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# On the Generalizations of Identity-Based Encryption <br> Hidden Vector Encryption and Inner-Product 

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## Abstract

Nowadays, public-key cryptographic is widely deployed and successfully used but still a major drawback exists. In fact, from encrypted data a party can either decrypt or cannot learn anything at all about the message other than the intentionally leaked information such as its length. In the recent years, the cloud computing paradigm has emerged as the new standard to use computing resources, such as storage devices, that are delivered as a service over a network. In such a scenario, the notion of public key cryptography is not enough. It would be desirable to specify a decryption policy in the encrypted data in such a way that only the parties who satisfy the policy can decrypt. In a more general form, we may want to only give access to a function of the message, depending on the decryptor's authorization.

Thus, in the last decade researchers have started looking at a more sophisticated type of encryption called functional encryption. A functionality $F$ is a function $F: K \times M \rightarrow \Sigma$ where $K$ is the key space and $M$ is the message space. Then, a functional encryption scheme, in the public-key setting, for $F$ is a special encryption scheme in which, for every key $k \in K$, the owner of the master secret key $m s k$ associated with the master public key $m p k$ can generate a special secret-key $s k_{k}$ that allows the computation of $F(k, m)$ from a ciphertext of $m \in M$ computed under public key $m p k$. In other words, whereas in traditional encryption schemes decryption is an all-or-nothing affair, in functional encryption it is possible to finely control the amount of information that is revealed by a ciphertext. One of the most notable example of functional encryption is identity-based encryption first introduced by Shamir as an alternative to the standard notion of public-key encryption.

In this thesis, we discuss several instantiations of function encryption that can all be seen as generalizations of identity-based encryption. We improve on previous constructions in terms of capabilities and security guarantees.

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## Chapter 1

## Introduction

A set of parties can securely exchange information over an insecure network or storage service by using encryption. In the simpler scenario, called symmetric setting or private-key setting, the parties must first agree on a common secret key $k$ (using a secondary channel assumed to be secure), before they can start any secret communication. In this setting, the secret key $k$ is involved in both the encryption and decryption process. Such a solution can be practical for a small organization but it is not feasible for larger networks such as Internet where millions of parties are involved.

In the 1970s, with the advent of public-key cryptography the issue of having an a prior mutual secret (and thus a secure communication channel) has been solved radically. In the asymmetric setting or public-key setting, the decryption key (also called secret key) differs from the encryption key (also called public key) and it is infeasible to find the decryption key, given the encryption key. In this setting, a party $P$ generates a pair of keys and keeps the decryption key. The encryption key is then published and now any party can send $P$ private messages by encrypting them using $P$ 's encryption key. Then, only $P$ will be able to recover those messages by using its decryption key.

Nowadays, public-key cryptographic is widely deployed and successfully used. But still some issues remain. In fact, from encrypted data a party can either decrypt or cannot learn anything at all about the message other than the intentionally leaked information such as its length. In the recent years, the cloud computing paradigm has emerged as the new standard to use computing resources, such as storage devices, that are delivered as a service over a network. In such a scenario, the notion of public key cryptography is not enough. It would be desirable to specify a decryption policy in the encrypted data in such a way that only the parties who satisfy the policy can decrypt. In a more general form, we may want to only give access to a function of the message, depending on the decryptor's authorization.

With these motivations in mind, researchers have started looking at a more sophisticated type of encryption called functional encryption. A functionality $F$ is a function $F: K \times M \rightarrow \Sigma$ where $K$ is the key space and $M$ is the message space. Then, a functional encryption scheme, in the public-key setting, for $F$ is a special encryption scheme in which, for every key $k \in K$, the owner of the master secret key $m s k$ associated with the master public key $m p k$ can generate a special secret key (also called token) $s k_{k}$ that allows the computation of $F(k, m)$ from a ciphertext of $m \in M$ computed under public key $m p k$. In other words, whereas in traditional encryption schemes decryption is an all-or-nothing affair, in functional encryption it is possible to finely control the amount of information that is revealed by a ciphertext.

One of the most notable example of functional encryption is identity-based encryption (IBE, for short) first introduced by Shamir [Sha85] as an alternative to the standard notion of publickey encryption. In an IBE scheme, the public key associated with a user can be an arbitrary
identity string, such as his email address, and others can send encrypted messages to a user using this arbitrary identity without having to rely on a public-key infrastructure. In these systems, there also exists a trusted central authority capable of generating secret keys for any user in the system. The first practical constructions of an IBE are due to Boneh and Franklin [BF03] and Cocks [Coc01]. While the Boneh-Franklin IBE scheme is based on elliptic curve groups equipped with efficiently computable bilinear maps, the scheme by Cocks is based on the quadratic residuosity problem [Coc01]. In addition to the scheme, Boneh and Franklin also introduced a formal security definition for IBE schemes to model privacy, known as indistinguishability under chosen-identity-attacks, and showed that their scheme was provably secure under this notion based on a new computational problem, known as bilinear Diffie-Hellman problem, in the random oracle model [BR93]. However, access to the data itself is still inherently all-or-nothing. The decryptor either will be able to decrypt the data and learn everything or will not and thus learn nothing. This has motivated the study of anonymous identity-based encryption schemes, where the ciphertext does not leak the identity of the recipient. The Boneh-Franklin IBE scheme was already known to be inherently anonymous but the first anonymous IBE without random oracles was proposed by Boyen and Waters [BW06]. Moreover, anonymity can be leveraged to construct Public key Encryption with Keyword Search (PEKS) schemes, as observed by Boneh et al. [ $\overline{\mathrm{BDOP} 04}]$ and later formalized by Abdalla et al. $\left[\widehat{\left.\mathrm{ABC}^{+} 08\right]}\right.$. Following the work of Boneh and Franklin, Horwitz and Lynn [HL02] proposed the concept of Hierarchical IBE (HIBE, for short), in which users are organized in a tree of depth $L$, with the root being the trusted central authority. In a HIBE scheme, intermediate nodes in the tree are capable of generating secret keys for any of their descendants. Since its introduction, several practical constructions of HIBE schemes have been proposed in the literature (e.g., [GS02, CHK03, BB04, BBG05]), with the first being due to Gentry and Silverberg [GS02], who proposed a scheme based on the BonehFranklin IBE scheme and proved it secure in the random-oracle model based on the bilinear Diffie-Hellman problem.

After the realization of the first Identity-Based Encryption schemes a number of new cryptosystems have provided increasing functionality and expressiveness of decryption capabilities. In particular, one of the most relevant generalization of identity based encryption, known in the literature as Attribute-Based Encryption (ABE, for short), was proposed by Sahai and Waters [SW05] to express complex access policies. Subsequently, Goyal, Pandey, Sahai and Waters [GPSW06] refined this concept into two different formulations of ABE: Key Policy ABE and Ciphertext-Policy ABE. In a key policy ABE, a ciphertexts contains a vector $\boldsymbol{v}=\left(v_{1}, \ldots, v_{n}\right)$ of $n$ Boolean values and a secret key contains a poly-sized boolean formula $\phi$ over $n$ Boolean variables. The formula $\phi$ must evaluates to true over $\boldsymbol{v}$ to be able to recover the encrypted message. On the other side, two notable generalizations of anonymous identity based encryption are known in the literature as hidden vector encryption (HVE, for short) and inner-product encryption (IPE, for short). In an hidden vector encryption scheme, first proposed by Boneh and Waters [BW07], a ciphertext contains a vector $\boldsymbol{v}=\left(v_{1}, \ldots, v_{\ell}\right)$ of $\ell$ elements in $\{0,1\}^{*}$ and a secret key contains a vector $\boldsymbol{w}=\left(w_{1}, \ldots, w_{\ell}\right)$ of $\ell$ elements in $\{0,1\}^{*} \cup\{*\}$ where we refer to $*$ as a wildcard character. The decryptor can then compute a predicate of $\boldsymbol{v}$ and $\boldsymbol{w}$ that is satisfied if and only if for each $i=1, \ldots, \ell, v_{i}=w_{i}$ whenever $v_{i} \neq *$. Applications of such encryption scheme include conjunctive and range searches. Later, Katz, Sahai and Waters [KSW08] proposed inner-product encryption, a more general system to test if a dot product operation over the ring $\mathbb{Z}_{N}$ is equal to 0 , where $N$ is the product of three random primes chosen by the setup algorithm. Despite this apparently restrictive structure, IPE can support conjunction, subset and range queries on encrypted data as well as disjunction, polynomial evaluation, and CNF and DNF formulas. Following the work of Katz, Sahai and Waters, Okamoto and Takashima [OT09] proposed the concept of Hierarchical IPE (HIPE, for short). All the previous constructions of

IPE and HIPE schemes were based on elliptic curve groups equipped with efficiently computable bilinear maps. More recently, Agrawal et al. AFV11] proposed a lattice-based functional encryption scheme for inner product whose security follows from the difficulty of the learning with errors problem (LWE, for short). The LWE problem, first introduced by Regev [Reg05], has received a lot of attention in the recent years due to its versatility and because it is also conjectured to be hard to solve even for quantum adversaries.

