

# Garbled Circuits from an SCA Perspective

## Free XOR can be Quite Expensive. . .

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**Abstract.** Garbling schemes, invented in the 80's by Yao (FOCS'86), have been a versatile and fundamental tool in modern cryptography. A prominent application of garbled circuits is constant round secure two-party computation, which led to a long line of study of this object, where one of the most influential optimizations is Free-XOR (Kolesnikov and Schneider ICALP'08), introducing a global offset  $\Delta$  for all garbled wire values where XOR gates are computed locally without garbling them. To date, garbling schemes were not studied per their side-channel attacks (SCA) security characteristics, even though SCA pose a significant security threat to cryptographic devices. In this research we, demonstrate that adversaries utilizing advanced SCA tools such as horizontal attacks, mixed with advanced hypothesis building and standard (vertical) SCA tools, can jeopardize garbling implementations.

Our main observation is that garbling schemes utilizing a global secret  $\Delta$  open a door to quite trivial side-channel attacks. We model our side-channel attacks on the garbler's device and discuss the asymmetric setting where various computations are not performed on the evaluator side. This enables dangerous leakage extraction on the garbler and renders our attack impossible on the evaluator's side.

Theoretically, we first demonstrate on a simulated environment, that such attacks are quite devastating. Concretely, our attack is capable of extracting  $\Delta$  when the circuit embeds only 8 input non-linear gates with fifth/first-order attack Success-Rates of 0.65/0.7. With as little as 3 such gates, our attack reduces the first-order Guessing Entropy of  $\Delta$  from 128 to  $\sim 48$ -bits. We further demonstrate our attack via an implementation and power measurements data over an STM 32-bit processor software implementing circuit garbling, and discuss their limitations and mitigation tactics on logical, protocol and implementation layers.

**Keywords:** Secure Computation · Garbled Circuits · Free-XOR · Side-channel analysis · Horizontal Attacks · Single Trace

## 1 Introduction

**Background.** Secure multi-party computation (MPC) allows a set of  $n$  parties  $P_i$  to jointly compute a function  $f$  on their inputs such that nothing beyond the output of that function is revealed. Privacy of the inputs and correctness of the outputs need to be guaranteed even if some subset of the parties is corrupted by an adversary. Such an adversary (either passive or active), is assumed to break into the devices of a subset of parties, learning their entire state which includes their secret inputs to the computation and their internal randomness. Efficient and privacy-preserving multi-party computation is an important goal of modern society, organizations and computational platforms as a whole. The vast research effort was spent to provide protocols that are concretely efficient and practical; see [BCS19, HIMV19, YWZ20, BCO<sup>+</sup>21] for recent implementation efforts.

Garbled circuits, invented in the 80's by Yao [Yao86], have been a versatile and fundamental tool in modern cryptography with a large body of work building up on Yao's construction, providing optimizations, abstractions, variations, and indeed several new

applications of garbled circuits. One of the prominent applications of garbled circuits is constant round secure two-party computation protocols (2PC) [LP09]. Such a protocol proceeds by designating one party as a garbler and the other as the evaluator. The garbler generates an encoded version of the computed function (viewed as a Boolean circuit), using its private randomness. This is referred to as a *garbled circuit* (GC). The GC is then sent to the evaluator along with input labels corresponding to the garbler’s input bits. Next, the garbler and evaluator engage in an oblivious transfer (OT) protocol, through which the evaluator receives the input labels corresponding to its own input to the function. Finally, the evaluator uses all the input labels and the GC to compute the function’s output which it can share with the garbler. Garbled circuits have found many additional applications both in theory and in practice such as one-time programs [GKR08], KDM-security [BH10], verifiable computation [GGP10], multi-hop homomorphic computations [GHV10], zero-knowledge argument systems [JKO13] and more.

Secure garbling of circuits and corresponding ways of reducing the garbling size has been the aim of a long line of research, e.g., [BMR90, NPS99, KS08, LP09]. The most common paradigm for garbling a circuit operates at the gate level where for each gate in the circuit, each line in the truth table of the gate functionality is encrypted separately (this is also known as the ‘gate-by-gate paradigm’). The garbling procedure follows by choosing a pair of labels for each wire in the circuit corresponding to the values 0 and 1. Then for each gate and each row of the truth table, two labels, one for each input wire, are used as the joint key for masking the output label. The underlying primitive for realizing this mask is a symmetric-key algorithm (e.g., a pseudorandom function (PRF) or a correlation robust hash function) which yields extremely fast algorithms. This paradigm led to a long sequence of successful optimizations in computation and communication, that established garbled circuits as a practical tool for achieving 2PC; see e.g., [PSSW09, KMR14, ZRE15].

As it stands today, all garbling implementations have found expression in MPC environments, where both the garbler and the evaluator are part of a distributed computation that is typically carried out on powerful devices (such as cloud servers). However, with the technology outbreak, platforms such as advanced Systems-On-a Chip (SoCs), smartphones, Automotive, IoTs and standard desktops are powerful enough to execute efficiently such schemes. Specifically, with the growing popularity of applications that require dedicated hardware or use such platforms, which can benefit from garbling schemes and secure computation, designing and executing special purpose GC-hardware is inevitable [JKSS10, HJP13]. For some examples we can consider one-time programs and obfuscation [GKR08, JKSS10, KSNO17, ZCD<sup>+</sup>19], post-quantum signature schemes [JKO13, AHIV17, CDG<sup>+</sup>17, KKW18] and verifiable computation [GGP10, LMR22]. Nevertheless, such hardware devices are susceptible to side-channel attacks (SCA).

**Side Channel Attacks.** Indeed, it is always possible to protect against leakage attacks at the primitive level (such as permutation, encryption, hashing) with the assumption that if we protect these lower levels primitives, higher abstractions such as GCs, which are instantiated with such secure primitives, will be protected as well. Nevertheless, protection against side-channels attacks induces significant costs with respect to latency, energy etc. For instance, the masking countermeasure may introduce an overhead between  $2x - 10x$  [GMK17, GM18, CGLS20] (as a function of the security level required). Therefore adding such protections should be well justified. The folklore, which is further supported by our paper, is that the evaluator’s algorithm is relatively protected against SCA [GKR08]. The garbler’s algorithm, on the other hand, has not been explored to date as sensitive to such attacks. In this paper, we push forward on these aspects and showcases in which special care should be given to SCA and additional countermeasures should be considered.

To be concrete, we consider a setting where garbling schemes are embedded within a small chip that realizes the garbler’s role. With our attacker model (as discussed in detail below) we assume it got hold of side-channel access to the garbler’s device. We further

assume that the adversary can eavesdrop on public communication channels, such as ones used for sending the circuit description, if not agreed upon, and public garbled values.

## 1.1 Free-XOR and Our Contribution

In more detail, in the setting of garbled circuits, the garbler generates the garbled circuit from the Boolean representation of a  $\lambda$ -gates circuit  $C$ , typically represented by AND and XOR gates. This process follows where each wire  $w_i$  for  $i < \lambda + m$  (where  $m$  is the input length), which can take up a Boolean value in  $\{0, 1\}$ , is labeled by two randomly generated  $n$ -bit values corresponding to the logical ‘0’ and ‘1’ values (where  $n$  typically serves as the security parameter). These labels are denoted by  $W_i^j$  for  $j \in \{0, 1\}$ . Assuming that all gates have fan-in 2, then their truth table includes 4 entries while enabling the recovery of one of the two possible garbled output labels without learning its semantic Boolean value. Naively, each gate is associated with 4 entries (or ciphertexts) [LP07] where each entry requires a symmetric-key operation denoted by  $\mathcal{E}_{W_i^j, W_k^{j'}}(W_l^{j''})$  where  $j''$  corresponds to the gate operation applied to  $j$  and  $j'$ . Followup optimizations further reduce the number of sent ciphertexts; see Section 2.1.

One of the most influential optimizations introduced for garbling schemes is the Free-XOR technique proposed by Kolesnikov and Schneider [KS08], which reduces the computation and communication resources. Namely, it allows XOR gates to be locally evaluated by the Boolean XOR of their input labels/keys *with no communication*. Consequentially, a considerable effort is put in order to decompose the circuit computation efficiently to a topology with a minimal number of non-linear operations (in  $\text{GF}(2)$ ), on the account of (potentially significantly) more linear ones [BCCM20, TSR<sup>+</sup>20, SZ13]<sup>1</sup>. As it stands today, this optimization is embedded within all garbled circuits implementations. In their construction, Kolesnikov and Schneider proposed to choose only one label/key associated with each wire independently at random. Namely, if wire numbers  $i$  and  $j$  are the input wires to an XOR gate and  $k$  is the output wire number, then the keys for the input wires are set by the pairs  $\{W_{i/j}^0, W_{i/j}^0 \oplus \Delta\}$  where  $\Delta$  is a global offset used across all wire labels in the circuit. Following this approach, the output label is the XOR result of the two input labels implying that it will be set either to  $\{W_i^0 \oplus W_j^0\}$  or  $\{W_i^0 \oplus W_j^0 \oplus \Delta\}$ . Thus, garbled XOR gates do not require cryptographic building blocks such as PRFs and are communication free. For non-linear operations, e.g., ANDs as illustrated in Figure 2b, it is understood that  $\Delta$  appears both within the input keys and plaintexts (which correspond to the output keys), where each input key is used to compute the two ciphertexts that correspond to the logical value of that key; see more details in Section 2.1. As a result, the security of Free-XOR relies on a non-standard assumption of circular security [CKKZ12].

*The Free-XOR optimization is our starting point, introducing the correlation we take advantage of in our side-channel attacks.* More concretely, in this work we show that SCA attacks on the garbler’s device are viable and devastating.

Our technical contribution relates to building an attack that is based on the observation that garbling schemes utilizing a global secret  $\Delta$  generate side-channel attacks sensitivity. Specifically, our attacks merge information from multiple time samples within a leakage trace, but more importantly from multiple **non-linear input-layer garbled gates** processed in the garbler associated device, utilizing publicly known labels and other publicly available information provided by the protocol to extract  $\Delta$ . We demonstrate that such attacks are quite devastating; technically, capable of extracting  $\Delta$  when the circuit embeds only 8 input non-linear gates with fifth/first-order attack Success-Rates of 0.65/0.7. With as little as 3 such gates, our attack reduces the first-order Guessing Entropy of  $\Delta$  from 128 to  $\sim 48$ -bits. We further demonstrate our attack via an implementation and

<sup>1</sup>Interestingly, it is similar to the masked SCA context [BDMD<sup>+</sup>20] where ANDs are much more expensive than XORs with high masking orders.

measurements data over an STM 32-bit processor software implementing circuit garbling. We discuss the limitations of our attacks in terms of miss-classification of leakages, needed in our attack procedure, and various other parameters of the attack, such as the enumeration complexity of the adversary. Lastly, we list a spectrum of possible countermeasures.

