# On the Impossibility of Virtual Black-Box Obfuscation in Idealized Models 

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#### Abstract

The celebrated work of Barak et al. (Crypto'01) ruled out the possibility of virtual blackbox (VBB) obfuscation for general circuits. The recent work of Canetti, Kalai, and Paneth (TCC'15) extended this impossibility to the random oracle model as well assuming the existence of trapdoor permutations (TDPs). On the other hand, the works of Barak et al. (Crypto'14) and Brakerski-Rothblum (TCC'14) showed that general VBB obfuscation is indeed possible in idealized graded encoding models. The recent work of Pass and Shelat (Cryptology ePrint $2015 / 383$ ) complemented this result by ruling out general VBB obfuscation in idealized graded encoding models that enable evaluation of constant-degree polynomials in finite fields.

In this work, we extend the above two impossibility results for general VBB obfuscation in idealized models. In particular we prove the following two results both assuming the existence of trapdoor permutations: - There is no general VBB obfuscation in the generic group model of Shoup (Eurocrypt'97) for any abelian group. By applying our techniques to the setting of Pass and Shelat we extend their result to any (even non-commutative) finite ring. - There is no general VBB obfuscation in the random trapdoor permutation oracle model. Note that as opposed to the random oracle which is an idealized primitive for symmetric primitives, random trapdoor permutation is an idealized public-key primitive.


Keywords: Virtual Black-Box Obfuscation, Idealized Models, Graded Encoding, Generic Group Model.

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## Contents

1 Introduction ..... 1
1.1 Our Results ..... 2
1.2 Technical Overview ..... 3
1.2.1 Generic Group Model: Proving Theorem 1.1 ..... 4
1.2.2 Low-Degree Graded Encoding Model: Proving Theorem 1.2 ..... 5
1.2.3 Random Trapdoor Permutation Model: Proving Theorem 1.3. ..... 6
2 Virtual Black-Box Obfuscation ..... 7
3 Impossibility of VBB in Generic Algebraic Models ..... 8
3.1 Preliminaries ..... 8
3.2 Generic Group Model ..... 8
3.2.1 The Construction ..... 10
3.3 Degree-O(1) Graded Encoding Model ..... 15
3.3.1 Proving Theorem 3.25 ..... 16
4 Impossibility of VBB in the Random TDP Model ..... 18
4.1 The Construction ..... 19
4.1.1 How to Obfuscate ..... 19
4.1.2 How to Execute ..... 20
4.2 Completeness and Soundness ..... 20
4.3 Extension to hierarchical random TDP ..... 22

## 1 Introduction

Obfuscating programs to make them "unintelligible" while preserving their functionality is one of the most sought after holy grails in cryptography due to its numerous applications. The celebrated work of Barak et al. [BGI $\left.{ }^{+} 01\right]$ was the first to launch a formal study of this notion in its various forms. Virtual Black-Box (VBB) obfuscation is a strong form of obfuscation in which the obfuscated code does not reveal any secret bit about the obfuscated program unless that information could already be obtained through a black-box access to the program. It was shown in [BGI $\left.{ }^{+} 01\right]$ that VBB obfuscation is not possible in general as there is a family of functions $\mathcal{F}$ that could not be VBB obfuscated. Roughly speaking, $\mathcal{F}$ would consist of circuits $C$ such that given any obfuscation $B=O(C)$ of $C$, by running $B$ over $B$ itself as input one can obtain a secret $s$ about $C$ that could not be obtained through mere black-box interaction with $C$. This strong impossibility result, however, did not stop the researchers from exploring the possibility of VBB obfuscation for special classes of functions, and positive results for special cases were presented (e.g., [Can97, Wee05]) based on believable computational assumptions.

The work of Lynn, Prabhakaran and Sahai [LPS04] showed the possibility of VBB obfuscation for certain class of functions in the random oracle model (ROM). The work of [LPS04] left open whether general purpose obfuscator for all circuits could be obtained in the ROM or not. Note that when we allow the random oracle to be used during the obfuscation phase (and also let the obfuscated code to call the random oracle) the impossibility result of $\left[\mathrm{BGI}^{+} 01\right]$ no longer applies, because the proof of $\left[\mathrm{BGI}^{+} 01\right]$ requires the obfuscated algorithm to be a circuit in the plain model where no oracle is accessed. In fact, despite the impossibility of general VBB obfuscation in the plain model, a construction for VBB obfuscation in the ROM could be used as a practical heuristic obfuscation mechanism once instantiated with a strong hash function such as SHA3. This would be in analogy with the way ROM based constructions of other primitives are widely used in practice despite the impossibility results of [CGH04].

On a different route, the breakthrough of Garg et al. $\left[\mathrm{GGH}^{+} 13 \mathrm{~b}\right]$ proposed a candidate indistinguishability obfuscation (iO), a weaker form of obfuscation compared to VBB for which no impossibility results were (are) known, relying on the so called "approximate multi-linear maps" (MLM) assumption [GGH13a]. Shortly after, it was proved by Barak et al. [ $\left.\mathrm{BGK}^{+} 14\right]$ and Brakerski and Rothblum [BR14] that the construction of [GGH $\left.{ }^{+} 13 \mathrm{~b}\right]$ could be used to get even VBB secure obfuscation (rather than the weaker variant of iO ) in an idealized form of MLMs, called the graded encoding model. The VBB obfuscation schemes of $\left[\mathrm{BGK}^{+} 14, \mathrm{BR} 14\right]$ in idealized models raised new motivations for studying the possibility of VBB obfuscation in such models including the ROM.

Canetti, Kalai, and Paneth [CKP15] proved the first impossibility result for VBB obfuscation in a natural idealized model by ruling out the existence of general purpose VBB obfuscators in the random oracle model, assuming the existence of trapdoor permutations. Their work resolved the open question of [LPS04] negatively. At a technical level, [CKP15] showed how to compile any VBB obfuscator in the ROM into an approximate VBB obfuscator in the plain model which preserves the circuit's functionality only for "most" of the inputs. This would rule out VBB obfuscation in plain model (assuming TDPs) since Bitansky and Paneth [BP13] had shown that no approximate VBB obfuscator for general circuits exist if trapdoor permutations exist.

Pass and shelat [Pas15] further studied the possibility of VBB obfuscation in idealized algebraic models in which the positive results of $\left[\mathrm{BGK}^{+} 14, \mathrm{BR} 14\right]$ were proved. [Pas15] showed that the existence of VBB obfuscation schemes in the graded encoding model highly depends on the degree of polynomials (allowed to be zero tested) in this model. In particular they showed that VBB obfuscation of general circuits is impossible in the graded encoding model of constant-degree polynomials. Their work nicely complemented the positive results of $\left[\mathrm{BGK}^{+} 14, \mathrm{BR} 14\right]$ that were
proved in a similar (graded encoding) model but using super-constant (in fact polynomial in security parameter) polynomials.

We shall emphasize that proving limitations of VBB obfuscation or any other primitive in generic models of computation such as the generic group model of Shoup [Sho97] are strong lower-bounds (a la black-box separations [RTV04, IR89]) since such results show that for certain crypto tasks, as long as one uses certain algebraic structures (e.g., an abelian group) in a black-box way as the source of computational hardness, there will always be a generic attack that (also treats the underlying algebraic structure in a black-box way and) breaks the constructed scheme. The fact that the proposed attack is generic makes the lower-bound only stronger.

### 1.1 Our Results

In this work we extend the previous works [CKP15, Pas15] on the impossibility of VBB obfuscation in idealized models of computation and generalize their results to more powerful idealized primitives. We first focus on the generic group model of Shoup [Sho97] (see Definitions 3.5 and 3.6) and rule out the existence of general VBB obfuscation in this model for any finite abelian group.

Theorem 1.1 (Informal). Assuming trapdoor permutations exist, there is no virtual black-box obfuscation for general circuits in the generic group model for any finite abelian group.

The work of [Pas15] implies a similar lower bound for the case of abelian groups of prime order. We build upon the techniques of [CKP15, Pas15] and extend the result of [Pas15] to arbitrary (even noncyclic) finite abelian groups. See the next section for a detailed description of our techniques for proving this theorem and the next theorems described below.

We then apply our techniques designed to prove Theorem 1.1 to the setting of graded-encoding model studied in [Pas15] and extend their results to arbitrary finite rings (rather than fields) which remained open after their work. Our proof even handles noncommutative rings.

Theorem 1.2 (Informal). Assuming trapdoor permutations exist, there is no virtual black-box obfuscation for general circuits in ideal degree-O(1) graded encoding model for any finite ring.

Finally, we generalize the work of [CKP15] beyond random oracles by ruling out general VBB obfuscation in random trapdoor permutations (TDP) oracle model. Our result extends to an arbitrary number of levels of hierarchy of trapdoors, capturing idealized version of primitives such as hierarchical identity based encryption [HL02].

Theorem 1.3 (Informal). Assuming trapdoor permutations exist, there is no virtual black-box obfuscation for general circuits in the random trapdoor permutation model, even if the oracle provides an unbounded hierarchy of trapdoors.

Note that the difference between the power of random oracles and random TDPs in cryptography is usually huge, as random oracle is an idealized primitive giving rise to (very efficient) symmetric key cryptography primitives, while TDPs could be used to construct many public-key objects. Our result indicates that when it comes to VBB obfuscation random TDPs are no more powerful than just random oracles.

Connection to black-box complexity of iO. In a very recent follow-up work by the authors, Rafael Pass, and abhi shelat $\left[\mathrm{MMN}^{+} 15\right]$ it is shown that the results of this work and those of [Pas15] could be used to derive lower-bounds on the assumptions that can be used in a black-box way to construct indistinguishability obfuscation. In particular, let $\mathcal{P}$ be a primitive implied by (i.e.
exist relative to) random trapdoor permutations, generic abelian group model, or the degree- $O(1)$ graded encoding model; this includes powerful primitives such as exponentially secure TDPs or exponentially secure DDH-type assumptions. $\left[\mathrm{MMN}^{+} 15\right]$ shows that basing iO on $\mathcal{P}$ in a black-box way is either impossible, or it is at least as hard as basing public-key cryptography on one-way functions (in a non-black-box way). Whether or not public-key encryption can be based on one-way functions has remained as one of the most fundamental open questions in cryptography.

### 1.2 Technical Overview

The high level structure of the proofs of our results follows the high level structure of [CKP15], so we start by recalling this approach. The idea is to convert the VBB obfuscator $O^{\mathcal{I}}$ in the idealized model to an approximate VBB obfuscation $\widehat{O}$ in the plain model which gives the correct answer $C(x)$ with high (say, $9 / 10$ ) probability over the choice of random input $x$ and randomness of obfuscator. The final impossibility follows by applying the result of [BP13] which rules out approximate VBB in the plain model. Following [CKP15] our approximate VBB obfuscator $\widehat{O}$ in the plain model has the following high level structure.

1. Obfuscation emulation. Given a circuit $C$ emulate the execution of the obfuscator $O^{\mathcal{I}}$ in the idealized model over input $C$ to get circuit $B$ (running in the idealized model). ${ }^{1}$
2. Learning phase. Emulate the execution of $B$ over $m$ random inputs for sufficiently large $m$. Output $B$ and the view $z$ of the $m$ executions above as the obfuscated code $\widehat{B}=(B, z)$.
3. Final execution. Execute the obfuscated code $\widehat{B}=(B, z)$ on new random points using some form of "lazy evaluation" of the oracle while only using the transcript $z$ of the learning phase (and not the transcript of obfuscator $O$ which is kept private) as the partially fixed part of the oracle. The exact solution here depends on the idealized model $\mathcal{I}$, but they all have the following form: if the answer to a new query could be "derived" from $z$ then use this answer, otherwise generate the answer from some simple distribution.

VBB property. As argued in [CKP15], the VBB property of $\widehat{O}$ follows from the VBB property of $O$ and that the sequence of views $z$ could indeed be sampled by any PPT holding $B$ in the idealized model (by running $B$ on $m$ random inputs), and so it is simulatable (see Lemma 2.2).

Approximate correctness. The main challenge is to show the approximate correctness of the new obfuscation procedure in the plain model. The goal here is to show that if the learning phase is run for sufficiently large number of rounds, then its transcript $z$ has enough information for emulating the next (actual) execution consistently with but without knowing the view of $O$. In the case that $\mathcal{I}$ is a random oracle [CKP15] showed that it is sufficient to bound the probability of the "bad" event $E$ that the final execution of $\widehat{B}=(B, z)$ on a random input $x$ asks any of the "private" queries of the obfuscator $O$ which is not discovered during the learning phase. The work of [Pas15] studies graded encoding oracle model where the obfuscated code can perform arbitrary zero-test queries for low degree polynomials $p(\cdot)$ defined over the generated labels $s_{1}, \ldots, s_{k}$. The oracle will return true if $p\left(s_{1}, \ldots, s_{k}\right)=0$ (in which case $p(\cdot)$ is called a zero polynomial) and returns false otherwise. Due to the algebraic structure of the field here, it is no longer enough to learn the heavy queries of the obfuscated code who might now ask its oracle query $p(\cdot)$ from some "flat" distribution while its answer is correlated with previous answers.