One of the most important achievements of the modern theory of Cryptography is the formalization of the notion of a secure cryptosystem given in the seminal work of Goldwasser and Micali [GM84]. There, two notions of security were proposed, one is a game-based notion and the other is simulation-based, and they were shown to be equivalent in the sense that an encryption scheme meets one security definition if and only if it meets the other. The study of simulation-based notions of security for functional encryption was initiated by Boneh, Sahai, and Waters [BSW11] and O'Neill [O’N10]. Specifically, in [BSW11] Boneh et al. showed that there exists a clearly insecure functional encryption scheme that is nonetheless deemed secure by the game-based notion of security. This motivated the study of a simulation-based notion of security for functional encryption. In the same paper [BSW11], Boneh et al. gave impossibility results in the non-programmable random oracle model for simple instances of functional encryption for which, instead, game-based security had been shown to be achievable (again, under appropriate complexity assumptions). Specifically, they showed that it is not possible to construct Identity Based Encryption that is secure against an adaptive adversary that can ask to see the encryption of an unbounded number of messages. Informally, this is because any simulation-based definition that allows the adversary to query for secret keys after seeing the challenge ciphertext must achieve something very similar to non-interactive non-committing encryption for which impossibility results are already known [Nie02].

The impossibility result of [BSW11] was recently shown to hold also in the standard model by Bellare and O'Neill in [BO12] under an additional complexity assumption. Recently, Gorbunov, Vaikuntanathan, and Wee [GVW12], building on the construction of Sahai and Seyalioglu [SS10], gave a functional encryption scheme for all polynomial size circuits that is simulationbased secure against adaptive adversaries that have access to a bounded number of tokens. This is complemented by the recent result of Agrawal, Gorbunov, Vaikuntanathan, and Wee [AGVW12] that shows that there exists no functional encryption scheme for the functionality of poly-size circuits that is simulation-based secure against non-adaptive adversaries that can see one ciphertext and ask for an unbounded number of tokens. O'Neill [O’N10] gave a different notion of simulation-based security and introduced a simple condition on a functionality (called pre-image samplability). O'Neill showed that, for pre-image sampleable functionalities, the game-based and the simulation-based notions of non-adaptive security coincide (just as it happens in regular encryption schemes).

### 1.1 Summary of Our Results

## Generalized Key Delegation for Wildcarded Identity-Based and Inner-Product Encryp-

tion. One of the main applications of IBE and HIBE schemes is email encryption, where users can encrypt a message to the owner of the email address without having to obtain a certified copy of the owner's public key first. Motivated by the fact that many email addresses correspond to groups of users rather than single individuals, Abdalla et al. [ACD ${ }^{+} 06$ ] introduced the concept of identity-based cryptography with wildcards (WIBE, for short). In a WIBE scheme, decryption keys are issued exactly as in a standard HIBE scheme and the main difference lies in the encryption process. More specifically, in a WIBE scheme, the sender can encrypt messages to more general patterns consisting of identity strings and wildcards so that any identity matching the
given pattern can decrypt. For instance, by encrypting a message to the pattern UNISA.DI.*, then any user belonging to the DI department at the University of Salerno would be able to decrypt it.

Unfortunately, like in standard HIBE schemes, the hierarchical key derivation of a WIBE scheme has its limitations. In particular, it does not allow any deviation from the hierarchical structure or prevent from further deriving keys below their identities. In order to overcome these limitations, Abdalla, Kiltz, and Neven [AKN07] introduced the concept of identity-based encryption with wildcard key derivation (WKD-IBE, for short) to generalize the key delegation mechanism of HIBE schemes. In a WKD-IBE primitive, which can be seen as the dual of a WIBE, secret keys are associated with patterns consisting of identity strings and wildcards and the owner of a key can derive keys for any identity that matches the pattern associated with this secret key. In addition to introducing the concept of WKD-IBE, Abdalla, Kiltz, and Neven also proposed several instantiations of the new primitive, both in the standard and random-oracle models, along with new applications such as identity-based broadcast encryption (IBBE, for short) and wildcard signatures. They also indicated that the concepts of WKD-IBE and WIBE could be combined into a universal primitive which allows for more general patterns to be used in both the encryption and key derivation algorithms. We refer to this WKD-IBE with generalized key delegation as WW-IBE. In Chapter 4 we show how to build a fully secure anonymous WWIBE scheme.

Lattice-based Hierarchical Inner Product Encryption. In Chapter 5 , we consider the problem of constructing hierarchical inner-product encryption scheme based on lattices assumptions. To achieve this goal, we extend the lattice-based IPE scheme by Agrawal et al. [AFV11] to the hierarchical setting by employing basis delegation technics by Peikert et al. [CHKP10] and by Agrawal et al. [ABB10]. As the underlying IPE scheme, our new scheme is shown to be weak selective secure based on the difficulty of the learning with errors problem in the standard model, as long as the total number of levels in the hierarchy is a constant.

Adaptive Simulation-Based Secure Constructions for Functional Encryption. In Chapter 6, we consider the problem of designing simulation-based secure functional encryption schemes. We look both at general functionalities (the functionality corresponding to the class of poly-sized circuits) and at a more specialized functionality.

Specifically, we give a general transformation that takes a game-based functional encryption scheme for all polynomial-size circuits and constructs a functional encryption scheme for the same functionality that is simulation-based secure against adaptive adversaries that can ask for a bounded number of tokens and can see one ciphertext. Our transformation is both black-box and for the standard model and is inspired by the work of Feige, Lapidot and Shamir [FLS90]. We remark that by the recent impossibility results of [AGVW12], the restriction to bounded number of tokens is necessary and thus our construction is essentially the best one can hope for.

Then, we show how to construct a HVE scheme whose simulation-based security can be proved under standard assumptions in the bilinear pairing setting in the standard model. Our construction is shown secure against adaptive adversaries obtaining one ciphertext and asking an unbounded number of tokens. Again, this is the best one can hope for in the standard model, given the impossibility result of [BSW11, BO12] for IBE. The only previous simulation-based construction for IBE secure against adaptive adversary was given in [BSW11] and is in the programmable random oracle model and imposes no bound on the number of tokens and ciphertexts obtained by the adversary.

## Chapter 2

## Functional Encryption

We start this chapter by describing the syntactic and security definition of functional encryption. To do so, we will follow the work of Boneh, Sahai, and Waters [BSW11] who initiated the formal study of functional encryption by giving precise definitions of the concept and its security.

Definition 2.0.1 [Functionality] A functionality $F=\left\{F_{n}\right\}$ is a family of functions $F_{n}: K_{n} \times$ $X_{n} \rightarrow \Sigma$ where $K_{n}$ is the key space for parameter $n, X_{n}$ is the plaintext space for parameter $n$ and $\Sigma$ is the output space. Sometimes we will refer to functionality $F$ as a function from $F: K \times X \rightarrow \Sigma$ with $K=\cup_{n} K_{n}$ and $X=\cup_{n} X_{n}$. For simplicity, and without loss of generality, we assume that $\Sigma=\{0,1\}$.

An easy example of a functionality is that for the evaluation of Boolean circuits. Specifically, the Circuit functionality is defined as follow.

Definition 2.0.2 [Circuit Functionality] The Circuit functionality has key space $K_{n}$ equals to the set of all $n$-input Boolean circuits and plaintext space $X_{n}$ the set $\{0,1\}^{n}$ of $n$-bit strings. For $C \in K_{n}$ and $x \in X_{n}$, we have

$$
\operatorname{Circuit}(C, x)=C(x)
$$

Then, informally, a functional encryption scheme for a functionality F defined over ( $K, X$ ) enables the decryptor to evaluate $F(k, x)$ given the encryption of $x \in X$ and a secret key $s k_{k}$ for $k \in K$.

Formally, a functional encryption scheme can be defined as follows.