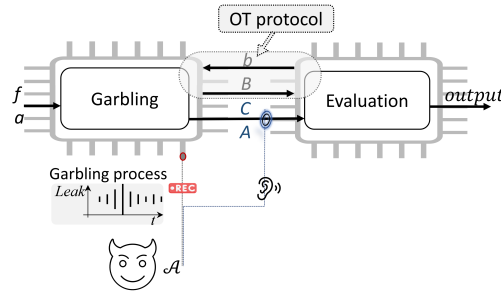
The main differences between our attack and generic SCA attacks on PRFs (instantiated as AES) are due to the underlying cryptographic primitive used for the garbling, the circuit topology and further optimizations such as FreeXOR. In particular, the way the key/input material are processed by the primitive is different from standard SCA (e.g.,  $\text{AES}_k(p)$ ) where the same key is repeatedly used. (1) Whereas our attack requires additional hypotheses as all leakage samples associated with the four truth-table entries appear per gate and the adversary is uncertain regarding the matching between the leakage and the correct table entry. (2) Moreover, only a few garbled gates are exploited by our attacks. Therefore, only a few leakage traces are provided to the adversary per key, where our tactic to remedy this and extract more information from the leakage is by utilizing horizontal attacks, and finally (3) we show a specific attack owing to the ultra-popular Free-XORs optimization while extracting the globally shared  $\Delta$ .

## 2 Technical Background

**Notations:** In this manuscript, variables are denoted with capital letters, sampled values with lowercase letters, functions with sans serif fonts and vectors with bold letters. We use standard notations for mean ( $\mu$ ), standard deviation ( $\sigma$ ), and denote by  $f_{\mu,\sigma}(x)$  the probability density of a normal distribution  $\mathcal{N}(\mu, \sigma^2)$  with the random variable  $X$  with realization  $x$ . We further denote by  $\Pr(x \leq \alpha)$  the (cumulative) probability and by  $\Pr(x \leq \alpha | \beta)$  the conditional (cumulative) probability with standard notations for parameters ( $\alpha$  and  $\beta$ ). In our notations  $\parallel$  denotes concatenation, and  $\oplus$  and  $\wedge$  represent addition and multiplication in  $\text{GF}(2)$  i.e., Boolean XOR and AND gates. Similarly,  $\cdot$  and  $+$  represent addition and multiplication over the reals,  $\mathbb{R}$ .

**Adversary and Threat Model:** The adversary  $\mathcal{A}$  is modeled by a Probabilistic Polynomial Time (PPT) algorithm. In practice, in the proposed SCA attack, the computation complexity required from  $\mathcal{A}$  is tiny (typically around  $2^{11}$  and upper bounded by  $2^{16}$  to extract  $\Delta$  and consequently break the privacy of the protocol). The adversary has eavesdropper access to the communication channel as abstractly illustrated in Figure 1, and is completely passive, and does not affect the garbler’s computation nor communicated data. In addition, all public communication within the protocol is known to  $\mathcal{A}$  such as the function description  $f$  and the topology of the circuit. We emulate the trusted garbler device in an untrusted environment, namely, one which grants an adversary SCA traces access. In our attack, we assume that  $\mathcal{A}$  thus has physical access to the garbler’s device as well as the ability to perform measurements which means that he may need knowledge on the whereabouts of some (e.g.,) power-supply pin. The letter is not regarded as knowledge strictly required from  $\mathcal{A}$ , as such knowledge is quite trivial to extract. Alternatively,  $\mathcal{A}$  can perform any other SCA measurement apparatus such as passive electro-magnetic (EM) SCA which does not require that information. More generally, either the adversary has these abilities or it can get access to such abilities/information.  $\mathcal{A}$  does not utilize any assumptions about the internals of a device, nor does the attack require any further internal implementation or architectural knowledge. In this sense, our attack requires nothing more than a conventional SCA attack as illustrated in Figure 1. The goal we set for the adversary in our attack is to extract  $\Delta$  from the SCA leakage samples by building leakage hypotheses utilizing publicly known or communicated information in the protocol.

Finally, in our implemented attacks, we pick randomly-chosen input labels and randomly generated  $\Delta$ s’ and model (or measure) leakages from the computations of gates truth-tables. The experiments are repeated hundreds of times with fresh randomness to evaluate the



**Figure 1:** Attack and Adversarial ( $\mathcal{A}$ ) Model where  $a$  and  $b$  denote the respective inputs of the garbler and the evaluator.  $A$  and  $B$  denote their garbled version.  $b$  and  $B$  are privately transferred through the OT channel and only abstractly denoted on communication lines.

attack Success-Rate and Guessing-Entropy. The detailed apparatus of our attack, which describes how we build SCA hypotheses, is provided in Algorithm 1. The concrete list of hypothesized/modeled intermediate values in the garbling-gates instantiation, used by our attack, is listed in Section 5, e.g., consider **listing 1**.

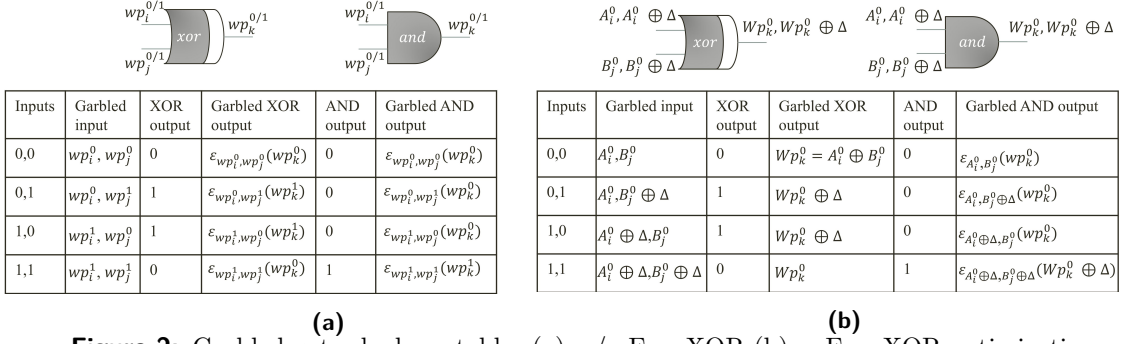
## 2.1 Garbling Schemes

We recall the abstraction of a garbling scheme from [BHR12]. That is, a garbling scheme is a tuple of algorithms  $GS = (\text{Gb}, \text{En}, \text{Ev})$  where the probabilistic garbling algorithm  $\text{Gb}$  takes the function description  $f$  and outputs an encoded representation  $F$  and an input encoding function  $e$ . The deterministic input encoding algorithm  $\text{En}$  gets  $e$  and the function input  $x$  and returns an encoded input representation  $X$ . Finally, the deterministic evaluation algorithm  $\text{Ev}$  takes  $F$  and  $X$  and outputs  $f(x)$  by evaluating the encoding. The classic Yao’s garbled circuits construction follows by choosing two secret keys (also denoted as labels) for each circuit wire, each is associated with a single bit 0/1. The garbling is a process for which the Boolean truth table of each gate is “encrypted”. In practice, this process is instantiated with symmetric-key cryptography e.g., pseudorandom functions (PRFs), double key PRFs or correlation robust hash functions. The input to the circuit  $x$  is encoded similarly. Finally, the evaluation process maintains the invariant that a single key per wire is obtained during the evaluation. This abstraction captures [LP09] and all its subsequent optimizations and is essential for proving security which is captured by the two properties of correctness and privacy. Where the former ensures that  $f(x)$  is computed by the garbled circuit and the latter ensures that nothing beyond that is revealed.

To avoid revealing any information about the internal wire values, the table rows are permuted in a way that enables output labels recovery without capturing their semantic values corresponding with ‘0’/‘1’. The most practiced approach is the point-and-permute (PaP) technique [BMR90]. To achieve this, an additional random bit is associated with each wire label, i.e., each wire label is composed out of  $n + 1$ -bits, denoted here by:  $Wp_i^j = \{W_i^j \| p_i^j\}$ ,  $j \in \{0, 1\}$ , where  $p_i^j \in \{0, 1\}$  is a random permutation bit as illustrated in Figure 2a). Applying the PaP optimization, the evaluator only needs to access just one table entry per gate. The garbled-row-reduction (GRR) [NPS99] is another optimization that enables reduction of the number of communicated ciphertexts per gate from 4 to 3.

Introducing the Free-XOR optimization, Kolesnikov and Schneider proved its security in the random oracle model [KS08] but also claimed that some form of correlation robustness would suffice. Following their work, Choi et al. [CKKZ12] have explored the security of garbled circuits with Free-XOR based on a stronger notion of circular correlation robustness and demonstrated that correlation robustness is insufficient as Free-XOR gives rise to related-key attacks which are not captured by correlation robustness.

With the aim of proving privacy under standard assumptions, Kolesnikov et al. proposed



**Figure 2:** Garbled gates look-up table: (a) w/o Free XOR (b) w Free XOR optimization.

FleXOR gates [KMR14] (or flexible XOR) in which inputs can be adjusted to target specific  $\Delta$  values. The FleXOR optimization is also compatible with the advanced row-reduction technique by Pinkas et al. [PSSW09] (denoted here by GRR2) which reduces the number of communicated ciphertexts per AND gate into two. Nevertheless, the GRR2 technique is not-compliant with the Free-XOR technique. Following that, Half-Gates were introduced by Zahur et al. [ZRE15], proposing means to evaluate AND gates with two ciphertexts (like GRR2) while being compliant with the Free-XOR optimization.