[^1]
### 1.2.1 Generic Group Model: Proving Theorem 1.1

To describe the high level ideas of the proof of our Theorem 1.1 it is instructive to start with the proof of [Pas15] restricted to zero testing degree-1 polynomials and adapt it to the very close model of GGM for $\mathbb{Z}_{p}$ when $p$ is a prime, since as noted in [Pas15] when it comes to zero-testing linear functions these two models are indeed very close. ${ }^{2}$

Case of $\mathbb{Z}_{p}$ for prime $p$ [Pas15]. When we go to the generic group model we can ask addition and labeling queries as well. It can be shown that we do not need to generate any labels during obfuscation and they can be emulated using addition queries. Then, by induction, all the returned labels $t_{1}, \ldots, t_{\ell}$ for addition queries are linear combinations of $s_{1}, \ldots, s_{k}$ with integer coefficients ${ }^{3}$ and that is how we represent queries. Suppose we get an addition query $\mathbf{a}+\mathbf{b}$ and want to know the label of the group element determined by (the coefficients) $\mathbf{a}+\mathbf{b}=\mathbf{x}$. Suppose for a moment that we know $s$ is the label for a vector of integers $\mathbf{c}$, and suppose we also know that the difference $\mathbf{x}-\mathbf{c}$ evaluates to zero. In this case, similarly to [CKP15], we can confidently return the label $s$ as the answer to $\mathbf{a}+\mathbf{b}$. To use this idea, at any moment, let $W$ be the space of all (zero) vectors $\boldsymbol{\alpha}-\boldsymbol{\beta}$ such that we have previously discovered same labels for $\boldsymbol{\alpha}$ and $\boldsymbol{\beta}$. Now to answer $\mathbf{a}+\mathbf{b}=\mathbf{x}$ we can go over all previously discovered labels $(\mathbf{c} \mapsto s)$ and return $s$ if $\mathbf{x}-\mathbf{c} \in \operatorname{span}(W)$, and return a random label otherwise. The approximate correctness follows from the following two points.

- The rank argument. First note that if $\mathbf{x}-\mathbf{c} \in \operatorname{span}(W)$ then the label for the vector $\mathbf{a}+\mathbf{b}=\mathbf{x}$ is indeed $s$. So we only need worry about cases where $\mathbf{x}-\mathbf{c} \notin \operatorname{span}(W)$ but $\mathbf{x}-\mathbf{c}$ is still zero. The rank argument of [Pas15] shows that this does not happen too often if we repeat the learning phase enough times. The main idea is that if this "bad" event happens, it increases the rank of $W$, but this rank can increase only $k$ times.
- Gaussian elimination. Finally note that the test $\mathbf{x}-\mathbf{c} \notin \operatorname{span}(W)$ can be implemented efficiently using Gaussian elimination when we work in $\mathbb{Z}_{p}$.

Case of general cyclic abelian groups. We first describe how to extend the above to any cyclic abelian group $\mathbb{Z}_{m}$ (for possibly non-prime $m$ ) as follows.

- Alternative notion for rank of $W$. Unfortunately, when we move to the ring of $\mathbb{Z}_{m}$ for nonprime $m$ it is no longer a field and we cannot simply talk about the rank of $W$ (or equivalently the dimension of $\operatorname{span}(W)$ ) anymore. ${ }^{4}$ More specifically, similarly to [Pas15], we need such (polynomial) bound to argue that during most of the learning phases the set $\operatorname{span}(W)$ does not grow. To resolve this issue we introduce an alternative notion which here we call $\widetilde{\operatorname{rank}(W)}$ that has the following three properties even when vectors $\mathbf{w} \in W$ are in $\mathbb{Z}_{m}^{k}(1)$ If $\mathbf{a} \in \operatorname{span}(W)$ then $\widetilde{\operatorname{rank}}(W)=\widetilde{\operatorname{rank}}(W \cup\{\mathbf{a}\})$, and (2) if $\mathbf{a} \notin \operatorname{span}(W)$ then $\widetilde{\operatorname{rank}}(W)+1 \leq \widetilde{\operatorname{rank}}(W \cup\{\mathbf{a}\})$, and (3) $1 \leq \widetilde{\operatorname{rank}}(W) \leq k \cdot \log \left|\mathbb{Z}_{m}\right|=k \cdot \log m$. In particular in Lemma 3.21 we show that the quantity $\widehat{\operatorname{rank}}(W):=\log |\operatorname{span}(W)|$ has these three properties. These properties together show that $\operatorname{span}(W)$ can only "jump up" $k \cdot \log m$ (or fewer) times during the learning phase, and that property is enough to be able to apply the argument of [Pas15] to show that sufficiently large number of learning phases will bound the error probability by arbitrary $1 / \operatorname{poly}(n)$.

[^2]- Solving system of linear equations over $\mathbb{Z}_{m}$. Even though $m$ is not necessarily prime, this can still be done using a generalized method for cyclic abelian groups [McC90].

Beyond cyclic groups. Here we show how to extend the above argument beyond cyclic groups to arbitrary abelian groups. First note that to solve the Gaussian elimination algorithm for $\mathbb{Z}_{m}$, we first convert the integer vectors of $W$ into some form of finite module by trivially interpreting the integer vectors of $W$ as vectors in $\mathbb{Z}_{m}^{k}$. This "mapping" was also crucially used for bounding $\widetilde{\operatorname{rank}(W)}$.

- Mapping integers to general abelian $G$. When we move to a general abelian group $G$ we again need to have a similar mapping to map $W$ into a "finite" module. Note that we do not know how to solve these questions using integer vectors in $W$ efficiently. In Lemma 3.4 we show that a generalized mapping $\rho_{G}(\cdot): \mathbb{Z} \mapsto G$ (generalizing the mapping $\rho_{\mathbb{Z}_{m}}(x)=(x$ $\bmod m)$ for $\mathbb{Z}_{m}$ ) exists for general abelian groups that has the same effect; namely, without loss of generality we can first convert integer vectors in $W$ to vectors in $G^{k}$ and then work with the new $W$.
- The alternative rank argument. After applying the transformation above over $W$ (to map it into a subset of $G^{k}$ ) we can again define and use the three rank-like properties of $\operatorname{rank}(\cdot)$ (instead of $\operatorname{rank}(W)$ ) described above, but here for any finite abelian group $G$. In particular we use $\operatorname{rank}(W):=\log \left|\operatorname{span}_{\mathbb{Z}}(W)\right|$ where $\operatorname{span}_{\mathbb{Z}}(\cdot)$ is the module spanned by $W$ using integer coefficients. Note that even though $G$ is not a ring, multiplying integers with $x \in G$ is naturally defined (see Definition 3.3).
- System of linear equations over finite abelian groups. After the conversion step above, now we need to solve a system of linear equation $\mathbf{x} W=\mathbf{a}$ where elements of $W$, a are from $G$ but we are still looking for integer vector solutions $\mathbf{x}$. After all, there is no multiplication defined over elements from $G$. See the full version of this paper in which we give a reduction from this problem (for general finite abelian groups) to the case of $G=\mathbb{Z}_{m}$ which is solvable in polynomial time [McC90].


### 1.2.2 Low-Degree Graded Encoding Model: Proving Theorem 1.2

To prove Theorem 1.3 for general finite rings, we show how to use the ideas developed for the case of general abelian generic groups discussed above and apply them to the framework of [Pas15] for low-degree graded encoding model as specified in Theorem 1.2. Recall that here the goal is to detect the zero polynomials by checking their membership in the module span $(W)$. Since here we deal with polynomials over a ring (or field) $R$ the multiplication is indeed defined. Therefore, if we already know a set of zero polynomials $W$ and want to judge whether a is also (the vector corresponding to) a zero polynomial, the more natural approach is to solve a system of linear equations $\mathbf{x} W=\mathbf{a}$ over ring $R$.

Searching for integer solutions again. Unfortunately we are not aware of a polynomial time algorithm to solve $\mathbf{x} \cdot W=\mathbf{a}$ in general finite rings and as we mentioned above even special cases like $R=\mathbb{Z}_{m}$ are nontrivial [McC90]. Our idea is to try to reduce the problem back to the abelian groups by somehow eliminating the ring multiplication. Along this line, when we try to solve $\mathbf{x} \cdot W=\mathbf{a}$, we again restrict ourselves only to integer solutions. In other words, we do not multiply inside $R$ anymore, yet we take advantage of the fact that the existence of integer solution to $\mathbf{x} \cdot W=\mathbf{a}$ is still sufficient to conclude $\mathbf{a}$ is a zero vector. As we mentioned above, we indeed know how to find
integer solutions to such system of linear equations in polynomial time (see [McC90] and the full version).

Finally note that, we can again use our alternative rank notion of $\widetilde{\operatorname{rank}}(W)$ to show that if we run the learning phase of the obfuscation (in plain model) $m$ times the number of executions in which $\operatorname{span}_{\mathbb{Z}}(W)$ grows is at most poly $(n)$ (in case of degree- $O(1)$ polynomials). This means that we can still apply the high level structure of the arguments of [Pas15] for the case of finite rings without doing Gaussian elimination over rings.

### 1.2.3 Random Trapdoor Permutation Model: Proving Theorem 1.3.

Here we give the high-level intuition behind our result for the random TDP oracle model.

Recalling the case of Random Oracles [CKP15]. Recall the high level structure of the proof of [CKP15] for the case of random oracles described above. As we mentioned, [CKP15] showed that to prove approximate correctness it is sufficient to bound the probability of the event $E$ that the final execution of $\widehat{B}=(B, z)$ on a random input $x$ asks any of the queries that is asked by emulated obfuscation $O$ of $B$ (let $Q_{O}$ denote this set) which is not discovered during the learning phase. So if we let $Q_{E}, Q_{B}, Q_{O}$ denote the set of queries asked, respectively, in the final execution, learning, and obfuscation phases, the bad event would be $Q_{E} \cap\left(Q_{O} \backslash Q_{B}\right) \neq \varnothing$. This probability could be bound by arbitrary small 1 / poly by running the learning phase sufficiently many times. The intuition is that all the "heavy" queries which have a 1 / poly-chance of being asked by $\widehat{B}=(B, z)$ (i.e., being in $Q_{E}$ ) on a random input $x$ would be learned, and thus the remaining unlearned private queries (i.e., $Q_{O} \backslash Q_{B}$ ) would have a sufficiently small chance of being hit by the execution of $\widehat{B}=(B, z)$ on a random input $x$.

Warm-up: Random Permutation Oracle. We start by first describing the proof for the simpler case of random permutation oracle. The transformation technique for the random oracle model can be easily adapted to work in the random permutation model as follows. For starters, assume that the random permutation is only provided on one input length $k$; namely $R:\{0,1\}^{k} \mapsto$ $\{0,1\}^{k}$. If $k=O(\log n)$ where $n$ is the security parameter, then it means that the whole oracle can be learned during the obfuscation and hardcoded in the obfuscated code, and so $R$ cannot provide any advantage over the plain model. On the other hand if $k=\omega(\log n)$ it means that the range of $R$ is of super-polynomial size. As a result, the same exact procedure proposed in [CKP15] (that assumes $R$ is a random oracle and shows how to securely compile out $R$ from the construction) would also work if $R$ is a random permutation oracle. The reason is that the whole transformation process asks poly $(n)$ number of queries to $R$ and, if the result of the transformation does not work when $R$ is a random permutation, then the whole transformation gives a $\operatorname{poly}(n)=q$ query attack to distinguish between whether $R$ is a random permutation or a random function. Such an attack cannot "succeed" with more than negligible probability when the domain of $R$ has super-polynomial size $q^{\omega(1)}$ in the number of queries $q^{5}$.

Random TDP Model. Suppose $T=\left(G, F, F^{-1}\right)$ is a random trapdoor permutation oracle in which $G$ is a random permutation for generating the public index, $F$ is the family of permutations evaluated using public index, and $F^{-1}$ is the inverse permutation computed using the secret key (see

[^3]Definition 4.1 for formal definition and notation used). When the idealized oracle is $T=\left(G, F, F^{-1}\right)$, we show that it is sufficient to apply the same learning procedure used in the random oracle case over the normalized version of the obfuscated algorithm $B$ to get a plain-model execution $\widehat{B}(x)$ that is statistically close to an execution $B^{T}(x)$ that uses oracle $T$. This, however, requires careful analysis to prove that inconsistent queries specific to the TDP case occur with small probability.

