

Improved Higher-Order Side-Channel Attacks with FPGA Experiments

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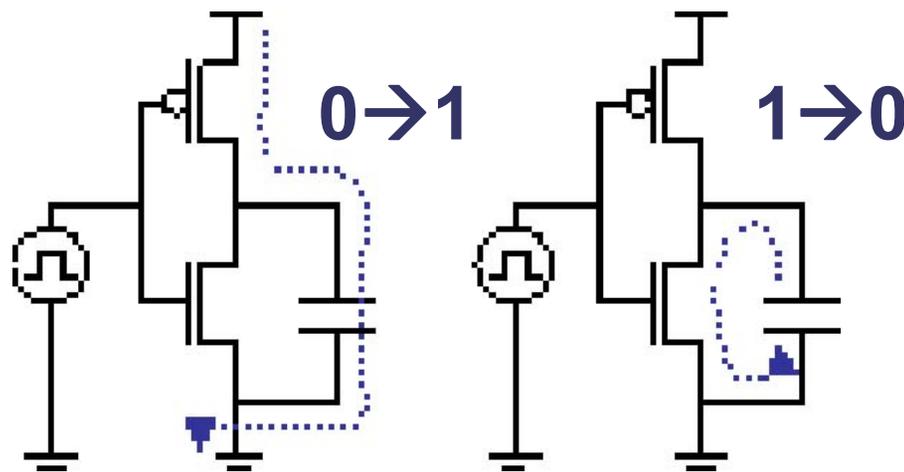
Outline

- Power Analysis: CMOS models
- Block Cipher
- Boolean Masking
- High-Order Power Attack
 - With perfect measurements
 - With real measurements
- Comparison
- Conclusions