In what follows, we specifically discuss side-channel leakage in garbling schemes that share a global secret (i.e.,  $\Delta$ ). *In that spirit, the alarming aspect is precisely that the most prominent GC optimization is what, in essence, makes SCA sensitive.*

## 2.2 Side Channel Attacks (SCA)

Side-channel attacks have repeatedly demonstrated the sensitivity of implemented cryptographic schemes. As such, current National Institute of Standardization and Technology (NIST) competitions for future symmetric-key, e.g., Authenticated-Encryption [TMC<sup>+</sup>21] and public-key Post-Quantum schemes [AASA<sup>+</sup>20], are considering SCA security and its associated electronic cost factors as very important aspects. Successful single leakage-trace SCA attacks were shown possible for public-key encryption/ digital signatures schemes, both on hardware and software implementations [PPM17, PKH<sup>+</sup>21, KAA21]. In this section, we briefly recall some of the necessary SCA basics and the details needed for this manuscript on SCA security evaluation metrics and horizontal/single-trace attacks.

Side channel attacks enable the extraction of secret values manipulated by the hardware by exploiting secret-key dependent computations variations in some physically measurable quantity. Most reports focus on side-channel leakage measured through power or electromagnetic radiation channels, owing to their ease of access and rather a high signal-to-noise ratio. These leakage samples originate mainly as physical outcomes of dynamic (switching) current dissipation of microelectronic devices. Correlation or Differential power analysis (C/DPA) [KJJ99, BCO04] are powerful side-channel attacks that follow a divide-and-conquer approach: an estimation on distinct parts of the key (denoted by sub-keys) takes place, called hypotheses, and these hypotheses are checked for correlations with the measured leakage from the device through multiple tests. Typically, various pre-processing techniques and tests are required to average out the noise and reveal underlying correlations between the hypothesized sub-key dependent internal secret value and the leakage.

In simulated environments, we typically estimate the leakage model for some intermediate values manipulated by the device. A leakage model is aimed at representing to some extent the actual physical behaviour measured in the SCA paradigm. The Hamming Weight (HW) leakage model is commonly utilized as a practiced leakage model, which typically represents software implementations leakage, by which an intermediate variable  $y$  of  $n$ -bits leaks:  $\alpha \cdot \text{HW}(y) + \beta + \mathcal{N}(\mu, \sigma)$ . Where, the  $\alpha$  factor may represent some scaling

owing to (e.g.,) current dividers on the main power supply path originating from Maxwell and Kirchhoff's Current Law (KCL). The  $\beta$  factor may represent some DC offset owing to some noisy element in the electronic system or current consuming elements which are data-independent. These factors highly vary between different implementations.  $\mathcal{N}(\mu, \sigma)$  is the modeled noise distribution.

The SCA adversary is modeled as one which can thus get access or perform such measurements. Clearly, it does not necessarily imply physical proximity to the device, as measurements can be broadcast through networks, extracted from near or far-field electromagnetic emanation channels, and can be completely passive [GS15, SGMT18, CPM<sup>+</sup>18, CFS20]. Over the last few decades, the most powerful SCA attacks were statistical, meaning many such (plaintext-encryption/ ciphertext-decryption) pairs were retrieved by the adversary prior to the attack procedure. Then, the adversary follows a modeling phase of a key-dependent internal computation in the algorithm, and further models the effect this value has on the (e.g.,) encryption-leakage. Then, a statistical distinguisher is used, with the goal of eliminating the large measurements noise to extract the hypothetical key from the leakage samples (relying on the laws of large numbers). We next recall necessary background on model- and profiled-based attacks used in this paper.

### 2.2.1 Model-Based Attacks

Let's assume a device that performs encryption and any randomness used by the protocol is public (e.g., AES-CTR mode). I.e., a classical SCA attack scenario. With conventional CPA [BCO04], we assume multiple measurements are available to the adversary under the same secret-key. Then an attack works as follows.

Let  $l_{x,k}$  be a leakage trace measurement under a plaintext/key  $x = \{x_0 || x_1 || \dots || x_{15}\} / k = \{k_0 || k_1 || \dots || k_{15}\}$  of  $n$ -bits where for example each  $x_i$  or  $k_i$  represents one byte of say  $n = 128$  bits. To perform a CPA attack, one should choose a target computation to hypothesize (such as the first AES round Sbox output). That is, some logical manipulation of the known plaintext (byte) by a deterministic function and the secret-key (byte) (I.e.,  $y_i = \text{Sbox}(x_i \oplus k_i)$ ). Next, the adversary builds a hypothesis table per each hypothesized secret-key  $k_i \in \{0, 1\}^8$  and all  $x_i$  related possible measurements of size  $N_{tr}$  as abstractly illustrated in Fig. 3a. The leakage model can, e.g., follow the HW of the intermediate hypothesized  $y_i$  value. Next, a distinguisher can be used to find the hypothesized model which depends on the secret-key and most correlates with the measured leakage, e.g., a simple and commonly used distinguisher is Pearson Correlation Coefficient:

$$\rho_{k^h} = \rho(\mathbf{y}_i^{k^h}, \mathbf{l}_{x,k}) = \frac{\text{cov}(\mathbf{y}_i^{k^h}, \mathbf{l}_{x,k})}{\sigma(\mathbf{y}_i^{k^h}) \cdot \sigma(\mathbf{l}_{x,k})} \quad (1)$$

where cov and  $\sigma$  are the covariance and standard deviation, respectively, and  $\mathbf{y}_i^{k^h}$  is  $\mathbf{y}_i$  computed under an hypothetical key  $k^h$ , i.e.,  $\mathbf{y}_i^{k^h} = \text{Sbox}(x_i \oplus k^h)$ .

Eventually, the secret-key (byte)  $k^*$  that maximizes the correlation is chosen, namely  $k^* = \text{argmax}_{k^h} \rho_{k^h}$ . In practice, each leakage trace is a vector over time  $l_{x_i, k_i}^t$  where  $t \in \{0, \dots, \#\text{Samples}\}$  and the correlation in Eq. 1 takes the maximum value found over time per hypothesized key, estimated in a point-of-interest (POI) in time where the secret value  $y_i$  is being manipulated by the device [SMY09, LPB<sup>+</sup>15].

### 2.2.2 Template Attacks

An alternative approach for trying to model the leakage function of a device by apriori or a simple mathematical model, is template attacks [CRR02]. These are performed in two consequent (or interleaved) phases of *profiling* and *attack*. It is assumed that the adversary got hold of one device for which it can program (or control) the secret-key and therefore profile the leakage and another target device from which it tries to extract additional information on the underlying key.

Similarly to model-based attacks, to perform a template attack, one chooses a target intermediate variable to template the probability density function (PDF),  $f(x_i, k_i)$ , as-

sociated with a known/chosen plaintext chunk  $x_i$  and the secret-key subset  $k_i$  at some POI. A set of  $\mathcal{L}_p$  profiling traces of size  $N_p$  is first used in order to estimate the leakage distribution parameters for each intermediate value of  $y_i$ , denoted as  $\hat{M}_{y_i}$ . For the attack phase, a fresh set of new traces from possibly a different device  $\mathcal{L}_{att}$  of size  $N_{att}$  are used. In most cases in the literature (and in practice) the PDF is assumed to follow a Gaussian distribution,  $f(l_{y_i} | y_i) = \mathcal{N}(\hat{\mu}_{l_{y_i}}, \hat{\sigma}_{l_{y_i}})$ , i.e., in this case, the statistical estimated model  $\hat{M}_{y_i}$  contains a tuple  $\{\hat{\mu}_{l_{y_i}}, \hat{\sigma}_{l_{y_i}}\}$ . In practice, distributions can follow more complex structures, especially if countermeasures are embedded within the device. However, the simplicity of such a model, manifested by only two parameters per internal variable value, induces low model storage requirements, low computational effort and thus efficient, making it a very popular *de-facto* tool. Finally, the secret-key (byte/chunk)  $k^*$  is estimated to be the one which maximizes the uni-variate Maximum Likelihood (denoted by ML):  $k_i^* = \underset{k_i}{\operatorname{argmax}} \operatorname{ML}(k_i) = \underset{k_i}{\operatorname{argmax}} \prod_{i=1}^{N_{att}} (f(l_i | y_i))$ . Due to practical computational reasons and numerical errors, the log-likelihood (LLH) is typically used [FDLZ14]  $k_i^* = \underset{k_i}{\operatorname{argmax}} \operatorname{LLH}(k_i) = \underset{k_i}{\operatorname{argmax}} \sum_{i=1}^{N_{att}} \log(f(l_i | y_i))$ .

It is noteworthy that many other (distinguisher,model)-pairs can be used instead of ML and the a-priori Gaussian modeled probabilities (E.g., [GBTP08, PR10, MS16]).

### 2.2.3 Success Rate and Guessing Entropy

An evaluator typically uses a set of attack traces of size  $N_{tr}$  or  $N_{att}$  in the modeled or profiled settings, respectively, and computes a score vector per key based on the number of actual traces utilized out of the entire set. Based on the scores of the correct key and the number of repetitions of the experiment, a Success-Rate and Guessing-Entropy can be computed. I.e., following an attack, the adversary obtains  $N_{sk}$  lists of sub-key rankings, where  $N_{sk}$  is the number of the sub-keys following the divide-and-conquer approach; 16 subkeys in the AES example above with an 8-bit length. In the case of profiled template attacks, the adversary directly captures rankings in terms of probabilities (likelihoods). In the case of model-based correlation attacks, correlation values can be easily converted to probabilities, i.e., utilizing Base formulae or conversion formulae, for some examples see [SLP05, SMY09]. That is, each of the  $N_{sk}$  lists of size  $|\mathcal{K}|$  (e.g.,  $2^8$ ) contains a value,  $\Pr(k_i | \mathbf{x}_i, \mathbf{l}_{x,k})$ . The rank  $R$  of the correct key is defined as its position in the ordered list of probabilities. The Success-Rate of order  $Ord$ , denoted by  $\operatorname{SR}^{Ord}$ , being computed over  $N_{tr}$  leakage measurements per single experiment, is defined as [SMY09]:

$$\operatorname{SR}^{Ord}(N_{tr}) = \Pr(R < Ord). \quad (2)$$

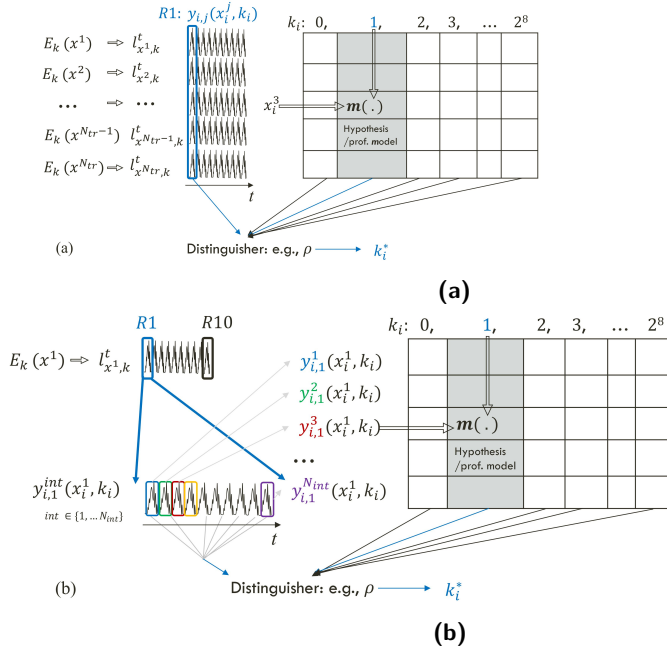
Simply put,  $\operatorname{SR}^{Ord}$  is the probability that the correct key is ranked among the  $Ord$  first most probable keys. The Guessing Entropy (GE) can also be defined as the expected rank of the correct key directly linking the average enumeration effort of the adversary [SMY09].

