Implementing Cryptographic Program Obfuscation

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Abstract

Program obfuscation is the process of making a program "unintelligible" without changing the program's underlying input/output behavior. Although there is a long line of work on heuristic techniques for obfuscation, such approaches do not provide any cryptographic guarantee on their effectiveness. A recent result by Garg et al. (FOCS 2013), however, shows that cryptographic program obfuscation is indeed possible based on a new primitive called a *graded encoding scheme*.

In this work, we present the first implementation of such an obfuscator. We describe several challenges and optimizations we made along the way, present a detailed evaluation of our implementation, and discuss research problems that need to be addressed before such obfuscators can be used in practice.

1 Introduction

The goal of *program obfuscation* is to make programs "unintelligible" while preserving their original behavior. Formally, an obfuscator $\mathcal{O}$ is a compiler taking as input a program f (expressed, say, in C code or as a Boolean circuit) and outputting an obfuscated program $f' = \mathcal{O}(f)$ such that $f'(x) = f(x)$ for all inputs x. Intuitively, the security requirement is that the obfuscated code f' reveals nothing about the internal structure of the original program f other than what can be inferred from its input/output behavior and (a bound on) its size/running time.

For decades, program obfuscation has been suggested for, e.g., protecting intellectual property [23] or avoiding static or dynamic code analysis [22, 28, 30]. Generally, researchers have approached the problem by applying a series of static program transformations, hoping to obtain an incomprehensible (but equivalent) version of the original program. Unfortunately, such approaches do not offer any cryptographic guarantee of security. However, the recent landmark result of Garg et al. [15] demonstrates that cryptographic program obfuscation can be achieved, assuming a *graded encoding scheme*. Since then, many papers have been written improving various aspects of the original construction [3, 5, 10, 24] and utilizing the technique for many interesting applications [9, 14, 19, 25].

The construction of Garg et al. [15] satisfies a weaker notion of obfuscation than the *virtual black-box obfuscation* notion described intuitively above (which is known to be impossible to achieve in general [6]). This weaker notion, called *indistinguishability obfuscation*, ensures that for any two equivalent programs f and g (i.e., programs such that $f(x) = g(x)$ for all x), an obfuscation of f is indistinguishable (in a cryptographic sense) from an obfuscation of g. While weaker than virtual

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black-box obfuscation, indistinguishability obfuscation is surprisingly powerful. Applications of indistinguishability obfuscation include deniable encryption [25], efficient traitor tracing [9], two-round multiparty computation with succinct messages [14], full domain hash without a random oracle [19], and more. In addition, many obfuscation constructions have been shown to be virtual black-box obfuscators in a generic model [3, 5, 10], and it appears that such constructions may indeed be virtual black-box obfuscators for particular classes of functions, such as *evasive* functions [4].

Yet despite the fast growing body of literature on cryptographic program obfuscation, the idea has thus far remained entirely theoretical. It was not even clear whether these new constructions are even close to practically feasible.

Contributions. This work investigates the practicality of cryptographic program obfuscation through implementation and experimentation.

- We implement a full program-obfuscation toolchain for the first time, based on schemes from the literature [5, 10] and using the formula-to-branching-program construction of Ananth et al. [3]; see Section 2 and Section 3. Our code is publicly available for the benefit of the research community.[1]
- We investigate the *practicality* of program obfuscation through experiments; see Section 4. We find that program obfuscation is still far from being deployable, with the most complex functionality we are able to obfuscate being a 16-bit point function (i.e., a function that outputs 1 if the input matches some secret value, and 0 otherwise).
- We discuss the implications of our work and the bottlenecks that researchers should focus on to make obfuscation more practical; see Section 5.

2 Overview

In this section, we provide a self-contained overview of the techniques underlying cryptographic program obfuscation and our implementation. Our obfuscator $\mathcal{O}$ takes as input a program represented as a *Boolean formula*. (We only implement "basic" obfuscation and not the "bootstrapping" step which requires obfuscating an FHE logic implemented purely by Boolean *formulas* to obfuscate generic Boolean *circuits*. Note that any circuit can be expressed as a formula; although this results in exponential blowup asymptotically, for moderately-sized circuits converting the circuit to a formula and using basic obfuscation should be more efficient than applying bootstrapping to the original circuit.) A Boolean formula is a binary tree where each vertex is labeled with some Boolean gate type — in our implementation, AND, NOT, or XOR — and each leaf is labeled with some input bit x_i for $i \in \{1, \ldots, n\}$; the root of the tree corresponds to the (1-bit) output.[2] The *depth* d of a formula is the number of vertices on the longest path from any leaf to the root, and the *size* s of a formula is the number of vertices in the entire tree.

Given a Boolean formula f, our obfuscator proceeds through a sequence of transformations (see Figure 1) before producing the final obfuscated program $\mathcal{O}(f)$. We give a general outline of these steps before giving further details:

[1] https://github.com/amaloz/ind-obfuscation

[2] In order to obfuscate a function with k output bits one would need to construct k Boolean formulas, with the ith formula computing the ith output bit.

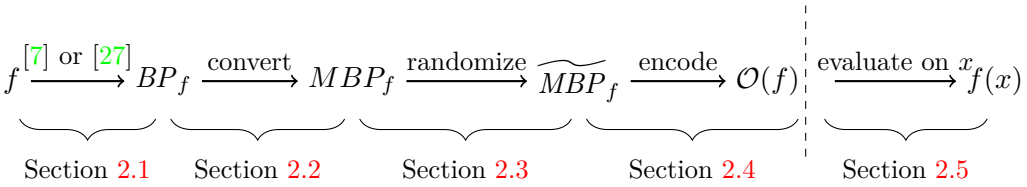

Figure 1: Workflow diagram for obfuscation. Starting from a Boolean formula f, we first convert f to a *branching program* BP_f using one of two methods. We then convert this to a *matrix branching program* MBP_f. Next, MBP_f is randomized to give $\widetilde{MBP}_f$. Finally, each matrix element in $\widetilde{MBP}_f$ is encoded using a graded encoding scheme. The output of this procedure is an obfuscation $\mathcal{O}(f)$ of f. Any party given $\mathcal{O}(f)$ can evaluate it on any input x of their choice to compute $\mathcal{O}(f)(x) = f(x)$.

1. The Boolean formula f is converted into a functionally equivalent *branching program* BP_f, written as a directed, acyclic graph (Section 2.1.1). (Branching programs are defined below.)

2. The branching program BP_f is mapped to a *matrix branching program* MBP_f, which consists of a sequence of pairs of matrices (Section 2.2).

3. The matrix branching program MBP_f is *randomized* to give $\widetilde{MBP}_f$ (Section 2.3). Intuitively, this step (in combination with the next step) prevents an attacker from mixing-and-matching different pieces of the computation together.

4. The randomized matrix branching program $\widetilde{MBP}_f$ is then encoded, matrix element by matrix element, using a graded encoding scheme (Section 2.4). This step can be viewed, naively, as encrypting MBP_f using a scheme that enables homomorphic evaluation of MBP_f, and is the only step that relies on a cryptographic hardness assumption. It has the effect of hiding from the attacker all details of the matrix branching program as well any information about the computation being performed (other than the final output).

We now describe each of these steps in more detail. In order to illustrate the concepts, we use the running example of obfuscating a single AND gate, i.e., the Boolean formula $f(x_1, x_2) = x_1 \wedge x_2$.

Notation. We let λ denote the security parameter. That is, an obfuscation with security parameter λ is designed to bound the probability of successful attacks by $2^{-\lambda}$. We let $[k] = \{1, \ldots, k\}$, use bold letters to denote vectors, and use capital letters to denote matrices.

