Trivially and Efficiently Composing Masked Gadgets with Probe Isolating Non-Interference

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Abstract—We revisit the analysis and design of masked cryptographic implementations to prevent side-channel attacks. Our starting point is the (known) observation that proving the security of a higher-order masked block cipher exhaustively requires unrealistic computing power. As a result, a natural strategy is to split algorithms in smaller parts (or gadgets), with as main objectives to enable both simple composition (as initiated by Barthe et al. at CCS 2016) and efficient implementations.

We argue that existing composition strategies allow either trivial composition with significant overheads or optimized composition with more analysis efforts. As a result, we first introduce a new definition of Probe Isolating Non-Interference (PINI) that allows both trivial composition and efficient implementations. We next prove general composition theorems for PINI gadgets that considerably simplify the analysis of complex masked implementations. We finally design efficient multiplication gadgets that satisfy this definition. As additional results, we exhibit a limitation of existing compositional strategies for the analysis of Multiple-Inputs / Multiple-Outputs (MIMO) gadgets, extend Barthe et al.’s definition of Strong Non-Interference (SNI) to deal with this context, and describe an optimization method to design efficient MIMO-SNI (sub)circuits. Our results allow proving the security of a recent masked AES implementation by Goudarzi and Rivain (EUROCRYPT 2017). From the implementation viewpoint, PINI implementations reach the level of performance of the best composable masking schemes for the AES Rijndael, and outperform them by significant factors for lightweight ciphers.

I. INTRODUCTION

Side-channel attacks such as differential power analysis are a significant threat to security devices implementing cryptographic functionalities. Masking is among the most popular countermeasures to prevent such attacks. Its working principle is to split each sensitive data $x$ manipulated by an implementation into a randomized sharing $(x_0, \ldots, x_{d-1})$ such that $x = x_0 \oplus \cdots \oplus x_{d-1}$, and to perform the computations on those shares only by replacing each operation (e.g., Boolean gate) by a gadget that performs the operation over randomized sharings. Under now well understood (noise and independence) leakage assumptions, masking guarantees that the security of a masked implementation against any side-channel attack grows exponentially in the number of shares.

a) State-of-the-art: In the current state-of-the-art, masking schemes usually come with a security proof in the so-called probing model. In its simplest definition, $t$-probing security requires that the observation of up to $t$ intermediate computations in the implementation does not reveal anything about the sensitive variables. It has later been shown that while locally sufficient, this definition does not ensure secure composition. Admittedly, a security proof in the (abstract) probing model is only a first step in the analysis of a masked implementation. Various physical defaults can contradict probing security. Yet, it is a necessary first step since an insecurity in the probing model usually leads to powerful concrete attacks.

Many solutions have been introduced in order to mitigate this issue, but none of them is both generic and efficient. Genericity (that we also name trivial composition) means that a solution can be easily applied to any circuit for any $t$ using a simple circuit transformation that maps each logic gate to a masked gadget implementing that kind of gate (independently of the structure of the circuit). Efficiency is harder to characterize in a binary way, hence we rely on two criteria: the number of shares $d$ should be minimal (i.e., $d = t + 1$), and linear operations should be implemented trivially, by implementing these operations share by share.

Based on this observation, the problem we tackle in this paper is: Can we define generic composition rules for $d = t + 1$ masking such that the trivial implementation of linear functions is directly composable (i.e., does not need to be refreshed) without causing significant overheads for the non-linear operations?

b) Contribution: We answer this question positively by introducing a new security notion that we denote as Probe Isolating Non-Interference (PINI), which is satisfied by linear operations and enjoys the useful property that any composition of PINI gadgets is PINI. In contrast with the (Strong) Non-Interference (NI/SNI) definitions of Barthe et al. that rely on the the number of probes in a target implementation, PINI rather relies on their position (i.e., the shares’ indices). We then show that this approach is applicable by designing multiplication gadgets that are PINI. As a result, we can trivially analyze complex masked circuits, where all linear operations are trivially implemented and all non-linear operations are PINI. We apply our results to analyze the composition strategy of Goudarzi and Rivain, which can be viewed as a trivial composition that uses as non-linear element a special “double-SNI” multiplication gadget. We prove that this gadget is PINI, leading to a direct formal proof that the composition strategy of is secure.

In order to confirm that the trivial PINI composition also leads to efficient implementations, we next compare the performances of masked block cipher implementations based on PINI with other published solutions. Since it is currently the most efficient approach for masking, we use bit-level imple-
mportant investigations (such as the software bitslice AES by Goudarzi and Rivain [19] for this purpose, which leads us to a couple of additional observations of independent interest.

First, we analyze the limitations of the SNI definition given in [3] for composition in such bit-level implementations (such limitations apply for any gadget with multiple inputs and multiple outputs). We define Multiple-Input Multiple-Output Strong Non-Interference (MIMO-SNI) as the natural extension of SNI that allows composition in this case. Interestingly, it turns out that such cases are not problematic in the PINI framework, and we show that any MIMO-SNI gadget is actually PINI.

Second, the optimized composition of complex circuits such as bit-level AES S-boxes requires significantly more efforts than its counterpart in $\mathbb{F}_{256}$ studied by Belaïdi et al. [6]. We propose a solution to this problem, by representing the circuit to mask as a “computation graph”, and describe how to express the definitions of NL, SNI and MIMO-SNI as graph properties, leading to an algorithm (implemented as an open-source tool) to minimize the number of SNI gadgets.

We finally illustrate that PINI gadgets lead to excellent performance often improving over state-of-the-art solutions, by comparing various masked block ciphers according to the use of only SNI gadgets, the double-SNI strategy, our MIMO-SNI optimization, the recent Tight Private Circuit (TPC) approach [7] and the PINI framework. We use the AES block cipher as well as some lightweight block ciphers (Noekeon [16], PRESENT [8] and Fantomas [21]). The results allow us to conclude that our tools enable both efficient composition and efficient masked implementations. We use abstract metrics (amount of randomness, operation count) to measure algorithmic improvements, independently of the implementation (e.g., software, FPGA, ASIC). For AES, the TPC approach and the PINI framework lead to the best performances. For lightweight ciphers, the PINI approach outperforms all existing solutions by significant factors (2 to 4), making it a particularly relevant solution for lightweight ciphers submitted to the ongoing NIST competition.

This paper is organized as follows. Section II recalls the notion of masked circuit and gadget, along with the relevant security notions. We also explain the so-called “probe propagation framework” used to build intuition about those notions. In Section III we introduce the PINI definition and its properties. We instantiate it with two gadgets in Section IV, which completes the main contribution of this paper. Section V details the limitations of SNI, introduces the notion of MIMO-SNI and covers the design of the optimized MIMO-SNI AES S-box. Related work is reviewed in Section VI. Finally, Section VII briefly compares the performance of various implementation strategies and concludes.

II. PRELIMINARIES

A. Masked gadgets

We work with circuits and use the definition of [22]. A deterministic circuit $C$ is a Directed Acyclic Graph (DAG) whose vertices are gates, inputs or outputs, and whose edges are wires carrying elements of $\mathbb{F}_q$. A randomized circuit is a circuit augmented with random gates. A random gate is a gate with fan-in 0 that produces a random output, uniformly and independently of everything else afresh for each invocation of the circuit.

Let $x_*$ be a $d$-sharing of a sensitive variable $x$ if $x = \sum_{i=0}^{d-1} x_i$. The index $i$ is named the share index. A set $A$ is a set of share indices if $A \subseteq \{0, \ldots, d-1\}$, and we denote $x_A := \{x_i : i \in A\}$. We say that the sharing $x_*$ is fresh if any tuple of at most $d-1$ of its shares is distributed uniformly and independently of any other element under consideration.