## 2.3 Horizontal and Combined SCA Attacks

Horizontal attacks make use of hypotheses on a variety of intermediate values within a single execution, i.e., multiple hypotheses of various intermediate variables which relate to the same secret value being hypothesized. Such attacks combine the collective informativeness/ correlations with multiple leakage samples to extract secrets. Fig. 3a illustrates the classic CPA procedure where leakages from the same time-sample are jointly grouped (vertically) to a vector which is then correlated with the leakage-model vector, denoted on the figure by  $m(\cdot)$ . This context is regarded here as *vertical* SCA (V-SCA). In this scenario, the adversary collects multiple measurements, each with possibly a different input indexed with a superscript  $x^i$ , and performs a single test on the power trace to extract sub-key correlations and filter out the noise as abstractly illustrated.

In contrast, in *horizontal* SCA attacks, H-SCA [BJPW13, BCPZ16, PPM17, PZS17,





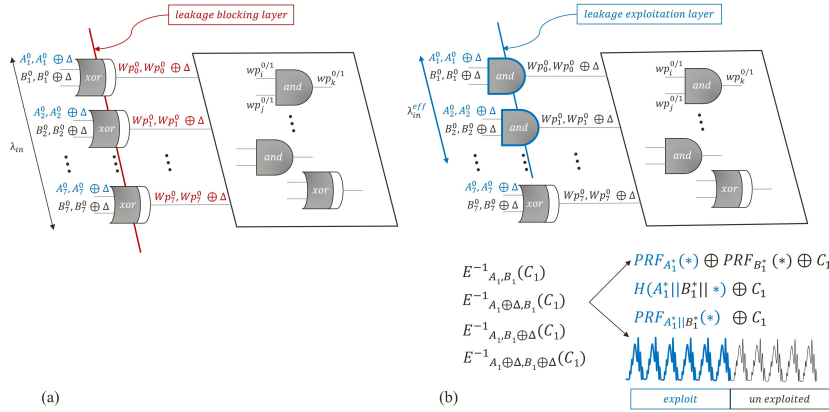
**Figure 3:** Conventional SCA apparatus: (a) CPA attack (b) Horizontal attack.

[KAA21, PKH<sup>+</sup>21], as illustrated in Fig. 3b, the adversary targets various computations under the same  $x^i$  and thus builds a model for multiple samples from the same leakage trace. Then these collective hypotheses are combined with the goal of extracting the sub-key from (e.g.) a single measurement as illustrated in the figure. Horizontal SCAs focus on multiple intermediate computations,  $int \in \{1, \dots, N_{int}\}$  that all depend on the public input and the same sub-key. The main challenge of applying horizontal SCA is in grouping multiple internal computations within a single computation, that leak over the same sub-key in a meaningful way (information sense), and be able to nicely allocate the POIs each is associated with. Even though this attack is slightly more complex, as an adversary needs to characterize a set of associated POIs, it is more powerful and can theoretically extract more information or jeopardize keys with far fewer measurements.

**Combined Attacks.** V-SCA and H-SCA can be combined. That is, multiple horizontal hypotheses can be utilized over multiple (vertical) traces collected while processing different data. In what follows below, our discussed attack merges V-SCA and H-SCA over the leakage extracted from gates garbling.

### 3 Attacking the Garbler and the Evaluator

In this section, we detail topological aspects which make the proposed attack viable. We assume that the adversary can monitor the communication channel and get a hold of the garbler associated labels ( $A_i^j$ 's), and introduce the notion of leakage-blocking layer and leakage exploitation layer in garbled circuits as illustrated in Fig. 4. In this figure, we consider a general garbled circuit which is comprised of multiple input-layer gates and 'deeper' gates fanning-out from the input layer. Considering Fig. 4(a) which schematically illustrates some GC instantiation, one can identify the layer of input gates size  $\lambda_{in}$ . In this example all input gates are XOR gates; thus, considering a globally shared  $\Delta$ . Nevertheless, as the XOR operation is locally computed not by cryptographic means and without communication, extracting some information from the leakage would be hard as the unknown  $B$  labels serve as a one-time-pad. Namely, the output labels are masked so that an adversary cannot utilize them to further build a leakage model for subsequent



**Figure 4:** GC illustration: (a) a leakage blocking layer and (b) a leakage exploitation layer.

gates. Thus, such a linear (XOR) input layer is considered here as a leakage-blocking layer.

Considering Fig. 4(b) which schematically illustrates another GC instantiation, in this example, a subset of the input gates are non-linear (e.g., ANDs) and are therefore being implemented based on a cryptographic object. Let the *effective* number of such gates in the input layer be denoted by  $\lambda_{in}^{eff}$ . In prior work such as [NPS99, KAA21, BHKR13], the cryptographic instantiations which implement garbling gates are done in a way where parts of the computation can be associated with public information (say  $\{A_i^j\}$ ) while other parts with private-information (Say  $\{B_i^j\}$ ). These can be processed and attacked separately (blue highlight), e.g., as in  $PRF_{A_1^*}(\ast) \oplus PRF_{B_1^*}(\ast) \oplus C_1$  [NPS99] or  $PRF_{A_1^* || B_1^*}(\ast) \oplus C_1$  [KSS12], where  $A_1$  and  $B_1$  are the input labels, and  $C_1$  is the output label. Namely, one part of the leakage will only depend on (blue) public labels and public data, and the rest of the leakage (black) depends on secret labels or a mixture. *These are the main openings which empower the discussed attack:* (1) incorporating a globally shared secret  $\Delta$  as specified by Free-XOR and, (2) the fact that it is possible to partition the computation’s leakage to several parts which are associated with public information only (public labels) and a shared secret ( $\Delta$ ), enabling building statistical model. *This is why such a non-linear (AND) input layer is considered here as a leakage exploitation layer.*

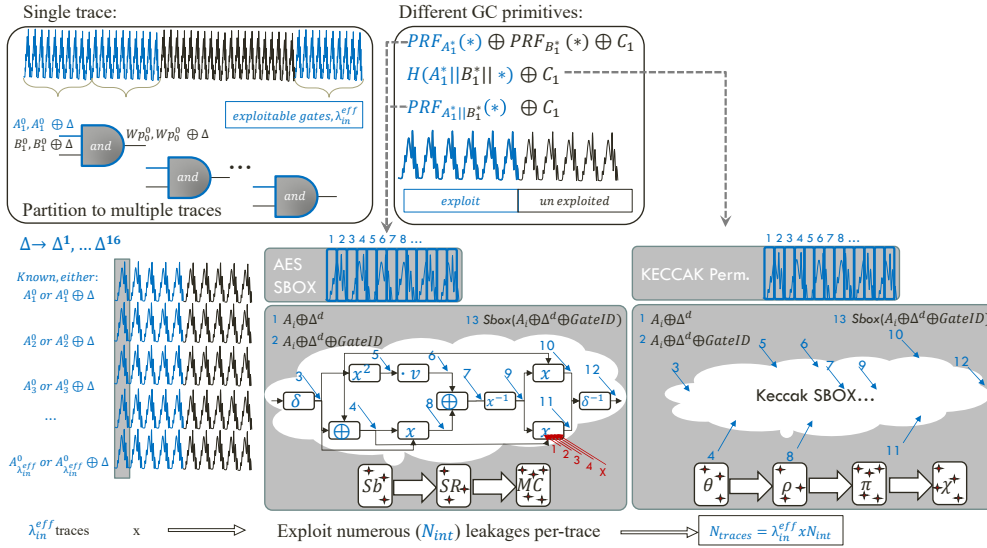
Before introducing our attack we recall that the adversary does not know the labels’ association i.e., if  $A$  or  $A \oplus \Delta$  was communicated. This information is not needed in order to build the leakage profile as we build a hypothesis for it (see Section 2.2 for hypothesis definition in the SCA context); we refer to this specific point below.

### 3.1 Running Example, Observations and Further Notations

In this subsection, we discuss our running example and further highlight some observations relating to the type of primitive used to implement the garbled gates output/truth table. We consider Fig. 5 and refer to the discussion in the previous subsection on the leakage exploitation layer. We assume that the adversary got hold of a (long) leakage trace taken from the garbling device, as illustrated on the top-left part of the figure where the publicly known values are highlighted in blue. Assuming that the circuit is known to the attacker, it can then follow with partitioning the (long) leakage to shorter snapshots, each associated with a non-linear input gate processing. Following this partition, the leakages are abstractly stacked one above the other<sup>2</sup>, so as to be utilized in a conventional SCA attack, as illustrated in the lower-left part of the figure.

The literature proposes several different primitives to implement the cryptographic

<sup>2</sup>Or placed in an array, where each gate is associated with the leakage in a different row.