2.1 Branching Programs

Let $f : \{0,1\}^n \to \{0,1\}$ be some Boolean formula, and let $x \in \{0,1\}^n$ be the input with x_i denoting the ith bit of x. A *branching program* is a directed acyclic graph whose vertices are partitioned into disjoint layers, defined as follows:

Definition 1 *A **branching program** of width w and length m is a directed acyclic graph with s vertices, such that the following constraints hold:*

- *Each vertex is either a* source *vertex, an* internal *vertex or a* final *vertex.*
- *Each non-final vertex is labeled with some $\ell \in \{`x_1', \ldots, `x_n'\}$ and has out-degree two, with one of the outgoing edges labeled '0' and the other '1'.*

- *There are two final vertices; both have out-degree zero and one is labeled '0' and the other '1'.*
- *There is a unique source vertex with in-degree zero.*
- *The vertices are partitioned into m layers L_j, where $|L_j| \leq w$. All vertices in a layer have the same label 'x_i'.*
- *Edges starting in layer L_j end in some layer $L_{j'}$ with $j' > j$.*
- *Both final vertices are in the same layer, L_m.*

Note that the notation 'x_i' stresses it is the *symbol*, rather than the bit value represented by the symbol, which is associated with a vertex.

We can evaluate a branching program as follows. On input $x = x_1 \cdots x_n$, start from the source node in the branching program and traverse the edges corresponding to the Boolean value of the current node's label. For example, suppose the source node has label 'x_j'. Then, on an input x where $x_j = 1$, we would take the outgoing edge labeled '1'. The 0/1 label of the final vertex is the output of the program.

2.1.1 From Formulas to Branching Programs

We investigate two approaches for compiling Boolean formulas into branching programs: using Barrington's theorem [7] (as done in the work of Garg et al. [15], among others), or using the approach of Sauerhoff et al. [27] (as proposed in the work of Ananth et al. [3]).[3] Barrington's theorem shows how to convert any Boolean formula of depth d into a branching program of width 5 and length at most 4^d, whereas the approach of Sauerhoff et al. converts formulas of size s into branching programs of width at most $2s + 4$ and length at most s. However, the approach of Sauerhoff et al. is more efficient both asymptotically [3] and in practice, and thus we focus on this approach in what follows.[4]

The transformation of Sauerhoff et al. from formulas to branching programs is inductive. The base case is a "trivial" branching program—one for each input wire—consisting of only three vertices: the source node, the reject node, and the accept node. There is a directed edge labeled '1' from source to accept, and a directed edge labeled '0' from source to reject.

We then proceed inductively. Given a branching program BP_f for some formula f, we can construct the branching program for the formula $\neg f$ by swapping the labels of the accept and reject nodes in BP_f.

Given two branching programs BP_f and BP_g for the formulas f and g, we can "AND" the two formulas, producing the branching program $BP_{f \wedge g}$, as follows: First, we merge the accept node of BP_f with the source node of BP_g. Second, we merge the reject node of BP_f with the reject node of BP_g. Finally, we let the source node of BP_f be the source node of $BP_{f \wedge g}$, and let the accept and reject nodes of BP_g be the accept and reject nodes, respectively, of $BP_{f \wedge g}$. Figure 2 shows the branching program obtained by applying this approach to a single AND gate.

Lastly, given two branching programs BP_f and BP_g for the formulas f and g, we can "XOR" the two formulas, producing the branching program $BP_{f \oplus g}$ as follows: Without loss of generality, assume BP_g is the smaller of the two branching programs (otherwise we can swap the order of f and g). We first duplicate BP_g and use the NOT-transformation to produce $BP_{\neg g}$. Next, we merge

[3] Recently, two new approaches [26, 31] to constructing obfuscators present asymptotic improvements to the approaches discussed in this work. In addition, later versions of the work of Ananth et al. [3] propose an alternative approach than that of Sauerhoff et al. We leave implementations of these new approaches as future work.

[4] Note that our implementation contains both approaches.

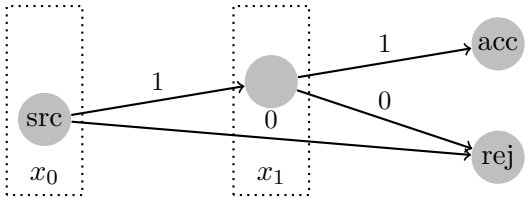

Figure 2: Branching program for an AND gate. To evaluate on input x_0x_1, look at each layer of the graph (denoted by the dotted rectangles). If the layer corresponds to the ith input bit, then remove all outgoing edges from that layer that are not labeled by the value x_i. The output is 1 iff there is a path from src to acc in the resulting graph.

the accept node of BP_f with the source node of $BP_{\neg g}$ and merge the reject node of BP_f with the source node of BP_g. Finally, we merge the accept nodes (and, respectively, the reject nodes) of BP_g and $BP_{\neg g}$. The branching program $BP_{f \oplus g}$ has the same source node as BP_f and the same accept and reject nodes as BP_g (equivalently, $BP_{\neg g}$).

2.2 Matrix Branching Programs

The next step is to compile the graph-based branching program BP_f into a functionally equivalent matrix branching program MBP_f, which can be evaluated by iterated matrix multiplication. We utilize the notion of a matrix branching program, defined as follows:

Definition 2 *A **matrix branching program** of width w and length m for n-bit inputs is given by a tuple*

$$MBP_f = \left(\mathsf{inp}, \mathbf{s}, (B_{j,0}, B_{j,1})_{j \in [m]}, \mathbf{t}\right)$$

where $B_{j,b} \in \{0,1\}^{w \times w}$, for $j \in [m]$ and $b \in \{0,1\}$, $\mathsf{inp} : [m] \to [n]$ is a function mapping layer indices j to input-bit indices i, $\mathbf{s} := (1, 0, \ldots, 0)$, and $\mathbf{t} := (0, \ldots, 0, 1)^T$. The output of MBP_f on input $x \in \{0,1\}^n$ is defined as follows:

$$MBP_f(x) = \mathbf{s} \cdot \left(\prod_{j=1}^{m} B_{j, x_{\mathsf{inp}(j)}} \right) \cdot \mathbf{t} \in \{0,1\}.$$

Note that the output of a matrix branching program is the 0/1 entry in the first row and last column of the iterated matrix product, and the vectors $\mathbf{s}$ and $\mathbf{t}$ simply select this entry as the output.

Thus, for each layer L_j of BP_f, we produce two w-by-w 0/1-valued, integral matrices $B_{j,0}, B_{j,1}$. Specifically, the matrix $B_{j,b}$ is the adjacency matrix composed of edges leaving layer L_j with label b. Additionally, the diagonal entries are marked 1 so that $B_{j,b}$ is a full-rank matrix [3], which is important for the subsequent randomization step.

In Figure 3 we show the matrix branching program generated from the graph in Figure 2.