Definition 2.0.3 [Functional Encryption Scheme] A functional encryption scheme for a functionality $F$ defined over $(K, X)$ is a tuple (Setup, KeyGen, Enc, Dec) of 4 algorithms with the following syntax:

Setup $\left(1^{\lambda}\right)$ outputs public and master secret keys $(m p k, m s k)$ for security parameter $\lambda$;

KeyGen $(m s k, k)$ on input a master secret key $m s k$ and $k e y k \in K$, outputs token $s k_{k}$;
$\operatorname{Enc}(m p k, x)$ on input public key $m p k$ and plaintext $x \in X$, outputs ciphertext Ct ;
$\operatorname{Dec}\left(m p k, s k_{k}, \mathrm{Ct}\right)$ outputs a string $y$.

Correctness. We require that for all $k \in K$ and $x \in X$, and for $\operatorname{all}(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}\right)$, $s k_{k} \leftarrow \operatorname{KeyGen}(m s k, k)$ and $\mathrm{Ct} \leftarrow \operatorname{Enc}(m p k, x)$, then $\operatorname{Dec}\left(m p k, s k_{k}, \mathrm{Ct}\right)=F(k, x)$ with very high probability.

In the above definition the plaintext space and the key space only depend on the security parameter. In a more general definition both sets could also depend on the public key $m p k$ as it is the case for most known candidate constructions of (functional and plain) public key encryption schemes. Nonetheless, since our results do not depend on this definitional detail, we choose not to overburden our notations and definitions.

In the next section, we will see two sub-classes of functional encryption well recognized in the literature. These sub-classes are characterized by having a plaintext space with additional structure.

### 2.1 Predicate Encryption

In the literature, many functional encryption schemes are defined for functionality whose plaintext space $X$ consists of two subspaces $I$ and $M$ called respectively index space and message space. In this case, the functionality $F$ is defined in terms of a polynomial-time predicate $P: K \times I \rightarrow\{0,1\}$ as follows:

$$
F(k,(\mathfrak{i n d}, \mathfrak{m}))= \begin{cases}\mathfrak{m} & \text { if } P(k, \mathfrak{i n d})=1 \\ \perp & \text { if } P(k, \mathfrak{i n d})=0\end{cases}
$$

where $k \in K, \mathfrak{i n d} \in I$ and $\mathfrak{m} \in M$.

Definition 2.1.1 [Predicate Encryption Schemes] A predicate encryption scheme for a functionality $F$ defined over $(K, X=(I, M))$ for predicate $P: K \times I \rightarrow\{0,1\}$ is a tuple (Setup, KeyGen, Enc, Dec) of 4 algorithms with the following syntax:

Setup $\left(1^{\lambda}\right)$ outputs public and master secret keys ( $m p k, m s k$ ) for security parameter $\lambda$;
KeyGen $(m s k, k)$ on input a master secret key $m s k$ and key $k \in K$, outputs token $s k_{k}$;
$\operatorname{Enc}(m p k,(\mathfrak{i n d}, \mathfrak{m}))$ on input public key $m p k$, an index $\mathfrak{i n d} \in I$ and message $\mathfrak{m} \in M$, outputs ciphertext Ct ;
$\operatorname{Dec}\left(m p k, s k_{k}, \mathrm{Ct}\right)$ outputs a string $\mathfrak{m}$ or $\perp$.

Correctness. We require that for all $k \in K$ and $(\mathfrak{i n d}, \mathfrak{m}) \in X$, and for all $(m p k, m s k) \leftarrow$ $\operatorname{Setup}\left(1^{\lambda}\right), s k_{k} \leftarrow \operatorname{KeyGen}(m s k, k)$ and $\mathrm{Ct} \leftarrow \operatorname{Enc}(m p k,(\mathfrak{i n d}, \mathfrak{m}))$, then $\operatorname{Dec}\left(m p k, s k_{k}, \mathrm{Ct}\right)=$ $\mathfrak{m}$ if and only if $P(k, \mathfrak{i n d})=1$.

Predicate Only Encryption. A predicate only encryption scheme is a predicate encryption where the plaintext space coincides with the index space meaning that the ciphertext does not contain any message. More formally, in this case the functionality $F$ is defined as follows:

$$
F(k, \mathfrak{i n d})=P(k, \mathfrak{i n d}) \in\{0,1\}
$$

where $k \in K, \mathfrak{i n d} \in I$.

Predicate Encryption with Public Index. In a predicate encryption scheme with public index, the ciphertext always leaks the index. More formally, in this case the functionality $F$ is defined as follows:

$$
F(k,(\mathfrak{i n d}, \mathfrak{m}))= \begin{cases}(\mathfrak{i n d}, \mathfrak{m}) & \text { if } P(k, \mathfrak{i n d})=1 \\ (\mathfrak{i n d}, \perp) & \text { if } P(k, \mathfrak{i n d})=0\end{cases}
$$

where $k \in K, \mathfrak{i n d} \in I$ and $\mathfrak{m} \in M$.

### 2.1.1 Identity-Based Encryption

In an IBE scheme, ciphertexts and private keys are associated with identities, or strings, and a secret key can decrypt a ciphertext if they encode the same identity. An IBE scheme can then be formally defined as a predicate encryption scheme with public index in the following way. The keys space $K$ and the index space $I$ are the sets of binary strings and the IBE predicate $P$ over $(K, I)$ is defined as follows:

$$
P(k, \mathfrak{i n d})=\left\{\begin{array}{ll}
1 & \text { if } k=\mathfrak{i n d} \\
0 & \text { if } k \neq \mathfrak{i n d}
\end{array} .\right.
$$

If the index is hidden by the ciphertext then we recover the definition of Anonymous IBE.

Hierarchical IBE. Following the work of Boneh and Franklin, Horwitz and Lynn HLO2] proposed the concept of Hierarchical IBE (HIBE, for short), in which users are organized in a tree of depth $L$, with the root being the trusted central authority.

An HIBE scheme for hierarchy of depth $L$ can then be formally defined as a predicate encryption scheme with public index in the following way. The keys space $K$ and the index space $I$ are the sets of vectors $\left(\mathrm{ID}_{1}, \ldots, \mathrm{ID}_{\ell}\right)$ where $\ell \leq L$ and each $\mathrm{ID}_{i} \in\{0,1\}^{*}$. Then, the HIBE predicate $P_{L}$ over $(K, I)$ is defined as follows:
$P_{L}\left(k=\left(\mathrm{ID}_{1}^{k}, \ldots, \mathrm{ID}_{\ell}^{k}\right), \mathfrak{i n d}=\left(\mathrm{ID}_{1}^{c}, \ldots, \mathrm{ID}_{\ell^{\prime}}^{c}\right)\right)=\left\{\begin{array}{ll}1 & \text { if } \ell \leq \ell^{\prime} \text { and for } i=1 \ldots \ell, \mathrm{ID}_{i}^{k}=\mathrm{ID}_{i}^{c} \\ 0 & \text { otherwise }\end{array}\right.$.
If the index is hidden by the ciphertext then we recover the definition of Anonymous HIBE.

### 2.1.2 Hidden Vector Encryption

An HVE scheme, first proposed by Boneh and Waters [BW07], is a generalization of IBE with the support of a special character $*$ we refer to as a wildcard character. Formally, an HVE scheme can be defined as a predicate encryption scheme in the following way. The keys space $K$ consists of all the $\ell$-tuple $\left(v_{1}, \ldots, v_{\ell}\right)$ where each $v_{i} \in\{0,1\}^{*} \cup\{*\}$ and the index space $I$ consists of all the $\ell$-tuple $\left(w_{1}, \ldots, w_{\ell}\right)$ of $\ell$ where each $w_{i} \in\{0,1\}^{*}$ and the HVE predicate $P_{\ell}$ over $(K, I)$ is defined as follows:

$$
P_{\ell}\left(k=\left(v_{1}, \ldots, v_{\ell}\right), \mathfrak{i n d}=\left(w_{1}, \ldots, w_{\ell}\right)\right)= \begin{cases}1 & \text { if } v_{i}=w_{i} \text { whenever } v_{i} \neq \star \\ 0 & \text { otherwise }\end{cases}
$$

Applications of such encryption scheme include conjunctive and range searches.