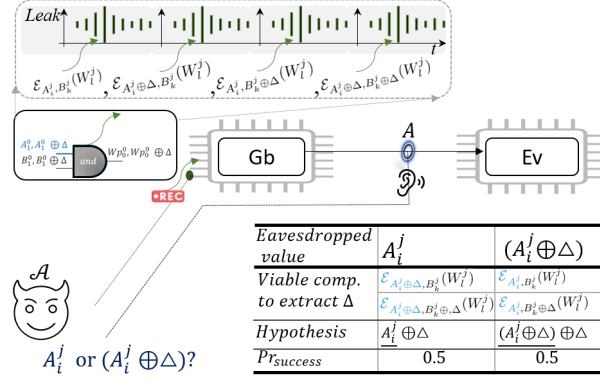
**Figure 5:** Attack illustration and internal leakages per trace, viable for advanced leakage model building (hypothesis). Where GateID denotes a public tweak known to the attacker.

garbled gate operation such as PRFs, double key PRFs and correlation robust hash functions. As different options are sensitive to the proposed attack, we do not wish to limit the discussion at this stage and therefore several such primitives are listed in the upper-right part of the figure. Our only restriction is that the underlying primitive follows the popular confusion and diffusion paradigm, where iterations of smaller permutations are used to construct the larger permutations. In this case, even if public and private information are bound by concatenation ( $A_1^* || B_1^*$ ), it will still be trivially sensitive to SCA divide-and-conquer paradigm due to the parallel confusion layer.

Next, we consider the popular and widely studied garbled gate instantiation based on AES, owing to its proven high-security level, performance and support on virtually all platforms. We discuss an attack by a weak divide-and-conquer adversary targeting only an 8-bit enumeration complexity, namely growing with  $2^8$ , and claim that even such a weak adversary is empowered by a large amount of information from the leakage traces. In our example, Fig. 5, and relating to the H-SCA discussion in subsection 2.3, an adversary can choose to exploit several internal computations within an AES. From the algorithmic perspective, only two intermediate variables with 8-bit complexity exist in the first AES round before the code diffuses and is manipulated by other sub-keys, these are the XOR and the Sbox operations. Each of these variables depends on parts of the secret  $\Delta$  and the garbler transmitted labels  $\{A_i^j\}$ . Our attack relies on the following fact: numerous leakage samples in various POIs can be extracted from correlations with these two internal variables alone. We next discuss the mechanisms which enable this property:

- **Algorithmic access** - many parts of the first-round computations are highly correlated with a selected Sbox output in various time-samples: memory access to an Sbox lookup table, further shift-rows and mix columns stages manipulations being read from the register file (RF) etc. These generate multiple leakage samples with independent noise.
- **Adversary's abilities** - when the sample rate of the measurement equipment is high, many proximate points in time can leak significantly on the same variable.

In the H-SCA context, all the above-mentioned leakages, associated with different time samples on the same internal variables, are profiled and used in the attack procedure. Below, in the experimental part of the paper, we show that by only targeting these two intermediate variables, each with associated sets of as little as 5 POIs (and up to 100),  $\Delta$


 Figure 6:  $\mathcal{A}$ 's dilemma.

can be easily recovered. Moreover, we point out that other internal variables can be used to extract additional information. For example, taking the efficient and popular, tower-field  $\mathbb{GF}(((2^2)^2)^2)$  AES-Sbox implementation, which does not utilize a memory look-up table for the Sbox, but implements it by a sequence of Boolean operations: an adversary can make use of multiple internal computations which manipulate both the global secret  $\Delta$  and the known garbler labels as illustrated in the bottom part of Fig. 5. In this case, it is easy to find at least 13 such 8-bit variables. It is worth noting that there are multiple other internal computations within the internal blocks of this Sbox such as the GF multiplier, inverter and squarer, which potentially leak. In that sense, each gate contributes multiple leakages which eventually aids the SCA adversary to average out the measurement noise, increasing the SNR, etc. Moreover, instead of targeting only 8-bit sub-keys in the divide-and-conquer approach, larger subkeys can be hypothesized leading to a larger set of internal variables independent in other subkeys, extracting more information per trace.

The above example is not only restricted to PRF-based constructions. The garbling instantiation  $H_{A^*||B^*}(\ast) \oplus C_1$  [LPS08] that utilizes a hash function also suffers from a similar sensitivity. Considering the SHA-3 construction based on the KECCAK permutation, it is still trivial to identify the same scenario and leak numerous intermediate values in various POIs in the same fashion as illustrated in the bottom right part of the figure.

### 3.2 Attacking the Garbler

Broadly speaking, the adversary captures the leakage from the garbler's hardware. This leakage is associated with a permuted garbled table as illustrated in Fig. 6.  $\mathcal{E}_{A_i^0, B_j^0}(W_l^0)$ ,  $\mathcal{E}_{A_i^0 \oplus \Delta, B_j^0}(W_l^0)$ ,  $\mathcal{E}_{A_i^0, B_j^0 \oplus \Delta}(W_l^0)$  and  $\mathcal{E}_{A_i^0 \oplus \Delta, B_j^0 \oplus \Delta}(W_l^1)$  where the permutation is unknown to the adversary. Here the input labels are denoted as  $A$  and  $B$  instead of  $W_i^j$  and  $W_k^j$  so as to highlight their association with the garbler or the evaluator, respectively. Nevertheless, in the SCA context, only two specific computations out of these four are exploitable as they reflect computations that expose  $\Delta$  to  $\mathcal{A}$ . In more detail, the SCA hypothesis, namely the internal values computed while hypothesizing  $\Delta^i$ , are computed as follows. The adversary receives either  $A_i^0$  or  $A_i^0 \oplus \Delta$  without knowing which of which and aims at associating this value with one of the four leakages, viable to extract  $\Delta$ . Therefore, in case the received garbled value is  $A_i^0$  there would exist two options of correctly associating their leakage in order to build SCA hypotheses, namely,  $\mathcal{E}_{A_i^0 \oplus \Delta, B_j^0}(W_l^0)$  and  $\mathcal{E}_{A_i^0, B_j^0 \oplus \Delta}(W_l^1)$ . Whereas, if the received garbled value is  $A_i^0 \oplus \Delta$ , the two other garbled table options would correctly associate leakages that are viable for information extraction as illustrated in the figure. The hypothesized value would anyway be the exclusive or of the received label with a hypothetical  $\Delta$ . I.e., either  $A_i^0 \oplus \Delta$  or  $(A_i^0 \oplus \Delta) \oplus \Delta$ .

Therefore, an easy technique can be to exhaust all four leakage options from the

truth-table entries to correctly select these associated leakages. The success probability of doing that per gate is half and therefore the complexity for correctly associating the leakages of all  $\lambda_{in}^{eff}$  gates are  $2^{-\lambda_{in}^{eff}}$  in total in the worst case<sup>3</sup>. The adversary utilizes each associated garbled label per non-linear gate to build a hypothesis. Generally, the attack is a conventional ‘vertical’ CPA augmented with multiple internal variables per-trace (horizontal attack components of size  $N_{int}$ ). Thus, the internal variables hypotheses can be kept of ultra-low complexity depending on a sub-key of size  $|\Delta^i|$ . For example, the attacks in Section 5 below consider values of 8 and 16 bit-lengths. In this spirit, we denote our attack by a V/H prefix for vertical/horizontal. For instance, for a model-based CPA or Template based log-likelihood distinguisher, we use V/H-CPA or V/H-TA.

The pseudo-algorithm of such an attack is listed in Algorithm 1. In the algorithm, first, the long leakage trace associated with the entire computation the garbler performs is measured (Step 1). Then, this leakage trace is partitioned into smaller leakages associated with the particular gates in accordance with the topology of the circuit (Step 2), where the adversary discards all but the leakages associated with the non-linear input gates (Step 3). In Step 4, the global secret  $\Delta$  is then partitioned into sub-keys ( $\Delta^i$ ’s). In the next step, *the adversary chooses one out of four leakage samples per each of the  $\lambda_{in}^{eff}$  gates, and runs the attack over each such selection, jointly for all gates (loop in Step 6)*. Assuming the adversary knows  $A_i^0$ , one of these choices must contain only leakages associated with computations of the form:  $\mathcal{E}_{A_i^0 \oplus \Delta, B_j^0}(W_l^0)$ , this is the easiest attack scenario. Notably, the attack will also work if the leakage that is associated with the computation that is defined by  $\mathcal{E}_{A_i^0 \oplus \Delta, B_j^0 \oplus \Delta}(W_l^1)$  is chosen instead, owing to the leakage part associated with the sensitive joint computation highlighted in light blue as discussed above. In Step 7, a leakage model is built per each gate-associated leakage and for multiple leakage samples within it ( $N_{int}$ ). In this case, for some constructions, a publicly known value such as GateID, which indicates the topological order/placement of a gate in the circuit C, may be required [BHKR13]. The comparison in Step 8 of the algorithm can be implemented using either LLH, correlation, etc.

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**Algorithm 1** Attacking the Garbler
 

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- 1: **Capture** a leakage trace from the garbler’s device.
- 2: **Partition** leakage trace to garbling leakages associated with each gate.
- 3: **Pre-processing:** keep leakage traces associated only with non-linear input gates, there exist four such encryption-leakage traces per  $\lambda_{in}^{eff}$  gates.
- 4: **Pre-processing:** partition  $\Delta$  to  $N_{sk}$  sub-keys ( $\Delta^i, i \in \{1..N_{sk}\}$ ).
- 5: **Loop:** over  $\Delta^i$ .
- 6: **Loop:** over a selection which comprises one leakage out of four leakages associated with each gate,  $4^{\lambda_{in}^{eff}}$  such options.
- 7: **Model:** build a leakage model of all internal computations used to extract information by using all of the garbler labels sent via communication and gateID. In the model, hypothesize all possible  $\Delta^i$  values of all sub-keys.
- 8: **Compare** the leakage model vector (size  $\lambda_{in}^{eff} \cdot N_{int}$ ) with the set of selected leakages
- 9: **Store** comparison results per  $\Delta^i$  and chosen leakages.
- 10: **Find:** per  $\Delta^i$  search for best comparison amongst the possible chosen leakages.

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**Error-rate:** An important aspect of our attack is that it can succeed even if some percentage of the gates associated with the leakage samples are ‘wrong’. Therefore, the attack complexity in practice can be considerably reduced below the complexity described above. In the evaluation section below, we explore the success rate (SR) of our attack as a function of the number of gates wrongly associated with leakages out of  $\lambda_{in}^{eff}$ ; denoted by “Error in leakage selection” in percentage. The concrete list of intermediate values in the algorithm (step 7) used in our attack apparatus for 8-bit chunks is the first AES round

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<sup>3</sup>Where small  $\lambda_{in}^{eff}$  values are needed for a successful attack as demonstrated below, with  $\lambda_{in}^{eff} > 3$ .