A gadget $G$ with $m$ inputs and $n$ outputs working with $d$ shares implementing a function $f : \mathbb{F}_{q}^{m} \rightarrow \mathbb{F}_{q}^{n} : (x_0, \ldots, x_{m-1}) \rightarrow (y_0, \ldots, y_{n-1})$ is a circuit with $md$ inputs grouped into $m$ $d$-sharings denoted $(x_{s,0}, \ldots, x_{s,m-1})$ and $nd$ outputs grouped into $n$ sharings denoted $(y_{s,0}, \ldots, y_{s,n-1})$. A gadget must be correct. That is, if $x_j = \sum_{i=0}^{d-1} x_{i,j}$ for all $j$, then $y_j = \sum_{i=0}^{d-1} y_{i,j}$ for all $j$ and for any value of the outputs of the random gates.

We additionally use the following notations: $x_{i,*} = \{x_{i,j} : 0 \leq j \leq m-1\}$, $x_{A,*} = \{x_{i,j} : i \in A, 0 \leq j \leq m-1\}$ where $A$ is a set of share indices, and $x_{i,*} = \{x_{i,j} : 0 \leq i \leq d-1, 0 \leq j \leq m-1\}$. When it is not clear from the context, we explicitly denote the gadget $G$ to which the inputs or the outputs are related with a superscript as $x_{i,j}^G$, $y_{i,*}^G$.

For any linear function $f$, there is a trivial implementation gadget which requires no random gates and consists in applying the function independently to each share: $y_{i,*} = f(x_{i,*})$ for $i = 0, \ldots, d-1$. This also applies to affine functions (subtracting the offset $(d-1)f(0)$ to one share).

In this article, we are primarily interested in composing gadgets, that is, connecting gadgets together to build more complex gadgets.

Definition 1 (Gadget composition): A gadget composition $G$ over $d$ shares is a directed acyclic graph (DAG) whose vertices are composing gadgets (which are gadgets over $d$ shares) or inputs/outputs, and edges are connections between those gadgets. For each composing gadget, there is a one-to-one mapping between its $m$ inputs and the incoming edges of the associated vertex. Furthermore, each outgoing edge is associated to an output of the gadget (there can be multiple edges associated to the same output). Output vertices (resp., input vertices) have one (resp., zero) incoming edge and zero (resp., any number of) outgoing edges(s).

A gadget composition can be instantiated by mapping each vertex to the corresponding gadget or $d$ inputs/outputs, and each edge to $d$ wires (which connect the composing gadgets). The inputs and outputs of the composing gadgets are erased in the instantiation process. We use the term composite gadget to refer to the instantiation of a gadget composition.

B. Probing model and security definitions

In the $t$-probing model, an adversary can choose a set $P$ of $t$ probes, which are wires in the target circuit. It has then access to the values carried by each of the chosen wires. A gadget $G$ is $t$-probing secure if the output of any $t$-probing adversary is independent of the sensitive variables $(x_0, \ldots, x_{m-1})$.
Fig. 1: Simulatability examples. All outputs (bold, red) are probed. Inputs needed for simulation are circled in blue.

Fig. 2: Masked multiplication gadget with \( d = 2 \) shares. The arrows indicate the adversarial probes (there is thus one internal probe in the multiplication in (a), and one output probe in (b) and (c)). The red snake wires are the inputs needed for simulation. Each internal probe requires knowledge of one share of each input to be simulated in both NI and SNI cases (2b, 2c). Simulating an output probe requires knowledge of one share of each of the inputs for the NI gadget (2d) while it requires no input knowledge for the SNI gadget (2d).

all the input sharings are fresh. The parameter \( t \) is known as the security order. In the best situation, which is often studied in the literature and which we target, \( t = d - 1 \) (e.g., \cite{28}, \cite{6} in software, \cite{13, 20} in hardware), but this is not always the case: \( t \) can also be smaller than \( d - 1 \), or in order to mitigate physical defaults such as glitches \cite{27}). The trivial implementation of any linear function with \( d \) shares is always \( d - 1 \)-probing secure.

One important limitation of \( t \)-probing security is that it is not sufficient to ensure composability: the connection of \( t \)-probing secure gadgets is not necessarily \( t \)-probing secure \cite{15}. This has lead Barthe et al. to introduce stronger notions of security that enable composability in \cite{3}. In order to define them, we use the simulatability framework put forward by Belaid et al. in \cite{6}, illustrated in Figure 1. Intuitively, a set of probes is simulatable knowing some input shares if there exists a simulator (which is specific to that set of probes) that (knowing the said input shares) can generate simulated probes that have the same statistical distribution as the true probes. Simulatability means that the true probes are independent of the input values that are not given to the simulator. As extreme cases, if probes can be simulated using no input values, then the probes are independent on the inputs of the gadgets. On the other hand, any set of probes can trivially be simulated using all the input values: the simulator can run the gadget itself. Note that the notion of \((\mathcal{I}, \mathcal{O})\)-Non-Interference introduced in \cite{3} is equivalent.

**Definition 2 (Simulatability):** Let \( P = \{p_1, \ldots, p_l\} \) be a set of \( l \) probes of a gadget \( C \) and \( C_P \) the tuple of values of the probes for an execution of \( C \). Let \( I = \{(i_1, j_1), \ldots, (i_k, j_k)\} \subset \{0, \ldots, d-1\} \times \{0, \ldots, m-1\} \) be a set of input wires of \( C \). A simulator is a randomized function \( S : \mathbb{F}_q^d \rightarrow \mathbb{F}_q^d \). The set of probes \( P \) can be simulated with the set of input wires \( I \) if and only if there exists a simulator \( S \) such that for any inputs \( x_\ast \), the distributions \( C_P(x_\ast) \) and \( S(x_{i_1, j_1}, \ldots, x_{i_k, j_k}) \) are equal, where the probability is over the random coins in \( C \) and \( S \).

We can now define Non-Interfering (NI) gadgets and Strong Non-Interfering (SNI) gadgets. We note that for now, those notions are limited to gadgets with only one output. We take the definitions from \cite{6} and denote probes on shares of a gadget’s outputs as output probes, and probes on any wire of the gadget including inputs and outputs as internal probes (therefore any output probe is also an internal probe). The definitions are illustrated in Figure 2.

**Definition 3 (Non-Interference):** A gadget with one output sharing is \( t \)-NI if and only if every set of \( t' \leq t \) internal probes can be simulated with at most \( t' \) shares of each input.

For a gadget with \( d > t \) shares, satisfying \( t \)-NI is strictly stronger than \( t \)-probing security. Indeed, the inputs required by the simulator are independent of the sensitive variables, which implies that the simulated probes are likewise independent of the sensitive variables, and the adversarial probes have the same distribution as the simulated probes thanks to indistinguishability. \( t \)-NI is however not a necessary condition for probing security, because it requires indistinguishable simulation for any value of the input shares, not only for any value of the sensitive variables, which sometimes makes the simulation of probing secure gadgets impossible (e.g., in first-order threshold implementations, where non-linear gadgets leverage the input shares in order to reduce the randomness requirements \cite{27}). The trivial implementation of any linear function is NI: the simulator can use the \( x_i \) values for all the \( i \)'s for which there is a probe in the evaluation of \( f(x_i, \ast) \).

Next, \( t \)-SNI gadgets guarantee independence between the input and output shares, even in presence of a \( t \)-probing adversary.

**Definition 4 (Strong Non-Interference):** A gadget with one output sharing is \( t \)-SNI if and only if for every set \( \mathcal{I} \) of \( t_1 \) internal probes and every set \( \mathcal{O} \) of \( t_2 \) output probes such that \( t_1 + t_2 \leq t \), the set of probes \( \mathcal{I} \cup \mathcal{O} \) can be simulated with \( t_1 \) shares of each input.

There are many designs of gadgets that implement elementary field operations and are NI or SNI. The most studied ones are NI and SNI field multiplication and SNI refresh gadgets (which implement the identity function in a SNI fashion) \cite{22, 6}. Barthe et al. showed that it is possible to build secure composite gadgets based on NI and SNI gadgets \cite{3}.