### 2.1.3 Inner Product Predicate

Katz, Sahai and Waters [KSW08] proposed inner-product encryption (IPE, for short), a more general system to test if a dot product operation over the ring $\mathbb{Z}_{N}$ is equal to 0 , where $N$ is the product of three random primes chosen by the setup algorithm. Later, Okamoto and Takashima [OT09] and Lewko et. al. [LOS ${ }^{+}$10] gave constructions over the field $F_{p}$. Formally, an IPE scheme can be defined as a predicate encryption scheme in the following way. Let $p$ be a prime of length $\lambda$, chosen at setup time where $\lambda$ is the security parameter. Then, the keys space $K$ and the index space $I$ consist of all the $n$-tuple $\left(v_{1}, \ldots, v_{n}\right)$, where each $v_{i} \in F_{p}$, and the IPE predicate $P_{p, n}$ over $(K, I)$ is defined as follows:

$$
P_{p, n}\left(k=\left(v_{1}, \ldots, v_{n}\right), \mathfrak{i n d}=\left(w_{1}, \ldots, w_{n}\right)\right)= \begin{cases}1 & \text { if } \sum_{i=1}^{n} v_{i} \cdot w_{i}=0 \\ 0 & \text { otherwise }\end{cases}
$$

Despite this apparently restrictive structure, IPE can support conjunction, subset and range queries on encrypted data as well as disjunction, polynomial evaluation, and CNF and DNF formulas.

Hierarchical IPE. Following the work of Katz, Sahai and Waters, Okamoto and Takashima [OT09] proposed the concept of Hierarchical IPE (HIPE, for short). Specifically,
Definition 2.1.2 [Hierarchical Format] Let $\boldsymbol{\mu}$ be a tuple of positive integers $\boldsymbol{\mu}=\left(n, d ; \mu_{1}, \ldots, \mu_{d}\right)$ such that $\mu_{0}=0<\mu_{1}<\cdots<\mu_{d}=n$. For $i=1, \ldots, d$, let $\Sigma_{i}=\mathbb{F}_{p}^{\mu_{i}-\mu_{i-1}} \backslash\left\{\mathbf{0}_{\mu_{i}-\mu_{i-1}}\right\}$ the set of attributes. Let $\Sigma$ be the hierarchical attributes $\Sigma=\cup_{i=1}^{d}\left(\Sigma_{1} \times \ldots \times \Sigma_{i}\right)$, where the union is a disjoint union. Them, we call $\boldsymbol{\mu}$ an hierarchical format of depth $d$ for the attribute space $\Sigma$. Sometimes, we will use notation $\Sigma_{\mid t}=\cup_{i=1}^{t}\left(\Sigma_{1} \times \ldots \times \Sigma_{i}\right)$

Let $p$ be a prime of length $\lambda$, chosen at setup time where $\lambda$ is the security parameter. Then, an HIPE scheme can be formally defined as a predicate encryption scheme in the following way. For a given hierarchical format $\boldsymbol{\mu}$, the keys space $K$ and the index space $I$ correspond to the hierarchical attributes set $\Sigma$, and the HIPE predicate $P_{p, \mu}$ over $(K, I)$ is defined as follows:

$$
P_{p, \boldsymbol{\mu}}\left(k=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{\ell}\right), \mathfrak{i n d}=\left(\boldsymbol{w}_{1}, \ldots, \boldsymbol{w}_{\ell^{\prime}}\right)\right)= \begin{cases}1 & \text { if } \ell \leq \ell^{\prime} \text { and for each } i,\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}\right\rangle=0 \\ 0 & \text { otherwise }\end{cases}
$$

### 2.2 Security Definitions

One of the most important achievements of the modern theory of Cryptography is the formalization of the notion of a secure cryptosystem given in the seminal work of Goldwasser and Micali [GM84]. There, two notions of security were proposed, one is a game-based notion and the other is simulation-based, and they were shown to be equivalent in the sense that an encryption scheme meets one security definition if and only if it meets the other. The study of simulation-based notions of security for functional encryption was initiated by Boneh, Sahai, and Waters [BSW11] and O'Neill [O’N10]. Specifically, in [BSW11] Boneh et al. showed that there exists a clearly insecure functional encryption scheme that is nonetheless deemed secure by the game-based notion of security. This motivated the study of a simulation-based notion of security for functional encryption. In the same paper [BSW11], Boneh et al. gave impossibility results in the non-programmable random oracle model for simple instances of functional encryption for which, instead, game-based security had been shown to be achievable (again, under appropriate complexity assumptions).

In the next sections, we give formal definitions of game-based and simulation-based security for functional encryption.

### 2.2.1 Game-based security

The game-based notion of security for functional encryption, also called indistinguishability under-chosen message attack (IND-CPA, for short), can be formalized by means of the following game between an adversary $\mathcal{A}$ and a challenger $\mathcal{C}$. Let $\mathcal{F E}$ be a functional encryption scheme for functionality $F$ over $(K, X)$ and let $\mathcal{A}$ be a probabilistic polynomial-time adversary. Game $\operatorname{IND}-\mathrm{CPA}_{\mathcal{A}}^{\mathcal{F} \mathcal{E}}\left(1^{\lambda}\right)$, where $\lambda$ is the security parameter, is defined as follows.

Setup: $\mathcal{C}$ generates public key and master secret key by invoking the setup algorithm on input the security parameter $\lambda$ given in unary. Specifically, $(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}\right)$. Then, $\mathcal{C}$ starts the interaction with $\mathcal{A}$ on input $m p k$.

Query Phase 1: $\mathcal{A}$ adaptively submits key queries. On input a key $k \in K, \mathcal{C}$ generates a secret key for $\mathrm{k}, s k_{k}=\operatorname{KeyGen}(m s k, k)$, and gives $s k_{k}$ back to $\mathcal{A}$.

Challenge: $\mathcal{A}$ submits two equal length plaintexts $x_{0}, x_{1} \in X . \mathcal{C}$ chooses random $\eta \in\{0,1\}$ and generates ciphertext $\mathrm{Ct}^{\star}=\operatorname{Enc}\left(m p k, x_{\eta}\right) . \mathcal{C}$ sends back $\mathrm{Ct}^{\star}$ to $\mathcal{A}$.

Query Phase 2: It is the same as Query Phase 1.
Guess: Eventually, $\mathcal{A}$ submits its guess $\eta^{\prime}$.
Winning Condition: $\mathcal{A}$ wins the game if the following conditions are satisfied:

1. $\eta=\eta^{\prime}$.
2. For all $k$ queried by $\mathcal{A}$, it holds that $F\left(k, x_{0}\right)=F\left(k, x_{1}\right)$.

The advantage of $\mathcal{A}$ in the above game is defined as

$$
\mathbf{A d v}_{\text {IND-CPA }}^{\mathcal{F E}, \mathcal{A}}(\lambda)=\operatorname{Prob}\left[\operatorname{IND}-\operatorname{CPA}_{\mathcal{A}}^{\mathcal{F} \mathcal{E}}\left(1^{\lambda}\right)=1\right]-1 / 2
$$

Definition 2.2.1 A functional encryption scheme $\mathcal{F E}$ is IND-CPA-secure if any polynomialtime adversaries making at most a polynomial number of queries to the key derivation oracle only has a negligible advantage in the IND-CPA game described above, i.e., $\mathbf{A d v}_{\text {IND-CPA }}^{\mathcal{F} \mathcal{E}, \mathcal{A}}(\lambda)$ is negligible function of $\lambda$.

CCA Security. In a chosen-ciphertext attack (IND-CCA, for short), the adversary is additionally given access to a decryption oracle that for a given key $k \in K$ and a given ciphertext Ct returns the decryption $\operatorname{Dec}(m p k, \operatorname{KeyGen}(m s k, k), \mathrm{Ct})$.

Selective Security. It is a weaker security model, introduced by [CHK03, BB04] to obtain the first IBE scheme secure in the standard model, where the adversary must commit to its challenge before seeing the public key. We refer to this definition of security as selectiveindistinguishability under chosen-message attack (chosen-ciphertext-attack), sIND-CPA (sINDCCA) for short. In the literature, the non-selective model of security (IND-CPA) is also sometimes called full-security model.

Predicate Encryption. In this case, Definition 2.2.1 is sufficient. In particular in the INDCPA game, the adversary's challenge consists of two indexes $\mathfrak{i n d}{\underset{d}{0}}, \mathfrak{i n d}_{1}$ and two messages $m_{0}, m_{1}$. This models the privacy of the index and that of the message. This definition of security is also know in the literature as anonymity under chosen-plaintext attack (ANO-CPA, for short). Moreover, in the literature a weaker security model has been defined. It is known as
weakly attribute-hiding model (wAH-IND-CPA or wAH-ANO-CPA for short) where the winning conditions is relaxed. Specifically, it is required that, for all $k$ queried by $\mathcal{A}$, it holds that $F\left(k, x_{0}\right)=F\left(k, x_{1}\right)=0$, meaning that the adversary cannot see secret keys that can be used to decrypt the ciphertext. This relaxation is still reasonable and has been used to achieve fullsecurity.