Indeed, since the three algorithms (emulation, learning, and final execution) are correlated, there is a possibility that the execution of $B$ on the new random input might ask a new query that is not in $Q_{O}$, and yet still be inconsistent with some query in $Q_{O} \backslash Q_{B}$. For example, assume we have a query $q$ of the form $G(s k)=p k$ that was asked during the obfuscation emulation phase (and thus is in $Q_{O}$ ) but was missed in the learning phase (and thus is not in $Q_{B}$ ) and assume that a query of the form $F[p k](x)=y$ was asked during the learning phase (so it is in $Q_{B}$ ). Then, it is possible that during the evaluation of the circuit $B$, it may ask a query $q^{\prime}$ of the form $F^{-1}[s k](y)$ and since this is a new query undetermined by previously learned queries, the plain-model circuit $\widehat{B}$ will answer with some random answer $y^{\prime}$. Note that in this case, $y^{\prime}$ would be different from $y$ with very high probability, and thus even though $q \neq q^{\prime}$, they are together inconsistent with respect to oracle $T$.

As we show in our case-by-case analysis of the learning heavy queries procedure for the case of trapdoor permutation (in Section 4.2), the only bad events that we need to consider (besides hitting unlearned $Q_{O}$ queries, which was already shown to be unlikely) will be those whose probability of occurring are negligible (we use the lemmas from [GKLM12] as leverage). Due to our normalization procedure, the rest of the cases will be reduced to the case of not learning heavy queries, and this event is already bounded.

## 2 Virtual Black-Box Obfuscation

Below we give a direct formal definition for approximately correct virtual black-box (VBB) obfuscation in idealized models. The (standard) definition of VBB is equivalent to 0 -approximate VBB in the plain model where no oracle is accessed.

Definition 2.1 (Approximate VBB in Idealized Models [BGK+13, CKP15]). For a function $\epsilon(n) \in$ $[0,1]$, a PPT algorithm $O$ is called an $\epsilon$-approximate general purpose VBB obfuscator in the $\mathcal{I}$-ideal model if the following is satisfied:

- Approximate Functionality: For any circuit $C$ of size $n$ and input size $m$

$$
\operatorname{Pr}_{x \leftarrow\{0,1\}^{m}}\left[O^{\mathcal{I}}(C)(x) \neq C(x)\right] \leq \epsilon(n)
$$

where the probability is over the choice of input $x$, the oracle $\mathcal{I}$, and the internal randomness of $O$.

- Virtual Black-Box: For every PPT adversary $A$, there exists a PPT simulator $S$ and a negligible $\sigma(n)$ such that for all $n \in \mathbb{N}$ and circuits $C \in\{0,1\}^{n}$ :

$$
\left|\operatorname{Pr}\left[A^{\mathcal{I}}\left(O^{\mathcal{I}}(C)\right)=1\right]-\operatorname{Pr}\left[S^{C}\left(1^{n}\right)=1\right]\right| \leq \sigma(n)
$$

where the probability is over $\mathcal{I}$ and the randomness of $A, S$, and $O$.
The following lemma is used in [CKP15], and here we state it in an abstract form considering only the VBB security and ignoring the completeness.

Lemma 2.2 (Preservation of VBB Security). Let $O$ be a PPT algorithm in the $\mathcal{I}$-ideal model that satisfies VBB security, and let $U$ be a PPT algorithm also in the $\mathcal{I}$-ideal model that given input $B=O^{\mathcal{I}}(C)$ for some circuit $C \in\{0,1\}$ of size $n$, outputs circuit $B^{\prime}$, and suppose $\widehat{O}$ is some plainmodel PPT algorithm that given circuit $C$, outputs $\widehat{B}$. If it holds that conditioned on any given $C$, the statistical distance between $B^{\prime}$ and $\widehat{B}$ are negligible, then $\widehat{O}$ satisfies the VBB security.

## 3 Impossibility of VBB in Generic Algebraic Models

In this section we will formally state and prove our Theorems 1.1 and 1.2 for the generic group and graded encoding models.

### 3.1 Preliminaries

We start by stating some basic group theoretic notation, facts, and definitions. By $\mathbb{Z}$ we refer to the set of integers. By $\mathbb{Z}_{n}$ we refer to the additive (or maybe the ring) of integers modulo $n$. When $G$ is an abelian group, we use + to denote the operation in $G$. A semigroup $(G, \odot)$ consists of any set $G$ and an associative binary operation $\odot$ over $G$.

Definition 3.1. For semi-groups $\left(G_{1}, \odot_{1}\right), \ldots,\left(G_{k}, \odot_{1}\right)$, by the direct product semi-group $(G, \odot)=$ $\left(G_{1} \times \cdots \times G_{k}, \odot_{1} \times \cdots \times \odot_{k}\right)$ we refer to the group in which for $g=\left(g_{1}, \ldots, g_{k}\right) \in G, h=$ $\left(h_{1}, \ldots, h_{k}\right) \in G$ we define $g \odot h=\left(g_{1} \odot_{1} h_{1}, \ldots, g_{k} \odot_{k} h_{1}\right)$. If $G_{i}$ 's are groups, their direct product is also a group.

The following is the fundamental theorem of finitely generated abelian groups restricted to case of finite abelian groups.

Theorem 3.2 (Characterization of Finite Abelian Groups). Any finite abelian group $G$ is isomorphic to some group $\mathbb{Z}_{p_{1}^{\alpha_{1}}} \times \cdots \times \mathbb{Z}_{p_{d}^{\alpha_{d}}}$ in which $p_{i}$ 's are (not necessarily distinct) primes and $\mathbb{Z}_{p_{i}^{\alpha_{i}}}$ is the additive group $\bmod p_{i}^{\alpha_{i}}$.

Definition 3.3 (Integer vs in-group multiplication for abelian groups). For integer $a \in \mathbb{Z}$ and $g \in G$ where $G$ is any finite abelian group by $a \cdot g$ we mean adding $g$ by itself $|a|$ times and negating it if $a<0$. Now let $g, h \in G$ both be from abelian group $G$ and let $G=\mathbb{Z}_{p_{1}^{\alpha_{1}}} \times \cdots \times \mathbb{Z}_{p_{d}}$ where $p_{i}$ 's are primes. If not specified otherwise, by $g \cdot h$ we mean the multiplication of $g, h$ in $G$ interpreted as the multiplicative semigroup that is the direct product of the multiplicative semigroups of $\mathbb{Z}_{p_{i}}{ }^{\alpha_{i}}$ 's for $i \in[d]$ (where the multiplications in $\mathbb{Z}_{p_{i}^{\alpha_{i}}}$ are $\bmod p_{i}^{\alpha_{i}}$ ).

Lemma 3.4 (Mapping integers to abelian groups). Let $G=\mathbb{Z}_{p_{1}^{\alpha_{1}}} \times \cdots \times \mathbb{Z}_{p_{d}^{\alpha_{d}}}$. Define $\rho_{G}: \mathbb{Z} \mapsto G$ as $\rho_{G}(a)=\left(a_{1}, \ldots, a_{d}\right) \in G$ where $a_{i}=a \bmod p_{i}^{\alpha_{i}} \in \mathbb{Z}_{p_{i}^{\alpha_{i}}}$. Also for $\mathbf{a}=\left(a_{1}, \ldots, a_{k}\right) \in \mathbb{Z}^{k}$ define $\rho_{G}(\mathbf{a})=\left(\rho_{G}\left(a_{1}\right), \ldots, \rho_{G}\left(a_{k}\right)\right)$. Then for any $a \in \mathbb{Z}$ and $g \in G=\mathbb{Z}_{p_{1}^{\alpha_{1}}} \times \cdots \times \mathbb{Z}_{p_{d} \alpha_{d}}$ it still holds that $a \cdot g=\rho_{G}(a) \cdot g$ where the first multiplication is done according to Definition 3.3, and the second multiplication is done in $G$. More generally, if $\mathbf{a}=\left(a_{1}, \ldots, a_{k}\right) \in \mathbb{Z}^{k}$, and $\mathbf{g}=\left(g_{1}, \ldots, g_{k}\right) \in G$, then $\sum_{i \in[k]} a_{i} g_{i}=\langle\mathbf{a}, \mathbf{g}\rangle=\left\langle\rho_{G}(\mathbf{a}), \mathbf{g}\right\rangle$.

### 3.2 Generic Group Model

We start by formally defining the generic group model.

Definition 3.5 (Generic Group Model [Sho97]). Let $(G, \odot)$ be any group of size $N$ and let $S$ be any set of size at least $N$. The generic group oracle $\mathcal{I}[G \mapsto S]$ (or simply $\mathcal{I}$ ) is as follows. At first an injective random function $\sigma: G \mapsto S$ is chosen, and two type of queries are answered as follows.

- Type 1: Labeling Queries. Given $g \in G$ oracle returns $\sigma(g)$.
- Type 2: Addition Queries. Given $y_{1}, y_{2}$, if there exists $x_{1}, x_{2}$ such that $\sigma\left(x_{1}\right)=y_{1}$ and $\sigma\left(x_{2}\right)=y_{2}$, it returns $\sigma\left(x_{1} \odot x_{2}\right)$. Otherwise it returns $\perp$.
Definition 3.6. [Generic Algorithms in Generic Group Model] Let $A^{\mathcal{I}}$ be an algorithm (or a set of interactive algorithms $A=\left\{A_{1}, A_{2}, \ldots\right\}$ ) accessing the generic group oracle $\mathcal{I}[G \mapsto S]$. We call $A^{\mathcal{I}}$ generic if it never asks any query (of the second type) that is answered as $\perp$. Namely, only queries are asked for which the labels are previously obtained.
Remark 3.7 (Family of Groups). A more general definition allows generic oracle access to a family of groups $\left\{G_{1}, G_{2}, \ldots\right\}$ in which the oracle access to each group is provided separately when the index $i$ of $G_{i}$ is also specified as part of the query and the size of the group $G_{i}$ is known to the parties. Our negative result of Section 3 directly extends to this model as well. We use the above "single-group" definition for sake of simplicity.

Remark 3.8 (Stateful vs Stateless Oracles and the Multi-Party Setting). Note that in the above definition we used a stateless oracle to define the generic group oracle, and we separated the generic nature of the oracle itself from how it is used by an algorithm $A^{\mathcal{I}}$. In previous work (e.g., Shoup's original definition [Sho97]) a stateful oracle is used such that: it will answer addition queries only if the two labels are already obtained before. ${ }^{6}$

Note that for "one party" settings in which $A^{\mathcal{I}}$ is a single algorithm, $A^{\mathcal{I}}$ "knows" the labels that it has already obtained from the oracle $\mathcal{I}$, and so w.l.o.g. $A^{\mathcal{I}}$ would never ask any addition queries unless it has previously obtained the labels itself. However, in the multi-party setting, a party might not know the set of labels obtained by other parties. A stateful oracle in this case might reveal some information about other parties' oracle queries if the oracle does not answer a query ( $y_{1}, y_{1}$ ) (by returning $\perp$ ) just because the labels for $y_{1}, y_{2}$ are not obtained so far.

Remark 3.9 (Equivalence of Two Models for Sparse Encodings). If the encoding of $G$ is sparse in the sense that $|S| /|G|=n^{\omega(1)}$ where $n$ is the security parameter, then the probability that any party could query a correct label before it being returned by oracle through a labeling (type 1) query is indeed negligible. So in this case any algorithm (or set of interactive algorithms) $A^{\mathcal{I}}$ would have a behavior that is statistically close to a generic algorithm that would never ask a label in an addition query unless that label is previously obtained from the oracle. Therefore, if $|S| /|G|=n^{\omega(1)}$, we can consider $A^{\mathcal{I}}$ to be an arbitrary algorithm (or set of interactive algorithms) in the generic group model $\mathcal{I}$. The execution of $A$ would be statistically close to a "generic execution" in which $A^{\mathcal{I}}$ never asks any label before obtaining it.

In light of Remarks 3.8 and 3.9, for simplicity of the exposition we will always assume that the encoding is sparse $|S| /|G|=n^{\omega(1)}$ and so all the generic group model are automatically (statistically close to being) generic.

Theorem 3.10 (Theorem 1.1 Formalized). Let $G$ be any abelian group of size at most $2^{\operatorname{poly}(n)}$. Let $O$ be an obfuscator in the generic group model $\mathcal{I}[G \mapsto S]$ where the obfuscation of any circuit followed by execution of the obfuscated code (jointly) form a generic algorithm. If $O$ is an $\epsilon$-approximate $V B B$ obfuscator in the generic group model $\mathcal{I}[G \mapsto S]$ for poly-size circuits, then for any $\delta=1 / \operatorname{poly}(n)$ there exists an $(\epsilon+\delta)$-approximate $V B B$ obfuscator $\widehat{O}$ for poly-size circuits in the plain model.

[^4]Remark 3.11 (Size of $G$ ). Note that if a poly $(n)$-time algorithm accesses (the labels of the elements of) some group $G$, it implicitly means that $G$ is at $\operatorname{most}$ of $\exp (n)$ size so that its elements could be names with poly ( $n$ ) bit strings. We chose, however, to explicitly mention this size requirement $|G| \leq 2^{\text {poly }(n)}$ since this upper bound plays a crucial role in our proof for general abelian groups compared to the special case of finite fields.
Remark 3.12 (Sparse Encodings). If we assume a sparse encoding i.e., $|S| /|G|=n^{\omega(1)}$ (as e.g., is the case in [Pas15] and almost all prior work in generic group model) in Theorem 3.10 we no longer need to explicitly assume that the obfuscation followed by execution of obfuscated code are in generic form; see Remark 3.9.