**Proposition 1:** A composite gadget \( G \) is \( t \)-NI if all its composing gadgets are \( t \)-NI, and all outputs of composing gadgets (and input vertices) are connected to at most one edge not connected to the input of a SNI refresh gadget.
Fig. 3: Implementation of \((x + y)(x + z)\) masked with \(d = 2\) shares, with input sharings \((x_0, x_1)\), \((y_0, y_1)\) and \((z_0, z_1)\). The circuit is made of a 1-SNI multiplication with linear circuits at its inputs (two trivial implementations of the addition). The circuits illustrate (a) the limitation of SNI input composability and (b) a simple fix. The arrows indicate the adversarial probes (there is thus one internal probe in the multiplication) and the red snake wires are the propagated probes. The R box is a 1-SNI refresh gadget.

This result gives a simple way to securely compose gadgets. However, it usually requires the use of many (expensive) refresh gadgets\(^\text{[19]}\). One important reason of these overheads, which is also the seed of our following investigations, is that the trivial implementation of a linear function is not SNI. It naturally suggests the quest for a security definition that allows simple composition, as Proposition\(^\text{[1]}\) while also leveraging the efficiency and probing security of trivial implementations for linear functions, as an interesting research challenge.

C. Probe propagation framework

In this section, we describe an intuitive interpretation of the simulatability definition first used in \([6]\), that we next call the probe propagation framework, and discuss its application to the NI and SNI notions.

We start with the simple circuit example of Figure 3 which performs a multiplication of dependent values (masked with \(d = 2\) shares). There is one adversarial (internal) probe in the SNI multiplication gadget and we show how to prove (or fail to prove) that the probe is not an attack in the 1-probing model (i.e., that it is independent of the sensitive inputs) by demonstrating that it is possible to simulate it using at most one share of each of the inputs. (Analyzing all possible sets of probes would prove 1-probing security. More efficient ways of making such proofs are discussed in the following sections.)

According to the SNI definition, it is possible to perfectly simulate the adversarial probe by knowing one share of each of the inputs of the SNI multiplication. Let those required shares be the red snake wires in the circuit (the set of wires shown is an arbitrary example, the shares required by the simulator depend on the position of the adversarial probe). Those wires are called propagated probes. The proof then works by observing that if it is possible to simulate the propagated probes, then the adversarial probe can be simulated.

In our example of Figure 3a, we can propagate the probes one step further: a probe at the output of an addition can be simulated with probes on the two inputs of the addition. But then, we obtain all two propagated probes on the input sharing of \(x\). As a result, we cannot prove that the circuit is probing secure.

In order to circumvent this impossibility, the circuit of Figure 3b (which has the same functionality as the circuit of Figure 3a), implements a simple fix: there is a 1-SNI refresh gadget on one of the inputs of the multiplication gadget. The propagated probe at the output of the refresh gadget can then be simulated using no input of the gadget (thanks to the SNI property), which makes the circuit secure against this probe. As a result, this composite gadget is 1-NI (and thus 1-probing secure) thanks to Proposition\(^\text{[1]}\).

Summarizing, the main idea of the probe propagation framework is to prove the security of an implementation by replacing the adversarial probes with propagated probes that can be used to simulate the adversarial probes, and by iterating the process until the propagated probes are all at the inputs of the circuit. The conclusion is then easy. More precisely, the propagation of probes always happens backwards in the circuit (probes on the outputs of a gadget are simulated by probes on the inputs of the gadget), and the definitions of NI and SNI can be expressed with the following set of simple rules.

a) Probe propagation rules:
- For a NI gadget with \(n_o\) probes on its output shares and \(n_i\) probes inside the gadget, there are propagated probes on \(n_0 + n_i\) shares of each input.
- For a SNI gadget with \(n_o\) probes on its output shares and \(n_i\) probes inside the gadget, there are propagated probes on \(n_0 + n_i\) shares of each input. Hence, SNI gadgets (and SNI refreshes) stop the propagation of probes.

There are then three probe propagation conditions to verify, in order to guarantee security against the considered adversarial probes (this is an application of the type system from [5]).

b) Probe propagation security conditions:
1) For some parameter \(t < d\), all the gadgets must satisfy \(t\)-NI (or multiple-output variants discussed later such as PINI or MIMO-SNI).
2) For any edge in the gadget composition graph, there cannot be propagated probes on more than \(t\) shares (out of the maximum \(d\)).
3) For all SNI gadgets, the following must hold: \(n_i + n_o \leq t\) (\(n_i\) and \(n_o\) are the number of internal and output probes, respectively). Note: For NI gadgets, this condition is redundant with the second probe propagation condition applied to input sharings.

III. TRIVIAL COMPOSITION & PINI

In this section, we introduce a new definition of Probe Isolating Non-Interference (PINI) which is directly satisfied by the trivial implementation of any linear function and enjoys a simple and practical composition property: any composite gadget whose composing gadgets are all PINI is itself PINI.

We then show that this new definition allows us to build a trivial masking compiler that only requires a PINI implementation for each of the non-linear gates of interest. This compiler instantiates the given PINI non-linear gadgets, trivial implementations of the linear functions, and then makes the appropriate connections. In addition to simplicity, this...
technique is cost-efficient, since it uses no refresh gadgets. In particular, for bit-level implementations, this compiler only needs a PINI multiplication gadget (i.e., an AND gate) for which we provide efficient instances in the next section. Another interest of the PINI definition is that it applies (and its properties apply) easily to gadgets with multiple outputs, which is not the case for NI/SNI, as will be discussed in Section V-A.

A. Intuition and probe propagation framework

The main idea behind the PINI definition is to take into account not the number of probes (or of required inputs for simulation) as in the NI/SNI definitions, but instead their position (i.e., the shares’ indices). The whole circuit can then be cut into $d$ circuit shares that are not interconnected, except for non-linear gadgets. If we neglect those gadgets, the circuit is $t$-probing secure (for $t < d$): the adversary can only probe $t$ of the circuit shares, hence it has no information about at least $d - t \geq 1$ circuit shares, which contains at least one share of each input. Non-linear PINI gadgets then behave in the probing model as if they had no connection between circuit shares (i.e., they can be simulated as such), which allows implementing non-linear functions while keeping the previous circuit sharing intuition.

In the probe propagation framework, probes propagate through PINI gadgets in a way that respects the isolation of the circuit shares. Output probes propagate to all input shares with the same share index (i.e., inside the same circuit share). Internal probes in non-linear PINI gadgets are more subtle since they cannot be trivially associated to a circuit share, because there is no circuit share isolation inside those gadgets (the isolation is only simulated). However, we can let those probes carry the same intuition as the output probes: each internal (adversarial) probe gives knowledge of at most one circuit share to the adversary. This preserves the feature that the adversary has knowledge of at most $t$ of the $d$ circuit shares, which we formalize with a new probe propagation rule.

- Each output probe on a PINI gadget propagates to all the input shares that are in the same circuit share as the output probe. Each internal probe propagates to all the input shares that are in one additional circuit share (this circuit share may depend on the position of the probes).

No new probe propagation security condition is needed: if there are too many probes (internal and output), they propagate to the inputs (thanks to the previous rule), and violate the second security condition.

The way PINI works is illustrated in Figure 4 which takes the case discussed in the previous section (where a refresh was needed to prove security), and a new case not handled by (S)NI definitions: a gadget with multiple outputs.

In Figure 4a there is one internal probe which propagates to one share of each input of the multiplication as it is the case for (S)NI multiplications. However, the propagated probes have the same share index (they are in the same circuit share), hence probe propagation through the linear operation does not violate the second probe propagation security condition.

In Figure 4b the two propagated probes at the output of the S-box have the same share index, hence they propagate to only one circuit share.

B. Formalization: definition and properties

We now give the formal definition of PINI, and prove security and composability properties. For this purpose, we first show the link between the notion of circuit share and the notations of Section II. Namely, for a gadget with inputs $x_{i,j}$ and outputs $y_{i,j}$, all the inputs and outputs in the circuit share $i$ are $x_{i,*}$ and $y_{i,*}$.

In the following definition, the set $A$ is the set of share indices (i.e., the circuit shares) that are probed through output probes, and $B$ is the set of circuit shares requested to simulate the internal probes. The set of shares required to simulate all the probes is thus $A \cup B$.