Instead, for public index predicate encryption schemes it is enough to model only the privacy of the message because the index is public. Thus in the IND-CPA game, the adversary's challenge consists of an index $\mathfrak{i n d}$ and two messages $m_{0}, m_{1}$. In this case, the weakly attributehiding model is a minimum requirement because otherwise the adversary would be forced to submit a challenge with $m_{0}=m_{1}$.

### 2.2.2 Simulation-based security

In the context of functional encryption, the intuition that we would like to be captured by a simulation-based definition of security is that getting the secret key $s k_{k}$ corresponding to the key $k \in K$ should only reveal $F(k, x)$ when given an encryption of $x$. The following formal definition, largely based on the recent works of Boneh, Sahai and Waters [BSW11] and O'Neill [ $\left.\mathrm{O}^{\prime} \mathrm{N} 10\right]$, tries to capture this intuition avoiding to fall into the impossibility results of [BSW11], namely we restrict the adversary to request the encryption of a single message.

Definition 2.2.2 A functional encryption scheme $\mathcal{F E}$ for functionality $F$ over $(K, X)$ is adaptively simulation-based secure (SIM-Secure, for short) if there exists a simulator algorithm $\operatorname{Sim}=\left(\operatorname{Sim}_{0}, \operatorname{Sim}_{1}, \operatorname{Sim}_{2}\right)$ such that for all adversary algorithms $\mathcal{A}=\left(\mathcal{A}_{0}, \mathcal{A}_{1}\right)$ the output of the following two experiments are computationally indistinguishable.

```
RealExp \({ }^{\mathcal{F E}, \mathcal{A}}\left(1^{\lambda}\right)\)
\((m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}\right) ;\)
\((x\), aux \() \leftarrow \mathcal{A}_{0}^{\text {KeyGen }(m s k, \cdot)}(m p k) ;\)
\(\mathrm{Ct} \leftarrow \operatorname{Enc}(m p k, x) ;\)
\(\alpha \leftarrow \mathcal{A}_{1}^{\text {KeyGen }(m s k, \cdot)}(m p k, \mathrm{Ct}, \mathrm{aux})\);
Output: \((m p k, x, \alpha)\)
```

```
IdealExp \(\underset{\operatorname{Sim}}{\mathcal{F E}, \mathcal{A}}\left(1^{\lambda}\right)\)
\((m p k, m s k) \leftarrow \operatorname{Sim}_{0}\left(1^{\lambda}\right) ;\)
\((x\), aux \() \leftarrow \mathcal{A}_{0}^{\text {KeyGen }(m s k, \cdot)}(m p k) ;\)
\(\left(\mathrm{Ct}^{\prime}, \mathrm{aux}^{\prime}\right) \leftarrow \operatorname{Sim}_{1}\left(m p k,|x|,\left(k_{i}, F\left(k_{i}, x\right), s k_{i}\right)_{i=1}^{q_{1}}\right) ;\)
\(\alpha \leftarrow \mathcal{A}_{1}^{\mathcal{O}\left(m s k, \mathrm{aux}^{\prime}, \cdot\right)}\left(m p k, \mathrm{Ct}^{\prime}\right.\), aux \()\);
Output: \((m p k, x, \alpha)\)
```

Here, $k_{1}, \ldots, k_{q_{1}}$ are the keys for which $\mathcal{A}_{0}$ has asked to see a secret key in the first query phase for some $q_{1}=\operatorname{poly}(\lambda)$. Moreover, oracle $\mathcal{O}\left(m s k\right.$, aux $\left.^{\prime}, \cdot\right)$ is algorithm $\operatorname{Sim}_{2}\left(m s k\right.$, aux $\left.{ }^{\prime}, \cdot, F(\cdot, x)\right)$ that answers to the adversary's queries of the second phase. In particular, $\operatorname{Sim}_{2}$ receives as third argument the key $k$ for which the adversary has asked a secret key and as fourth argument the value $F(k, x)$, along with the master secret key $m s k$ and current state $\mathrm{aux}^{\prime}$ of the simulator.

We observe that the simulator algorithm $\operatorname{Sim}_{2}$ is stateful in that after each invocation, it updates the state aux ${ }^{\prime}$ which is carried over to its next invocation. Also, we note that the notion of nonadaptive SIM-Security (NA-SIM-Secure, for short), is obtained by giving $\mathcal{A}_{1}$ no oracle access in both experiments.

Recently, Agrawal et al. [AGVW12] showed that for adversaries that can obtain an unbounded number of tokens, Definition 2.2.2 cannot be achieved for incompressible functionalities (i.e., weak pseudo-random functions). Thus, by restricting the range of adversaries to those that make an a-priori bounded number of queries (counting both first stage and second stage token queries), we obtain the following weaker notion.

Definition 2.2.3 A functional encryption scheme $\mathcal{F E}$ is $q$-bounded simulation-based secure ( $q$ -SIM-Secure, for short) if there exists a simulator algorithm Sim such that for all adversary algorithms $\mathcal{A}$ that make at most $q$ token queries, the output of experiments RealExp ${ }^{\mathcal{F E}, \mathcal{A}}$ and IdealExp $\operatorname{Eim}_{\mathcal{F} \mathcal{E}, \mathcal{A}}$ are indistinguishable.

## Chapter 3

## Tools and Notations

In this chapter we introduce the mathematical tools we will use in our constructions.

### 3.1 Notations

We say a function $f(n)$ is negligible if it is $O\left(n^{-c}\right)$ for all $c>0$, and we use $\operatorname{negl}(n)$ to denote a negligible function of $n$. We say $f(n)$ is polynomial if it is $O\left(n^{c}\right)$ for some $c>0$, and we use $\operatorname{poly}(n)$ to denote a polynomial function of $n$.

If $D$ is a probability distribution, the writing " $x \leftarrow D$ " means that $x$ is chosen according to $D$. If $D$ is a finite set, the writing " $x \leftarrow D$ " means that $x$ is chosen according to uniform probability on $D$. If $A$ is an algorithm then $A\left(x_{1}, x_{2}, \ldots\right)$ denotes the probability distribution of the output of $A$ when $A$ is run on input $\left(x_{1}, x_{2}, \ldots\right)$ and randomly chosen coin tosses. Instead $A\left(x_{1}, x_{2}, \ldots ; R\right)$ denotes the output of $A$ when run on input $\left(x_{1}, x_{2}, \ldots\right)$ and (sufficiently long) coin tosses $R$. All algorithms, unless explicitly noted, are probabilistic polynomial time (PPT, for short) and all adversaries are modeled by non-uniform polynomial time algorithms. If $A$ is an algorithm and $B$ is an algorithm with access to an oracle then $A^{B}$ denotes the execution of $A$ with oracle access to $B$. We say an event occurs with overwhelming probability if its probability is $1-\operatorname{negl}(n)$.

If $q>0$ is an integer the $[q]$ denotes the set $\{1, \ldots, q\}$. The notation $\lfloor x\rceil$ denotes the nearest integer to $x$, rounding towards 0 for half-integers. For any integer $q \geq 2$, we let $\mathbb{Z}_{q}$ denote the ring of integers modulo $q$ and we represent $\mathbb{Z}_{q}$ as integers in $(q / 2, q / 2]$. We let $\mathbb{Z}_{q}^{n \times m}$ denote the set of $n \times m$ matrices with entries in $\mathbb{Z}_{q}$.

We use bold capital letters (e.g. $\boldsymbol{A}$ ) to denote matrices, bold lowercase letters (e.g. w) to denote vectors. The notation $\boldsymbol{A}^{\top}$ denotes the transpose of the matrix $\boldsymbol{A}$. When we say a matrix defined over $\mathbb{Z}_{q}$ has full rank, we mean that it has full rank modulo each prime factor of $q$. If $\boldsymbol{A}_{1}$ is an $n \times m$ matrix and $\boldsymbol{A}_{2}$ is an $n \times m^{\prime}$ matrix, then $\left[\boldsymbol{A}_{1} \| \boldsymbol{A}_{2}\right]$ denotes the $n \times\left(m+m^{\prime}\right)$ matrix formed by concatenating $\boldsymbol{A}_{1}$ and $\boldsymbol{A}_{2}$. If $\boldsymbol{w}_{1}$ is a length $m$ vector and $\boldsymbol{w}_{2}$ is a length $m^{\prime}$ vector, then we let $\left[\boldsymbol{w}_{1} \| \boldsymbol{w}_{2}\right]$ denote the length $\left(m+m^{\prime}\right)$ vector formed by concatenating $\boldsymbol{w}_{1}$ and $\boldsymbol{w}_{2}$. However, when doing matrix-vector multiplication we always view vectors as column vectors. The norm of a matrix $\boldsymbol{R} \in \mathbb{R}^{k \times m}$ is defined as $\|\boldsymbol{R}\|:=\sup _{\|\boldsymbol{u}\|=1}\|\boldsymbol{R} \boldsymbol{u}\|$.