XOR and Sbox values, and for 16-bit chunks are listed in **listing 1** in Section 5.

### 3.3 The Conceptual Hardness of Attacking the Evaluator

Admittedly, by using similar tactics, attacking the evaluator via leakage is impossible:

1. First as a general note, less computation takes place on the evaluator’s side, thus there exist fewer leakage opportunities and less information to extract and accumulate. In particular, the evaluator only needs to decrypt a single ciphertext (due to PaP), which implies a single usage of the input-key labels and no correlations to extract.
2. Moreover, differently from the garbler attack scenario, here the computation is directly performed on “known” garbled values. Namely, the adversary does not have access to any leakage computed over the known labels masked with  $\Delta$  for which building a hypothesis is possible.

Our findings are consistent with a prior conjecture that this algorithm is leakage resilient.

## 4 Modeling and Simulated Attacks

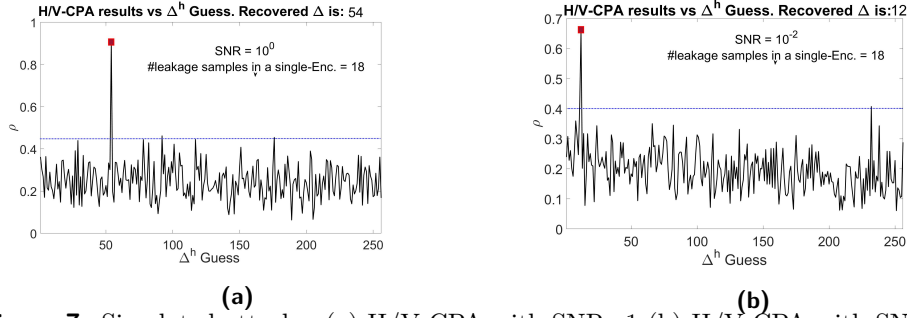
This section deals with simulating, modeling and evaluating our attack’s Success-Rate before we follow with the actual measurements in Section 5. Our starting point is simulating the leakage in a rather general setting. For that purpose, we set some assumptions:

- As discussed in Section 3 and illustrated in Fig. 4b we do not restrict the structure of the garbled circuit and we set the leakage exploitation layer size  $\lambda_{in}^{eff}$  to 8. This parameter setting implies that only  $8 \cdot 4$  AES leakage traces are effectively captured by the adversary, thus highlighting the strength of our attack. In Section 5, we show that this value can be much smaller for the attack to succeed in practice.
- We assume the garbled gate scheme utilizes a PRF which is instantiated by AES. In particular, we follow the garbling construction  $\text{PRF}_{A_1^*}(\ast) \oplus \text{PRF}_{B_1^*}(\ast) \oplus C_1$  [NPS99]. Nevertheless, as discussed above we believe that the same results and conclusions hold for other constructions based  $\text{PRF}_{A_1^* \parallel B_1^*}(\ast) \oplus C_1$  [KSS12], or  $H_{A_1^* \parallel B_1^* \parallel \ast}(\ast) \oplus C_1$  [LPS08], where  $(\ast)$  denotes a public knowledge.
- We assume that the garbler’s device is under attack and that the leakage samples are captured by monitoring it as well as exploiting the communicated garbler labels.

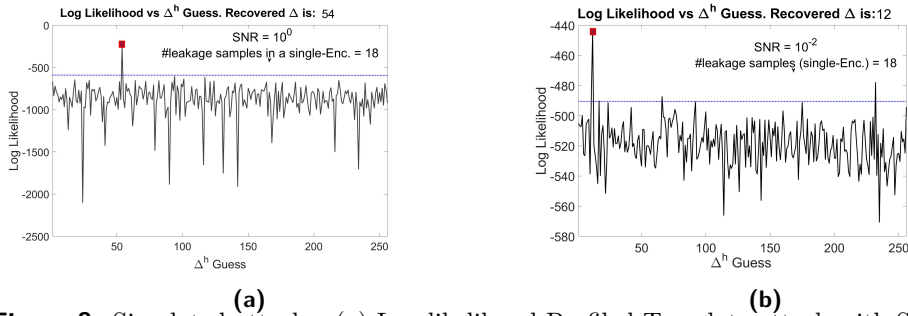
In simulating the leakages, we have used standard AES software packages to generate AES internal computation and exported internal states and various internal variables computed within the algorithm. The 8 AES-based garbled gates were asserted by randomly chosen labels and one joint  $\Delta$ . As conventional with the side-channel divide-and-conquer approach, we partition  $\Delta$  into  $N_{sk}$  sub-keys  $(\Delta^i, i \in \{1..N_{sk}\})$ . Per the computed intermediate variable, we have computed its Hamming Weight (HW) as the noiseless leakage model and added Gaussian noise with an SNR parameter [Man04] to represent the leakage as discussed in Section 2.2. The number of internal variables used in the attack per each of the above-mentioned 8 gates is denoted by  $N_{int}$  as defined above. We have experimented with various scenarios and different  $N_{int}$  values, some examples are listed below:

1. **With  $N_{sk}=16$ :** we build a hypothesis over 8-bit secret  $\Delta$  chunks. Therefore, the natural approach would be to take  $N_{int}=2$  modeling the first AES round XOR and Sbox output. Nevertheless, as discussed in Section 3.1 and shown in Section 5, in practice there exist numerous leakage points associated with either of these values, and each internal variable correlates greatly with the leakage in various time samples.
2. **With  $N_{sk}=8$ :** we build a hypothesis over 16-bit secret  $\Delta$  chunks. Therefore, we can build various hypotheses of several 8-bit XORs’, Sbox outputs and various internal processing in the shift-rows and mix-columns stages in the first round which combine these values. Practically, allowing  $N_{int}$  values  $\in \{1, \dots, 10\}$  as specified in **listing 1**.

We next discuss the results from a synthetic example which remarkably agrees with the experimental section below. Our goal here is to provide a general discussion of our



**Figure 7:** Simulated attacks: (a) H/V-CPA with SNR=1 (b) H/V-CPA with SNR= $10^{-2}$ .



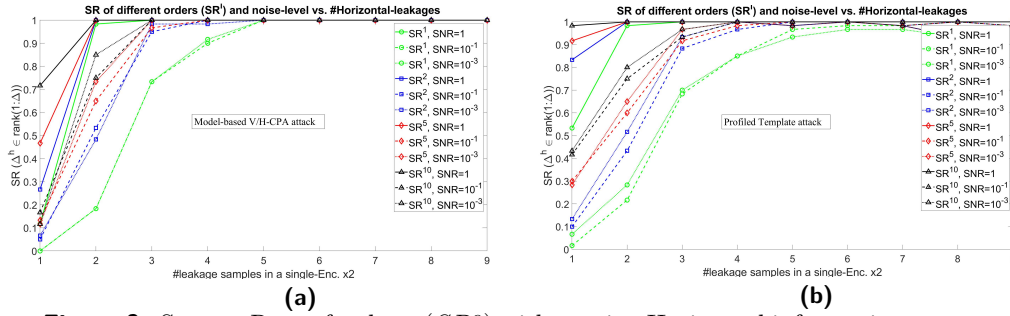
**Figure 8:** Simulated attacks: (a) Log-likelihood Profiled Template attack with SNR=1 (b) Log-likelihood Profiled Template attack with SNR= $10^{-2}$ .

simulation. In our example, for simplicity, we have assumed scenario (1) from the list above. This scenario ideally implies  $N_{int}=2$ . However, to comply with real-life scenarios, where leakage samples from multiple time instances correlate with internal variables as discussed above, we have duplicated each of these two internal variables (per trace) several times and added a fresh noise per each. Namely, instead of modeling some leakage of an internal value  $y$  by  $HW(y) + n$ , we have repeated the leakage from the same internal value with fresh randomness,  $HW(y) + n_1, \dots, HW(y) + n_r$ . In the figures below,  $N_{int}$  or verbally, the *#leakage samples in a single encryption trace*, encompass these samples as well.

We have utilized the following two discussed distinguishers over the simulated traces:

- **Model-based H/V-CPA attack** with HW leakage model as discussed in Section 2.2. This attack is augmented with modeling Horizontal leakages with various  $N_{int}$  values, as explained in Section 2.3. Our  $\lambda_{in}^{eff}$  parameter implies that the number of attack traces used in our setting is  $N_{att}=8$  and no profiling traces were used.
- **Profiled Gaussian Template attacks** with a modeled mean and standard-deviation as discussed in Section 2.2. This attack is augmented with profiled horizontal leakages as explained in Section 2.3. In this case, the number of attack traces is also  $N_{att}=8$ . However, relating to the profiled attack apparatus discussed in Section 2.2.2, we now assume that the adversary can profile a similar device to model the leakage samples and find POIs. The number of profiling traces was in the range of 100-1000, all showing successful results. Here we show results with  $N_p=500$ .

Fig 7 shows the advanced H/V-CPA attack results of a single 8-bit  $\Delta^i$ . Fig 7a shows the correlation peak with the hypothetical correct  $\Delta^i$  with a high SNR level of 1, approaching  $\rho=1$ , whereas Fig 7b shows still the correlation peak with the hypothetical correct  $\Delta^i$  with a low SNR level of  $10^{-2}$ , reducing  $\rho$ 's peak of the correct value. Fig 8 demonstrates the advanced log-likelihood-based distinguisher Gaussian template attack results of a single



**Figure 9:** Success Rate of order  $o$  ( $SR^o$ ) with varying Horizontal information per encryption ( $N_{int} \in \{1, \dots, 18\}$ ): (a) Model-based V/H-CPA (b) Profiled Gaussian Template.

8-bit  $\Delta^i$ . That is augmented with many horizontal modeled leakages. Fig 8a shows the correct  $\Delta^i$  peak with high SNR level of 1 and Fig 8b shows still the correct  $\Delta^i$  peak with a low SNR level of  $10^{-2}$ . In both examples,  $N_{int}$  was chosen to be 18. Nevertheless, as shown below, in software scenarios this value can be far higher<sup>4</sup>, making the attack much easier. In particular, the interesting question we were facing was “what is the bare minimum number of leaking time points per internal variable modeled, required for a successful attack?” That is, how many internal *horizontal* leakage samples suffice?