2.2.1 Oblivious Branching Programs

Depending on the setting, it may also be necessary to ensure that the inp function does not leak information about the underlying formula f. (This is needed in order to satisfy the formal definition of obfuscation. But in settings where some information about f is known anyway—e.g., when f is

$$\left(\underbrace{\begin{bmatrix}1\\0\\0\\0\end{bmatrix}}_{s}, \underbrace{\left(\begin{bmatrix}1&0&1&0\\0&1&0&0\\0&0&1&0\\0&0&0&1\end{bmatrix}, \begin{bmatrix}1&1&0&0\\0&1&0&0\\0&0&1&0\\0&0&0&1\end{bmatrix} \right)}_{x_0}, \underbrace{\left(\begin{bmatrix}1&0&0&0\\0&1&1&0\\0&0&1&0\\0&0&0&1\end{bmatrix}, \begin{bmatrix}1&0&0&0\\0&1&0&1\\0&0&1&0\\0&0&0&1\end{bmatrix} \right)}_{x_1}, \underbrace{\begin{bmatrix}0\\0\\0\\1\end{bmatrix}}_{t} \right)$$

Figure 3: Matrix branching program for an AND gate. To evaluate on input x_0x_1, proceed as follows. If $x_i = 0$, choose the first matrix; otherwise choose the second matrix. Multiply the matrices; if the $[1, 4]$ entry of the resulting matrix is 1 then the output is 1; otherwise, the output is 0.

known to be a point function, and only the hidden point must remain secret—this step may not be needed.) This can be done by fixing inp independently of f, setting inp to cycle through each of the input bits $x_1, x_2, \ldots, x_n$ in order, m times. "Dummy" layers are filled in with a pair of identity matrices, thus preserving correctness of the iterated matrix multiplication.

Note, however, that when using the approach of Sauerhoff et al., the width w is a function of the *type* of gates used. For example, a standalone XOR gate under Sauerhoff et al.'s approach equates to a width 5 matrix branching program whereas a standalone AND gate equates to a width 4 matrix branching program [27]. Thus, we need to pad the graph with "dummy" vertices *before* transforming the graph into a matrix branching program to prevent the attacker from learning information about what gate is being computed by simply looking at the width of the matrix branching program.

2.3 Randomized Matrix Branching Programs

The first step toward hiding the underlying computation is to *randomize* the matrices [20]. This randomization ensures that an attacker cannot "mix-and-match" steps of the computation once the matrices have been encoded under the graded encoding scheme.

Let g be a prime, and define the procedure $\mathsf{rand}_g(MBP_f)$ as follows:

1. Uniformly sample $m + 1$ invertible matrices from $\mathbb{Z}_g^{w \times w}$, denoted $R_0, \ldots, R_m$.
2. Let $\widetilde{B}_{j,b} := R_{j-1} B_{j,b} R_j^{-1}$ for all $j \in [m], b \in \{0, 1\}$.
3. Compute two "bookend" vectors $\widetilde{\mathbf{s}} := (1, 0, \ldots, 0) \cdot R_0^{-1}$ and $\widetilde{\mathbf{t}} := R_m \cdot (0, \ldots, 0, 1)^T$.
4. Output $\widetilde{MBP}_{f,g} = \left(\mathsf{inp}, \widetilde{\mathbf{s}}, \left(\widetilde{B}_{j,0}, \widetilde{B}_{j,1} \right)_{j \in [m]}, \widetilde{\mathbf{t}} \right)$.

Indeed, we have that for all inputs $x \in \{0, 1\}^n$,

$$\begin{aligned} \widetilde{MBP}_{f,g}(x) &= \widetilde{\mathbf{s}} \cdot \left(\prod_{j=1}^{m} \widetilde{B}_{j, x_{\mathsf{inp}(j)}} \right) \cdot \widetilde{\mathbf{t}} = \mathbf{s} \cdot R_0^{-1} \cdot \left(\prod_{j=1}^{m} R_{j-1} B_{j,b} R_j^{-1} \right) \cdot R_m \cdot \mathbf{t} \\ &= \mathbf{s} \cdot \left(\prod_{j=1}^{m} B_{j, x_{\mathsf{inp}(j)}} \right) \cdot \mathbf{t} = MBP_f(x), \end{aligned}$$

where matrix multiplication arithmetic is performed modulo g. For notational convenience, we drop the g subscript and write $\widetilde{MBP}_f$ when the modulus is understood.

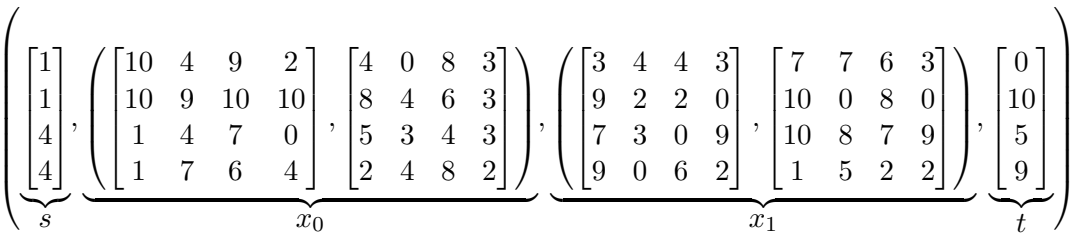

Figure 4: Randomized (with prime modulus 11) matrix branching program for an AND gate. Evaluation happens as in the non-randomized setting (cf. Figure 3), except all operations are done modulo 11.

Figure 4 shows the randomization of the matrix branching program (from Figure 3) for an AND gate, using "toy" prime modulus 11.

2.4 Graded Encoding Schemes

While the prior randomization step ensures that a valid sequence of matrices must be multiplied in order, we still need to ensure that (1) only valid sequences of matrices give meaningful output, and (2) the matrix values themselves are hidden—or more generally, that inspecting the obfuscated program does not leak distinguishing information about the underlying implementation. Concretely, observe that randomized matrix branching programs can suffer from attacks that try to compute something other than the original function $f(x)$. For example, if two layers L_j and $L_{j'}$ of $\widetilde{MBP}_f$ are labeled with the same input bit x_i, it is possible to include $\widetilde{B}_{j,0}$ and $\widetilde{B}_{j',1}$ in the product and obtain some output where the random R_j terms still cancel out. This equates to computing a function f' that is different from f, which enables the attacker to do things that are impossible merely from access to f.

To thwart such attacks, we need to use some cryptographic mechanism to hide the entries of the matrices and to prevent an attacker from using inconsistent matrices for some input bit. The cryptographic primitive used is a *graded encoding scheme* [13].

In a graded encoding scheme, producing an encoding of a plaintext message is a private method that can only be performed by a party that knows the scheme's secret parameters. But, similar to fully homomorphic encryption (FHE), any party, given an encoding u of message m and an encoding u' of a message m', can add $u + u'$ to get a new encoding of the message $m + m'$. Similarly, multiplying encodings $u \cdot u'$ gives a new encoding of the message $m \cdot m'$.

Unlike FHE, however, each encoding u is defined relative to some subset S of a universe $[Z]$. These subsets *restrict* the way in which encodings may be added and multiplied. In particular, two encodings may only be added if they are encoded relative to the *same* set S, and their sum is encoded relative to S as well. Moreover, two encodings may only be multiplied if they are encoded relative to two *disjoint* sets S and S', and their product is encoded relative to the *union* of S and S'.

Finally, and again in contrast to FHE, there is a public "zero-test parameter" p_{zt} that is used to test if an encoding u encodes the plaintext 0. However, zero-testing only works on encodings relative to the *entire universe* $[Z]$.