### 3.2 Bilinear Settings

The study of cyclic groups in cryptography has been motivated by the assumed hardness of taking discrete logarithms on which the security of several cryptosystems (most notably the ElGamal encryption scheme) and cryptographic protocols has been based. The observation
that some cyclic groups admit bilinear maps has been used at first to break cryptosystems, see [MOV91, FMR99, FR94]. Later, it was discovered that they could also be used to build cryptosystems, so rather than avoiding this additional properties of the cyclic groups, one can exploit them to construct new schemes. The Boneh and Franklin's identity-based encryption scheme [BF01] is one of the most famous examples of what can be achieved using bilinear maps. Extending the basic idea leads to identity-based schemes with additional useful properties such as authenticated or hierarchical identity-based encryption [HLO2]. Moreover identity-based signature [BLS01, Hes03] identity-based identification [KH04, KH06] schemes have been constructed using bilinear maps.

Definition 3.2.1 [Bilinear Setting] Let $\mathbb{G}_{1}, \mathbb{G}_{2}$ and $\mathbb{G}_{T}$ be cyclic groups of prime-order $N$. Let $g_{1}$ be a generator of $\mathbb{G}_{1}$ and $g_{2}$ be a generator of $\mathbb{G}_{2}$. A bilinear pairing or bilinear map is an efficiently computable function $\hat{e}: \mathbb{G}_{1} \times \mathbb{G}_{2} \rightarrow \mathbb{G}_{T}$ such that:

1. (Bilinearity) $\forall a, b \in \mathbb{Z}_{N}$ it holds that $\hat{e}\left(g_{1}^{a}, g_{2}^{b}\right)=\hat{e}\left(g_{1}, g_{2}\right)^{a b}$.
2. (Non-degeneracy) $\hat{e}\left(g_{1}, g_{2}\right) \neq 1$.

Then, we call the tuple $\left(N, \mathbb{G}_{1}, \mathbb{G}_{2}, g_{1}, g_{2}, \mathbb{G}_{T}, \hat{e}\right)$ an asymmetric bilinear setting. Moreover, if $\mathbb{G}_{1}=\mathbb{G}_{2}=\mathbb{G}$ and $g$ is a generator of $\mathbb{G}$ then the tuple $\left(N, \mathbb{G}, \mathbb{G}_{T}, g, \hat{e}\right)$ is called symmetric bilinear setting.

Finally, we suppose the existence of an efficient algorithm BSGen which takes in input the security parameter $\lambda$ and outputs a description of a bilinear setting over cyclic groups whose order is of $\Theta(\lambda)$ bits.

The cryptographic relevance of bilinear maps stem from the fact that in cyclic groups that admit such a mapping the decisional Diffie-Hellman assumption does not hold. Indeed, given $g, g^{x}, g^{y}, g^{z}$ it is possible to check if $z=x y$ (and thus solve the decisional DH problem) by testing $\hat{e}\left(g, g^{z}\right)$ and $\hat{e}\left(g^{x}, g^{y}\right)$ for equality.

Decisional Linear Assumption. One of the most well studied assumption in the contest of prime-order bilinear settings is the Decisional Linear Assumption (DL, for short). It can be described in the following way.

For a given prime-order bilinear setting generator BSGen, define the following distribution: Pick a random bilinear setting $\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{BSGen}\left(1^{\lambda}\right)$, and then pick

$$
g, f, v \leftarrow \mathbb{G} \text { and } c_{1}, c_{2}, c_{3} \in \mathbb{Z}_{N}
$$

and set $D=\left(N, g, f, v, f^{c_{1}}, v^{c_{2}}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking the DL Assumption is

$$
\operatorname{Adv}_{\mathrm{DL}}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, g^{c_{1}+c_{2}}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, g^{c_{1}+c_{2}+c_{3}}\right)=1\right]\right|
$$

Definition 3.2.2 [Decisional Linear Assumption] We say that the DL Assumption holds for generator $\operatorname{BSGen}$ if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \mathbf{A d v}_{\mathrm{DL}}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

### 3.2.1 Composite-Order Bilinear Setting

Composite-order bilinear settings were first used in cryptography by [BGN05] (see also [Bon07]) and then widely used to construct functional encryption schemes and more.

Definition 3.2.3 [Composite-Order Bilinear Setting] Let $\mathbb{G}$ and $\mathbb{G}_{T}$ be cyclic groups of composite order $N$ product of $m>1$ primes, $N=p_{1} \cdot \ldots \cdot p_{m}$. Let $g$ be a generator of $\mathbb{G}$ and $\hat{e}: \mathbb{G} \times \mathbb{G} \rightarrow \mathbb{G}_{T}$ an efficiently computable non-degenerate bilinear map. Then, we call the tuple $\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right)$ a composite-order bilinear setting.

Moreover, we suppose the existence of an efficient algorithm CBSGen which takes as input the security parameter $\lambda$ and the number of primes $m>1$, and outputs a description of a composite-order bilinear setting over cyclic groups whose order $N$ is the product of $m$ primes of $\Theta(\lambda)$ bits each.

For notational convenience, for each non-empty subset $S \subseteq[m]$ we denote by $\mathbb{G}_{S}$ the subgroup of order $\prod_{i \in S} p_{i}$ in the bilinear group $\mathbb{G}$.

Orthogonality. From the fact that the group is cyclic, it is easy to verify that if $g$ and $h$ are group elements of co-prime orders then $\hat{e}(g, h)=1$. This is called the orthogonality property and is a crucial tool in many proof of full security for functional encryption schemes.

General Subgroup Decision Assumption. In the contest of composite-order bilinear settings, one of the most common assumption is the General Subgroup Decision Assumption (GSD, for short) introduced by Bellare et al. [BWY11]. It can be described in the following way.

For a given composite-order bilinear setting generator CBSGen, define the following distribution: Pick a random bilinear setting $\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{CBSGen}\left(1^{\lambda}, 1^{m}\right)$. Let $S_{0}, S_{1}, S_{2}$, $\ldots, S_{k}$, for constant $k$, be non-empty subsets of $[m]$ such that for each $2 \leq j \leq k$, either $S_{j} \cap S_{0}=S_{j} \cap S_{1}=\emptyset$ or $S_{j} \cap S_{0} \neq \emptyset$ and $S_{j} \cap S_{1} \neq \emptyset$, and then pick

$$
\begin{gathered}
Z_{2} \leftarrow \mathbb{G}_{S_{2}}, \ldots, Z_{k} \leftarrow \mathbb{G}_{S_{k}} \\
T_{0} \leftarrow \mathbb{G}_{S_{0}}, T_{1} \leftarrow \mathbb{G}_{S_{1}}
\end{gathered}
$$

and set $D=\left(N, Z_{2}, \ldots, Z_{k}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking the GSD Assumption is

$$
\mathbf{A d v}_{\mathrm{GSD}}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, T_{0}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, T_{1}\right)=1\right]\right|
$$

Definition 3.2.4 [General Subgroup Decision Assumption] We say that the GSD Assumption holds for generator CBSGen if, for all $S_{0}, \ldots, S_{k}$ that satisfy the conditions above and for all probabilistic polynomial-time algorithms $\mathcal{A}, \operatorname{Adv}_{\mathrm{GSD}}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

We will use the following algorithm to sample a random instance of the GSD Assumption.
Algorithm GSDGen on input the security parameter $1^{\lambda}$, the number of primes $1^{m}$ and $k+1$ nonempty subsets of $\left[m\right.$ ], $S_{0}, S_{1}, S_{2}, \ldots, S_{k}$, does the following: Pick a random compositeorder bilinear setting $\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{CBSGen}\left(1^{\lambda}, 1^{m}\right)$, and then pick

$$
\begin{gathered}
Z_{2} \leftarrow \mathbb{G}_{S_{2}}, \ldots, Z_{k} \leftarrow \mathbb{G}_{S_{k}} \\
T_{0} \leftarrow \mathbb{G}_{S_{0}}, T_{1} \leftarrow \mathbb{G}_{S_{1}}
\end{gathered}
$$

Then the algorithm picks a random $\beta \in\{0,1\}$ and returns the tuple $\left(N, T_{\beta}, Z_{2}, \ldots, Z_{k}\right)$.
For notational convenience, sometimes we will omit the parameter $m$ when it is clear from the context.