Definition 5 (Probe-Isolating Non-Interference): Let $G$ be a gadget over $d$ shares and $P$ a set of $t_1$ probes on wires of $G$ (called internal probes). Let $A$ be a set of $t_2$ share indices. $G$ is $t$-Probe-Isolating Non-Interfering ($t$-PINI) iff for all $P$ and $A$ such that $t_1 + t_2 \leq t$, there exists a set $B$ of at most $t_1$ share indices such that probes on the set of wires $P \cup y^G_{A,*}$ can be simulated with the wires $x^G_{A,B,*}$.

The following proposition shows that any PINI gadget is probing secure.

Proposition 2: Any $t$-PINI gadget (with a number of shares $d > t$) is $t$-probing secure.

Proof: Any set of at most $t$ probes can be simulated with at most $t$ shares of each input. Thanks to the independent input encodings, this set of input shares is independent of all the sensitive input values.

We now look at composability properties for PINI gadgets.

Proposition 3 (PINI composability): Any composite gadget made of $t$-PINI composing gadgets is $t$-PINI.

Proof: We build a PINI simulator that takes as input a set of probes $P$ and a set of share indices of probed output shares $A$. Without loss of generality, we assume that there is no probe on wires connecting composing gadgets (hence there are only output probes and probes inside composing gadgets), since such probes can be considered to be inside one of the gadgets connected to the wire.

Let us order gadgets from output to input (in reverse topological sort order for the gadget composition DAG) and...
denote them $G_1, \ldots, G_l$. We build iteratively the sets $A_i$ which are sets of circuit shares needed to simulate probes inside gadgets $G_1, \ldots, G_{i-1}$ and output probes.

Let $P_i$ be the set of probes inside the gadget $G_i$. Let $A_1 = A$ be the set of share indices of the output probes. By induction, let $A_{i+1} = A_i \cup B_i$, where $B_i$ is obtained by running the PINI simulator for gadget $G_i$, with set of probes $P_i$ and output share indices set $A_i$ (this is possible since $|P_i| + |A_i| \leq t$). Let $B = A_{i+1}\setminus A$ be the set of input share indices required to the oracle for the simulation (in addition to the shares with indices in $A$, always given). Simulation is performed from $G_1$ to $G_l$: the PINI simulator for $G_i$ simulates $y_i^{G_i}$, and has access to $x_{A_{i+1}}^{G_i}$ (obtained from oracle and/or other PINI simulators). We next have to show that the simulator respects the cardinality constraints of PINI, that is: $|B| \leq t_1 = \sum_{i=1}^l |P_i|$. This is obtained by induction on the inequality $|A_{i+1}| \leq |A_i| + |B_i| \leq |A_i| + |P_i|$ and by the observation that $A \subset A_{i+1}$. We finally prove that the trivial implementations of linear gadgets are PINI.

Proposition 4: The trivial implementation of a linear function is $t$-PINI for any $t$.

Proof: The simulator requires access to inputs from all the circuit shares in which there is a probe. Simulation is then trivial.

IV. PINI MULTIPLICATION GADGETS

Given the results in the previous section, the main remaining challenge to leverage the trivial composition of PINI gadgets is to instantiate PINI field multiplications. We next propose two solutions for this purpose. The first one is based on the composition of existing (SNI) gadgets while the second one is a new design.

A. Goudarzi and Rivain double-SNI gadget

This gadget is based on an implementation strategy used in [19], where only SNI multiplications are used, and one input of every multiplication gadget is refreshed in a SNI manner. Goudarzi and Rivain claim (without proof) that linear operations can then be implemented in the trivial way, leading to a trivial composition using double-SNI multiplications (Algorithm 1) for every non-linear gadget. We next show that the AES implementation using this strategy is secure by showing that the double-SNI multiplication gadget is PINI.

\begin{algorithm}
\caption{Double-SNI multiplication gadget over $d$ shares. $R_d$ and $Mul_d$ are SNI $t$-refresh and $t$-SNI multiplication gadgets over $d$ shares, respectively.}
\begin{algorithmic}
\Require $(a_i), (b_i) \in \mathbb{F}_d^d$ such that $\sum a_i = a, \sum b_i = b$.
\Ensure Output $(c_i) \in \mathbb{F}_d^d$ such that $c_i = a \cdot b$.
\State $(x_i) \gets R_d((a_i));$
\State $(c_i) \gets Mul_d((x_i), (b_i));$
\end{algorithmic}
\end{algorithm}

Intuitively, this result comes from the fact that each probe inside the gadget of Algorithm 1 propagates to at most one share of one of the inputs, and output probes do not propagate, which implies PINI.

Proposition 5: Any double-$t$-SNI multiplication gadget is $t$-PINI.

Proof: Let us assume that there is a set $P_1$ of probes on the output of the gadget (i.e., probes on $(c_i)$), a set $P_2$ of probes inside the SNI multiplication gadget and a set $P_3$ of probes inside the SNI refresh gadget, such that $|P_1| + |P_2| + |P_3| \leq t$. Any probe on $(x_i)$ is counted as a probe inside the SNI multiplication gadget.

Using the simulator for the SNI multiplication gadget, we can simulate the sets of probes $P_1$ and $P_2$ using a set $P_4$ of shares of $(c_i)$ and a set $P_5$ of shares of $(b_i)$, such that $|P_4| \leq |P_2|$ and $|P_5| \leq |P_2|$. Then, using the simulator for the SNI refresh gadget, we can simulate the sets of probes $P_3$ and $P_4$ using a set $P_6$ of shares of $(a_i)$, such that $|P_6| \leq |P_3|$ (since $|P_3| + |P_4| \leq t$). Overall, we can simulate all the adversarial probes using the sets of input shares $P_5$ and $P_6$.

Since $|P_2| + |P_3|$ is the number of internal probes and $|P_5| + |P_6| \leq |P_2| + |P_3|$, the PINI simulator can request knowledge of all the input shares whose index is the one of a share in $P_5$ or $P_6$. Hence, the PINI simulator knows the values of the shares in $P_5$ and $P_6$ and can use the two SNI simulators to complete the simulation.

B. PIN1\(_1\): a more efficient field multiplication

We next introduce in Algorithm 2 a new PINI multiplication gadget for $d$ shares. It is a variation of the ISW multiplication [22] and has the same randomness requirement (i.e., $d(d-1)/2$ field elements). Compared to the double-SNI multiplication gadget, it is thus more efficient.

We defer the proof that this algorithm is $d-1$-PINI to Appendix A, but we give here the intuition behind it. The only probes that violate the PINI definition (i.e., probes that require knowledge of inputs from more than one circuit share to be simulated) in the ISW multiplication are partial products $a_ib_j$, which are intermediate values of the computation of $z_{ij} = r_{ij} + a_ib_j$. This issue can be solved by using a fresh random element $r'_{ij}$ and computing the same result as $z_{ij} = (r_{ij} + a_i r'_{ij}) + a_i (r'_{ij} + b_j)$. None of the intermediate values in this computation require the knowledge of both $a_i$ and $b_j$ to be simulated. Such a technique requires more random field elements, but it can be optimized by using only the $r_{ij}$’s and not fresh $r'_{ij}$’s, since the computation $z_{ij} = (1 + a_i)r_{ij} + a_i(r_{ij} + b_j)$ enjoys the same security and correctness as the previous expression.

This algorithm can be adapted to require only linear memory by re-ordering the operations: for each generated $r_{ij}$, compute directly $z_{ij}$ and $z_{ji}$, and update $c_{ij}$ and $c_{ji}$. The intermediate values depending on $r_{ij}$ can then be dropped. This does not change the set of possible probes, hence the security proof is still valid.

V. COMPOSITION OF LARGER GADGETS

In this section, we aim to study a representative example of larger gadget to estimate the cost of using our trivial composition for small (PINI) gadgets compared to other state-of-the-art solutions. We take the bit-level (i.e., manipulated elements are in $\mathbb{F}_2$, not $\mathbb{F}_{256}$) S-box of Boyar, Matthews and
and we prefer extension (ii), which we next call MO-SNI. Hence, the extension (i) of the SNI definition is not sufficient, propagating to $2^2$ probes on one output of the linear operation. The probes are propagated to the non-linear gadget (depicted in Figure 5). The adversary has order.