Definition 3 *A **graded encoding scheme** is a tuple of probabilistic polynomial time algorithms* (InstGen, Enc, Add, Mult, isZero) *that behave as follows:*

- ***Instance Generation:*** $(\mathsf{sp}, \mathsf{pp}) \leftarrow \mathsf{InstGen}(1^\lambda, 1^\kappa)$**.**
  $\mathsf{InstGen}$ *takes as input the security parameter* λ *and multilinearity parameter* κ*, and outputs secret parameters* sp *and public parameters* pp*. The secret parameters* sp *contain an integer* Z *with* $\kappa \leq Z \leq 2\kappa$*, primes* $g_1, \ldots, g_N$*, where* $N = poly(\lambda, \kappa)$*, and a collection of sets* $\{E_S^m : m \in \mathbb{Z}_{g_1} \times \cdots \times \mathbb{Z}_{g_N}, S \subseteq [Z]\}$*. We view* E_S^m *as the set of possible encodings of the value* m *with respect to the set* S*. The public parameters* pp *enable the public operations on encodings described below.*

- ***Encoding:*** $u \leftarrow \mathsf{Enc}(\mathsf{sp}, m, S)$**.**
  Enc *takes as input the secret parameters* sp*, a plaintext value* $m \in \mathbb{Z}_{g_1} \times \cdots \times \mathbb{Z}_{g_N}$*, and a set* $S \subseteq [Z]$*, and outputs a randomized encoding of* m *with respect to the set* S*, denoted* $u \in E_S^m$*.*

- ***Addition:*** $u \leftarrow \mathsf{Add}(\mathsf{pp}, u, u')$**.**
  Add *takes as input the public parameters* pp *and encodings* $u \in E_S^m, u' \in E_{S'}^{m'}$*, and outputs an encoding* $u \in E_S^{m+m'}$ *if* $S = S'$*, and* $\perp$ *otherwise.*

- ***Multiplication:*** $u \leftarrow \mathsf{Mult}(\mathsf{pp}, u, u')$**.**
  Mult *takes as input the public parameters* pp *and encodings* $u \in E_S^m, u' \in E_{S'}^{m'}$*, and outputs an encoding* $u \in E_{S \cup S'}^{m \cdot m'}$ *if* $S \cap S' = \emptyset$*, and* $\perp$ *otherwise.*

- ***Zero Test:*** $b \leftarrow \mathsf{isZero}(\mathsf{pp}, u)$**.**
  isZero *takes as input the public parameters* pp *and an encoding* u *and outputs 1 if* $u \in E_{[Z]}^{\boldsymbol{0}}$ *(i.e.,* u *is an encoding of* $\boldsymbol{0}$ *with respect to the entire universe* $[Z]$*), and 0 otherwise.*

We defer a discussion of security until later, but informally, a graded encoding scheme is secure if (1) the only way an evaluator can produce an encoding u^* of the all-zeros plaintext with respect to the entire universe $[Z]$ is by combining *given* encodings $\{u_1, u_2, \ldots\}$ by addition and multiplication, and (2) only such encodings u^* pass the zero-test.

2.4.1 Choosing a Set System

It remains to describe the system of sets S under which each matrix element of $\widetilde{MBP}_f$ is encoded in order to produce $\mathcal{O}(f)$. Intuitively, we want to leverage the fact that zero-testing requires an element encoded with respect to the *entire* universe $[Z]$ to force the evaluator of $\mathcal{O}(f)$ to commit to a fixed setting of each input bit $x_i \in \{0, 1\}$. In particular, we ensure that only combinations of matrices in $\widetilde{MBP}_f$ that are consistent with some well-defined input x produce an encoding that forms an exact set cover of $[Z]$. Thus, we associate each matrix of $\widetilde{MBP}_f$ with a distinct subset $S \subset [Z]$. This means that for any fixed matrix of $\widetilde{MBP}_f$, we encode each of its individual elements using the same set S. The vectors $\mathbf{s}$ and $\mathbf{t}$ are each encoded using a distinct element in $[Z]$.

More concretely, we group the matrices of $\widetilde{MBP}_f$ according to their input bit label 'x_i'. For example, denote the matrices associated with x_1 as $\widetilde{B}_{j_1,0}, \widetilde{B}_{j_1,1}, \ldots, \widetilde{B}_{j_k,0}, \widetilde{B}_{j_k,1}$ (for some k). We assign sets S to these matrices as follows:

- We assign the matrices $\widetilde{B}_{j_1,0}, \widetilde{B}_{j_2,0}, \ldots, \widetilde{B}_{j_k,0}$ the sets $\{1, 2\}, \{3, 4\}, \ldots, \{2k-3, 2k-2\}, \{2k-1\}$ (in that order).
- We assign the matrices $\widetilde{B}_{j_1,1}, \widetilde{B}_{j_2,1}, \ldots, \widetilde{B}_{j_k,1}$ the sets $\{1\}$, $\{2, 3\}$, $\ldots$, $\{2k-2, 2k-1\}$ (in that order).

We continue this process for each such group of matrices (associated with the remaining labels 'x_2', ..., 'x_n') using increased set-indices, until every matrix is assigned a distinct subset $S \subset [Z]$. This process allows us to determine the value of Z needed in our implementation, which we increase by two to account for encoding the vectors $\mathbf{s}$ and $\mathbf{t}$.

Note the attacker cannot use inconsistent values for a given variable since sets corresponding to the inconsistent values are not disjoint (as valid multiplications require disjoint sets). The attacker cannot run a zero-test on a partially evaluated program either, because the encoding of a partial result is always formed with respect to a strict subset of Z whereas zero-testing only works for encodings regarding the (full) set Z.

2.5 Executing Obfuscated Programs

Putting everything together, an evaluator can use the Add and Mult procedures of the graded encoding scheme to evaluate the iterated matrix multiplication of an underlying, randomized matrix branching program $\widetilde{MBP}_f$ on some input x, retrieving $\mathsf{Enc}(\mathbf{s} \cdot \prod_{j=1}^{m} B_{j,x_{\mathsf{inp}(j)}} \cdot \mathbf{t}) = u$. The evaluator can then use isZero to zero-test this result; namely, if $\mathsf{isZero}(\mathsf{pp}, u) = 1$ then $f(x) = 0$, otherwise $f(x) = 1$.

3 Implementation

We have implemented an obfuscator using both Barrington's theorem [5] and the approach of Sauerhoff et al. [3] for converting formulas to branching programs. Our implementation is written in a combination of Python and C, using Sage [29] for algebraic operations, the GNU Multiple Precision Arithmetic Library [1] for efficient large number computations, and OpenMP [2] for parallelization. We program all the computationally expensive steps in C, and thus our use of Python does not significantly affect the overall performance.[5]

As shown in Figure 1, our tool takes as input a Boolean formula f and produces as output its obfuscated form $\mathcal{O}(f)$, which can then be evaluated by anyone who receives a copy of it. In the rest of this section, we provide details about our implementation and discuss some challenges encountered during implementation along with the various optimizations we made to address them.

3.1 Implementing Graded Encodings

We utilize the graded encoding scheme of Coron et al. [12] (the "CLT scheme"), which has better space efficiency than the scheme based on ideal lattices [13]. We note that a recent result constructs a graded encoding scheme from lattices using an assumption similar to the Learning With Errors assumption [16]. We leave implementing and optimizing obfuscation using this new scheme as interesting future work.

The CLT graded encoding scheme is instantiated as follows. (We discuss the concrete values of all parameters in Section 3.1.3.)

- $\mathsf{InstGen}$: Let Z be the desired size of the set system (cf. Section 2.4.1), and N a value that depends on Z and the security parameter λ (see Section 3.1.3). First, we generate

[5] For example, when obfuscating and evaluating $f(x_0, x_1) = x_0 \wedge x_1$ with security parameter 128 we find that 98.5% of the execution time is spent executing C code. This percentage only increases as the size of the Boolean formula increases.

N (large) secret primes $\{p_i\}_{i=1}^N$ and compute their product $q = \prod_{i=1}^N p_i$. In addition, we generate N (small) secret primes $\{g_i\}_{i=1}^N$, N random integers $\{h_i\}_{i=1}^N$, and Z random values $z_1, \ldots, z_Z \in \mathbb{Z}_q$.