Power Consumption Model

- Most IC's → CMOS



$$O \approx H[d \oplus d']$$

perfect
real



Different Contexts

- **Hamming Weight** model

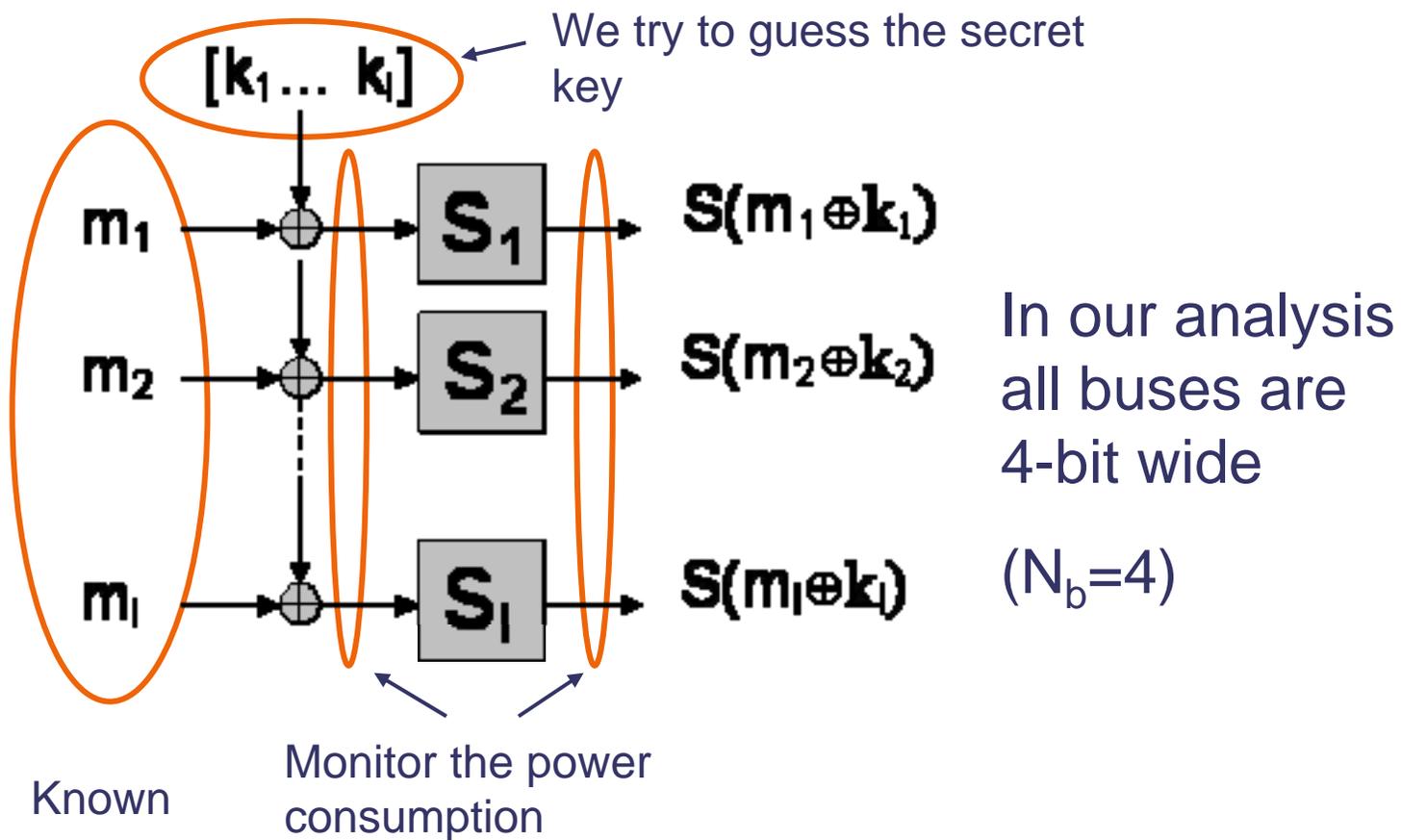
$$O = H[d]$$

- **Hamming Distance** model: depends on d and d'

$$O = H[d \oplus d']$$



First Round of a Block Cipher



Targeting FPGA

- Pipelined Implementation