To answer this question we have repeated our modeled attacks with a varying  $N_{int}$  and a varying SNR level. Each experiment was repeated 200 times with random garbler labels and a random  $\Delta$ , where the attack Success-Rate of order  $Ord$  ( $SR^{Ord}$ ) was computed for different orders. Fig. 9a shows the success rates of the V/H-CPA attack. It is possible to capture from the figure that first, with as little as 10 effective leakage samples per trace all attacks succeed, i.e., 5 leakage samples were correlated with each of the 8-bit XOR and Sbox for each of the 8 garbled gates. That is, totaling 80 leakage samples, all attacks are successful with the probability of 1. With only two samples, i.e., one of the XOR and one of the Sbox, the attack effectively reduces the entropy of the  $\Delta^i$  to 3 bits out of 8. Clearly, progressively smaller orders,  $SR^{Ord}$ , converge with more samples. Fig. 9b shows our results from the same experiment over the profiled template attacks, showing similar trends. Yet, it is possible to see that with a smaller number of samples (horizontal leakage points in the x axis), the attack achieves higher success rates owing to the better (profiled) leakage model. Clearly, we can expect that in the experimental part the profiled setting will be much more powerful compared to the model-based CPA attack, as actual leakage functions of devices do not completely follow the HW leakage model.

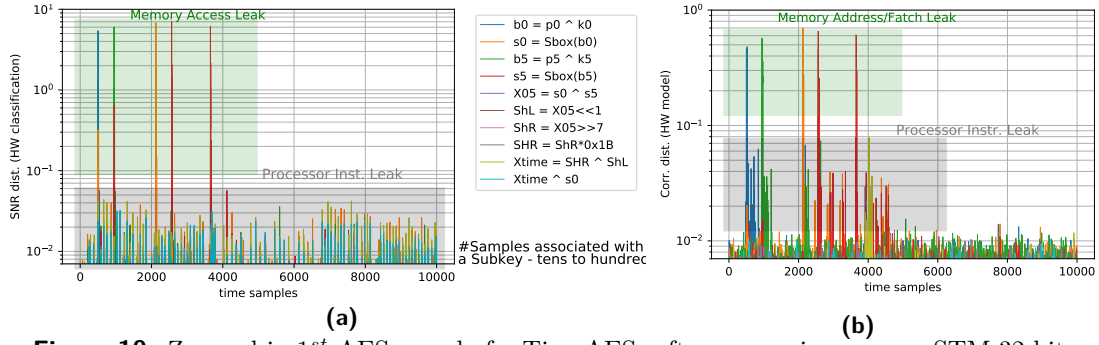
## 5 Implementation and Attacks

In this section, we follow with concrete power measurement data and remove synthetic modeling assumptions from our attacks. We utilize power measurements from an STM 32-bit software implementation of AES while garbling the circuit. Specifically, a ChipWhisperer CW308 UF0 test-board (STM32F415RGT6). On the device internally a Tiny-AES C was loaded. A series-6 PicoScope was used for the acquisition of dedicated measurement points in this platform. Additionally, we have exercised another setup with a modified Arduino Due platform, with Atmel SAM3X8E ARM Cortex-M3 CPU. In this setup we have disconnected the four Power pins of the chip and supplied power externally, measuring with a Tektronics CT-2 current probe, amplifying with a low noise RF amplifier and measured with the same PicoScope. As measurements in both setups shown similar results and were of similar quality (low SNR) we stick with the first from this point on.

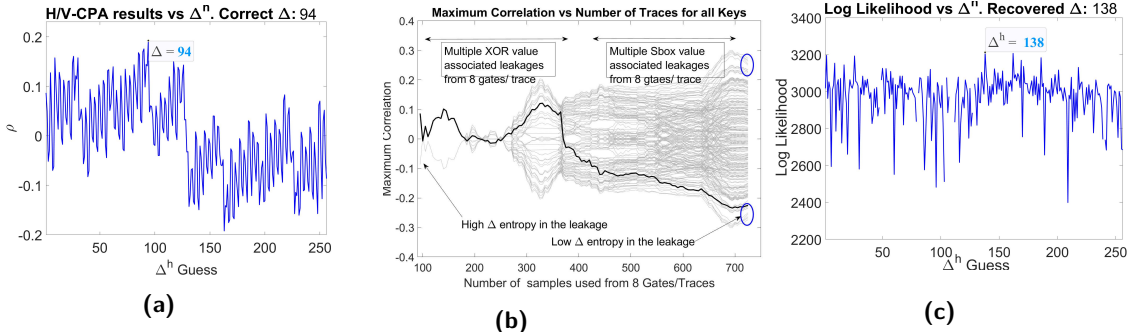
We follow with the same example of an exploitable leakage layer of eight non-linear

<sup>4</sup>We evaluated the SR of the attack with up to 100 internal leakages per internal variable in a trace.





**Figure 10:** Zoomed-in 1<sup>st</sup> AES round of a Tiny-AES software running over an STM 32-bit processor. Leakages feature numerous samples associated with internal variables depending on a single key byte: (a) SNR with HW classification (b) Correlation distinguisher.



**Figure 11:** Attacks results over a Tiny-AES SW implementation running on an STM 32-bit processor, examples: (a) H/V-CPA vs.  $\Delta^h$  guess (b) H/V-CPA vs. the number of samples used by the attack (c) Log-likelihood Profiled Template attack vs.  $\Delta^h$  guess.

gates. Our measurements-related assumptions are as follows:

- **Synchronization** - Following a circuit garbling we assume the adversary can perfectly “cut” independent leakages in a synchronized fashion. That is, we eliminate jitter and provide a clean and aligned traces scenario. The motivation to evaluate this scenario corresponds with the high SNR typically perceived in software leakages. This makes (unprotected) leakages rather easy to decently align, i.e., with techniques like elastic alignment [vWWB11] and cross-correlation [LCWY09]. We further discuss below cases of clock jitter/drifts, software interrupts or deliberate shuffling countermeasures.
- **Permutations** - In the first part below we assume the adversarial best case, namely, it specifically associates the correct exploitable leakages out of each possible 4-tuple leakages per gate. Later, we further evaluate error rates regarding this assumption.
- **Sniffed labels** - We assume the adversary got hold of the needed associated labels.

Random labels and random  $\Delta^h$  (but fixed per attack instantiation) were generated in our attack using measurements from a Tiny-AES software implementation over the New-AE Chipwesperer Pro with an STM32F target device taken from [B<sup>+</sup>19].

In this experimental part of the paper we have evaluated two attack scenarios (1)  $N_{sk}=16$  with an 8-bit attack and (2)  $N_{sk}=8$  with a 16-bit attack. We start with discussing the first case. We have performed both modeled and profiled attacks. For both the profiled and non-profiled attacks the adversary needs to find sets of POIs per internal variable. To get these, we used up to 1000 traces with random input data and compute both the SNR [Man04] and the univariate correlation [BCO04], to be used as distinguishers to

identify these POIs' sets per internal variable. Fig. 10a and 10b show the resulting SNR and correlation, respectively over a zoom-in to the first AES round leakage. In this illustration, for exposition purposes, we show SNR and correlation trends for various internal-variables as listed below for the more comprehensive 16-bit attack case.

Modeled (hypothesized) internal variables in the 16-bit  $|\Delta^i|$  case are (**listing 1**):

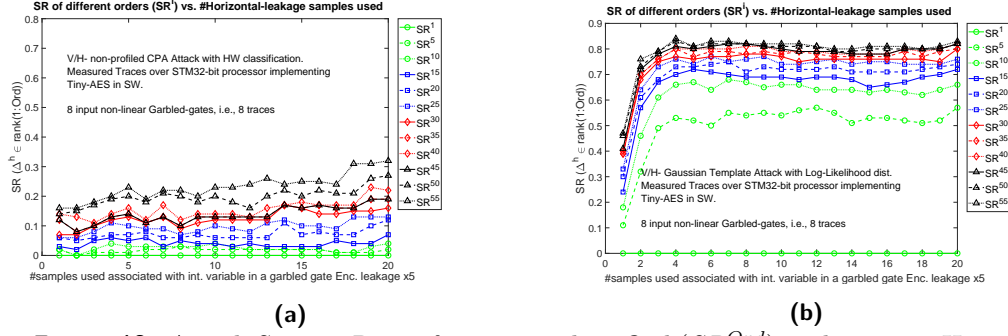
1.  $x_0 = A_0^* \oplus \Delta^0 \oplus (*)$  equivalent in the AES to the computation:  $p_0 \oplus k_0$ .
2.  $s_0 = \text{Sbox}(A_0^* \oplus \Delta^0 \oplus (*))$  equivalent in the AES to the computation:  $\text{Sbox}(p_0 \oplus k_0)$ .
3.  $x_4 = A_4^* \oplus \Delta^4 \oplus (*)$  equivalent in the AES to the computation:  $p_4 \oplus k_4$ .
4.  $s_4 = \text{Sbox}(A_4^* \oplus \Delta^4 \oplus (*))$  equivalent in the AES to the computation:  $\text{Sbox}(s_4 \oplus s_4)$ .
5.  $xs_{0\_4} = (s_4 \oplus s_0)$  equivalent AES computation.
6.  $SL = 2 \cdot xs_{0\_4} \bmod 2^8$  representing the modulus shift left in the Tiny AES.
7.  $SR = \lfloor xs_{0\_4} / 2^8 \rfloor$  representing the shift right in the Tiny AES.
8.  $Val1 = SR \cdot 27$  representing the conditional primitive polynomial reduction.
9.  $Val2 = SL \oplus Val1$  representing the mix columns internal computation.
10.  $Val3 = Val2 \oplus s_4$  representing the mix columns internal computation.

Where, as discussed above  $(*)$  represents public information,  $A^*$  represents one of the gates' labels and  $A_0^*$  indicates the zero byte chunk of this label, and  $p_0$  represents the plaintext chunk. Per each of these internal values 100 POIs' were evaluated, this step can be easily profiled with as little as a few hundred of traces with the SNR or correlation distinguishers and with other simple and univariate dimensionality reduction tools.