Next, we construct a zero-test parameter, defined as the integer[6]

$$p_{zt} = \sum_{i=1}^{N} h_i \cdot \left(\prod_{j=1}^{Z} z_j \cdot g_i^{-1} \bmod p_i \right) \cdot \prod_{i' \neq i} p_{i'} \bmod q.$$

The secret parameters are $\left(\{z_i\}_{i=1}^Z, \{g_i\}_{i=1}^N, \{p_i\}_{i=1}^N\right)$; the public parameters are (p_{zt}, q).

- Enc: An encoding of a message $m = (m_1, \ldots, m_N) \in \mathbb{Z}_{g_1} \times \cdots \times \mathbb{Z}_{g_N}$ relative to the set $S \subseteq [Z]$ is a value $u \in \mathbb{Z}_q$ such that for all $1 \leq i \leq N$,

$$u \equiv \frac{r_i \cdot g_i + m_i}{\prod_{j \in S} z_j} \pmod{p_i} \tag{1}$$

  for (small, random) integers r_i.

- Add: Given $u, u' \in \mathbb{Z}_q$ with

$$\forall i: \; u \equiv \frac{r_i \cdot g_i + m_i}{\prod_{j \in S} z_j} \pmod{p_i}, \quad u' \equiv \frac{r'_i \cdot g_i + m'_i}{\prod_{j \in S} z_j} \pmod{p_i},$$

  it is easy to see that

$$\forall i: \; u + u' \equiv \frac{(r_i + r'_i) \cdot g_i + (m_i + m'_i)}{\prod_{j \in S} z_j} \pmod{p_i},$$

  which lies in $E_S^{m+m'}$ assuming $(r_i + r'_i) \cdot g_i + (m_i + m'_i) < p_i$ for all i.

- Mult: Let S and S' be sets such that $S \cap S' = \emptyset$. Given $u, u' \in \mathbb{Z}_q$ with

$$\forall i: \; u \equiv \frac{r_i \cdot g_i + m_i}{\prod_{j \in S} z_j} \pmod{p_i}, \quad u' \equiv \frac{r'_i \cdot g_i + m'_i}{\prod_{j \in S'} z_j} \pmod{p_i}$$

  it is easy to see that

$$\forall i: \; u \cdot u' \equiv \frac{(r_i r'_i + r_i m'_i + r'_i m_i) \cdot g_i + m_i m'_i}{\prod_{j \in S \cup S'} z_j} \pmod{p_i},$$

  which lies in $E_{S \cup S'}^{m \cdot m'}$ assuming $(r_i r'_i + r_i m'_i + r'_i m_i) \cdot g_i + m_i m'_i < p_i$ for all i.

- isZero: Using p_{zt}, we zero-test an element u encoded with respect to $[Z]$ by first computing

$$\omega := p_{zt} \cdot u \pmod{q}$$

[6]Note that here we are utilizing the heuristic as presented by Coron et al. [12, §6] of using a single zero-testing element p_{zt} rather than a vector of elements.

where

$$
\begin{aligned}
p_{zt} \cdot u = &\left(\sum_{i=1}^{N} h_i \cdot \left(\prod_{j=1}^{Z} z_j \cdot g_i^{-1} \bmod p_i \right) \cdot \prod_{i' \neq i} p_{i'} \right) \\
&\cdot \left(\sum_{i=1}^{N} (r_i \cdot g_i + m_i) \cdot \left(\prod_{j=1}^{Z} z_j^{-1} \bmod p_i \right) \cdot \left(\prod_{i' \neq i} p_{i'} (p_{i'}^{-1} \bmod p_i) \right) \right) \pmod q \\
= &\sum_{i=1}^{N} h_i \cdot \left(\prod_{j=1}^{Z} z_j \cdot g_i^{-1} \bmod p_i \right) \cdot (r_i \cdot g_i + m_i) \cdot \left(\prod_{j=1}^{Z} z_j^{-1} \bmod p_i \right) \cdot \prod_{i' \neq i} p_{i'} \pmod q \\
= &\sum_{i=1}^{N} h_i \cdot \Big(r_i + m_i \cdot (g_i^{-1} \bmod p_i) \Big) \cdot \prod_{i' \neq i} p_{i'} \pmod q.
\end{aligned}
$$

Then, we test whether $\omega \in \mathbb{Z}$ is "small" compared to q or not. (Concretely, ω is small if $|\omega| < q \cdot 2^{-\nu-\lambda-2}$.) In particular, ω is small if $m_i = 0$ for all $i \in [N]$, as otherwise for some i, the relatively large $g_i^{-1} \bmod p_i$ term does not vanish in the final expression.

3.1.1 Using the CLT Graded Encoding Scheme

We make use of the CLT graded encoding scheme as follows:

1. We first construct a (non-randomized) matrix branching program MBP_f. Using this we can derive the multilinearity κ of the graded encoding scheme, and thus run $\mathsf{InstGen}$ to generate the public and secret parameters.

2. Using the g_is generated in the previous step, we can now construct λ randomized matrix branching programs $\widetilde{MBP}_{f,g_1}, \ldots, \widetilde{MBP}_{f,g_\lambda}$ corresponding to the first λ "(mod g_i)-slots" of the plaintext space (see Section 3.1.2 for an explanation of why we generate λ branching programs instead of, say, one).

3. Finally, we encode the matrices element by element, where each element is the vector corresponding to the same-position matrix element in the λ randomized matrix branching programs.

4. To evaluate on an input x, we multiply together the matrices (as specified by x and inp) along with the bookend vectors, and zero-test the resulting value.

3.1.2 Attack on a Naive Implementation

At first glance, it seems we only need to make use of *one* slot of the plaintext space rather than the λ number of slots mentioned above. That is, instead of constructing, say, λ randomized matrix branching programs, one for each prime g_i, it seems plausible to construct a *single* matrix branching program with respect to g_1 only. Then, we could encode an element $m_1 \in \mathbb{Z}_{g_1}$ as $m \stackrel{\text{def}}{=} (m_1, 0, \ldots, 0) \in \mathbb{Z}_{g_1} \times \cdots \times \mathbb{Z}_{g_N}$. While this approach preserves correctness, it proves to be insecure, in the sense that we can learn secret prime g_1 (although it is not clear how to "break" the obfuscation with this additional knowledge). We discuss the attack below. (We note that Gentry

et al. [18, Appendix B.6] encounter a similar vulnerability in CLT encodings in a different algebraic context.)

When utilizing only a single slot of the plaintext space, the output of the zero-test procedure on an element u encoding a non-zero message m is close to a somewhat small multiple of the secret prime g_1 (or technically, g_1^{-1}, although the effect is the same). It turns out that this isolates g_1 in a way that allows one to recover g_1 with high probability using the Euclidean-GCD algorithm (to solve the shortest vector problem exactly in two dimensions).