- Structure fed with a new message every clock cycle

- Hamming Distance model

- Consumption before the S-box

$$O = H[m \oplus k \oplus m' \oplus k]$$

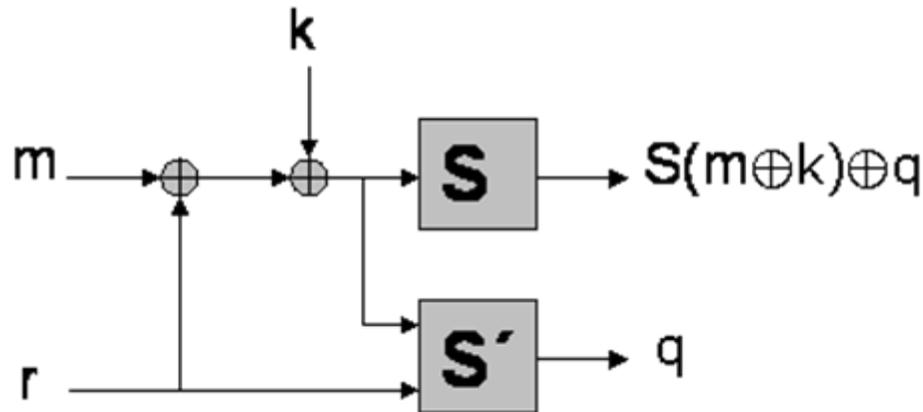
$$O = H[m \oplus m']$$

- Consumption after the S-box

$$O = H[S(m \oplus k) \oplus S(m' \oplus k)]$$



Boolean Masking

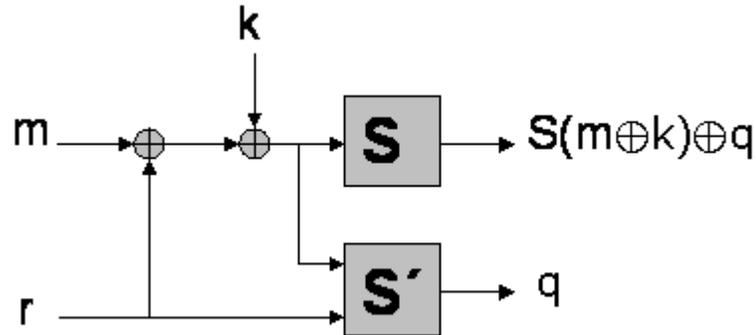


[Goubin et al. CHES '99]