Since [BP13] showed that (assuming TDPs) there is no ( $1 / 2-1$ / poly)-approximate VBB obfuscator in the plain-model for general circuits, the following corollary is obtained by taking $\delta=\epsilon / 2$.

Corollary 3.13. If TDPs exist, then there exists no $(1 / 2-\epsilon)$-approximate VBB obfuscator $O$ for general circuits in the generic group model for any $\epsilon=1 / \operatorname{poly}(n)$, any finite abelian group $G$ and any label set $S$ of sufficiently large size $|S| /|G|=n^{\omega(1)}$. The result would hold for labeling sets $S$ of arbitrary size if the execution of the obfuscator $O$ followed by the execution of the obfuscated circuit $O(C)$ form a generic algorithm.

Now we formally prove Theorem 3.10. We will first describe the algorithm of the obfuscator in the plain model, and then will analyze its properties.

Notation and w.l.o.g. assumptions. Using Theorem 3.2 w.l.o.g. we assume that our abelian group $G$ is isomorphic to the additive direct product group $\mathbb{Z}_{p_{1}^{\alpha_{1}}} \times \cdots \times \mathbb{Z}_{p_{d}^{\alpha_{d}}}$ where $p_{i}$ 's are prime. Let $e_{i} \in G$ be the vector that is 1 in the $i$ 'th coordinate and zero elsewhere. Note that $\left\{e_{1}, \ldots, e_{k}\right\}$ generates $G$. We can always assume that the first $d$ labels obtained by $O$ are the labels of $e_{1}, \ldots, e_{d}$ and these labels are explicitly passed to the obfuscated circuit $B=O(C)$. Let $k=\operatorname{poly}(n)$ be an upper bound on the running time of the obfuscator $O$ for input $C$ which in turn upper bounds the number of labels obtained during the obfuscation (including the the $d$ labels for $e_{1}, \ldots, e_{d}$ ). We also assume w.l.o.g. that the obfuscated code never asks any type one (i.e., labeling) oracle queries since it can use the label for $e_{1}, \ldots, e_{d}$ to obtain labels of any arbitrary $g=a_{1} e_{1}+\cdots+a_{d} e_{d}$ using a polynomial number of addition (i.e., type two) oracle queries. For $\sigma(g)=s, a \in \mathbb{Z}$, and $t=\sigma(a \cdot g)$ we abuse the notation and denote $a \cdot s=t$.

### 3.2.1 The Construction

Even though the output of the obfuscator is always an actual circuit, we find it easier to first describe how the obfuscator $\widehat{O}$ generates some "data" $\widehat{B}$, and then we will describe how to use $\widehat{B}$ to execute the new obfuscated circuit in the plain model. For simplicity we use $\widehat{B}$ to denote the obfuscated circuit.

## How to Obfuscate

The new obfuscator $\widehat{O}$. The new obfuscator $\widehat{O}$ uses lazy evaluation to simulate the labeling $\sigma(\cdot)$ oracle. For this goal, it holds a set $Q_{\sigma}$ of the generated labels. For any new labeling query $g \in G$ if $\sigma(g)=s$ is already generated it returns $s$. Otherwise it chooses an unused label $s$ from $S$ uniformly at random and adds the mapping $(g \rightarrow s)$ to $Q_{\sigma}$ and returns $s$. For an addition query $\left(s_{1}, s_{2}\right)$ it first finds $g_{1}, g_{2}$ such that $\sigma\left(g_{1}\right)=s_{1}$ an $\sigma\left(g_{2}\right)=s_{2}$ (which exist since the algorithm that calls the oracle is in generic form) and gets $g=g_{1}+g_{2}$. Now $\widehat{O}$ proceeds as if $g$ is asked as a labeling query and returns the answer. The exact steps of $\widehat{O}$ are as follows.

1. Emulating obfuscation. $\widehat{O}$ emulates $O^{\mathcal{I}}(C)$ to get circuit $B$.
2. Learning phase 1 (heavy queries): Set $Q_{B}=\varnothing$. For $i \in[d]$ let $t_{i}=\sigma\left(e_{i}\right)$ be the label of $e_{i} \in G$ which is explicitly passed to $B$ by the obfuscator $O(C)$ and $T=\left(t_{1}, \ldots, t_{d}\right)$ at the beginning. The length of the sequence $T$ would increase during the steps below but will never exceed $k$. Choose $m$ at random from $\ell=[\lceil 3 \cdot k \cdot \log (|G|) / \delta\rceil]$. For $i=1, \ldots, m$ do the following:

- Choose $x_{i}$ as a random input for $B$. Emulate the execution of $B$ on $x_{i}$ using the (growing) set $Q_{\sigma}$ of partial labeling for the lazy evaluation of labels. Note that as we said above, w.l.o.g. $B$ only asks addition (i.e., type two) oracle queries. Suppose $B$ (executed on $x_{i}$ ) needs to answer an addition query $\left(s_{1}, s_{2}\right)$. If either of the labels $u=s_{1}$ or $u=s_{2}$ is not already obtained during the learning phase 1 (which means it was obtained during the initial obfuscation phase) append $u$ to the sequence $T$ of discovered labels by $T:=(T, u)$. Using induction, it can be shown that for any addition query asked during learning phase 1 , at the time of being asked, we would know that the answer to this query will be of the form $\sum_{i \in[k]} a_{i} \cdot t_{i}$ for integers $a_{i}$. Before seeing why this is the case, let $\mathbf{a}_{i}=\left(a_{i, 1}, \ldots, a_{i, k}\right)$ be the vector of integer coefficients (of the labels $t_{1}, t_{2}, \ldots$ ) for the answer $s$ that is returned to the $i$ 'th query of learning phase 1 . We add $\left(\mathbf{a}_{i} \rightarrow s\right)$ to $Q_{B}$ for the returned label. To see why such vectors exist, let $\left(s_{1}, s_{2}\right)$ be an addition query asked during this phase, and let $s \in\left\{s_{1}, s_{2}\right\}$. If the label $s$ is obtained previously during learning phase 1 , then the linear form $s=\sum_{i \in[k]} a_{i} \cdot t_{i}$ is already stored in $Q_{B}$. On the other hand, if $s$ is a new label discovered during an addition (i.e., type two) oracle query which just made $T=\left(t_{1}, \ldots, t_{j-1}, t_{j}=s\right)$ have length $j$, then $s=a_{i} \cdot t_{i}$ for $a_{j}=1$. Finally, if the linear forms for both of ( $s_{1}, s_{2}$ ) in an addition oracle query are known, the linear form for the answer $s$ to this query would be the summation of these vectors. ${ }^{7}$

3. Learning phase 2 (zero vectors): This step does not involve executing $B$ anymore and only generates a set $W=W\left(Q_{B}\right) \subseteq G^{k}$ of polynomial size. At the beginning of this learning phase let $W=\varnothing$. Then for all $\left(\mathbf{a}_{1} \rightarrow s_{1}\right) \in Q_{B}$ and $\left(\mathbf{a}_{2} \rightarrow s_{2}\right) \in Q_{B}$, if $s_{1}=s_{2}$, let $\mathbf{a}=\mathbf{a}_{1}-\mathbf{a}_{2}$, and add $\rho_{G}(\mathbf{a})$ to $W$ where $\rho_{G}(\mathbf{a})$ is defined in Lemma 3.4.
4. The output of the obfuscation algorithm will be $\widehat{B}=\left(B, Q_{B}, W, T, r\right)$ where $T$ is the current sequence of discovered labels $\left(t_{1}, t_{2}, \ldots\right)$ as described in Lemma 3.4, and $r$ is a sufficiently large sequence of random bits that will be used as needed when we run the obfuscated code $\widehat{B}=\left(B, Q_{B}, W, T, r\right)$ in the plain model. ${ }^{8}$

## How to Execute

In this section we describe how to execute $\widehat{B}$ on an input $x$ using ( $\left.B, Q_{B}, W, T, r\right) .{ }^{9}$ Before describing how to execute the obfuscated code, we need to define the following algebraic problem.

Definition 3.14. [Integer Solutions to Linear Equations over Abelian Groups (iLEAG)] Let $G$ be a finite abelian group. Suppose we are given $G$ (e.g., by describing its decomposition factors according

[^5]to Theorem 3.2) an $n \times k$ matrix $A$ with components from $G$ and a vector $\mathbf{b} \in G^{k}$. We want to find an integer vector $\mathbf{x} \in \mathbb{Z}^{n}$ such that $\mathbf{x} A=\mathbf{b}$.

Remark 3.15 (Integer vs. Ring Solutions). Suppose instead of searching for an integer vector solution $\mathbf{x} \in \mathbb{Z}^{n}$ we would ask to find $\mathbf{x} \in G^{n}$ and define multiplication in $G$ according to Definition 3.3 and call this problem $G$-LEAG. Then any solution to iLEAG can be directly turned into a solution for $G$-LEAG by mapping any integer coordinate $x_{i}$ of $\mathbf{x}$ into $G$ by mapping $\rho_{G}\left(x_{i}\right)$ of Lemma 3.4. The converse is true also for $G=\mathbb{Z}_{n}$, since any $g \in \mathbb{Z}_{n}$ is also in $\mathbb{Z}$ and it holds that $\rho_{G}(g)=g \in G$. However, the converse is not true in general for general abelian groups, since there could be members of $G$ that are not in the range of $\rho_{G}(\mathbb{Z})$. For example let $G=\mathbb{Z}_{p^{2}} \times \mathbb{Z}_{p}$ for prime $p>2$ and let $g=(2,1)$. Note that there is no integer $a$ such that $a \bmod p^{2}=2$ but $a \bmod p=1$.

Executing $\widehat{B}$. The execution of $\widehat{B}=\left(B, Q_{B}, W, T, r\right)$ on $x$ will be done identically to to the "next" execution during the learning phase 1 of the obfuscation (as if $x$ is the $(m+1)$ 'st execution of this learning phase) and even the sets $Q_{B}, W=W\left(Q_{B}\right)$ will grow as the execution proceeds, with the only difference described as follows. ${ }^{10}$ Suppose we want to answer an addition (i.e., type two) oracle query ( $s_{1}, s_{2}$ ) where for $b=\{1,2\}$ we inductively know that $s_{b}=\sum_{i \in[k]} a_{b, i} \cdot t_{i}$. For $b=\{1,2\}$ let $\mathbf{a}_{b}=\left(a_{b, 1}, \ldots, a_{b, k}\right)$ and let $\mathbf{a}=\mathbf{a}_{1}+\mathbf{a}_{2}$.

- Do the following for all $(\mathbf{b} \rightarrow s) \in Q_{B}$. Let $\mathbf{c}=\mathbf{a}-\mathbf{b}$ and let $\overline{\mathbf{c}}=\rho_{G}(\mathbf{c}) \in G^{k}$ as defined in Lemma 3.4. Let $A$ be a matrix whose rows consists of all vectors in $W$. Run the polynomial time algorithm presented in the full version of this paper (based on an algorithm of [McC90] for $G=\mathbb{Z}_{n}$ ) to see if there is any integer solution $\mathbf{v}$ for $\mathbf{v} A=\overline{\mathbf{c}}$ as an instance of the iLEAG problem defined in Definition 3.14. If an integer solution $\mathbf{v}$ exists, then return $s$ as the result (recall $(\mathbf{b} \rightarrow s) \in Q_{B}$ ), break the loop, and continue the execution of $\widehat{B}$. If the loop ended and no such $(\mathbf{b} \rightarrow s) \in Q_{B}$ was found, choose a random label $s$ not in $Q_{B}$ as the answer, add $(\mathbf{a} \rightarrow s)$ to $Q_{B}$ and continue.


## Completeness and the Soundness

In this section we prove the completeness and soundness of the construction of Section 3.2.1.
Size of $S$. In the analysis below, we will assume w.l.o.g. that the set of labels $S$ has superpolynomial size $|S|=n^{\omega(1)}$. This would immediately hold if the labeing of $G$ is sparse, since it would mean even $|S| /|G| \geq n^{\omega(1)}$. Even if the labeling is not sparse, we will show that w.l.o.g. we can assume that $G$ itself has super-polynomial size (which means that $S$ will be so too). That is because otherwise all the labels in $G$ can be obtained by the obfuscator, the obfuscated code, and the adversary and we will be back to the plain model. More formally, for this case Theorem 3.10 could be proved through a trivial construction in which the new obfuscator simply generates all the labels of $G$ and plants all of them in the obfuscated code and they will be used by the obfuscated algorithm. More precisely, when the size of $G$ (as a function of security parameter $n$ ) is neither of polynomial size $|G|=n^{O(1)}$ nor super-polynomial size $|G|=n^{\omega(1)}$ we can still choose a sufficiently large polynomial $\gamma(n)$ and generate all labels of $G$ when $|G|<\gamma(n)$, and otherwise use the obfuscation of Section 3.2.1.