Factors

Require: Factors $a, b \in \mathbb{F}_q$ such that $\sum_i a_i = a$ and $\sum_i b_i = b$
Ensure: Output $(c_i)_i \in \mathbb{F}_q^n$ such that $\sum_i c_i = a \cdot b$.

for $i = 0$ to $d - 1$ do
  for $j = i + 1$ to $d - 1$ do
    $r_{ij} \in \mathbb{F}_q$; $r_{ji} \leftarrow r_{ij}$;
  for $i = 0$ to $d - 1$ do
    for $j = 0$ to $d - 1$, $j \neq i$ do
      $s_{ij} \leftarrow b_j + r_{ij}$;
      $p_{ij}^i \leftarrow (a_i + 1) \cdot r_{ij}$;
      $p_{ij}^j \leftarrow a_i \cdot s_{ij}$;
      $z_{ij} \leftarrow p_{ij}^i + p_{ij}^j$;
  for $i = 0$ to $d - 1$ do
    $c_i \leftarrow a_i \cdot b_i + \sum_{j=0, j \neq i}^{d-1} z_{ij}$;

Peñalta [9] that is used in [19] for this purpose. In order to obtain fair comparisons, we then follow the optimized approach to composition introduced by Belaïd et al. [6], which has been shown to provide better performances than the straightforward application of Proposition 1. Doing so, we face two additional challenges.

First, in order to exploit optimized S-boxes, we require that such S-boxes lead to a secure circuit when composed with the trivial implementation of the AES linear layer. We observe that the SNI definition is not sufficient to reach this goal, and we introduce a natural extension of SNI that allows such a composition. It essentially extends SNI to a multiple-input and multiple-output setting (hence the name MIMO-SNI for the proposed extension).

Next, we observe that the optimization of a complex circuit such as the bit-level S-box in [9] is computationally intensive, and cannot be exhaustively analyzed like the $\mathbb{F}_{2^{256}}$ S-box investigated in [6]. So we propose an automated heuristic based on linear programming to mitigate this limitation.

A. Multiple-Input-Multiple-Output SNI

a) Multiple-Output SNI (MO-SNI): A first issue with the SNI definition is its specialization to single-output gadgets (whereas a bit-level AES S-box has eight output bits). Therefore, we need to extend this definition to multiple-output gadgets. Two natural extensions can be considered for this purpose: (i) the gadget tolerates at most a total of $t$ probes for all the outputs, or (ii) it tolerates up to $t$ probes for each of the outputs.

The next example shows that the first option is not sufficient to ensure composability with linear layers. In order to simplify the discussion, we take a simple case of 2-bit non-linear functions (i.e., each gadget has two inputs and two outputs) masked at order $t = 1$ (i.e., $d = 2$), but our reasoning applies to any gadget (e.g., the 8-bit S-boxes of the AES) and any order.

We consider a linear operation between two outputs of a non-linear gadget (depicted in Figure 5). The adversary has $t$ probes on one output of the linear operation. The probes propagate to $2t$ probes on the output of the non-linear gadget. Hence, the extension (i) of the SNI definition is not sufficient, and we prefer extension (ii), which we next call MO-SNI.

Fig. 5: Multiple-output non-linear gadget with linear layer at the output: this circuit is insecure with SNI’s MO extension (i) for the non-linear gadget, while it is secure for extension (ii). Example for $d = 2$ and 2 bit gadget.

b) Multiple-Input (MI): A already discussed in Section IIIC in general a non-linear SNI gadget cannot have a linear layer at its input without refreshing some of these inputs. Therefore, we introduce a stronger constraint: the simulator must be able to simulate the probes using one input share per probe, while it is one input share per input and per probe for SNI.

c) Multiple-Input-Multiple-Output SNI (MIMO-SNI): The combination of the MO-SNI definition with the MI constraint gives the MIMO-SNI notion, that we first introduce in the probe propagation framework, and then define formally. There is a new probe propagation rule for MIMO-SNI gadgets (to cover the MI-SNI part of the definition):

- For a MIMO-SNI gadget with at most $n_o$ probes on each output and $n_i$ internal probes, there is a total of $n_i$ propagated probes on the input shares.

There is then a fourth probe propagation security condition, which is the condition for SNI gadgets adapted to MO-SNI:

4) For any MIMO-SNI gadget with at most $n_o$ probes on each output and $n_i$ internal probes, the following must hold: $n_i + n_o \leq t$.

Definition 6 (MIMO-SNI): Let $O_i$ be a set of share indices for $i = 0, \ldots, n - 1$. A gadget is $t$-MIMO-SNI if and only for any set $\mathcal{I}$ of $t_1$ internal probes and any sets $O_i$ such that there exists a $t_2$ that satisfies $t_1 + t_2 \leq t$ and $|O_i| \leq t_2$ for $i = 0, \ldots, n - 1$, the set of probes $\mathcal{I} \cup y_{O_0,0} \cup \cdots \cup y_{O_{n-1},n-1}$ can be simulated with at most $t_1$ input shares.

This definition is very strong, in fact it is strictly stronger than PINI, as shown in the next proposition.

Proposition 6: Any $t$-MIMO-SNI gadget is $t$-PINI.

Proof: The MIMO-SNI simulator can be used as a PINI simulator, with the set of share indices $B$ sent to the PINI oracle being made of the indices of the $t_1$ input shares required by the MIMO-SNI simulator.

This proposition shows that MIMO-SNI benefits from the same composability properties as PINI: a MIMO-SNI S-box can thus be trivially composed with linear layers. Despite it is stronger than PINI (which is already sufficient to compose securely), an interesting feature of MIMO-SNI is that this notion can be obtained by combining NI and SNI elementary gadgets in an optimized composition similar to the one proposed in [6] (see next).

Additional remark. The previous definitions are quite connected to the recent work of Belaïd et al. on Tight Private Circuits (TPC) [7], in which the AES S-box is probing-secure and its outputs are all outputs of SNI gadgets. This property is similar to MO-SNI: it guarantees independence between outputs, and between outputs and inputs; but it is
weaker than MO-SNI: it does not require simulatability. The authors show that despite the weaker definition, it can be securely composed with linear layers to build a t-probing-secure circuit. This weaker requirement enables more efficient implementations than [19]: S-boxes only requires 32 SNI multiplication gadgets, and no refresh gadgets in this case. However, it does not enable trivial composition. First, this composability result is limited to full linear layers (which results in a need to refresh the key scheduling even if it is linear). Second, the analysis of the S-box itself is not trivial (it requires circuit-specific analysis to prove its security).

B. Building large MIMO-SNI gadgets from small (S)NI gadgets

In order to fairly assess the interest of the PINI framework with respect to state-of-the-art solutions, and to allow sound performance comparisons in the next section, we now tackle the problem of building a large MIMO-SNI gadget by composition of smaller (S)NI gadgets. For a given functionality, we try to minimize the amount of SNI gadgets in the implementation in order to reduce its complexity. In other words, we investigate the optimized composition approach mentioned in the introduction as a natural competitor to the trivial one.

For this purpose, we first show how to express this optimization based on the properties of a graph describing the computations to perform. We then apply this optimization to the AES S-box in $\mathbb{F}_{256}$ (confirming the results in [6]) and to the bit-level AES S-box of Boyar, Matthews and Peralta [9], bringing significant improvements over the double-SNI strategy in [19].

1) Connecting composability to computation graph properties: We introduce a new computation graph model based on the gadget composition DAG in order to explicitly put into evidence the cases where a sharing is used multiple times. The computation graph restricts the gadget composition DAG by forbidding the connection of more than one edge to an output of a gadget or an input gate. To handle the forbidden cases, we add a new $\text{Split}_{n}$ gadget which has one input, $n$ identical outputs and performs no operations (it only connects input to outputs). For simplicity, we assume that all the composing gadget are NI operation gadgets with one output (in practice mostly additions and multiplications), SNI refresh gadgets (with one input and one output) or $\text{Split}_{n}$ gadgets. SNI gadgets are thus modeled as NI ones followed by a SNI refresh. Given a computation graph resulting from our optimization, an implementer can then replace NI multiplications followed by a SNI refresh by (sometimes more efficient) SNI multiplications. This modeling is without loss of generality since it is equivalent from the probing model viewpoint and the respective costs of the different gadgets of a private circuit are parameters of the optimizations.