- Focus on 2nd order Boolean masking
- Extendable to higher-order Boolean masking



HO Attack with Perfect Measurements



$$O = H[(S(m \oplus k) \oplus q) \oplus (S(m' \oplus k) \oplus q')] + H[q \oplus q']$$

- Let us define the **machine state**

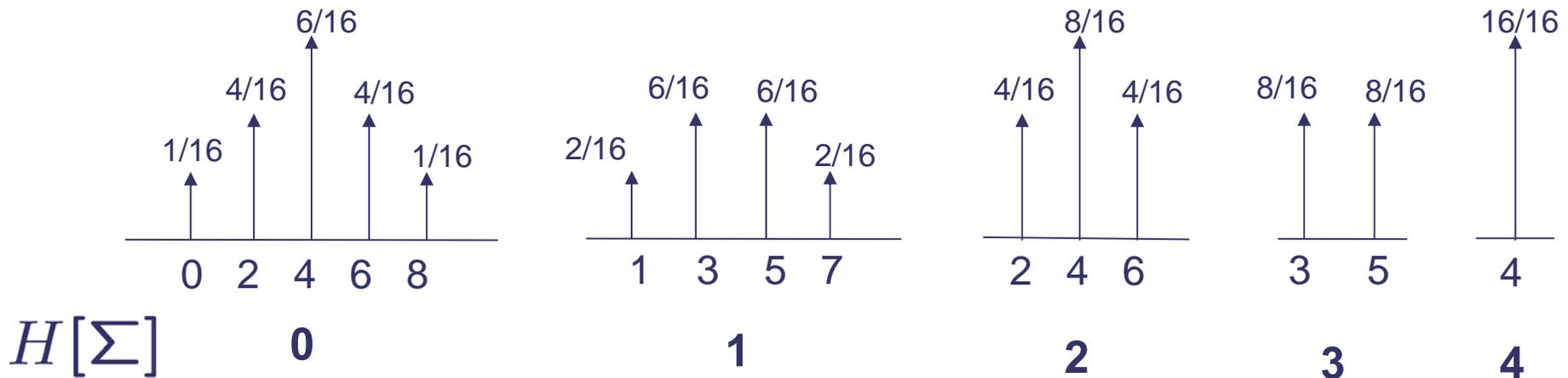
$$\Sigma = S(m \oplus k) \oplus S(m' \oplus k)$$

- And the random state $R = [q \oplus q']$



$$O(\Sigma, R) = H[\Sigma \oplus R] + H[R]$$

- **Machine state** Σ directly related to a Probability Density Functions (PDFs)



PDFs $P[O|\Sigma = \sigma_i]$ with $N_b = 4$



Maximum Likelihood Approach

$$\Sigma^*(k_0) = \{\sigma_1(k_0), \sigma_2(k_0), \dots, \sigma_n(k_0)\};$$

$$\Sigma^*(k_1) = \{\sigma_1(k_1), \sigma_2(k_1), \dots, \sigma_n(k_1)\};$$

$$\Sigma^*(k_2) = \{\sigma_1(k_2), \sigma_2(k_2), \dots, \sigma_n(k_2)\};$$

⋮

$$P[O^*|\Sigma^*(k_0)] = P[O = o_1|\Sigma = \sigma_1(k_0)] \times P[O = o_2|\Sigma = \sigma_2(k_0)] \times \dots$$

$$P[O^*|\Sigma^*(k_1)] = P[O = o_1|\Sigma = \sigma_1(k_1)] \times P[O = o_2|\Sigma = \sigma_2(k_1)] \times \dots$$

$$P[O^*|\Sigma^*(k_2)] = P[O = o_1|\Sigma = \sigma_1(k_2)] \times P[O = o_2|\Sigma = \sigma_2(k_2)] \times \dots$$

⋮

Decision in favor of k

$$k = \underset{\forall k_j}{\operatorname{argmax}} P[O^*|\Sigma^*(k_j)]$$



Example

- Correct key guess: '0001'
- Sequence of observation $O^* = \{5, 3, 6, 5, \dots\}$
- $M^* = \{5, 12, 10, 14, 0, \dots\}$
- $R^* = \{3, 11, 6, 12, 9, \dots\}$
- 4-bit S-box: $S = [9, 2, 15, 8, 3, 14, 5, 11, 13, 4, 10, 6, 0, 7, 1, 12]$



First Guess $k_0 = '0000'$?