Completeness: approximate functionality. Here we prove the the following claim.

[^6]Claim 3.16. Let $\widehat{B}=\left(B, Q_{B}, W, T, r\right)$ be the output of the obfuscator $\widehat{O}$ given input circuit $C$ with input length $\alpha$. If we run $\widehat{B}$ over a random input according to the algorithm described in Section 3.2.1, then it holds that

$$
\operatorname{Pr}_{x \leftarrow\{0,1\}^{\alpha}, \widehat{B} \leftarrow \widehat{O}(C)}[\widehat{B}(x) \neq C(x)] \leq \operatorname{Pr}_{x \leftarrow\{0,1\}^{\alpha}, B \leftarrow O^{\mathcal{I}[G \mapsto S]}(C)}\left[B^{\mathcal{I}[G \mapsto S]}(x) \neq C(x)\right]+\delta
$$

over the randomness of $\mathcal{I}[G \mapsto S]$, random choice of $x$ and the randomness of the obfuscators.
Proof. As a mental experiment, suppose we let the learning phase 1 always runs for exactly $\ell+1=$ $1+[\lceil 3 \cdot k \cdot \log (|G|) / \delta\rceil]$ rounds but only derive the components $\left(Q_{B}, W\left(Q_{B}\right), T\right)$ based on the first $m$ executions. Now, let $x_{i}$ be the random input used in the $i$ 'th execution and $y_{i}$ be the output of the $i$ 'th emulation execution the learning phase 1 . Since all the executions of the learning phase 1 are perfect simulations, for every $i \in[\ell]$, and in particular $i=m$, it holds that

$$
\operatorname{Pr}\left[B^{\mathcal{I}[G \mapsto S]}(x) \neq C(x)\right]=\operatorname{Pr}\left[y_{i} \neq C(x)\right]
$$

where probability is over the choice of inputs $x, x_{i}$ as well as all other randomness in the system. Thus, to prove claim 3.16 it will suffice to prove that

$$
\left|\operatorname{Pr}\left[y_{i} \neq C(x)\right]-\operatorname{Pr}\left[\widehat{B}\left(x_{i}\right) \neq C(x)\right]\right|<\delta .
$$

We will indeed do so by bounding the statistical distance between the execution of $\widehat{B}$ vs the $m+1$ 'st execution of the learning phase 1 over the same input $x_{i}$. Here we will rely on the fact that $m$ is chosen at random from $[\ell]$.
Claim 3.17. For random $[\ell]$ the statistical distance between the $m+1$ 'st execution of the learning phase 1 (which we call $B^{\prime}$ ) and the execution of $\widehat{B}$ over the same input $x_{i}$ is $\leq 2 \delta / 3+\operatorname{negl}(n)$.

To prove the above claim we will define three type of bad events over a joint execution of $B^{\prime}=B_{m+1}$ and $\widehat{B}$ when they are done concurrently and using the same random tapes (and even the input $x_{i}$ ). We will then show that (1) as long as these bad events do not happen the two executions proceed identically, and (2) the total probability of these bad events is at most $2 \delta / 3+\operatorname{negl}(n)$. In the following we suppose that the executions of $B^{\prime}$ and $\widehat{B}$ (over the same random input) has proceeded identically so far. Suppose we want to answer an addition (i.e., type two) oracle query $\left(s_{1}, s_{2}\right)$ where for $b=\{1,2\}$ we inductively know that $s_{b}=\sum_{i \in[k]} a_{b, i} \cdot t_{i}$. Several things could happen:

- If the execution of $\widehat{B}$ finds $(\mathbf{b} \rightarrow s) \in Q_{B}$ such that when we take $\mathbf{c}=\mathbf{a}-\mathbf{b}$ and let $\overline{\mathbf{c}}=\rho_{G}(\mathbf{c}) \in G^{k}$ and let $A$ be a matrix whose rows are vectors in (the current) $W$, there is an integer solution $\mathbf{v}$ to the iLEAG instance $\mathbf{v} A=\overline{\mathbf{c}}$. If this happens the execution of $\widehat{B}$ will use $\mathbf{b}$ as the answer. We claim that this is the "correct" answer as $B^{\prime}$ would also use the same answer. This is because by the definition of $W$ and Lemma 3.4 for all $\mathbf{w} \in W$ it holds that $\mathbf{w}=\left(w_{1}, \ldots, w_{k}\right)$ is a "zero vector in $G^{k}$ " in the sense that summing the (currently discovered labels in) $T$ with coefficients $w_{1}, \ldots, w_{k}$ (and multiplication defined according to Definition 3.3) will be zero. As a result, $\mathbf{v} A=\overline{\mathbf{c}}$ which is a linear combination of vectors in $W$ with integer coefficients will also be a zero vector. Finally, by another application of Lemma 3.4 it holds that $\left(c_{1}, \ldots, c_{k}\right)=\mathbf{c}=\mathbf{a}-\mathbf{b}$ is a "zero vector in $\mathbb{Z}^{k}$ in the sense that summing the (currently discovered labels in) $T$ with integer coefficients $c_{1}, \ldots, c_{k}$ (and multiplication defined according to Definition 3.3) will also be zero. Therefore the answer to the query defined by vector $\mathbf{a}$ is equal to the answer defined by vector $\mathbf{b}$ which is $s$.
- If the above does not happen (and no such $(\mathbf{b} \rightarrow s) \in Q_{B}$ is found) then either of the following happens. Suppose the answer returned for $\left(s_{1}, s_{2}\right)$ in execution of $B^{\prime}$ is $s^{\prime}$ :
- Bad event $E_{1}$ : $s^{\prime}$ is equal to one of the labels in $Q_{B}$. Note that in this case the executions will diverge because $\widehat{B}$ will choose a random label.
- Bad event $E_{2}: s^{\prime}$ is equal to one of the labels discovered in the emulation of $O^{\mathcal{I}}(C)$ (but not present in the current $Q_{B}$ ).
- Bad event $E_{3}: s^{\prime}$ is a new label, but the label chosen by $\widehat{B}$ is one of the labels used in the emulation of $O^{\mathcal{I}}(C)$. (Note that in this case the execution of $\widehat{B}$ will not use any previously used labels in $Q_{B}$.

It is easy to see that as long as none of the events $E_{1}, E_{2}, E_{3}$ happen, the execution of $B^{\prime}$ and $\widehat{B}$ proceeds statistically the same. Therefore, to prove Claim 3.17 and so Claim 3.16 it is sufficient to bound the probability of the events $E_{1}, E_{2}, E_{3}$ as we do below.
Claim 3.18. $\operatorname{Pr}\left[E_{3}\right]<\operatorname{negl}(n)$.
Proof. This is because (as we described at the beginning of this subsection above) the size of $S$ is $n^{\omega(1)}$ but the number of labels discovered in the obfuscation phase is at most $k=\operatorname{poly}(n)$. Therefore the probability that a random label from $S$ after excluding labels in $Q_{B}$ (which is also of polynomial size) hits one of at most $k$ possible labels is $\leq k /\left(|S|-\left|Q_{B}\right|\right)=\operatorname{negl}(n)$. Therefore, the probability that $E_{3}$ happens for any of the oracle quries in the execution of $\widehat{B}$ is also $\operatorname{negl}(n)$.

Claim 3.19. $\operatorname{Pr}\left[E_{2}\right]<\delta /(3 \log |G|)<\delta / 3$.
Proof. We will prove this claim using the randomness of $m \in[\ell]$. Note that every time that a label $u$ is discovered in learning phase 1 , this label $u$ cannot be discovered "again", since it will be in $Q_{B}$ from now on. Therefore, the number of possible indexes of $i \in[\ell]$ such that during the $i^{\prime}$ th execution of the learning phase 1 we discover a label out of $Q_{B}$ is at most $k$. Therefore, over the randomness of $m \leftarrow[\ell]$ the probability that the $m+1$ 'st execution discovers any new labels (generated in the obfuscation phase) is at most $k / \ell \leq \delta /(3 \log |G|)$.

Claim 3.20. $\operatorname{Pr}\left[E_{1}\right]<\delta / 3$.
Proof. Call $i \in[\ell]$ a bad index, if event $E_{3}$ happens conditioned on $m=i$ during the execution of $B^{\prime}$ (which is the $(m+1)$ 's execution of learning phase 1 ). Whenever $E_{3}$ happens at any moment, it means that the vector $\overline{\mathbf{c}}$ is not currently in $W\left(Q_{B}\right)$, but it will be added $W$ just after this query is made. We will show (Lemma 3.21 below) that the size of $\operatorname{span}_{\mathbb{Z}}(W)$ will at least double after this oracle query for some set $\operatorname{span}_{\mathbb{Z}}(W)$ that depends on $W$ and that $\operatorname{span}_{\mathbb{Z}}(W) \subseteq G^{k}$, and so $\left|\operatorname{span}_{\mathbb{Z}}(W)\right| \leq|G|^{k}$. As a result the number of bad indexes $i$ will be at most $\log |G|^{k}=k \log |G|$. Therefore, over the randomness of $m \in[\ell]$ the probability that $m+1$ is a bad index is at most $k \log |G| / \ell \leq \delta / 3$

Lemma 3.21. Let $W \subseteq G^{k}$ for some abelian group $G$. Let $\operatorname{span}_{\mathbb{Z}}(W)=\left\{\sum_{\mathbf{w} \in W} a_{\mathbf{w}} \mathbf{w} \mid a_{\mathbf{w}} \in \mathbb{Z}\right\}$ be the module spanned by $W$ using integer coefficients. If $\overline{\mathbf{c}} \notin \operatorname{span}_{\mathbb{Z}}(W)$, then it holds that

$$
\left|\operatorname{span}_{\mathbb{Z}}(W \cup\{\overline{\mathbf{c}}\})\right| \geq 2 \cdot\left|\operatorname{span}_{\mathbb{Z}}(W)\right| .
$$

Proof. Let $A=\operatorname{span}_{\mathbb{Z}}(W)$ and let $B=\left\{\overline{\mathbf{c}}+\mathbf{w} \mid \mathbf{w} \in \operatorname{span}_{\mathbb{Z}}(W)\right\}$ be $A$ shifted by $\overline{\mathbf{c}}$. It holds that $|A|=|B|$ and $A \cup B \subset \operatorname{span}_{\mathbb{Z}}(W \cup\{\overline{\mathbf{c}}\})$. It also holds that $A \cap B=\varnothing$, because otherwise then we would have: $\exists i, j: \mathbf{w}+\overline{\mathbf{c}}=\mathbf{w}^{\prime}$ for $\mathbf{w}, \mathbf{w}^{\prime} \in \operatorname{span}_{\mathbb{Z}}(W)$ which would mean $\overline{\mathbf{c}}=\mathbf{w}-\mathbf{w}^{\prime} \in \operatorname{span}_{\mathbb{Z}}(W)$ which is a contradiction. Therefore $\left|\operatorname{span}_{\mathbb{Z}}(W \cup\{\overline{\mathbf{c}}\})\right| \geq|A|+|B|=2 \cdot\left|\operatorname{span}_{\mathbb{Z}}(W)\right|$

Soundness: VBB Simulatability. To derive the soundness we apply Lemma 2.2 as follows. $O$ will be the obfuscator in the ideal model and $\widehat{O}$ will be our obfuscator in the plain model where $z^{\prime}=Q_{B}, W, T, r$ is the extra information output by $\widehat{O}$. The algorithm $U$ will be a similar algorithm to $\widehat{O}$ but only during its learning phase 1 and 2 starting from an already obfuscated $B$. However, $U$ will continue generating $z^{\prime}$ using the actual oracle $\mathcal{I}[G \mapsto S]$ instead of inventing the answers through lazy evaluation. Since the emulation of the oracle during the learning phases, and that all of $Q_{B}, W, T, R$ could be obtained by only having $B$ (and no secret information about the obfuscation phase are not needed) the algorithm $U$ also has the properties needed for Lemma 2.2.

Remark 3.22 (General abelian vs $\mathbb{Z}_{n}$.). Note that when $G=\mathbb{Z}_{n}$ is cyclic, the mapping $\rho_{G}: \mathbb{Z} \mapsto G$ of Lemma 3.4 will be equivalent to simply mapping every $a \in \mathbb{Z}$ to $(a \bmod n) \in G$. Therefore, Definition 3.3 generalizes the notion of $\mathbb{Z}_{n}$ as a ring to general abelian groups, since the multiplication $x \cdot y \bmod n$ in $\mathbb{Z}_{n}$ is the same as a multiplication in which $x$ is interpreted from $\mathbb{Z}$ (as in Definition 3.3) which is equivalent to doing the multiplication inside $G$ according to by Lemma 3.4.