Using the link between computation graph and the probe propagation framework and capitalizing on the fact that SNI refresh gadgets stop the propagation of probes, we can simply remove them (and their incident edges) from the graph to build a simplified graph. The probes inside the refresh gadgets can be reported to gadgets connected to their input, hence the simplified graph is equivalent to the original graph regarding security in the probing model.

Definition 7 (Simplified computation graph): The simplification of the computation graph $G$ is the graph that is obtained from $G$ by removing all SNI refresh vertices and their incident edges.

Let us now analyze under which condition a simplified computation graph represents a NI composite gadget, leveraging the probe propagation framework. Since we only have NI gadgets in a simplified computation graph, the only case where the probe propagation security conditions are not respected is when there are more than $t$ propagated probes on one share. Since there are at most $t$ adversarial probes, violation of the security condition means that a single adversarial probe is duplicated, that is, it propagates to the same share through different paths. We can thus derive a sufficient NI condition for simplified computation graphs: no probe should propagate backwards from a node to another one through two different paths. In other words, for any pair of vertices there should be at most one (directed) path between them. It can be seen that this condition cannot be weakened while guaranteeing security for any NI composing gadgets: if probes can propagate backwards through two paths from a node $A$ to a node $B$ and if the adversary has $t$ probes on the output of $A$, up to $2t$ shares of the output of $B$ could be required to simulate.

We now formalize this security condition with the following propositions (which generalize the proof that the AES inversion is $t$-SNI in [6]). For this purpose, we first define a property for composite gadgets and their simplified computation graph, which specifies the NI condition of the previous paragraphs.

Definition 8 (Single-Path-NI-Built gadget (SP-NIB)): A composite gadget $G$ is SP-NIB if it is implemented with only NI gadgets and SNI refreshes, and if for any pair of vertices $u$, $v$ in the corresponding simplified computation graph there exists at most one path from $u$ to $v$.

Proposition 7: Let $G$ be a composite gadget. If $G$ is SP-NIB (as per Definition 8), then it is $t$-NI.

Proof: For each edge $i$ in the computation graph, there is a number of adversarial probes $a_{i}$, a number of propagated probes $p_{i}$ and a total number of probes $s_{i}$. The sum of the $a_{i}$’s is at most $t$. For all $i$, $s_{i} = a_{i} + p_{i}$. For each edge, the number of propagated probes is either 0 if the node at the end of the edge is a refresh or an output gate, or the sum of the total number of probes of the outgoing edges of the vertex at the end of the edge otherwise (i.e., for split or NI gadgets). The probes inside a NI gadget are not considered since they can equivalently be replaced with probes on output shares of the gadget. Furthermore, the probes on output gates are modeled as adversarial probes on the edges connected to those gates.

If for each input edge $i$ (i.e., an edge connected to an input gate), a simulator knows $s_{i}$ well-chosen shares, then it can simulate all the probes of the adversary by using the simulator for each gadget in order to get the required intermediate values.

We now prove that the SP-NIB hypothesis implies that for all input edges $i$, $s_{i} \leq t$. This proves that the gadget is NI thanks to the previous observation.
We use a small lemma for this purpose: for all edges \(i\), \(p_i = \sum_j \alpha_{ij} a_j\) where \(\alpha_{ij}\) is the number of paths from the output node of \(i\) to the input node of \(j\) in the simplified computation graph. This can be proven by backwards induction on the graph: if all the outgoing edges of a node satisfy this property, it is also satisfied for all the incoming edges to this node if the node is a refresh, split or NI operation. As a base case, this is trivially satisfied for output edges (i.e., edges connected to an output gate).

Next, we observe that for any \(i\), \(\alpha_{ij} = 0\): there is no path from the output node of \(i\) to the input node of \(i\) (otherwise the graph is not a DAG). The previous lemma is thus strengthened to \(p_i = \sum_{j \neq i} \alpha_{ij} a_j\).

To conclude the main proof, we observe that the main hypothesis implies that \(\alpha_{ij} \leq 1\) for all pairs of edges \((i, j)\), hence \(s_i = a_i p_i = \sum_{j \neq i} \alpha_{ij} a_j \leq \sum a_j \leq t\).

We now give similar computation graph formalizations that guarantee SNI and MIMO-SNI. Intuitively, a composite gadget is SNI if no probes can propagate from any output to any input (i.e., there must be no path from an input to an output).

**Proposition 8:** Let \(G\) be a composite gadget. If the gadget is SP-NIB and if for any input node \(v\) and any output node \(v\), there is no path from \(u\) to \(v\), then the gadget is \(t\)-SNI.

**Proof:** Looking at the proof of Proposition 7 we observe that, under the current stronger hypothesis, \(\alpha_{ij} = 0\) for all input edges \(i\) and output edges \(j\). Hence for all input edges \(i\), \(s_i \leq t_1\) where \(t_1\) is the number of internal probes.

For MIMO-SNI, we additionally require that no probe can propagate to two inputs simultaneously (otherwise, \(t\) internal probes could propagate into strictly more than \(t\) input probes). Furthermore, there must be no composing gadget to which probes could propagate from two different outputs: otherwise, up to \(t\) probes could propagate from each output, resulting into up to \(2t\) propagated probes on the output of the composing gadget.

**Proposition 9:** A composite gadget \(G\) is \(t\)-MIMO-SNI if it satisfies the three following conditions. (i) \(G\) is SP-NIB. (ii) For any pair of output nodes \(u_1, u_2\) there is no node \(v\) such that there is a path from \(v\) to \(u_1\) and a path from \(v\) to \(u_2\). (iii) For any pair of input nodes \(u_1, u_2\) there is no node \(v\) such that there is a path from \(v\) to \(u_1\) and a path from \(v\) to \(u_2\).

**Proof:** We first have to prove that for all edges \(i, s_i \leq t\). Under the current hypothesis, the lemma from the proof of Proposition 7 is strengthened: for any edge \(i\), \(\sum_{j \in O_i} \alpha_{ij} \leq 1\) where \(O_i\) is the set of output edges. Furthermore, for any \(i, j, \alpha_{ij} \leq 1\). This implies that \(s_i \leq t_1 + t_2 \leq t\), taking the definitions of \(t_1\) and \(t_2\) from the MIMO-SNI definition.

Second, we have to prove that \(\sum_{i \in I} s_i \leq t_1\) where \(I_i\) is the set of input edges. We know that for all \(j, \sum_{i \in I_j} \alpha_{ij} \leq 1\) and for output edges \(j, \sum_{i \in I_j} \alpha_{ij} = 0\). Hence \(\sum_{i \in I_e} s_i \leq \sum_{i \in I_e} \alpha_{ij} = t_j\).

2) Optimizing the AES S-box in \(\mathbb{F}_{256}\): Using the previous graph formalization, we built a tool [11] that checks if a circuit is (MIMO)-(S)NI. If we want to build a SNI S-box with the multiplication chain from [6], there are 16 wires on which we could insert a refresh. This number is sufficiently small to make an exhaustive search, which confirms the result of [6] and shows that it is the only solution with only three refresh elements (up to the permutation of refresh gadgets with the \((\cdot)^2\) power gadgets): two refresh gadgets and one SNI multiplication. ([6] actually mentions two SNI multiplications are needed, but it was observed by Jean-Sébastien Coron that one is enough during Adrian Thillard’s PhD defense.) It also shows that two refresh gadgets is the minimum possible, even with all multiplications implemented as SNI gadgets.

3) Optimizing the bit-level AES S-box of Boyar et al.: We now optimize the implementation of a bit-level AES S-box. We take the logic circuit by Boyar et al. in [9] and search, starting from an implementation with NI gadgets, where it is required to add SNI refresh elements to get a MIMO-SNI implementation.