$O^* = \{5, 3, 6, 5, \dots\}$

$$H[\Sigma_1(k_0)] = 3 \longrightarrow \begin{array}{c} 8/16 \quad 8/16 \\ \uparrow \quad \uparrow \\ \hline 3 \quad 5 \end{array} \longrightarrow P[O_1 = 5 | \Sigma_1(k_0)] = \frac{8}{16}$$

$$H[\Sigma_2(k_0)] = 2 \longrightarrow \begin{array}{c} 8/16 \\ \uparrow \\ 4/16 \quad 4/16 \\ \uparrow \quad \uparrow \\ \hline 2 \quad 4 \quad 6 \end{array} \longrightarrow P[O_2 = 3 | \Sigma_2(k_0)] = 0$$

$$H[\Sigma_3(k_0)] = 3 \longrightarrow \begin{array}{c} 8/16 \quad 8/16 \\ \uparrow \quad \uparrow \\ \hline 3 \quad 5 \end{array} \longrightarrow P[O_3 = 6 | \Sigma_3(k_0)] = 0$$



Correct Guess $k_1 = '0001'$

$O^* = \{5, 3, 6, 5, \dots\}$

$$H[\Sigma_1(k_1)] = 1 \longrightarrow \begin{array}{c} \begin{array}{cccc} & 6/16 & & 6/16 \\ & \uparrow & & \uparrow \\ 2/16 & & & & 2/16 \\ \uparrow & & & & \uparrow \\ 1 & 3 & 5 & 7 \end{array} \\ \longrightarrow P[O_1 = 5 | \Sigma_1(k_1)] = \frac{6}{16} \end{array}$$

$$H[\Sigma_2(k_1)] = 1 \longrightarrow \begin{array}{c} \begin{array}{cccc} & 6/16 & & 6/16 \\ & \uparrow & & \uparrow \\ 2/16 & & & & 2/16 \\ \uparrow & & & & \uparrow \\ 1 & 3 & 5 & 7 \end{array} \\ \longrightarrow P[O_2 = 3 | \Sigma_2(k_1)] = \frac{6}{16} \end{array}$$

$$H[\Sigma_3(k_1)] = 2 \longrightarrow \begin{array}{c} \begin{array}{ccc} & 8/16 & \\ & \uparrow & \\ 4/16 & & 4/16 \\ \uparrow & & \uparrow \\ 2 & 4 & 6 \end{array} \\ \longrightarrow P[O_3 = 6 | \Sigma_3(k_1)] = \frac{4}{16} \end{array}$$



Compute the Likelihood for Each Key Guess

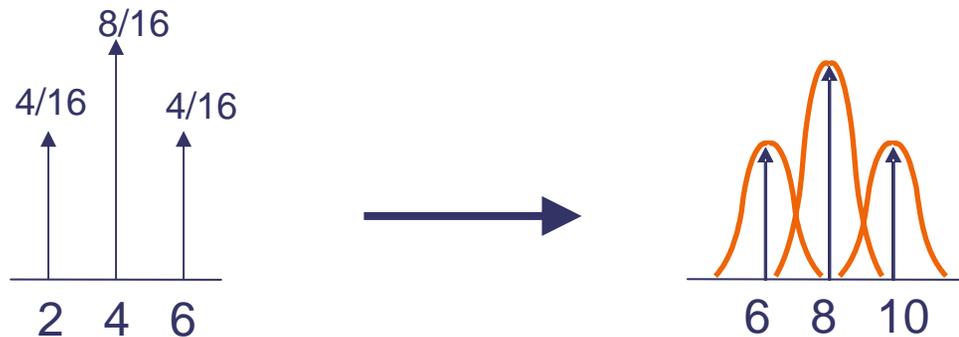
$$\begin{aligned} P[O^* | \Sigma^*(k_0)] &= P[O_1 = 5 | \Sigma_1(k_0)] \times P[O_2 = 3 | \Sigma_2(k_0)] \times \dots \\ &= \frac{8}{16} \times 0 \times \dots = 0 \end{aligned}$$