For an encoding u of the plaintext $m \stackrel{\text{def}}{=} (m_1, 0, \ldots, 0)$ with $m_1 \neq 0$, we have

$$u \equiv \frac{r_1 \cdot g_1 + m_1}{\prod_{j \in S} z_j} \pmod{p_1} \quad \text{and} \quad \forall i \neq 1, u \equiv \frac{r_i \cdot g_i}{\prod_{j \in S} z_j} \pmod{p_i}.$$

Thus upon zero-testing we recover

$$\begin{aligned}\omega &= \sum_{i=1}^{N} h_i \cdot \left(r_i + m_i \cdot (g_i^{-1} \bmod p_i)\right) \cdot \prod_{i' \neq i} p_{i'} \pmod{q} \\ &= \left(\sum_{i=1}^{N} h_i m_i (g_i^{-1} \bmod p_i) \prod_{i' \neq i} p_{i'}\right) + \left(\sum_{i=1}^{N} h_i r_i \prod_{i' \neq i} p_{i'}\right) \pmod{q}\end{aligned}$$

and since $m_i = 0$ for all $i \neq 1$, we have that

$$\omega = h_1 m_1 (g_1^{-1} \bmod p_1) \left(\prod_{i' \neq 1} p_{i'}\right) + \sum_{i=1}^{N} h_i r_i \prod_{i' \neq i} p_{i'} \pmod{q},$$

and although we do not have g_1, we know that

$$b \stackrel{\text{def}}{=} \omega \cdot g_1 = h_1 m_1 \left(\prod_{i' \neq 1} p_{i'}\right) + g_1 \sum_{i=1}^{N} h_i r_i \prod_{i' \neq i} p_{i'} \pmod{q}. \tag{2}$$

Note that this is quite close to the output of the zero-test on an element that in fact encodes 0. In particular, assuming that $h_1 m_1$ and $g_1 h_i r_i$ for all i are small relative to q, then b is small relative to q, allowing us to recover g_1 by applying the Euclidean-GCD algorithm.

Concretely, we first compute an integer Ω by sampling *fresh* values for the variables m_1 and h_i, r_i, g_i, p_i, for all i, and using these to compute an estimate of b as defined in Equation 2; we denote this estimated value as $\widehat{b}$. We then set $\Omega := \widehat{b}/\widehat{g_1}$, where $\widehat{g_1}$ is our freshly sampled guess at the value g_1 we are trying to recover. Finally, we compute a value $\omega = p_{zt} \cdot u$, where u is some non-zero encoding relative to the set $[Z]$ and p_{zt} is the (public) zero-test parameter.[7]

Now, consider the lattice $\mathcal{L}$ generated by the basis $\{(\Omega, \omega), (0, q)\}$, where q is the value given in the public parameters. The lattice $\mathcal{L}$ contains the vector $\vec{v} := (\Omega \cdot g_1, b)$ of length (roughly) at most $2\Omega \cdot g_1$. The determinant of $\mathcal{L}$ is $\Omega \cdot q$, which implies that all vectors in $\mathcal{L}$ not parallel to $\vec{v}$ must have length at least $\det(\mathcal{L})/||\vec{v}|| \geq (\Omega \cdot q)/(2\Omega \cdot g_1) = q/2g_1$. Then, when $\Omega < q/4g_1^2$, we have $2\Omega \cdot g_1 < q/2g_1$, which implies $\vec{v}$ is the unique, shortest vector in $\mathcal{L}$. Since the two

[7] We can compute u by choosing random encodings in each matrix and multiplying the values together; due to Kilian's randomization, with high probability this multiplication will not equal zero.

dimensional shortest vector problem can be solved exactly in polynomial time using the Euclidean-GCD algorithm, we recover $\vec{v} := (\Omega \cdot g_1, b)$ exactly. Dividing the (known) value Ω from the first coordinate of $\vec{v}$ yields the secret prime g_1, violating security of the graded encoding scheme.

To resolve this problem, we can run $\mathsf{rand}_{g_i}(MBP_f)$ for all $i \in [N]$, producing N independently randomized, same-size matrix branching programs, whose moduli match the plaintext slots of the graded encoding scheme. Then, for each matrix element of the original matrix branching program MBP_f, we produce a vector of integers $(m_1, \ldots, m_N) \in \mathbb{Z}_{g_1} \times \cdots \times \mathbb{Z}_{g_N}$ corresponding to the same-position matrix element in the N randomized matrix branching programs $\widetilde{MBP}_{f,g_i}$. Now, given an appropriate set system S as described in Section 2.4, we can use Equation 1 to encode the plaintext $(m_1, \ldots, m_N)$.

Alternatively, since the complexity of the attack presented above grows exponentially with the number of non-zero plaintext slots, we can choose to instead only fill λ-many slots and leave the remainder filled with 0's. This is, in fact, what we do in our implementation. The main impact on efficiency is the added cost of constructing λ randomized branching programs, which is negligible compared to the cost of encoding elements.

To validate the above attack, we implemented it and verified the ability to extract g_1 from obfuscated programs which use only one plaintext slot. As an example, running on a standard laptop we are able to recover g_1 from the obfuscation (utilizing only one slot of the plaintext space) of an AND gate with security parameter 48 in approximately 20 seconds.

3.1.3 Setting Parameters

The parameters given by Coron et al. [12] were chosen to support their particular use-case of multiparty key exchange, which requires a public mechanism for *re-randomizing* encodings. However, in the case of program obfuscation, there is no need to re-randomize. This allows us to optimize the parameters of the scheme to gain improved efficiency. That is, since Add and Mult give correct results as long as the growth of the noise terms r_i does not cause any of the respective numerators of an encoding to exceed the moduli p_i, we only need to set the size of the p_i sufficiently large to ensure we can perform all the operations required to evaluate the obfuscated program.

Similarly, we can bound ρ_f, the maximum size in bits of the noise terms r_i, at the maximum depth of multiplication (i.e., when κ initial encodings have been multiplied together) as $\rho_f = \kappa(2\lambda + 2)$. Using the parameters specified by Coron et al. would have given us $\rho_f \gg \kappa(5\lambda + 2)$. Reducing the size of ρ_f greatly improves efficiency, as it affects the size η of the primes p_i, which subsequently affects the vector dimension. Both of these parameters appear to be the main efficiency bottlenecks.

For completeness, we list all the parameters we use for the CLT scheme, and the reasoning behind our choices. Recall that λ denotes the security parameter. These settings are designed in order to achieve 2^λ security against known attacks.

- α (the bit-size of the primes g_i): λ, to prevent a brute-force attack on the plaintext space.
- β (the bit-size of the h_i values used to construct p_{zt}): λ, due to a GCD-based attack [12, §5.2].[8]
- ρ (the bit-size of the random r_i values): λ, to prevent a brute-force attack on the noise [11, 12].

[8] Recent work [21] demonstrates an attack on the zero-test parameter which necessitates increasing β to 4λ. However, that attack only applies to the full CLT scheme where there is a vector of zero-test parameters available to the attacker. As we (heuristically) only publish a single zero-test parameter, the presented attack does not apply in our setting.

κ	α	β	ρ	ρ_f	η	ν	N
10	52	52	52	1060	1276	115	7273
14	52	52	52	1484	1700	115	9690
18	52	52	52	1908	2124	115	12107

Table 5: Concrete parameter settings for various settings of κ. The security parameter λ is fixed at 52. Note that these are the parameters for the obfuscations presented in Table 7.

- η (the bit-size of the primes p_i): $\rho_f + \alpha + 2\beta + \lambda + 8$, due to [12, Lemma 3]. Recall that ρ_f is a bound on the noise after performing all the necessary multiplications.
- ν (for zero-testing; ω is "small" if the ν most-significant bits of ω are all zeroes): $\alpha + \beta + 11$, due to [12, Lemma 3].
- N (the vector dimension of the plaintext space): $\eta \log_2 \lambda$, to prevent an orthogonal lattice reduction attack [12, §5.1].

Table 5 presents concrete numbers for various settings of κ with a fixed security parameter of 52.

3.2 Security of Our Implementation

To date, the security of general-purpose obfuscators has been argued in three general ways:

1. By showing the virtual black-box (VBB) property in an ideal model [5, 10],
2. By an exponential number of "semantically-secure graded encoding" assumptions [8, 24], or
3. By a single assumption on multilinear groups (and CLT encodings' quality as an instantiation of multilinear groups) plus complexity leveraging [17].