### 3.3 Degree- $O(1)$ Graded Encoding Model

We adapt the following definition from [Pas15] restricted to the degree- $d$ polynomials. For simplicity, as in [Pas15] we also restrict ourselves to the setting in which only the obfuscator generates labels and the obfuscated code only does zero tests, but the proof directly extends to the more general setting of $\left[\mathrm{BGK}^{+} 14, \mathrm{BR} 14\right]$. We also use only one finite ring $R$ in the oracle (whose size could in fact depend on the security parameter) but our impossibility result extends to any sequence of finite rings as well.

Definition 3.23 (Degree- $d$ Ideal Graded Encoding Model). The oracle $\mathcal{M}_{R}^{d}=($ enc, zero) is stateful and is parameterized by a ring $R$ and a degree $d$ and works in two phases. For each $l \in[d]$, the oracle enc $(\cdot, l)$ is a random injective function from the ring $R$ to the set of labels $S=\{0,1\}^{3 \cdot|R|}$.

1. Initialization phase: In this phase the oracle answers enc $(v, l)$ queries and for each query it stores $(v, l, h)$ in a list $\mathcal{L}_{O}$.
2. Zero testing phase: Suppose $p(\cdot)$ is a polynomial whose coefficients are explicitly represented in $R$ and its monomials are represented with labels $h_{1}, \ldots, h_{m}$ obtained through enc $(\cdot, \cdot)$ oracle in phase 1 . Given any such query $p(\cdot)$ the oracle answers as follows:
(a) If any $h_{i}$ is not in $\mathcal{L}_{O}$ (i.e., it is not obtained in phase 1) return false.
(b) If the degree of $p(\cdot)$ is more than $d$ then return false.
(c) Let $\left(v_{i}, l_{i}, h_{i}\right) \in \mathcal{L}_{O}$. If $p\left(v_{1}, \ldots, v_{m}\right)=0$ return true; otherwise false.

Remark 3.24. Remarks 3.8 and 3.9 regarding the stateful vs stateless oracles and the sparsity of the encoding in the context of generic group model apply to the graded encoding model as well. Therefore, as long as the encoding is sparse (which is the case in the definition above whenever $|R|$ is of size $\left.n^{\omega(1)}\right)$ the probability of obtaining any valid label $h=\operatorname{enc}(v, l)$ through any polynomial
time algorithm without it being obtained from the oracle previously (by the same party or another party) becomes negligible, and so the model remains essentially equivalent (up to negligible error) even if the oracle does not keep track of which labels are obtained previously through $\mathcal{L}_{O}$.

We prove the following theorem generalizing a similar result by Pass and shelat [Pas15] who proved this for any finite field; here we prove the theorem for any finite ring.

Theorem 3.25. Let $R$ be any ring of size at most $2^{\operatorname{poly}(n)}$. Let $O$ be any $\epsilon$-approximate $V B B$ obfuscator for general circuits in the ideal degree-d graded encoding model $\mathcal{M}_{R}^{d}$ for $d=O(1)$ where the initialization phase of $\mathcal{M}_{R}^{d}$ happens during the obfuscation phase. Then for any $\delta=1 / \operatorname{poly}(n)$ there is an $(\epsilon+\delta)$-approximate obfuscator $\widehat{O}$ for poly-size circuits in the plain model.

See Section 3.3.1 for the proof of Theorem 3.25. As in previous sections, the following corollary is obtained from Theorem 3.25 by taking $\delta=\epsilon / 2$.

Corollary 3.26. If TDPs exist, then there exists no $(1 / 2-\epsilon)$-approximate VBB obfuscator $O$ for general circuits in the ideal degree-d graded encoding model $\mathcal{M}_{R}^{d}$ for any finite ring $R$ of at most exponential size $|R| \leq 2^{\operatorname{poly}(n)}$ and any constant degree d, assuming the initialization phase of $\mathcal{M}_{R}^{d}$ happens during the obfuscation phase.
[Pas15] state their theorem in a more general model where a sequence of fields of growing size are accessed. For simplicity, we state a simplified variant for simplicity of presentation where only one ring is accessed but we let the size of ring $R$ to depend on the security parameter $n$. Our proof follows the footsteps of [Pas15] but will deviate from their approach when $R \neq \mathbb{Z}_{p}$ by using some of the ideas employed in Section 3.

### 3.3.1 Proving Theorem 3.25

Here we sketch the proof assuming the reader is familiar with the proof of Theorem 3.10 from previous section. The high level structure of the proof remains the same.

Construction. The new obfuscator $\widehat{O}$ will have these phases:

- Emulating obfuscation. $\widehat{O}$ emulates $O^{\mathcal{M}_{R}^{d}}(C)$ to get circuit $B$.
- Learning heavy subspace of space of zero vectors: The learning phase here will be rather simpler than those of Section 3.2.1 and will be just one phase. Here we repeat the learning phase $m$ times where $m$ is chosen at random from $\ell=[[k \cdot \log (|G|) / \delta\rceil]$. The variables $W$ and $T$ will be the same as in Section 3.2.1 with the difference that $W$ will consist of the vector of coefficients for all polynomials whose zero test answer is true.
- The returned obfuscated code will be $\widehat{B}=(B, W, T, r)$ where $r$ is again the randomness needed to run the obfuscated code.
- Executing $\widehat{B}$. To execute $\widehat{B}$ on input $x$, we answer zero test queries as follows. For any query vector (of coefficients) a we test whether $\mathbf{a} \in \operatorname{span}_{\mathbb{Z}}(W) .{ }^{11}$ If $\mathbf{a} \in \operatorname{span}_{\mathbb{Z}}(W)$ then return true, otherwise return false.

[^7]
## Completeness and Soundness.

- The completeness follows from the same argument given for the soundness of Construction 3.2.1. Namely, the execution of $\widehat{B}$ is identical to the execution of the $m+1$ 's learning phase (as if it exists) up to a point where we return a wrong false answer to an answer that is indeed a zero polynomial. (Note that the converse never happens). However, when such event is about to happen, the size of $\operatorname{span}_{\mathbb{Z}}(W)$ will double. Since the size of $\operatorname{span}_{\mathbb{Z}}(W)$ is at most $|R|^{k}$, if we choose $m$ at random from $[\ell]$ the probability of the bad event (of returning a wrong false in $m+1$ 'st execution) is at most $k \log |R| / \ell=\delta$.
- The soundness follows from Lemma 2.2 similarly to the way we proved the soundness of the construction of Section 3.2.1.

Extension to avoid initialization. In Theorem 3.25 we have a restriction which says that the initialization phase must happen during the obfuscation phase only. We can extend the proof of Theorem 3.25 to the case that we don't have this restriction. This entails allowing the obfuscator $O$ and the obfuscated circuit $B$ to ask any type of query (be it initialization phase queries or zero-testing queries) during their execution. The reason that we can avoid this restriction is that, whenever the obfuscated circuit $B$ asks an initialization phase query enc $(v, l)$, we can treat it as a polynomial containing $v$. enc $(1, l)$ and using that we can find out whether we should answer this query randomly or using one of the previous labels. This is very similar to the method that we employed in the learning and execution phases of generic group model case.

Claim 3.27. Let $R$ be any ring of size at most $2^{\operatorname{poly}(n)}$. Let $O$ be any $\epsilon$-approximate VBB obfuscator for general circuits in the ideal degree-d graded encoding model $\mathcal{M}_{R}^{d}$ for $d=O(1)$, Then for any $\delta=1 / \operatorname{poly}(n)$ there is an $(\epsilon+\delta)$-approximate obfuscator $\widehat{O}$ for poly-size circuits in the plain model.

Proof. Suppose that obfuscated circuit is $B$, and let $\left\{h_{i}=\operatorname{enc}\left(v_{i}, l_{i}\right)\right\}_{1}^{n}$ be the obfuscator's queries. We already know that n is less than the running time of obfuscator. We might learn some pair of $\left(h_{i}, v_{i}\right)$ during the learning phase.

Construction. The new $\epsilon$-approximate obfuscator $\widehat{O}$ will have these phases:

- Emulating obfuscation. same as previous case.
- Learning obfuscator's queries and heavy subspace of space of zero vectors: We do exactly what we did in previous learning phase. Also if obfuscated circuit asked initialization phase queries, we memorize it.
- The returned obfuscated code will be $\widehat{B}=(B, W, T, r)$ where $r$ is again the randomness needed to run the obfuscated code.
- Executing $\widehat{B}$. To execute $\widehat{B}$ on input $x$, we do as follows. If we saw query enc $(v, l)$ : First we check, if we memorized query $\operatorname{enc}(v, l)$ before, we answer it using memorized queries list otherwise we answer it randomly. Also we treat enc(v,l) as a polynomial v.enc $(1, l)$. We answer zero test queries as follows. For any query vector (of coefficients) a we test whether $\mathbf{a} \in \operatorname{span}_{\mathbb{Z}}(W) .{ }^{12}$ If $\mathbf{a} \in \operatorname{span}_{\mathbb{Z}}(W)$, return true, otherwise return false.

[^8]
## Completeness and Soundness.

- The proof of completeness is same as previous case. The only difference is that here we need to be sure that we answer initialization phase query correctly (call it event $E$ ). Let $j_{i}$ be the index such that we saw the query $\operatorname{enc}\left(v_{i}, l_{i}\right)$ for the first time. $E$ happens if we hit one of the index $j_{i}$. Since we chose $m$ at random, we can always bound $\operatorname{pr}(E)$ by choosing the right $l$.
- The soundness is same as previous case.

Remark 3.28. Note that our proof of Theorem 3.25 does not assume any property for the multiplication (even the associativity!) other than assuming that it is distributive. Distributivity is needed by the proof since we need to be able to conclude that the summation of the vectors of the coefficients of two zero polynomials is also the vector of the coefficients of a zero polynomial; the latter is implied by distributivity.

## 4 Impossibility of VBB in the Random TDP Model

In this section we formally prove Theorem 1.3 showing that any obfuscator $O$ with access to a random trapdoor permutation oracle $T$ can be transformed into a new obfuscator $\widehat{O}$ in the plain model (no access to an ideal oracle) with some loss in correctness. We start by defining the random trapdoor permutation model and TDP query tuples followed by the formalization of Theorem 1.3.

Definition 4.1 (Random Trapdoor Permutation). For any security parameter $n$, a random trapdoor permutation (TDP) oracle $T_{n}$ consists of three subroutines ( $G, F, F^{-1}$ ) as follows:

- $G(\cdot)$ is a random permutation over $\{0,1\}^{n}$ mapping trapdoors $s k$ to a public indexes $p k$.
- $F[p k](x)$ : For any fixed public index $p k, F[p k](\cdot)$ is a random permutation over $\{0,1\}^{n}$.
- $F^{-1}[s k](y)$ : For any fixed trapdoor $s k$ such that $G(s k)=p k, F^{-1}[s k](\cdot)$ is the inverse permutation of $F[p k](\cdot)$, namely $F^{-1}[s k](F[p k](x))=x$.

Definition 4.2 (TDP query tuple). Given a random TDP oracle $T_{n}=\left(G, F, F^{-1}\right)$, a TDP query tuple consists of three query-answer pairs $\left(V_{G}, V_{F}, V_{F^{-1}}\right)$ where:

- $V_{G}=(s k, p k)$ represents a query to $G$ on input $s k$ and its corresponding answer $p k$
- $V_{F}=((p k, x), y)$ represents a query to $F[p k]$ on input $x$ and its corresponding answer $y$
- $V_{F^{-1}}=\left((s k, y), x^{\prime}\right)$ represents a query to $F^{-1}[s k]$ on $y$ and its corresponding answer $x^{\prime}$

We say that a TDP query tuple $\left(V_{G}, V_{F}, V_{F^{-1}}\right)$ is consistent if $x=x^{\prime}$.
Definition 4.3 (Partial TDP query tuple). A partial TDP query tuple is one where one or more of the elements of the tuple are unknown and we denote the missing elements with a period. For example, we say a query set $Q$ contains a TDP query tuple $\left(\cdot, V_{F}, \cdot\right)$ if it contains the query-answer pair $V_{F}=((p k, x), y)$ but is missing the query-answer pairs $V_{G}=(s k, p k)$ and $V_{F^{-1}}=\left((s k, y), x^{\prime}\right)$.

Theorem 4.4 (Theorem 1.3 formalized). Let $O$ be an $\epsilon$-approximate obfuscator for poly-size circuits in the random TDP oracle model. Then, for any $\delta=1 / \operatorname{poly}(n)$, there exists an $(\epsilon+\delta)$-approximate obfuscator $\widehat{O}$ in the plain model for poly-size circuits.

Before proving Theorem 4.4, we state a corollary of this theorem to rule out approximate VBB obfuscation in the ideal TDP model. Since [BP13] showed that assuming TDPs exist, ( $1 / 2-1 /$ poly $)$ approximate VBB obfuscator does not exist for general circuits, we obtain the following corollary by taking $\delta=\epsilon / 2$.