$$\begin{aligned} P[O^* | \Sigma^*(k_1)] &= P[O_1 = 5 | \Sigma_1(k_1)] \times P[O_2 = 3 | \Sigma_2(k_1)] \times \dots \\ &= \frac{6}{16} \times \frac{6}{16} \times \dots \end{aligned}$$

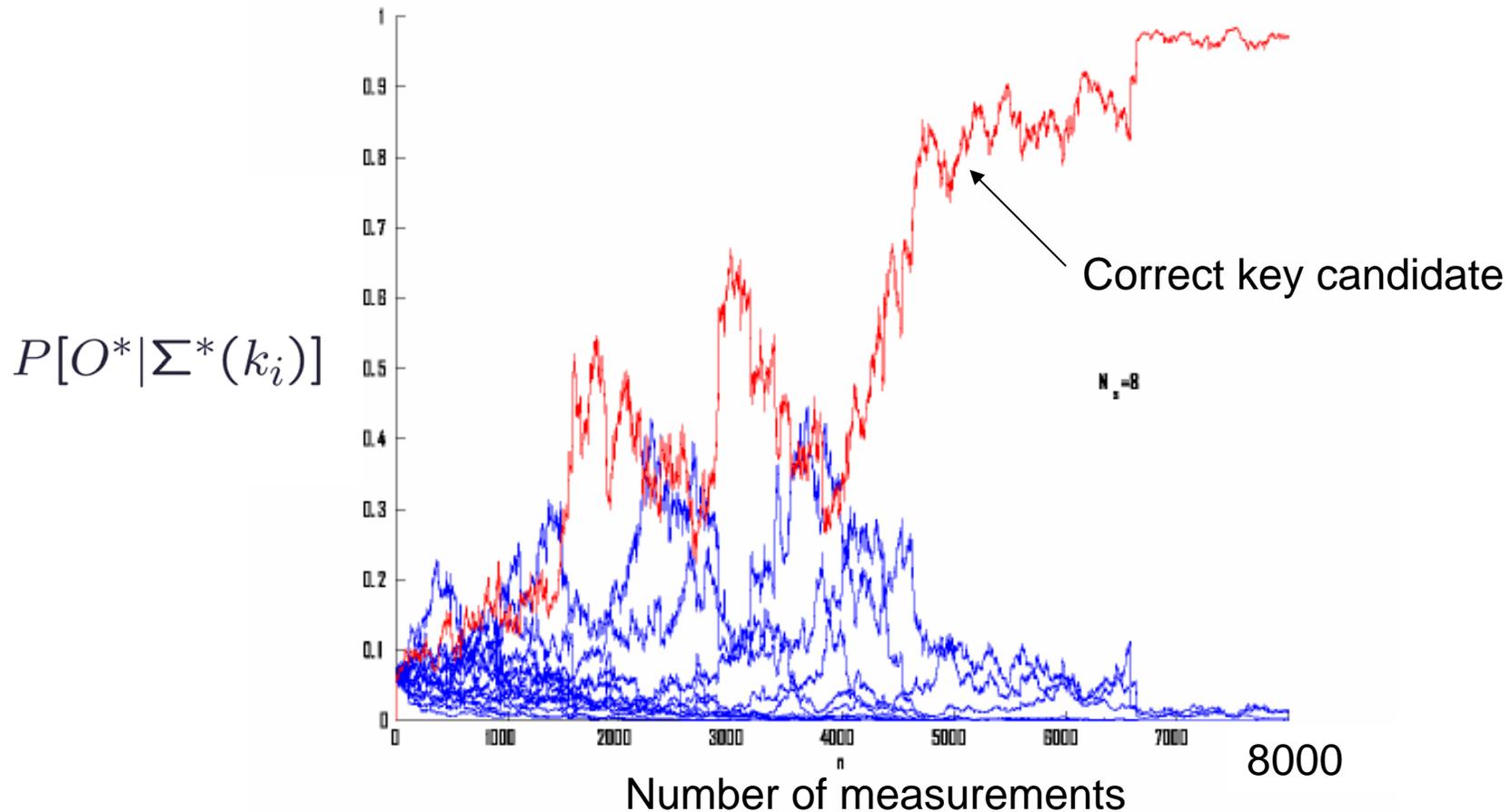


HO Attacks with Algorithmic Noise

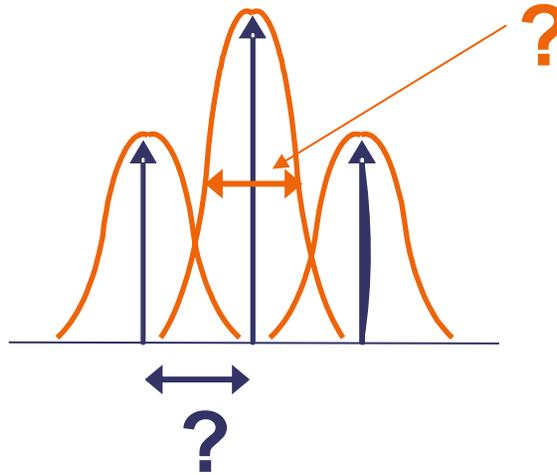
- Due to the bit transitions of non targeted S-boxes
- The noise is assumed to be Gaussian