An ideal model proof of VBB security for our scheme would follow that of Barak et al. [5], except that we do not use dual-input branching programs in our implementation (as an efficiency optimization). Consequently, we do not know how to directly prove Claim 5 in their work [5, §6], namely that, to use their terminology, profiles of all single-input elements are complete. Nonetheless, we observe that an ideal model proof of VBB security for our scheme should go through in a straightforward way assuming the Bounded Speedup Hypothesis [10].

In addition, as previously mentioned in Section 3.1.2, the resilience of our implementation of CLT graded encodings to cryptanalysis requires the same type of assumption as used by Gentry et al. [17]; namely, that it is safe to publish a set of encodings from which an attacker can reliably produce a zero-testable element that is non-zero in λ-many slots, rather than non-zero in all N slots.

4 Evaluation

We ran all of our experiments, unless stated otherwise, on Amazon EC2 using an r3.8xlarge instance (32 cores and 244 GB RAM).[9] Likewise, all our experiments use non-oblivious branching programs unless stated otherwise. For each set of experiments, we look at three things: the *time to generate* the obfuscation, the *size* of the resulting obfuscation, and the *time to evaluate* the obfuscation.

[9] All the experiments used the code from commit cac80b159ab99b8aab86cf776f0d48387db86ff4.

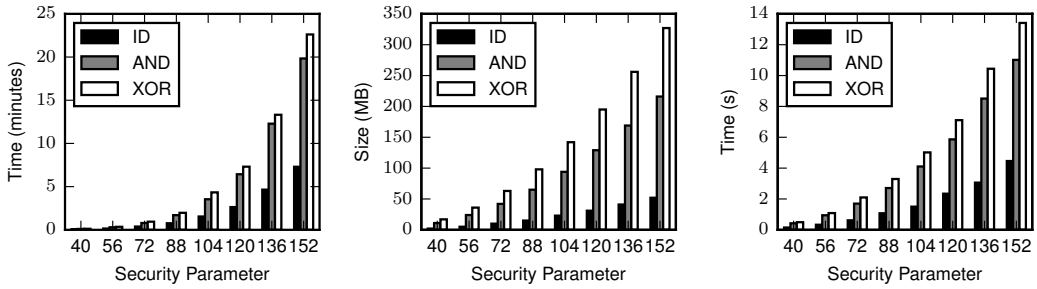

Figure 6: Obfuscation time (left), obfuscation size (center), and evaluation time (right) of single gate circuits with various security parameters.

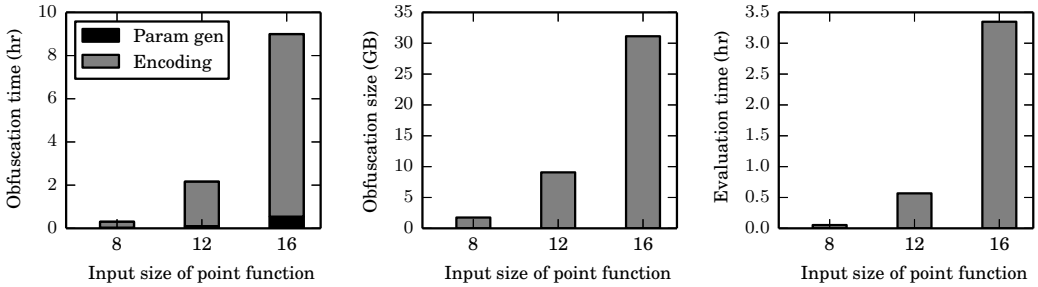

Figure 9: Obfuscation time (left), obfuscation size (center), and evaluation time (right) versus input size of the point function, with security parameter set to 52.

4.1 Varying the Security Parameter

We first look at the cost of obfuscating a *single* gate while varying the security parameter. While obviously a toy example, the goal here is to measure the impact of increasing the security parameter. We present results for three single-gate circuits: an identity (ID) gate, which takes a single bit as input and outputs that bit as the output, an AND gate, and an XOR gate; see Figure 6.

Note the increase in time and size between an AND gate and an XOR gate. Recall that a single AND gate maps to a width-4 branching program and a single XOR gate maps to a width-5 branching program. Here we quantify the cost of that increase. Using security parameter 152 as an example, an obfuscation of an AND gate has size 216.3 MB whereas an obfuscation of an XOR gate has size 327.4 MB, a 51.4% increase, which coincides reasonably well with the expected theoretical increase of $(5^2/4^2) = 56\%$.[10] We see a larger increase going from identity gates to AND gates due to the increase in the number of inputs, and thus the number of encoded matrices, when obfuscating an AND gate.

4.2 Point Obfuscation

In our next experiment, we obfuscated point functions of varying input lengths using a fixed security parameter in order to measure how the size of the function affects performance.

[10]This latter calculation comes from the fact that AND gates map to 4×4 matrices, and XOR gates map to 5×5 matrices.

#	CLT param gen					BP rand	Bookends enc		Layer enc		Obf time	RAM	Size
	p_i/g_i	CRT	z_i	p_{zt}	total		avg	total	avg	total		(GB)	(GB)
8	67.3	12.6	9.0	23.8	112.7	2.3	6.7	13.6	120.4	962.9	1091.8	9.2	1.8
12	260.6	32.3	17.8	53.2	364.0	7.5	19.3	38.8	613.8	7366.0	7776.9	21.5	9.1
16	1396.0	170.5	45.5	336.4	1948.4	19.4	56.8	113.9	1893.9	30302.9	32385.3	42.3	31.1

Table 7: Microbenchmarks for point function obfuscation, with security parameter $\lambda = 52$. All timings are in seconds. '#' denotes the input size of the point function. 'p_i/g_i' denotes the time to generate random p_is and g_is, as well as compute $q = \prod p_i$. 'CRT' denotes the time to compute CRT coefficients for the p_i's, which is used when encoding. Likewise, 'z_i' and 'p_{zt}' denote the time required to generate z_is and the zero test element p_{zt}. 'BP rand' denotes the time required to randomize λ branching programs. (This is the only computationally expensive step not implemented in C.) 'Bookend enc' denotes the time to construct and encode the bookend vectors. 'Layer enc' denotes the time to encode each layer of the randomized matrix branching program. 'Obf time' denotes the *total* time to obfuscate (which includes the CLT parameter generation time and encoding times). 'RAM' denotes the maximum amount of memory used during obfuscation. 'Size' denotes the size of the resulting obfuscation. See Table 5 for the concrete CLT parameters for these obfuscations.

#	1	2	3	4	5	6	7	8	9	10	Zero test	Total	RAM (GB)
8	0.11	9.1	12.5	21.6	26.7	29.3	33.7	45.9			12.7	191.7	3.7
12	0.39	36.7	72.8	86.6	128.0	152.8	158.1	178.4	255.8	283.3	41.0	2040.4	12.9
16	0.94	144.0	249.9	294.3	360.8	553.4	638.7	672.4	743.8	825.2	139.3	12053.4	40.7

Table 8: Microbenchmarks for point function evaluation. All times are in seconds. The number headings correspond to the index of the matrix being multiplied. For example, a heading of '6' corresponds to the time required to multiply the sixth matrix of the obfuscation with the existing matrix product up to that point. The heading of '1' corresponds to the time it takes to load the first matrix into memory. For the 12-bit and 16-bit point functions, we only display the running time of multiplying the first 10 matrices. 'Zero test' denotes the time to run the isZero method. 'Total' denotes the total running time of evaluation. 'RAM' denotes the maximum amount of memory (in GB) used during evaluation.