### 3.2.2 Dual Pairing Vector Spaces

Dual paring vector spaces (DPVS, for short) are rich mathematical structures introduced by Okamoto and Takashima [OT08, OT09] in the context of functional encryption. DPVS can be constructed on prime-order symmetric and asymmetric bilinear settings. We will focus on the symmetric setting.

Given a symmetric bilinear setting $\left(N, \mathbb{G}, \mathbb{G}_{\mathrm{T}}, g, \hat{e}\right)$, let $\mathbb{V}=\mathbb{G}^{n}$ be a vector space whose elements are expressed by vectors

$$
\boldsymbol{x}=\left(g^{x_{1}}, \ldots, g^{x_{n}}\right),
$$

where $x_{i} \in \mathbb{Z}_{N}$ for each $i \in[n]$. The canonical base of $\mathbb{V}$ is

$$
\mathbb{A}=\left(\boldsymbol{a}_{i}\right)_{i \in[n]},
$$

where $\boldsymbol{a}_{i}=(1, \ldots, 1, g, 1, \ldots, 1)$. Thus, a vector $\boldsymbol{x}$ can be expressed in base $\mathbb{A}$ as:

$$
\boldsymbol{x}=\left(x_{1}, \ldots, x_{n}\right)_{\mathbb{A}}=\sum_{i \in[n]} x_{i} \cdot \boldsymbol{a}_{i} .
$$

Then, a well-defined non-degenerate pairing operation between $\boldsymbol{x}, \boldsymbol{y} \in \mathbb{V}$, can be defined as follows:

$$
\hat{e}(\boldsymbol{x}, \boldsymbol{y})=\prod_{i \in[n]} \hat{e}\left(g^{x_{i}}, g^{y_{i}}\right)=\hat{e}(g, g)^{\sum_{i \in[n]} x_{i} \cdot y_{i}} .
$$

Notice that over $\mathbb{A}$ it is easy to decompose $x_{i} \boldsymbol{a}_{i}=\left(1, \ldots, 1, g_{i}^{x}, 1, \ldots, 1\right)$ from $\boldsymbol{x}=$ $\left(x_{1}, \ldots, x_{n}\right)_{\mathbb{A}}=\sum_{i \in[n]} x_{i} \cdot \boldsymbol{a}_{i}$. Thus, the canonical base $\mathbb{A}$ can be further randomized to obtain a new base $\mathbb{B}=\left(\boldsymbol{b}_{1}, \ldots, \boldsymbol{b}_{n}\right)$ of $\mathbb{V}$ by applying a random (regular) linear transformation $\boldsymbol{R}=\left(r_{i, j}\right)_{i \in[n], j \in[n]} \leftarrow \mathrm{GL}\left(n, \mathbb{Z}_{N}\right)$ as follows:

$$
\boldsymbol{b}_{i}=\sum_{j \in[n]} r_{i, j} \cdot \boldsymbol{a}_{j} .
$$

Moreover, if we apply to $\mathbb{A}$ randomization $\boldsymbol{R}$ to form base $\mathbb{B}$ and randomization $\left(\boldsymbol{R}^{\boldsymbol{\top}}\right)^{-1}$ to form base $\mathbb{B}^{\star}=\left(\boldsymbol{b}_{1}^{\star}, \ldots, \boldsymbol{b}_{n}^{\star}\right)$. Then $\left(\mathbb{B}, \mathbb{B}^{\star}\right)$ are dual orthonormal bases of $\mathbb{V}$ such that $\hat{e}\left(\boldsymbol{b}_{i}, \boldsymbol{b}_{j}^{\star}\right)=$ $\hat{e}(g, g)^{\delta_{i, j}}$ where $\delta_{i, j}=1$ if $i=j$, and 0 otherwise.

Definition 3.2.5 [Dual Pairing Vector Spaces] Let $\mathbb{G}$ and $\mathbb{G}_{T}$ be cyclic groups of prime order $N$. Let $g$ be a generator of $\mathbb{G}$ and $\hat{e}: \mathbb{G} \times \mathbb{G} \rightarrow \mathbb{G}_{T}$ an efficiently computable non-degenerate bilinear map. Let $\mathbb{V}=\mathbb{G}^{n}$, for integer $n>0$, be a vector space obtained as direct product of $\mathbb{G}$ and $\mathbb{A}=\left(\boldsymbol{a}_{i}\right)_{i \in[n]}$ its canonical base, and let $\left(\mathbb{B}, \mathbb{B}^{\star}\right)$ random dual orthonormal bases of $\mathbb{V}$. Let param $_{\mathbb{V}}=\left(N, \mathbb{V}, \mathbb{G}_{T}, \mathbb{A}\right)$. Then, we call the tuple $\left(\right.$ param $\left._{\mathbb{V}}, \mathbb{B}, \mathbb{B}^{\star}\right)$ a dual pairing vector space.

Finally, we suppose the existence of an efficient algorithm DPVSGen which takes in input the security parameter $\lambda$ and integer $n$, and outputs a description of a dual pairing vector space over cyclic groups whose order is of $\Theta(\lambda)$ bits. Notice that algorithm DPVSGen can be obtained by using algorithm BSGen and applying the randomization described above.

Decisional Subspace Problem. First introduced by [OT08]. It asks to distinguish between vector $\boldsymbol{u}=\left(r_{1}, \ldots, r_{m}\right)_{\mathbb{B}}$ and vector $\boldsymbol{v}=\left(\mathbf{0}_{n}, r_{n+1}, \ldots, r_{m}\right)_{\mathbb{B}}$ where $n+1<m$ and $r_{i} \leftarrow \mathbb{Z}_{N}$ for $i \in[m]$. Although the decisional subspace problem is assumed to be intractable, it can be efficiently solved by using trapdoor $\boldsymbol{t}^{\star} \in \operatorname{span}\left(\boldsymbol{b}_{1}^{\star}, \ldots, \boldsymbol{b}_{n}^{\star}\right)$. In fact $\hat{e}\left(\boldsymbol{u}, \boldsymbol{t}^{\star}\right) \neq 1$ and $\hat{e}\left(\boldsymbol{v}, \boldsymbol{t}^{\star}\right)=$ 1 with high probability.

### 3.3 Lattices

In 1994, Peter Shor [Sho96] showed that quantum computers can solve in polynomial time, integer factorization and discrete logarithms including elliptic curve discrete logarithms [BL95]. Because of the threat that Shor's algorithm poses to existing encryption techniques, there is a great deal of interest in new hard problems believed to resist against quantum attacks.

Lattice assumptions are hard assumptions that have received in the very recent years a lot of attentions both in cryptanalysis and cryptography. In this section, we will introduce the background relevant to this work and the main hardness assumption in the field, the Learning with Error problem formally introduced by Regev [Reg05]. Specifically, we collect, almost verbatim, results from [ABB10, AFV11, AP09, MR04, GPV08, CHKP10].

Lattice. An $m$-dimensional lattice $\Lambda$ is a full-rank discrete subgroup of $\mathbb{R}^{m}$. A basis of $\Lambda$ is a linearly independent set of vectors whose span is $\Lambda$. We will focus on integer lattices and among these we will focus on the $q$-ary lattices defined as follows: for any integer $q \geq 2$ and any $\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times m}$, we define

$$
\begin{aligned}
\Lambda_{q}^{\perp}(\boldsymbol{A}) & :=\left\{\boldsymbol{e} \in \mathbb{Z}_{m}: \boldsymbol{A} \cdot \boldsymbol{e}=0 \quad \bmod q\right\} \\
\Lambda_{q}^{\boldsymbol{u}}(\boldsymbol{A}) & :=\left\{\boldsymbol{e} \in \mathbb{Z}_{m}: \boldsymbol{A} \cdot \boldsymbol{e}=\boldsymbol{u} \quad \bmod q\right\} \\
\Lambda_{q}(\boldsymbol{A}) & :=\left\{\boldsymbol{e} \in \mathbb{Z}_{m}: \exists \boldsymbol{s} \in \mathbb{Z}_{q}^{m} \text { with } \boldsymbol{A}^{t} \cdot \boldsymbol{s}=\boldsymbol{e} \quad \bmod q\right\} .
\end{aligned}
$$

The lattice $\Lambda_{q}^{u}(\boldsymbol{A})$ is a coset of $\Lambda_{q}^{\perp}(\boldsymbol{A})$; namely, $\Lambda_{q}^{u}(\boldsymbol{A})=\Lambda_{q}^{\perp}(\boldsymbol{A})+\boldsymbol{t}$ for any $\boldsymbol{t}$ such that $\boldsymbol{A} \cdot \boldsymbol{t}=\boldsymbol{u} \bmod q$.