Simulated Attack With 8 S-boxes in Parallel: $N_S=8$



HO Attacks with Real Measurements

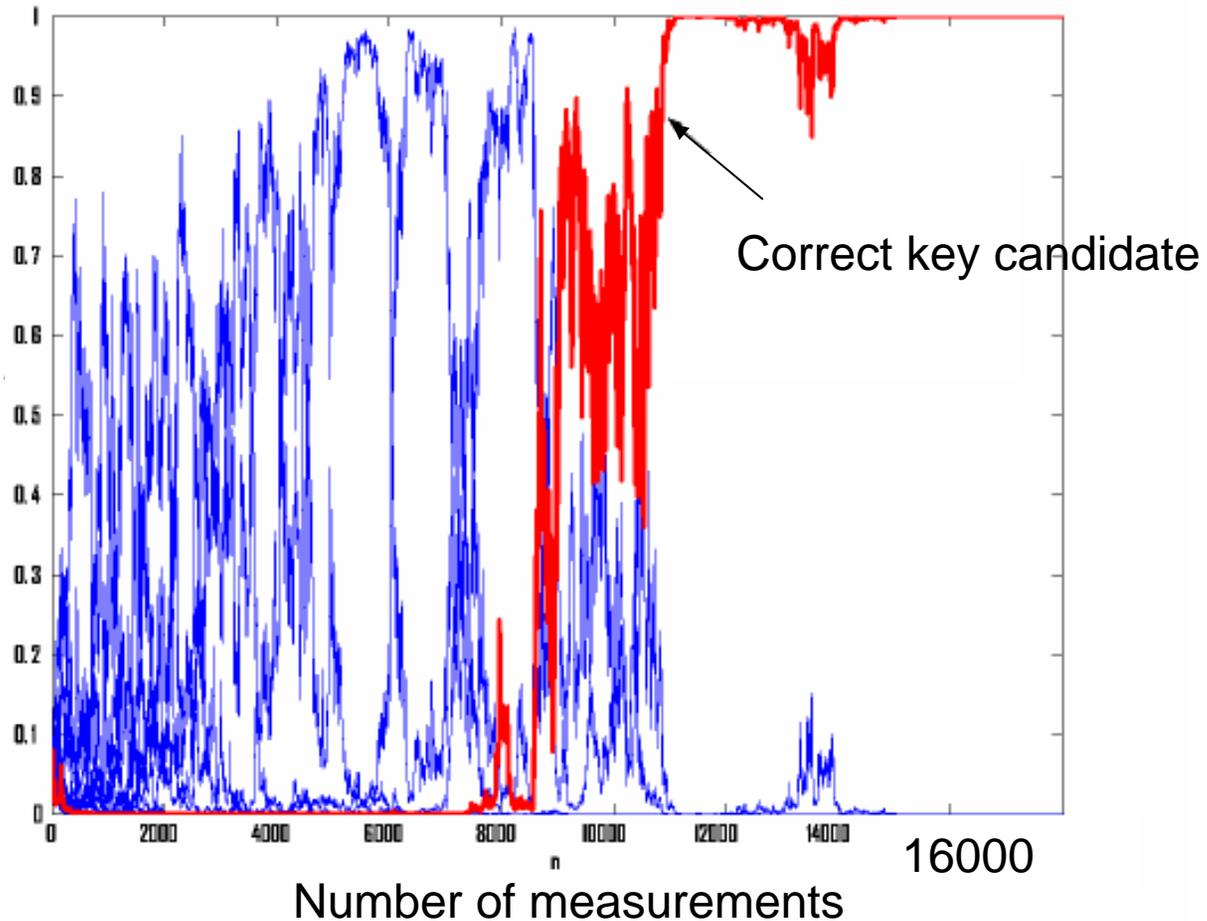


- PDFs? → Require **similar chip** to evaluate the mean and variance of each Gaussians
 - Or **machine learning** methods to extract the mixture of Gaussians' characteristics



FPGA Results

$$P[O^* | \Sigma^*(k_i)]$$



Comparison with Previous Work

- Second Order Power Attack on FPGA

Correlation

>>

Maximum likelihood

~130,000

~12,000

$N_S=1$

$N_S=8$

[Standaert et al. ITCC '05]

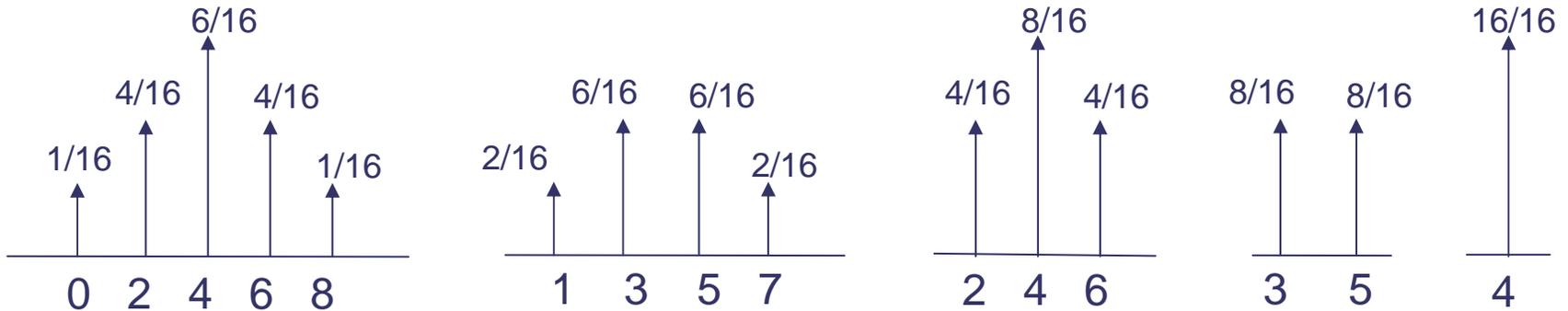
Ours

≈ Multiple bits Waddle's attack

[Waddle et al. CHES '04]



Remark



- $E[x^2]=E[x]^2+V[x]$ [Waddle et al. CHES '04]
- ➔ Squaring the trace yields key dependency since the variances are key dependent



Conclusions

- A key guess defines a succession of **machine states** related to different **PDFs**
 - ➔ We can compute the likelihood of the different key candidates from a sequence of observations
- Maximum likelihood is very efficient in this context
- FPGAs considered to be a challenging devices for SCA: noise
- Applicable to other devices: simulation with a 8-bit microprocessor ➔ ~ 50 measurements



Questions?