We obfuscate point functions taking 8, 12, and 16 bits of input, using security parameter 52 throughout.[11] These functions produce branching programs of widths 10, 14, and 18, respectively. Due to the time it takes to obfuscate and evaluate each function, we only conduct a single run for each. Table 7 breaks down the cost of each step in the obfuscation process, Table 8 breaks down the cost of evaluating the obfuscation, and Figure 9 depicts the obfuscation time, obfuscation size, and evaluation time of each point function.

As a first observation, note that the majority of time is spent encoding elements; the CLT parameter generation takes only 4–10% of the overall running time. Also note the large amount of RAM required to obfuscate. A large portion (90%+) of this is to compute and store the CRT coefficients which are needed to make encoding efficient. In terms of the evaluation time, we see that as we multiply matrices into the matrix product, the running time increases as well. This is because each multiplication increases the multilinearity level of the underlying graded encoding scheme and thus the size of the resulting encoding.

[11] We choose 52 simply because it is the smallest "reasonable" security parameter to experiment with.

#	CLT param gen					BP rand	Bookends enc	Layer enc	Obf time	Eval time
	p_i/g_i	CRT	z_i	p_{zt}	total					
8	47.7	35.7	29.1	27.7	41.9	4.2	25.7	40.8	40.7	43.5
12	49.4	42.7	37.1	39.9	47.1	1.3	49.2	45.0	45.1	46.8
16	30.8	49.0	47.7	45.1	36.1	-0.5	49.2	47.6	47.0	48.1

Table 10: Comparison of running times for point-function obfuscation when using 16 cores versus 32 cores. The numbers presented denote the *percentage decrease in running time* when going from 16 to 32 cores. Perfect parallelizability would result in a 50% decrease in running time. See Table 7 for column information.

Oblivious?	w	κ	η	N	CLT param gen	Encoding	Obf time	Obf size (MB)	Eval time
N	6	6	404	1852	1.9	2.6	4.5	26.8	1.0
Y	9	18	1004	4603	42.5	285.7	328.2	1436.8	250.4

Table 11: The effect of constructing oblivious branching programs. Both runs obfuscate the same circuit $f(x_0, x_1, x_2, x_3) = (x_0 \wedge x_1) \wedge (x_2 \wedge x_3)$ using (toy) security parameter 24, with the second run constructing oblivious matrix branching programs. 'w' is the width of the branching program, 'κ' denotes the multilinearity level of the graded encoding scheme, 'η' denotes the bit-size of the random primes p_i, and 'N' denotes the plaintext space for the CLT encodings. See Table 7 for details on the other columns.

4.3 Parallelizability

Notice that the CLT instantiation of InstGen algorithm involves generating a large number of prime numbers, which we can parallelize across multiple cores. Likewise, when obfuscating a matrix, note that each element in each matrix can be encoded independently of other elements, and thus we can easily parallelize the obfuscation process. Finally, when evaluating on obfuscated program, we can parallelize across the matrix computations. In order to test how parallelizable these steps are in practice, we re-run the point function obfuscations using an r3.4xlarge Amazon EC2 instance (16 cores and 122 GB RAM). Table 10 shows the results.

We can see that in general, obfuscation and evaluation are very parallelizable. The one step that has almost no parallelizability, the branching program randomization step ('BP rand') is also the only step done in Python, and thus we lose the ability to use OpenMP. Likewise, the reason some steps, e.g., the zero-test parameter generation ('p_{zt}'), are less parallelizable is because they require a synchronization step between threads to compute the result.

4.4 Oblivious Branching Programs

In the previous experiments we obfuscated *non*-oblivious matrix branching programs. Obfuscating non-oblivious branching programs makes sense for applications such as obfuscating point functions, where the structure of the program is known but an embedded secret needs to be hidden. For settings where the entire *program* needs to be hidden, however, the branching program needs to be made oblivious before being obfuscated (cf. Section 2.2.1). Table 11 measures the impact of this step on the efficiency of obfuscating the formula $f(x_0, x_1, x_2, x_3) = (x_0 \wedge x_1) \wedge (x_2 \wedge x_3)$. In this one example, we can see that the obfuscation size increases by 53.6× and the running time increases by 250×, and (unfortunately) these blow-ups in size and running time will only increase as the size of the formula and input length increases.

5 Discussion and Conclusion

In this work, we provide the first implementation of cryptographic program obfuscation. We discuss both challenges encountered and optimizations made over the course of our development, and present a detailed evaluation of the performance of such obfuscators. Although we show that obfuscation is still far from practical, we are still able to obfuscate some "meaningful" programs. We hope that the availability of an obfuscation prototype will spur further work in both making obfuscation more practical as well as understanding the underlying security primitives better.

The evaluation results from Section 4 show that work still needs to be done before program obfuscation is usable in practice. In fact, the most complex function we obfuscate with meaningful security is a 16-bit point function, which contains just 15 AND gates. Even such a simple function requires about 9 hours to obfuscate and results in an obfuscation of 31.1 GB. Perhaps more importantly (since the obfuscation time is a one-time cost), evaluating the 16-bit point obfuscation on a single input takes around 3.3 hours. However, it is important to note that the fact that we can produce *any* "useful" obfuscations at all is surprising. Also, both obfuscation and evaluation are embarrassingly parallel and thus would run significantly faster with more cores (the largest machine we experimented on had 32 cores). Hopefully our implementation will help spur further research into making obfuscation more practical.

Regarding the main bottlenecks in efficiency, the most immediate one is the efficiency of the underlying graded encoding scheme, and in particular, the size η of the primes and the vector dimension N. Recall that κ represents the multilinearity of our graded encoded scheme, and grows linear in the length of the matrix branching program. The size of η, which affects both the encoding time and overall size of the obfuscation, is roughly $O(\kappa\lambda)$. Likewise, N, which also greatly affects the encoding time and size, is roughly $O(\eta \log \lambda) = O(\kappa\lambda \log \lambda)$. Unfortunately, these parameters need to be set as such due to known attacks [12]. A possible research direction would be to analyze the practical impact of these attacks, and determine whether these values can be reduced in any way without impacting the overall security.

Ideally, we would even prefer to have a mechanism for graded encoding schemes that functions akin to bootstrapping or modulus switching for fully homomorphic encryption, allowing the evaluator to control the growth of the various system parameters mid-computation while retaining security. Unfortunately, all known noise management techniques for FHE implicitly rely on the plaintext space being public, while in the case of graded encoding schemes, the parameters that define the plaintext space — the primes g_i — must remain secret in order to hide the encoded values.

Another major bottleneck is the use of branching programs, which restrict us to obfuscating *single-bit Boolean formulas*. This (1) precludes computing more than $n - 1$ gates in one circuit, and (2) requires obfuscating the computation of every single output bit independently. While branching programs generated using Barrington's approach suffer from exponential growth with the circuit depth, the work of Ananth et al. [3] is a very important step in reducing the overhead due to branching programs. More recently, Zimmerman [31] devised a construction which avoids branching programs altogether; investigating the practical impact of this approach is left as future work.

Finally, to aid in the *cryptanalysis* of program obfuscation and the underlying cryptographic tool of graded encoding schemes, we have released a "challenge" to the community: we have published

an obfuscated point function with a hidden secret; the challenge is to determine the secret.[12] We hope this challenge will inspire cryptanalysts to focus on attacking the cryptographic constructs used in program obfuscation, both to improve confidence in the constructions themselves as well as to better understand the concrete parameters needed for the underlying graded encoding scheme in order to achieve real-world security.

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Changelog

- Version 1.0 (October 1, 2014): First release.