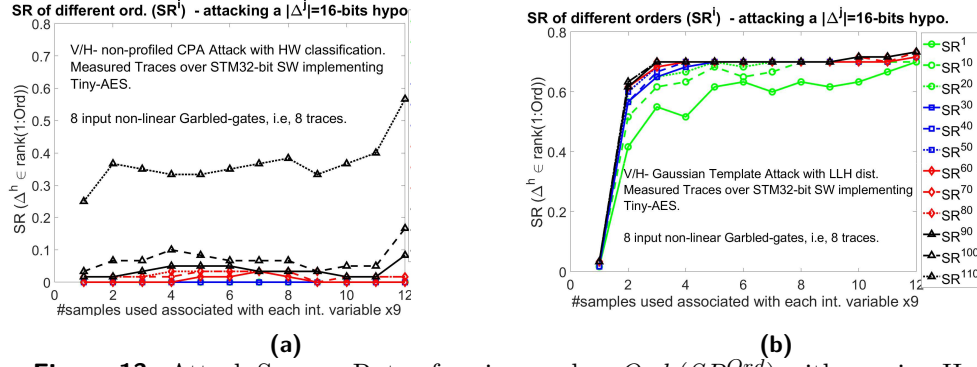
From Fig. 10 it is easily captured that numerous leakage samples correlate with the modeled internal values pinpointing and highlighting the power of horizontal attacks. For example, in the Tiny-AES implementation, as AES Sbox is stored in memory, leakages associated with memory access are clearly much leakier showing far higher SNR/Correlation levels. However, numerous other leakage samples associated with the same internal-variables leak in a round associated with Register-File access and ALU computations on these values. Though we have experimented with up to 100 samples per internal variable, a higher number of samples may exist. Furthermore, if one further zooms-in to each of these peaks in the plot, it is possible to see that it is actually smeared over time (depending on the sample-rate of the oscilloscope and the adversaries capabilities). This means that every single operation/computation leaks in many proximate samples which can enable repeated sampling and averaging. The sets of POIs were stored and utilized in the attack procedure.

Fig. 11 shows results from a single attack. That is, Fig. 11a and 11b show the results from an 8-bit H/V-CPA and log-likelihood template attack, respectively. As shown  $\Delta^i$ 's are correctly identified. In these cases, we took  $N_{int}=100$  just for illustration (i.e. maximum). The success rates while varying  $N_{int}$  are discussed below. Fig. 11b shows an interesting illustration of the H/V-CPA attack while varying  $N_{int}$ . That is, the figure shows the correlations of all hypothetical  $\Delta^i$ 's and from the left with no horizontal leakages, to the right with the maximum number of leakage samples:  $\#$  horizontal leakages (80 in this case)  $\cdot$   $\#$  of Gates (8)  $\cdot$   $\#$  of internal variables (2, XOR and Sbox). The figure highlights the reduction in guessing entropy of the key as more and more information is extracted from the leakage in our H/V apparatus. We have followed with computing the success rates of these attacks of both the modeled and profiled attacks and for both the 8-bit and 16-bit attack scenarios. Fig. 12 and 13 show the 8-bit and 16-bit scenarios, respectively. Each of these figures shows the model-based V/H-CPA case to the left (i.e. (a)) and the log-likelihood profiled Gaussian-Template attack case to the right (i.e. (b)). Quite different from the results of the modeling section, here we capture a rather poor behaviour of the model-less attacks which only provides significant  $\Delta^i$  entropy reduction with the maximum number of horizontal leakages ( $N_{int}$ ) or alternatively while hypothesizing 16-bits internal variables which enable hypothesis building of more internal variables in the algorithm.

The profiled Gaussian-Template attack shows a remarkable difference illustrating very high success rates with as little as 4-6 horizontal leakages per internal variable even for



**Figure 12:** Attack Success Rate of various orders  $Ord$  ( $SR^{Ord}$ ) with varying Horizontal information per encryption ( $N_{int} \in \{1, \dots, 100\}$ ), Profiled Gaussian Template based attacks. Attack over  $|\Delta^j| = 8$  bits: (a) V/H-CPA (b) Log-likelihood Gaussian Template.

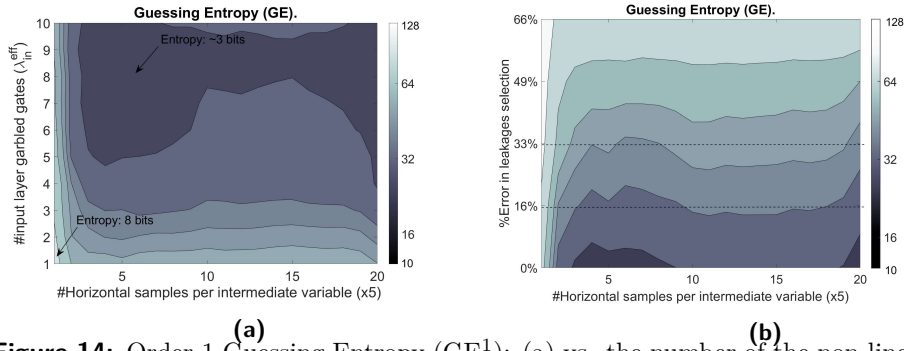


**Figure 13:** Attack Success Rate of various orders  $Ord$  ( $SR^{Ord}$ ) with varying Horizontal information per encryption ( $N_{int} \in \{1, \dots, 100\}$ ), Profiled Gaussian Template based attacks. Attack over  $|\Delta^j| = 16$  bits: (a) V/H-CPA (b) Log-likelihood Gaussian Template.

very low orders indicating the efficiency and viability of the garbler attack.

An interesting question we were facing was “what is the size of the minimal leakage exploitation layer?” Namely, what is the minimum number of non-linear input gates  $\lambda_{in}^{eff}$ , enabling an attack? We have implemented such an attack varying  $\lambda_{in}^{eff}$  from 1 to 10 gates. Fig. 14b shows the guessing-entropy (GE) of  $\Delta^i$  versus  $\lambda_{in}^{eff}$  (y axis) and the number of horizontal leakage samples utilized in the attack. The GE value is contoured with a color map showing that with as little as 5 horizontal samples per intermediate variable and even with three input gates, concrete reduction in entropy of  $\Delta^i$  is achieved with 4-bits, as compared to the lower-left corner (i.e., full 8-bit entropy). Therefore, even a very low number of non-linear input garbled gates can be quite dangerous.

Finally, in this experimental part of the manuscript we were interested to evaluate the success rate (or guessing-entropy) reduction owing to the fact that the adversary does not know whether the garbler’s associated labels, captured via the communication interface, reflect  $A$  or  $A \oplus \Delta$ . That is, we were interested to understand the influence of the ‘wrong’ gate’s associated leakages. The specific question we were interested to answer was “whether it implies an unsuccessful attack and at what stage?” If it does not considerably lower the attack success rate, the attack complexity in practice can clearly be considerably lower than the complexity described in Section 3.2. In Fig. 14b we explore the guessing-entropy as a function of the number of wrongly associated gates out of  $\lambda_{in}^{eff}$ ; denoted by “Error in leakage selection”, in %. It is possible to see that even with 16%/32% erroneous gates, significant entropy reduction is still achieved, i.e. 4-/5-bits, respectively.



**Figure 14:** Order-1 Guessing Entropy (GE<sup>1</sup>): (a) vs. the number of the non-linear garbled gates (i.e., the size of the leakage sensitive input layer,  $\lambda_{in}^{eff}$ ) (b) vs. the Error in correctly selecting  $A_j^i \oplus \Delta$  associated leakages out of  $\lambda_{in}^{eff}$ .

## 6 Future Research, Countermeasures and Conclusions

In this paper we have substantiated that if one aims to consider the attractive Free-XOR optimization for garbled circuits, SCAs are trivially viable and countermeasures are required. Below, we list potential future-research directions and several countermeasures, at the construction and protocol levels and the hardware/software implementation level. This paper emphasizes the need in protecting the sensitive secret-keys and countermeasures discussed here are given only as suggestions for follow-up research:

- **Binding:** bind both garbler’s and evaluator’s garbled keys. One such approach was demonstrated by Bellare et al. in [BHKR13] by proposing the garbled gate operation as  $\mathcal{E}_{A_i^j, B_k^j}(W_i^j) = PRF(const, 2A_i^j \oplus 4B_k^j \oplus gateID) \oplus 2A_i^j \oplus 4B_k^j \oplus gateID \oplus W_i^j$ .
- **Protocol:** adding a leakage blocking layer as discussed in Section 3 by (e.g.) introducing XOR gates at all input ports leading to non-linear gates. Either by manipulating the logic or by tying one of the inputs of the inserted XOR gates to a logically garbled ‘0’. This tied input should be associated with the evaluator’s input as well, thus preventing the adversary from propagating known inputs labels. For just one example, instead of performing  $a \wedge b$  one can compute  $a \wedge (a \oplus b) \wedge b \wedge (a \oplus b)$  processed from left to right, as to add a leakage blocking layer.
- **Implementation countermeasures:** clearly, embedding Hardware/Software Masking [ISW03, DCEM18, GMK18, BDMD<sup>+</sup>20, CGLS20] may make the already huge relative cost of GCs far larger. However, low-order masking approaches with or without other simple countermeasures such as instruction-shuffling [VCMKS12, LBS20] or amplitude-randomization techniques [LBBS20] would render attacks impractical, due to the relatively low number of observations captured by the adversary.
- **Protocol level shuffling:** another approach we identify as interesting would be to randomly shuffle the garbled gates execution at the protocol level while keeping the shuffling key a secret. The goal would be to increase the attack complexity and the error rate for incorrectly associating leakage, potentially exponentially with  $|C| = \lambda$ .

**Conclusions.** When GC are instantiated using hardware, the garbler’s device requires special attention as put forward for the first time in this paper. Our findings support devastating attacks on the scheme, clearly with standard dependency on the SNR of the SCA measurements. Admittedly, our software-based SNR level is not low, but quite standard for software implementations; Hardware garbling can naturally expect lower SNR, implying larger  $\lambda_{in}^{eff}$  required for a successful attack. However, our goal here was to put forward the general and concrete applicability of such an attack. I.e.,  $\Delta$  is jeopardized when there exist as little as 3 input layer non-linear garbled gates. We further reinforce conjectures from previous work on the leakage resiliency of the evaluator in SCA context.

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