Corollary 4.5. If TDPs exist, then there exists no $(1 / 2-\epsilon)$-approximate VBB obfuscator $O$ for general circuits in the ideal random TDP model for any $\epsilon=1 / \operatorname{poly}(n)$.

The proof of Theorem 4.4 now follows in the next two sections. We will first describe the algorithm of the obfuscator in the plain model, and then will analyze its completeness and soundness.

### 4.1 The Construction

We first describe how the new obfuscator $\widehat{O}$ generates some data $\widehat{B}$, and then we will show how to use $\widehat{B}$ to run the new obfuscated circuit in the plain model. We also let $l_{O}, l_{B}=\operatorname{poly}(n)$, respectively, be the number of queries asked by the obfuscator $O$ and the obfuscated code $B$ to the random trapdoor permutation oracle $T$. Note that, for simplicity of exposition, we assume the adversary only asks the oracle for queries of size $n$ (i.e. the domain of the permutations in $T$ are of fixed size $n$ ). However, as mentioned in Section 1.2.3, we can easily extend the argument to handle $O(\log (n))$-size or $\omega(\log (n))$-size queries to $T$.

### 4.1. 1 How to Obfuscate

The new obfuscator $\widehat{O}$ in plain model. Given an $\epsilon$-approximate obfuscator $O$ in the random TDP model, we construct a plain-model obfuscator $\widehat{O}$ such that, given a circuit $C \in\{0,1\}^{n}$, works as follows:

1. Emulation phase: Emulate $O^{T}(C)$. Let $Q_{O}$ represent the set of queries asked by $O^{T}$ and their corresponding answers. We initialize $Q_{O}=\varnothing$. For every query $q$ asked by $O^{T}(C)$, we would answer the query uniformly at random conditioned on the answers to previous queries.
2. Canonicalize $B$ : Let the obfuscated circuit $B$ be the output of $O(C)$. Modify $B$ so that, before asking any query of the form $F^{-1}[s k](y)$, it would first ask $G(s k)$ to get some answer $p k$ followed by $F^{-1}[s k](y)$ to get some answer $x$ then finally asks $F[p k](x)$ to get the expected answer $y$.
3. Learning phase: Set $Q_{B}=\varnothing$. Let the number of iterations to run the learning phase be $m=2 l_{B} l_{O} / \delta$ where $l_{B} \leq|B|$ represents the number of queries asked by $B$ and $l_{O} \leq|O|$ represents the number of queries asked by $O$. For $i=\{1, \ldots, m\}$ :

- Choose $x_{i} \stackrel{\$}{\leftarrow} D_{|C|}$
- Run $B\left(x_{i}\right)$. For every query $q$ asked by $B\left(x_{i}\right)$ :
- If $(q, a) \in Q_{O} \cup Q_{B}$ for some answer $a$, answer consistently with $a$
- Otherwise, answer $q$ uniformly at random and conditioned on the answers of previous related queries in $Q_{O} \cup Q_{B}$
- Let $a$ be the answer to $q$. If $(q, a) \notin Q_{B}$, add the pair $(q, a)$ to $Q_{B}$

4. The output of the obfuscation algorithm will be $\widehat{B}=\left(B, Q_{B}, R\right)$ where $R=\left\{r_{1}, \ldots, r_{|B|}\right\}$ is a set of (unused) oracle answers that are generated uniformly at random.

### 4.1.2 How to Execute

To execute $\widehat{B}$ on an input $x$ using $\left(B, Q_{B}, R\right)$ we simply emulate $B(x)$. For every query $q$ asked by $B(x)$, if $(q, a) \in Q_{B}$ for some $a$ then return $a$. Otherwise, answer randomly with one of the answers $a$ in $R$ and add $(q, a)$ to $Q_{B}$.

### 4.2 Completeness and Soundness

Completeness: Approximate functionality. Consider two separate experiments (real and ideal) that construct the plain-model obfuscator exactly as described in section 4.1 but differ when executing $\widehat{B}$. Specifically, in the real experiment, $\widehat{B}$ emulates $B(x)$ on a random input $x$ using $Q_{B}$ and $R$, whereas in the ideal experiment, we execute $\widehat{B}$ and answer $B(x)$ 's queries using the actual oracle $T$ instead. In essence, in the real experiment, we can think of the execution as $B^{\widehat{T}}(x)$ where $\widehat{T}$ is the TDP oracle simulated by $\widehat{B}$ using $Q_{B}$ and $R$ as the oracle's answers (without knowing $Q_{O}$, which is part of oracle $T$ ). We will contrast the real experiment with the ideal experiment and show that the statistical distance between these two executions is at most $\delta$. In order to achieve this, we will identify the events that differentiate between the executions $B^{T}(x)$ and $B^{\widehat{T}}(x)$, and to that end we will make use of the following two lemmas:

Lemma 4.6 ([GKLM12]). Let $B$ be a canonical oracle-aided algorithm that asks $t$ queries to a TDP oracle $T$. Let $E_{G}$ be the event that $B$ asks a query of the form $V_{G}=(s k, p k)$ after asking either query $V_{F}=((p k, x), y)$ and/or $V_{F-1}=((s k, y), x)$ from the TDP query tuple $\left(V_{G}, V_{F}, V_{F-1}\right)$. Then $\operatorname{Pr}\left[E_{G}\right] \leq O\left(t^{2} / 2^{n}\right)$.

Lemma 4.7 ([GKLM12]). Let $B$ be an oracle-aided algorithm that asks $t$ queries to a TDP oracle $T$ and let $Q$ be the set of queries that $B$ have issued. Then for any new query $x$, the answer is either $(\mathbf{1})$ determined completely by $Q$ or $(\mathbf{2})$ is drawn from a distribution with a statistical distance of $O\left(t / 2^{n}\right)$ away from the uniform distribution.

Now let $q$ be a new query that is being asked by $B^{\widehat{T}}(x)$. We present a case-by-case analysis of all possible queries to identify the cases that can cause discrepancies between the real and ideal experiments:

- Case 1: If $q$ is determined by the queries in $Q_{B}$ in the real experiment then it is also determined by $Q_{B}$ in the ideal experiment.
- Case 2: If $q$ is not determined by $Q_{B} \cup Q_{O}$ in the ideal experiment then it is also not determined by $Q_{B}$ in the real experiment. In the ideal experiment the query will be answered randomly and consistently with respect to $Q_{B} \cup Q_{O}$ whereas in the real experiment the query will be answered randomly and consistently with respect to $Q_{B}$. By Lemma 4.7 , the answers will be from a distribution that is statistically close to uniform.
- Case 3: If $q$ is not determined by $Q_{B}$ in the real experiment then, depending on the queries in $Q_{O}$, it may or may not be so the ideal experiment:
- Case 3a: The query $q$ is in $Q_{O}$. In that case, in the real experiment, the answer would be random whereas in the ideal experiment it would use the correct answer from $Q_{O}$.
- Case 3b: The query $q$ is of type $V_{G}=(s k, p k)$ and the corresponding partial TDP query tuple (., $\left.V_{F}, V_{F-1}\right)$ is in $Q_{O}$
- Case 3c: The query $q$ is of type $V_{F}=((p k, x), y)$ and the corresponding partial TDP query tuple $\left(V_{G}, ., V_{F^{-1}}\right)$ is in $Q_{O}$
- Case 3d: The query $q$ is of type $V_{F^{-1}}=((s k, y), x)$ and the corresponding partial TDP query tuple ( $\left.V_{G}, V_{F},.\right)$ is in $Q_{O}$
- Case 3e: The query $q$ is of type $V_{F}=((p k, x), y)$ and $V_{G}=(s k, p k)$ is in $Q_{B}$, but $V_{F^{-1}}=((s k, y), x)$ is in $Q_{O}$
- Case 3f: The query $q$ is of type $V_{F^{-1}}=((s k, y), x)$ and $V_{G}=(s k, p k)$ is in $Q_{B}$, but $V_{F}=((p k, x), y)$ is in $Q_{O}$
- Case 3g: The query $q$ is of type $V_{F}=((p k, x), y)$ and $V_{F^{-1}}=((s k, y), x)$ is in $Q_{B}$, but $V_{G}=(s k, p k)$ is in $Q_{O}$
- Case 3h: The query $q$ is of type $V_{F^{-1}}=((s k, y), x)$ and $V_{F}=((p k, x), y)$ is in $Q_{B}$, but $V_{G}=(s k, p k)$ is in $Q_{O}$

We note that the bad events that can cause any differences between the real and ideal experiments are case 2 and parts of case 3 . For case 2, Lemma 4.7 ensures that this event happens with negligible probability. For case $3 a$, learning heavy queries would diminish the effect of this event. For cases $3 b, 3 e$, and $3 f$, Lemma 4.6 ensures that this event happens with negligible probability since $V_{G}$ was issued after $V_{F}$ and/or $V_{F-1}$ was asked. For cases $3 c$ and $3 d$, the remaining query from the tuple would have been defined in $Q_{O}$ and is thus captured during the learning of heavy queries. For case $3 g$, if $V_{G}$ and $V_{F^{-1}}$ were asked during the emulation or learning phases, then $V_{F}$ would also be defined and thus can be learned. However, if $V_{F^{-1}}$ was asked during the execution phase then, due the canonicalization of $B$, it would have to ask $V_{G} \in Q_{O}$ which reduces to case $3 a$. Similarly, for case $3 h$, due the canonicalization of $B$, we would have to ask $V_{G} \in Q_{O}$ and this reduces to case $3 a$ once again.

For any $x$, define $E_{k}(x)$ to be the event that case $k$ happens and let event $E(x)=\left(E_{2}(x) \vee\right.$ $\left.E_{3 a}(x) \vee E_{3 b}(x) \vee E_{3 e}(x) \vee E_{3 f}(x)\right)$. Assuming that event $E$ does not happen, the output distributions of $B^{T}(x)$ and $B^{\widehat{T}}(x)$ are identical. More formally, the probability of correctness for $\widehat{O}$ is:

$$
\begin{aligned}
\operatorname{Pr}_{x}\left[B^{\widehat{T}}(x) \neq C(x)\right] & =\operatorname{Pr}_{x}\left[B^{\widehat{T}}(x) \neq C(x) \wedge \neg E(x)\right]+\operatorname{Pr}_{x}\left[B^{\widehat{T}}(x) \neq C(x) \wedge E_{1}(x)\right] \\
& \leq \operatorname{Pr}_{x}\left[B^{\widehat{T}}(x) \neq C(x) \wedge \neg E(x)\right]+\operatorname{Pr}_{x}[E(x)]
\end{aligned}
$$

By the approximate functionality of $O$, we have that:

$$
\operatorname{Pr}_{x}\left[O^{T}(C)(x) \neq C(x)\right]=\operatorname{Pr}_{x}\left[B^{T}(x) \neq C(x)\right] \leq \epsilon(n)
$$

Therefore,

$$
\operatorname{Pr}_{x}\left[B^{\widehat{T}}(x) \neq C(x) \wedge \neg E(x)\right]=\operatorname{Pr}_{x}\left[B^{T}(x) \neq C(x) \wedge \neg E(x)\right] \leq \epsilon
$$

We are thus left to show that $\operatorname{Pr}[E(x)] \leq \delta$. By Lemma 4.7, $\operatorname{Pr}\left[E_{2}(x)\right] \leq \operatorname{negl}(n)$ and by Lemma 4.6, $\operatorname{Pr}\left[E_{3 b} \vee E_{3 e}(x) \vee E_{3 f}(x)\right] \leq \operatorname{negl}(n)$ via a union bound. The probability of event $E_{3 a}$ was already given in [CKP15], but for the sake of completeness we show our version of the analysis here. As a result, we get that $\operatorname{Pr}[E(x)] \leq \delta / 2+\operatorname{negl}(n) \leq \delta$.

Claim 4.8. It holds that $\operatorname{Pr}_{x}\left[E_{3 a}(x)\right] \leq \delta / 2$.
Proof. Let $\left(q_{1}, \ldots, q_{l_{B}}\right)$ be the sequence of queries asked by $B^{\widehat{T}}(x)$ where $l_{B} \leq|B|$, and let $q_{i, j}$ be the $j^{\text {th }}$ query that is asked by $B^{T}\left(x_{i}\right)$ during the $i^{\text {th }}$ iteration of the learning phase. We define $E_{3 a}^{j}(x)$ to be the event that the $j^{\text {th }}$ query of $B(x)$ is in $Q_{O}$ but not in $Q_{B}$. We also define $p_{q, j}$ to be the
probability that $q_{j}=q$ for any query $q$ and $j \in\left[l_{B}\right]$. We can then write the probability of $E_{3 a}$ as follows:

$$
\begin{aligned}
\underset{x}{\operatorname{Pr}}\left[E_{3 a}(x)\right] & \leq \underset{x}{\operatorname{Pr}}\left[E_{3 a}^{1}(x) \vee \ldots \vee E_{3 a}^{l_{B}}(x)\right] \\
& =\sum_{j=1}^{l_{B}} \operatorname{Pr}_{x}\left[\neg E_{3 a}^{1}(x) \wedge \ldots \wedge \neg E_{3 a}^{j-1}(x) \wedge E_{3 a}^{j}(x)\right] \\
& \leq \sum_{j=1}^{l_{B}} \sum_{q \in Q_{O}} \operatorname{Pr}_{x}\left[q_{j}=q \wedge\left(q_{1, j} \neq q \wedge \ldots \wedge q_{m, j} \neq q\right)\right] \\
& =\sum_{j=1}^{l_{B}} \sum_{q \in Q_{O}} p_{q, j}\left(1-p_{q, j}\right)^{m} \leq \sum_{j=1}^{l_{B}} \sum_{q \in Q_{O}} \frac{1}{m} \leq \sum_{j=1}^{l_{B}} \frac{l_{O}}{m}=\frac{l_{B} l_{O}}{m} .
\end{aligned}
$$

Thus, given that $m=2 l_{B} l_{O} / \delta$, we get $\operatorname{Pr}\left[E_{3 a}(x)\right] \leq \delta / 2$.