Due to the large size of the non-linear part of the AES S-Box (more than 124 wires), it is not possible to apply exhaustive search as done for the S-Box in \(\mathbb{F}_{256}\). We instead re-write this problem as a mixed integer linear optimization problem, for which there exists solvers with efficient heuristics. The formulation of the optimization problem is explained in Appendix B and implemented in [11].

We ran the optimization with a uniform cost for all edges, which is sound if we assume that the cost of replacing a NI operation with a SNI one is the same as adding a SNI refresh. This assumption is valid for state-of-the-art gadgets at very high order (as confirmed Section VII). The optimization solver gave a solution with 41 SNI elements and a lower bound of 34 SNI elements (after two hours of running time). In comparison, the implementation of Goudarzi and Rivain in [19] uses two SNI elements per AND gate, totaling 64 SNI elements.

The same technique can be applied to other S-boxes, which gives 7 SNI elements for the Noekeon S-box (4 bit), 8 SNI elements for the Present S-box (4 bit) and 17 SNI elements for the Fantomas S-box (8 bit). Thanks to the lower gate count of those S-boxes, the solver is able to find an optimal solution.

VI. RELATED WORK

In this section, we compare our new composition strategies to other strategies from the literature.

First, the seminal masking transformation of Ishai et al. [22] is composable. The composition proof is based on the observation that if a share index \(i\) belongs to the set of shares \(I\) required for simulation (for both inputs of a multiplication gadget), then the output with share index \(i\) can be simulated. This constraint also appears in the PINI definition. However, Ishai et al. require a number of shares at least \(d = 2t + 1\) (for security at order \(t\)). The additional requirements in the PINI definition improve this bound to \(d = t + 1\).

All the composition schemes we discuss next use \(d = t + 1\) masking. The strategy of Goudarzi and Rivain [19], for instance, is based on trivial composition but uses a more complex multiplication gadget, namely the double-SNI construction, where both the multiplication and the refresh are based on the ISW multiplication gadget. However, they do not prove that the composite circuit is probing secure (see Section V-A).

A more formal and generic composition framework has been put forward by Barthe et al. in [3] through the NI and
SNI definitions. Their main composition result (Proposition 1) is simple, but leads to poor performance (see Section V-B1): it requires refresh gadgets to protect linear operations. The NI and SNI definitions are used by Belaïd et al. [6] as the basis for building a more efficient $F_{256}$ AES S-box circuit. This approach is the basis for the techniques developed in Section V-B.

Recently, Belaïd et al. [7] introduced the Tight Private Circuit (TPC) composition strategy. It is based on the use of SNI multiplication gadgets and a careful analysis of the structure of non-linear layers, in order to insert SNI refresh gadgets where needed. This approach allows to greatly reduce the number of refresh gadgets needed for masking the AES compared to previous approaches. It exploits the circuit shares isolation property of the linear operations, as well as the bijectivity of the XOR operations with respect to one of its inputs. One (slight) downside of the TPC strategy is that it is specialized to block ciphers with a given (admittedly very usual) structure. More importantly, it requires some specific optimizations leading to better or worse performances depending on the cipher to protect (see the next section). It also requires to add some refresh gadgets in linear key schedules (as frequently used in lightweight cryptography).

We note that for hardware implementations, other approaches have been introduced to deal with hardware particularities such as glitches, notably the Threshold Implementations [27] (TI) and the Domain Oriented Masking (DOM) [20]. Those approaches have no formal proof of composability. In the DOM approach, the notion of domain is similar to what we call a circuit share. They however analyze more the physical propagation of glitches inside or across domains whereas the probe propagation framework is based on statistical dependency propagation. We do not analyze glitch-resistance in the PINI framework and leave it as an interesting open problem.

A completely different approach for composing masked gadget is based direct verification of $t$-probing security of the composite circuit using automated tools [2]. Circuit verification is a computationally expensive problem, hence those tools are limited in circuit size and masking order.

The main features of these compositional strategies are summarized in Table I.

Finally, a related work [12] introduces other gadgets that satisfy the PINI definition, one of those (PINI$_2$) improving the randomness cost at very high orders.

### VII. PERFORMANCE COMPARISON

We conclude the paper by briefly discussing the performance aspects of various compositional strategies introduced in the literature (including ours).

For this purpose, we consider state-of-the art gadgets at each order (order-specific and generic constructions) for each property needed: SNI refresh, NI multiplication, SNI multiplication and PINI multiplication (whose costs are given in Appendix C). Namely, the refresh gadgets are taken from [4], the NI multiplications and SNI multiplication come from [22], [6], [5], the PINI$_1$ multiplication comes from this work (Algorithm 2), while the PINI$_2$ multiplication was introduced in [12]. For the construction of a SNI multiplication, we observe that for sufficiently high orders ($d > 12$), the multiplication of Belaïd et al. followed by a SNI refresh has a lower cost than the multiplication of Ishai, Sahai and Wagner, which justifies the assumptions made for the optimization in Section V-B.

Next, we analyze the strategies on various block ciphers. We take AES with the bit-level implementation of Boyar, Matthews and Peralta masked at various orders as a first realistic case study (since it is the basis for the best-reported masked software performances in [19]). We also consider lightweight block ciphers, which are arguably more relevant for embedded devices where masking is needed and when performance is a constraint: PRESENT (with 80 bit key) and Noekeon (with 4 bit S-boxes), and Fantomas (8 bit S-boxes).

A synthetic evaluation is shown for each compositional strategy and each block cipher in Figures 6 and 8 (the data is also given in tables in Appendix C for power-of-2 $d$). For this purpose, we use a simple model where the cost is the number of $F_2$ operations, assuming that the generation of one random $F_2$ element has the same cost as 80 $F_2$ operations. This model is admittedly abstract and based on the PRNG performances on a particular ARM Cortex M4 processor mentioned in [23]. We note that our conclusions are stable for a wide range of PRNG costs, and refer to [12] for more empirical confirmations of these comparisons.

The implementations considered use the following approaches: using only SNI gadgets (SNI multiplications and SNI refresh after each linear operation, Proposition 1); Tight Private Circuits (TPC) strategy from [7]; optimized MIMO-SNI S-boxes (see Section V-B); trivial composition using Double-SNI multiplication gadgets [19]; and trivial composition using PINI gadgets. For the PINI strategies, we included
These results confirm that PINI is a good strategy to build masked circuits. First, it enjoys the trivial composition property, which leads to simplicity (such as dealing with only one kind of gadget, while e.g., other strategies need rules to decide whether to use NI/SNI multiplications and SNI refresh gadgets). Second, PINI leads to high-performance implementations that reduce by 2 the cost of lightweight block cipher implementations. We believe those results are timely for enabling the comparison of side-channel resistant implementations in the ongoing NIST competition in lightweight cryptography, and are in general relevant to the security and efficiency of any embedded cryptographic implementation, which are important building blocks of many secure systems.

Finally, we recall that such performance comparisons (e.g., between the MIMO-SNI, TPC and PINI approaches) are dependent on the cost of the (state-of-the-art) gadgets used. Hence, the search of more optimized such gadgets is an interesting scope for further research, to further refine our understanding of cost-optimized masked implementations.

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References

Algorithm 3 Input shares chooser for the simulator of PINI$_1$ masking

Require: Set of probes $y_{d,A}^G \cup P$
\[ X \leftarrow \emptyset; \]
for $i = 0$ to $d - 1$ do
\[ \text{if } a_i, a_i + 1, b_i, a_i \cdot b_i \text{ or } c_i \text{ is probed then} \]
\[ X \leftarrow X \cup \{i\}; \]
\[ \text{else if there exists } k \text{ such that } \sum_{j=1}^k z_{ij} \text{ is probed then} \]
\[ X \leftarrow X \cup \{i,j\}; \]
for $j = 0$ to $d - 1$ do
\[ \text{if there are at least two probes on intermediate values of} \]
\[ \text{computation of } z_{ij} \text{ (these values are } r_{ij}, p_{ij}^s, s_{ij}^t \text{ and } z_{ij}) \text{ then} \]
\[ X \leftarrow X \cup \{i,j\}; \]
\[ \text{else if there is one probe on an intermediate value of the} \]
\[ \text{computation of } z_{ij} \text{ then} \]
\[ \text{if } i \in X \text{ or } j \in X \text{ then} \]
\[ X \leftarrow X \cup \{i,j\}; \]
\[ \text{else} \]
\[ X \leftarrow X \cup \{i\}; \]
\[ B \leftarrow X \setminus A; \]
Ensure: $|B| \leq |P|$

A bottom linear transformation. Since our goal is to have a probing secure implementation of the AES, we do not need to have a full MIMO-SNI S-box. Having only the middle nonlinear transformation MIMO-SNI is enough since the top and bottom linear transformations can be considered as combined with the other linear operations of the AES (i.e., ShiftRow, MixColumns and AddRoundKey) when applying MIMO-SNI composability.

