions:

$$\widehat{\varphi}_t^*(u) = \begin{cases} \widehat{\varphi}_t(u) & \text{if } u \neq 0 \\ 0 & \text{else} \end{cases} \quad \text{and} \quad \widehat{g}^*(u) = \begin{cases} \widehat{g}(u) & \text{if } u \neq 0 \\ 0 & \text{else} \end{cases} \quad (4)$$

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$$E(k)^2 = \sum_{u \in \{0,1\}^n} \widehat{\varphi}_t^*(u)^2 + \sum_{u \in \{0,1\}^n} \widehat{g}^*(u)^2 - 2 \sum_{u \in \{0,1\}^n} \widehat{\varphi}_t^*(u) \widehat{g}^*(u) (-1)^{u \cdot k} \quad (5)$$

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$$\implies \text{minimize } E(k)^2 \text{ amounts to maximize } F_t(k) = \sum_{u \in \{0,1\}^n} \widehat{\varphi}_t^*(u) \widehat{g}^*(u) (-1)^{u \cdot k}$$

The theoretical value of k^* is then given by:

$$\arg \max_k F_t(k) \quad (6)$$

Estimation of the reliability of the value k that is found

The reliability of the value k that is found can be estimated by:

$$r_t(k) = \frac{F_t(k)}{\|\widehat{\varphi}_t^*\| \cdot \|\widehat{g}^*\|} = \frac{\langle \widehat{\varphi}_t^*, \widehat{g}_k^* \rangle}{\|\widehat{\varphi}_t^*\| \cdot \|\widehat{g}_k^*\|}$$

This coefficient is the **Pearson correlation** coefficient of φ_t (the real consumption value) and g (the hypothetical consumption value):

$$r_t(k) = \rho(\varphi_t, g) \quad \text{with} \quad \rho(\varphi, \psi) = \frac{\langle \varphi - m_\varphi, \psi - m_\psi \rangle}{\|\varphi - m_\varphi\| \cdot \|\psi - m_\psi\|}$$

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Pearson correlation coefficient

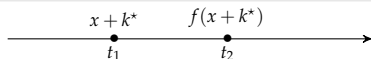
Let two random variables X and Y :

$$\rho(X, Y) = \frac{\text{cov}(X, Y)}{\sigma_X \cdot \sigma_Y} = \frac{\sum_{i=1}^n (X_i - \bar{X}) \cdot (Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2 \cdot \sum_{i=1}^n (Y_i - \bar{Y})^2}}$$

- $\rho(X, Y) \in [-1, 1]$
- $|\rho| = 1 \implies$ **high correlation**
- $\rho = 0 \implies$ **low correlation**

the correlation coefficient measures statistical relationship ("proportionality") between two random variables.

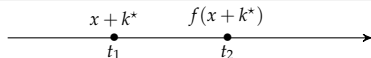
Multidimensional attack



multidimensional attack \implies attack that takes into account **consumption at several instants**

time		computed value	consumption	leakage model	error
t_1	\rightarrow	$f^1(x + k^*)$	φ^1	g^1	E^1
t_2	\rightarrow	$f^2(x + k^*)$	φ^2	g^2	E^2

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The most likely value of k^* :

$$\begin{aligned} & \text{minimize} \quad \|\vec{E}\| = \|(E^1, E^2)\| \\ \iff & \text{maximize} \quad F(k) = \sum_{u \neq 0} (\widehat{\varphi^1}(u) \widehat{g^1}(u) + \widehat{\varphi^2}(u) \widehat{g^2}(u)) (-1)^{u \cdot k} \end{aligned}$$

$$\Phi_2 : \{\vec{\varphi} : \{0, 1\}^n \rightarrow \mathbb{R}^2\}$$

$$\forall \vec{\varphi} = (\varphi^1, \varphi^2) \in \Phi_2$$

$$\forall \vec{\psi} = (\psi^1, \psi^2) \in \Phi_2$$

Euclidean scalar product in Φ_2

$$\langle \vec{\varphi}, \vec{\psi} \rangle = \langle \varphi^1, \psi^1 \rangle + \langle \varphi^2, \psi^2 \rangle$$

Euclidean norm in Φ_2

$$\|\vec{\varphi}\|^2 = \|\varphi^1\|^2 + \|\varphi^2\|^2$$

$$\text{reliability coefficient} \implies r_t(k) = \frac{\langle (\widehat{\varphi_t^1}, \widehat{\varphi_t^2}), (\widehat{g_k^1}, \widehat{g_k^2}) \rangle}{\|(\widehat{\varphi_t^1}, \widehat{\varphi_t^2})\| \cdot \|(\widehat{g_k^1}, \widehat{g_k^2})\|}$$

Experimental Results

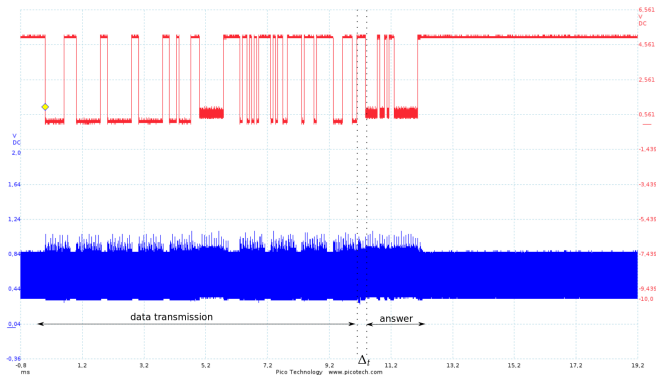
AES S-box (size of the S-box $n = 8$)

compute sequentially 4 times the S-box with different 8-bit keys k_1, k_2, k_3, k_4

record 256 traces of 40.000 samples

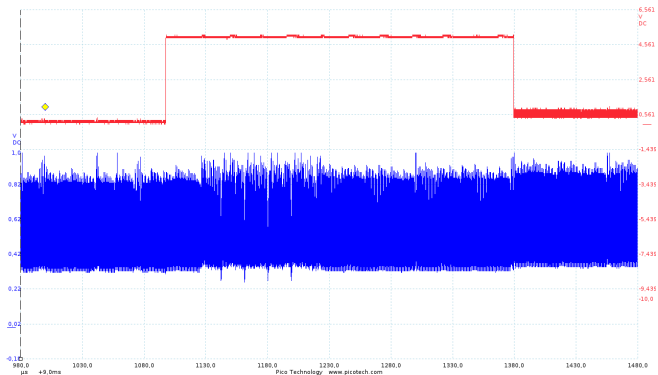
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power consumption and et I/O signal when running the encryption algorithm:



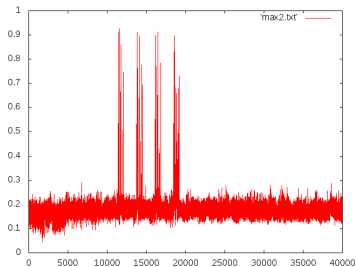
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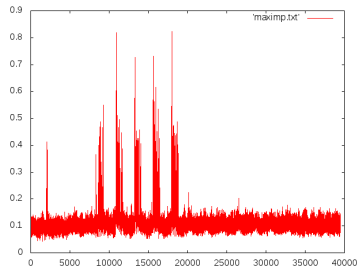


Experimental Results

Correlation between the real power consumption φ_t and the hypothetical power consumption g



One dimensional attack



Multidimensional attack

Conclusion

- ▶ we performed a CPA attack based on a Fourier Transform
- ▶ the approach allows multidimensional attack
- ▶ improvement: reduce the number of traces required to recover the right key

Thank you for your attention



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