Soundness: VBB Simulatability. To show that the security property is satisfied, it suffices to provide a PPT algorithm $U^{T}$ in the ideal TDP model that takes as input $O^{T}(C)$ for some circuit $C$ and outputs a distribution that is statistically close to the output distribution of $\widehat{O}$. If that is the case, we can invoke Lemma 2.2 and conclude that $\widehat{O}$ is also VBB-secure.

The description of $U$ is precisely the same as Steps 2-4 of the procedure detailed in Section 4.1 except that queries made by $B=O^{T}(C)$ are answered using oracle $T$ instead of being randomly simulated. If we let $\left(B, Q_{B}, R\right)$ be the output of $U^{T}\left(O^{T}(C)\right)$ then we can easily see that it is identically distributed to the output distribution of $\widehat{O}$ since, in both cases, $Q_{B}$ has query-answers with consistent and random TDP query tuples. They differ only by how these query answers are generated ( $U^{T}$ answers them using $T$, while $\widehat{O}$ simulates them using lazy evaluation with respect to some oracle $\widehat{T}$ distributed the same as $T$ ).

### 4.3 Extension to hierarchical random TDP

In this section, we reason that the proof for the ideal TDP case can be extended to hierarchical TDP oracles as well. We start by defining how the oracle for the random hierarchical trapdoor permutation primitive changes from Definition 4.1.
Definition 4.9 (Random Hierarchical Injective Trapdoor Functions). For any security parameter $n$ and $l=\operatorname{poly}(n)$, an $l$-level random hierarchical injective trapdoor function (HTDF) oracle $T_{n}^{l}$ consists of $2 l+3$ subroutines $\left(\left\{J_{i}\right\}_{i=1}^{l+1},\left\{K_{i}\right\}_{i=0}^{l+1}\right)$ defined as follows:

- $K_{i}\left[\mathrm{ID}_{i-2}, i d_{i-1}\right]\left(t d_{i}\right)$ : An injective function, indexed by identity vector $\mathrm{ID}_{i-2}=\left(i d_{0}, \ldots, i d_{i-2}\right)$ and $i d_{i-1}$, that accepts as input an $i$-level trapdoor $t d_{i} \in\{0,1\}^{m}$ and outputs a randomly chosen identity $i d_{i} \in\{0,1\}^{n}$ where $m=10 n l$ if $i \in[1, l]$ and $m=n$ (i.e. it is a permutation) if $i=\{0, l+1\}$.
- $J_{i}\left[\mathrm{ID}_{i-2}, t d_{i-1}\right]\left(i d_{i}\right):$ An injective function, indexed by identity vector $\mathrm{ID}_{i-2}=\left(i d_{0}, \ldots, i d_{i-2}\right)$ and $t d_{i-1}$ that, given the identity $i d_{i} \in\{0,1\}^{n}$, outputs the corresponding trapdoor $t d_{i} \in$ $\{0,1\}^{m}$ where $m=10 n l$ if $i \in[1, l]$ and $m=n$ (i.e. it is a permutation) if $i=\{0, l+1\}$.

Note that, for any fixed $\mathrm{ID}_{i-2}$, if $t d_{i}=J_{i}\left[\mathrm{ID}_{i-2}, t d_{i-1}\right]\left(i d_{i}\right)$ and $i d_{i-1}=$ $K_{i-1}\left[\mathrm{ID}_{i-3}, i d_{i-2}\right]\left(t d_{i-1}\right)$ then $i d_{i}=K_{i}\left[\mathrm{ID}_{i-2}, i d_{i-1}\right]\left(t d_{i}\right)$. In other words, we can think of $K_{i}$ as the inverse of $J_{i}$ only if the indices of the two functions match (that is, the trapdoor $t d_{i-1}$ indexing $J_{i}$ corresponds to the identity $i d_{i-1}$ indexing $K_{i}$ ).

Remark 4.10. It is also crucial to note that we used (sparse) injective functions for generating the intermediate levels of trapdoor. Such a change was made in order to obtain interesting primitives from this oracle, such as fully-secure hierarchical identity-based encryption (HIBE). If permutations were used instead, we would only achieve HIBE with security against adversaries that do not choose an identity for the permutation $F$ to attack. Furthermore, removing $K_{i}$ for $i \in[1, l]$ as a way to prevent this attack's capability hinders our ability to perform the canonicalization procedure for the obfuscated circuit.

Remark 4.11. For the special case of 1-level HTDF (i.e. TDP), we only have three permutations: $K_{0}, K_{1}\left[i d_{0}\right]$ and $J_{1}\left[t d_{0}\right]$, which correspond to permutations $G, F[p k]$, and $F^{-1}[s k]$, respectively in the language of TDP that we used in Definition 4.1. Note that here, we would refer to 0-level identities as master public keys and 0-level trapdoors as master secret keys.

We also present a variant of TDP query tuples that generalizes Definition 4.2 to work with hierarchical injective trapdoor functions.

Definition 4.12 (HTDF query tuple). Given a random $l$-level HTDF oracle $T_{n}^{l}=\left(\left\{J_{i}\right\}_{i=1}^{l+1},\left\{K_{i}\right\}_{i=0}^{l+1}\right)$, an $i$-level HTDF query tuple consists of three (possibly) related query-answer pairs ( $V_{K_{i-1}}, V_{K_{i}}, V_{J_{i}}$ ) where, for any fixed $\mathrm{ID}_{i-2}=\left(i d_{0}, \ldots, i d_{i-2}\right)$ :

- $V_{K_{i-1}}=\left(t d_{i-1}, i d_{i-1}\right)$ represents a query to $K_{i-1}\left[\mathrm{ID}_{i-3}, i d_{i-2}\right]$ on input $t d_{i-1}$ and its corresponding answer $i d_{i-1}$
- $V_{K_{i}}=\left(\left(i d_{i-1}, t d_{i}\right), i d_{i}\right)$ represents a query to $K_{i}\left[\mathrm{ID}_{i-2}, i d_{i-1}\right]$ on input $t d_{i}$ and its corresponding answer $i d_{i}$
- $V_{J_{i}}=\left(\left(t d_{i-1}, i d_{i}\right), t d_{i}^{\prime}\right)$ represents a query to $J_{i}\left[\mathrm{ID}_{i-2}, t d_{i-1}\right]$ on input $i d_{i}$ and its corresponding answer $t d_{i}^{\prime}$

We say that an $i$-level HTDF query tuple is consistent if $t d_{i}=t d_{i}^{\prime}$.
Remark 4.13. For the purposes of comparison, we note that, for the special case of 1-level HTDF (i.e. TDP), we only have TDP query tuples of the form $\left(V_{K_{0}}, V_{K_{1}}, V_{J_{1}}\right)=\left(V_{G}, V_{F}, V_{F}^{-1}\right)$. Thus, $V_{G}=(s k, p k)$ represents a query to $G$ on $s k=t d_{0}$ and the answer $p k=i d_{0}, V_{F}=((p k, x), y)$ represents a query to $F_{p k}$ on $x=t d_{1}$ and the answer $y=i d_{1}$, and $V_{F^{-1}}=\left((s k, y), x^{\prime}\right)$ represents a query to $F_{s k}^{-1}$ on $y$ and the answer $x^{\prime}$, which should be $x$ if the tuple is consistent.

Extension of the proof. The extension of the impossibility result to random HTDF is straightforward, so we will outline the main differences between the TDP case and describe how to resolve the issues that are related to this oracle. First, we still perform the normalisation procedure on $\widehat{O}$ and $B$ where the query behaviour of these algorithms are modified such that for any query $q$ of the form $J_{i}\left[\mathrm{ID}_{i-2}, t d_{i-1}\right]\left(i d_{i}\right)$, we first ask $K_{i-1}\left[\mathrm{ID}_{i-3}, i d_{i-2}\right]\left(t d_{i-1}\right)$ to get $i d_{i-1}$. This allows us to discover whether we have a query $K_{i}\left[\mathrm{ID}_{i-2}, i d_{i-1}\right]\left(t d_{i}\right)$ whose answer is $i d_{i}$, in which case we can answer $q$ with $t d_{i}$. This procedure ensures that all query tuples that contain $J_{i}$ queries are consistent.

We now turn to verifying whether the proof of approximate functionality for TDP holds in this case as well and, in particular, focus on the event $E(x)$ that was defined Section 4.2. The main issue that we have to consider, which is unique to the HTDF case, is the possibility that different consistent TDP query tuples can be related to each other, and an overlap between these queries may cause an inconsistency in one of the tuples. Specifically, an $i$-level TDP query tuple of the form $\left(V_{K_{i-1}}, \cdot, \cdot\right)$ might overlap with an $(i-1)$-level TDP query tuple $\left(\cdot, \cdot, V_{J_{i-1}}\right)$ from $Q_{O}$, where the
answer of $V_{K_{i-1}}$ is inconsistent with that of $V_{J_{i-1}}$. However, our normalisation procedure prevents precisely this issue as any TDP query tuple that contains $V_{J_{i-1}}$ must also have $V_{K_{i-1}}$, which means that the queries should not overlap otherwise event $E_{1}$ occurs leading to a contradiction to our initial assumption.

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[^1]:    ${ }^{1}$ The emulation here and in next steps would require the idealized model $\mathcal{I}$ to have an efficient "lazy evaluation" procedure. For example lazy evaluation for random oracles chooses a random answer (different from previous ones) given any new query.

[^2]:    ${ }^{2}$ More formally, using the rank argument of [Pas15] it can be shown that for the purpose of obfuscation, the two models are equivalent up to arbitrary small $1 / \operatorname{poly}(n)$ completeness error.
    ${ }^{3}$ Even though the summation is technically defined over the group elements, for simplicity we use the addition operation over the labels as well.
    ${ }^{4}$ Note that this is even the case for $\mathbb{Z}_{q}$ when $q$ is a prime power, although finite fields have prime power sizes.

[^3]:    ${ }^{5}$ In general, when the random permutation $R$ is available in all input lengths, we can use a mixture of the above arguments by generating all the oracle queries of length $c \log (n)$ (for a sufficiently large constant $c$ ) during the obfuscation (in the plain model) and representing this randomness in the obfuscated circuit. This issue also exists in the trapdoor permutation and the generic group models and can be handled exactly the same way.

[^4]:    ${ }^{6}$ So the oracle might return $\perp$ even if the two labels are in the range of $\sigma(G)$.

[^5]:    ${ }^{7}$ Note that although the sequence $T$ grows as we proceed in learning phase 1 , we already now that this sequence will not have length more than $d$ since all of these labels that are discovered while executing the obfuscated code has to be generated by the obfuscator, due to the assumption that the sequential execution of the obfuscator followed by the obfuscated code is in generic form. Therefore we can always consider $\mathbf{a}_{i}$ to be of dimension $k$.
    ${ }^{8}$ Note that even though $W\left(Q_{B}\right)$ could always be derived from $Q_{B}$, and even $T$ could be derived from an ordered variant of $Q_{B}$ (in which the order in which $Q_{B}$ has grown is preserved) we still choose to explicitly represent these elements in the obfuscated $\widehat{B}$ to ease the description of the execution of $\widehat{B}$.
    ${ }^{9}$ Note that we do not have access to the set $Q_{\sigma}$ that was used for consistent lazy evaluation of $\sigma(\cdot)$.

[^6]:    ${ }^{10}$ We even allow new labels $t_{i}$ to be discovered during this execution to be appended to $T$, even though that would indirectly lead to an abort!

[^7]:    ${ }^{11}$ Note that we do not solve a system of equations in $R$ and rather search only integer solutions to $\mathbf{x} W=\mathbf{a}$ as we did in Section 3.2.1.

[^8]:    ${ }^{12}$ Note that we do not solve a system of equations in $R$ and rather search only integer solutions to $\mathbf{x} W=\mathbf{a}$ as we did in Section 3.2.1.