The non-linear transformation is made of 30 XOR gates and 32 AND gates, hence it contains more than 2·(30+32) = 124 wires. This means that it is impossible to apply the exhaustive search used for the S-Box in F$_{256}$. We therefore reformulate our graph optimization problem into an integer linear programming problem, for which there exists numerous solvers. This does not guarantee that we can find an optimal solution with a reasonable amount of resources, but solvers have efficient heuristics to find good solutions and can prove lower bounds for the solution. Since we take care that our representation as an optimization problem admits as acceptable solutions all the possible implementations of the considered logic circuit, we are able to provide upper and lower bounds on the cost of the optimal implementation.

We write the linear optimization problem in the following way. A binary variable $e_i$ is associated to each edge $i$ of the graph, indicating if it is cut (i.e., if a refresh is inserted). All the paths in the graph are then computed and a binary variable $p_j$ is assigned to each path $j$, again indicating if it is cut. A path is cut if any edge in the path is cut. It implies a first general constraint $p_j \leq \sum_i e_i$ (the sum is over the edges in the path).

We can then add the various constraints related to Non-Interference properties. First, to enforce NI, for each pair of vertices $(u,v)$ all but one paths from $u$ to $v$ must be cut. Let $J$ be the set of paths from $u$ to $v$, $\sum_{j \in J} p_j \geq |J| - 1$. Next, to enforce SNI, when $u$ is an input node and $v$ an output node, the constraint becomes $\sum_{j \in J} p_j \geq |J|$. For the MI part we

APPENDIX A
PINI$_1$ SECURITY PROOF

Proposition 10: The multiplication gadget of Algorithm 2 is $d - 1$-PINI.

Proof: Let us build a PINI simulator: first, given a set of probes $P$ and probed output shares $A$, Algorithm 2 gives the set of required input shares $X = A \cup B$. All the probes that depend only on input shares with index in $X$ and on randomness are computed as specified by Algorithm 2 (required randomness is generated). If some $z_{ij}$ or $s_{ij}$ appears in the expression of some probe and cannot be computed from known inputs, it is assigned a fresh random (we say that the simulator cheats on that value) and evaluation continues until all probes are computed.

Let us now show that this algorithm produces the same probe distribution as Algorithm 2. The behavior of the simulator is identical to the behavior of the gadget, except for some $z_{ij}$ and $s_{ij}$. If the simulator cheats on $z_{ij}$, then $z_{ij}$ or a sum in which it appears is probed, and there is no probe on $z_{ij}$ (or their intermediate values, or a sum in which it appears) or on intermediate values of the computation of $z_{ij}$. Therefore, $r_{ij}$ is only observable through $z_{ij}$, which means that $z_{ij}$ is behaves like a uniform random variable independent of all other variables, which is what the simulator generates. For $s_{ij}$, the same argument applies: if the simulator cheats, then $r_{ij}$ is only observable through $s_{ij}$ (or probes depending on $s_{ij}$), hence $s_{ij}$ behaves as a uniform independent random variable.

APPENDIX B
MIMO-SNI S-BOX OPTIMIZATION PROBLEM

The AES S-Box of is made of three parts: a top linear transformation, a middle non-linear transformation and
need: for any node \( u \), let \( J \) be the set of paths from any input node to \( u \), \( \sum_{j \in J} p_j \geq |J| - 1 \). Finally, for the MO part we need: for any node \( u \), let \( J \) be the set of paths from \( u \) to any output node, \( \sum_{j \in J} p_j \geq |J| - 1 \).

The objective function to be minimized is a weighted sum of the \( e_i \) variables. The weight associated to each variable is the cost of adding a refresh on the corresponding edge. This cost can be any metric, such as the amount of random bits required, the computation time, etc. Since each edge has a distinct associated cost parameter, this is the point where we can take into account that the cost of adding a SNI refresh gadget may not be the same as replacing a NI multiplication with a SNI multiplication.

This simple way of writing our problem has two limitations. First, there are many paths in the computation graph (in the order of magnitude of \( 10^6 \) for the AES S-box) which leads to many variables and constraints in the optimization problem. This can be mitigated by grouping paths into clusters that share common parts and associating them to a single variable.

The second issue is related to split nodes: there are multiple trees of binary split nodes that represent the split of a value in more than two parts, and all these representations do not give equivalent possibilities for inserting refresh elements. Furthermore, no tree can provide all the optimization degrees of freedom. Since it would be impractical to run the optimization for all the possible trees, we instead modified the optimization problem. Each split node is replaced by a set of split nodes that form a fully connected DAG and constraints are set to ensure that a constant number of added edges is cut, which ensures that the added edges do not distort the objective function.

### Appendix C

#### Cost of Masked Implementations

The randomness and field operations cost for SNI refresh, NI & SNI multiplication, and PINI multiplication gadgets are given in Table II for some small orders. The cost formula for any order are given by:

- **Randomness** (the formula for \( R_{\text{ref}}(d) \) is \( O(d \log d) \)):
  \[
  R_{\text{ref}}^S(d) = 2 \lceil d/2 \rceil + R_{\text{mul}}^S(\lceil d/2 \rceil) + R_{\text{ref}}^S(\lceil d/2 \rceil)
  \]
  \[
  R_{\text{mul}}^S(d) = \lceil (d-1)^2/4 \rceil + d - 1
  \]
  \[
  R_{\text{ref}}^M(d) = \min\{d(d-1)/2, R_{\text{mul}}^M(d) + R_{\text{ref}}^S(d)\}
  \]
  \[
  R_{\text{mul}}^M(d) = (d-1)/2
  \]

- **Additions**:
  \[
  A_{\text{ref}}^S(d) = 2R_{\text{ref}}^S(d)
  \]
  \[
  A_{\text{mul}}^S(d) = 2R_{\text{mul}}^S(d) + d(d-1)
  \]
  \[
  A_{\text{ref}}^M(d) = 2R_{\text{ref}}^M(d) + d(d-1)
  \]
  \[
  A_{\text{mul}}^M(d) = 2R_{\text{mul}}^M(d) + 2d(d-1) + d
  \]
  \[
  A_{\text{mul}}^P(d) = 2R_{\text{mul}}^P(d) + 4d(d-1)
  \]

- **Multiplications**:
  \[
  M_{\text{ref}}^{\text{PINI}} = 0
  \]
  \[
  M_{\text{mul}}^{\text{PINI}} = d^2
  \]
  \[
  M_{\text{ref}}^{\text{PINI}} = d^2
  \]
  \[
  M_{\text{mul}}^{\text{PINI}} = 2d(2d-1) + d
  \]
  \[
  M_{\text{ref}}^{\text{PINI}} = 2d(2d-1) + d
  \]

The cost for whole encryptions are given in Tables III, IV and VII. The cost metric used is explained in Section VII.

### Table II: Randomness and field operations cost of SNI gadgets.

<table>
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<tr>
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<th>All-SNI</th>
<th>TPC</th>
<th>MIMO-SNI</th>
<th>Double-SNI</th>
<th>PINI</th>
<th>PINI</th>
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### Table III: Cost metric for AES 128 bit encryption.

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### Table IV: Cost metric for Noekeon encryption.

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### Table V: Cost metric for PRESENT 80 bit encryption.

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<th>PINI</th>
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### Table VI: Cost metric for Fantomas encryption.