Gram-Schmidt norm. Let $\boldsymbol{S}=\left\{\boldsymbol{s}_{1}, \ldots, \boldsymbol{s}_{k}\right\}$ be a set of vectors in $\mathbb{R}^{m}$. Let $\|\boldsymbol{S}\|$ denotes the length of the longest vector in $\boldsymbol{S}$, i.e., $\max _{1 \leq i \leq k}\left\|s_{i}\right\|$, and $\widetilde{\boldsymbol{S}}:=\widetilde{\boldsymbol{s}}_{1}, \ldots, \widetilde{\boldsymbol{s}}_{k} \subset \mathbb{R}^{m}$ denotes the Gram-Schmidt orthogonalization of the vectors $s_{1}, \ldots, s_{k}$. We refer to $\|\widetilde{\boldsymbol{S}}\|$ as the GramSchmidt norm of $\boldsymbol{S}$.

Gaussian distributions. Let $L$ be a discrete subset of $\mathbb{Z}^{n}$. For any vector $\boldsymbol{c} \in \mathbb{R}^{n}$ and any positive parameter $\sigma \in \mathbb{R}_{>0}$, let

$$
\rho_{\sigma, \boldsymbol{c}}(\boldsymbol{w}):=\exp \left(-\pi\|x-\boldsymbol{c}\|^{2} / \sigma^{2}\right)
$$

be the Gaussian function on $\mathbb{R}^{n}$ with center $\boldsymbol{c}$ and parameter $\sigma$. Let

$$
\rho_{\sigma, \boldsymbol{c}}(L):=\sum_{\boldsymbol{w} \in L} \rho_{\sigma, \boldsymbol{c}}(\boldsymbol{w})
$$

be the discrete integral of $\rho_{\sigma, c}$ over $L$, and let $\mathcal{D}_{L, \sigma, c}$ be the discrete Gaussian distribution over $L$ with center $\boldsymbol{c}$ and parameter $\sigma$. Specifically, for all $\boldsymbol{v} \in L$, we have

$$
\mathcal{D}_{L, \sigma, c}(\boldsymbol{v})=\frac{\rho_{\sigma, \boldsymbol{c}}(\boldsymbol{v})}{\rho_{\sigma, c}(L)}
$$

For notational convenience, $\rho_{\sigma, 0}$ and $\mathcal{D}_{L, \sigma, 0}$ are abbreviated as $\rho_{\sigma}$ and $\mathcal{D}_{L, \sigma}$ respectively. The following lemma captures standard properties of these distributions.

Lemma 3.3.1 Let $q \geq 2$ and let $\boldsymbol{A}$ be a matrix in $\mathbb{Z}_{q}^{n \times m}$ with $m>n$. Let $\boldsymbol{T}_{\boldsymbol{A}}$ be a basis for $\Lambda_{q}^{\perp}(\boldsymbol{A})$ and $\sigma \geq\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{A}}}\right\| \cdot \omega(\sqrt{\log m})$. Then for $\boldsymbol{c} \in \mathbb{R}^{m}$ and $\boldsymbol{u} \in \mathbb{Z}_{q}^{n}$ :

1. $\operatorname{Pr}\left[\|\boldsymbol{w}-\boldsymbol{c}\|>\sigma \sqrt{m}: \boldsymbol{w} \leftarrow \mathcal{D}_{\Lambda, \sigma, \boldsymbol{c}}\right] \leq \operatorname{negl}(n)$
2. A set of $O(m \log m)$ samples from $\mathcal{D}_{\Lambda_{q}^{\perp}(A), \sigma}$ contains a full rank set in $\mathbb{Z}^{m}$, except with negligible probability.
3. There is a PPT algorithm SampleGaussian $\left(\boldsymbol{A}, \boldsymbol{T}_{\boldsymbol{A}}, \sigma, \boldsymbol{c}\right)$ that returns $\boldsymbol{x} \in \Lambda_{q}^{\perp}(\boldsymbol{A})$ drawn from a distribution statistically close to $\mathcal{D}_{\Lambda, \sigma, c}$.
4. There is a PPT algorithm SamplePre $\left(\boldsymbol{A}, \boldsymbol{T}_{\boldsymbol{A}}, \boldsymbol{u}, \sigma\right)$ that returns $\boldsymbol{x} \in \Lambda_{q}^{\perp}(A)$ sampled from a distribution statistically close to $\mathcal{D}_{\Lambda_{q}^{u}(\boldsymbol{A}), \sigma}$, whenever $\Lambda_{q}^{u}(A)$ is not empty.

Norm of a random matrix. The following lemmata can be used to bound the norm of a random matrix in $\{-1,1\}^{m \times m}$.

Lemma 3.3.2 ( $\boxed{\mathrm{ABB} 10}$, Lemma 15]) Let $\boldsymbol{R}$ be a $k \times m$ matrix chosen at random from $\{-1,1\}^{k \times m}$. Then,

$$
\operatorname{Pr}[\|\boldsymbol{R}\|>12 \sqrt{k+m}]<e^{-(k+m)}
$$

Lemma 3.3.3 ([ABB10, Lemma 16]) Let $\boldsymbol{u} \in \mathbb{R}^{m}$ be some vector of norm 1. Let $\boldsymbol{R}$ be a $k \times m$ matrix chosen at random from $\{-1,1\}^{k \times m}$. Then

$$
\operatorname{Pr}[\|\boldsymbol{R} \boldsymbol{u}\|>\sqrt{k} \omega(\sqrt{\log k})]<\operatorname{negl}(k)
$$

Randomness extraction. We will use the following lemma, which follows from a generalization of the leftover hash lemma due to Dodis et al. [DRS04]. Agrawal, Boneh, and Boyen [ABB10] prove the lemma for prime moduli $q$ but that the result extends to square-free values of $q$ by the Chinese remainder theorem.

Lemma 3.3.4 [|ABB10, Lemma 13]] Suppose that $m>(n+1) \lg q+\omega(\log n)$ and that $q>2$ is square free. Let $\boldsymbol{R}$ be an $m \times k$ matrix chosen uniformly in $\{1,-1\}^{m \times k} \bmod q$ where $k=k(n)$ is polynomial in $n$. Let $\boldsymbol{A}$ and $\boldsymbol{B}$ be matrices chosen uniformly in $\mathbb{Z}^{n \times m}$ and $\mathbb{Z}^{n \times k}$ respectively. Then for all vectors $w \in \mathbb{Z}^{m}$, the distribution $\left(\boldsymbol{A}, \boldsymbol{A} \boldsymbol{R}, \boldsymbol{R}^{\top} w\right)$ is statistically close to the distribution $\left(\boldsymbol{A}, \boldsymbol{B}, \boldsymbol{R}^{\top} w\right)$.

Lattice problems. Two of the most well known computational problems on lattices are the shortest vector problem and the shortest independent vectors problem. Specifically, let $\lambda_{1}(\Lambda)$ denote the length of the shortest nonzero vector in the lattice $\Lambda$ and $\lambda_{n}(\Lambda)$ the minimum length of a set of $n$ linearly independent vectors from $\Lambda$, where the length of a set is defined as the length of longest vector in it, then the shortest vector problem and the shortest independent vectors problem are defined as follows:

Definition 3.3.5 [Shortest Vector Problem, GapSVP ${ }_{\gamma}$ ] GapSVP is a promise problem. Specifically, an instance of $\operatorname{GapSVP}_{\gamma}$, for $\gamma=\gamma(n)>1$, is given by an $n$-dimensional lattice $\Lambda$, and a number $d>0$. In YES instances, $\lambda_{1}(\Lambda) \leq d$ whereas in $N O$ instances $\lambda_{1}(\Lambda)>\gamma(n) \cdot d$.

Definition 3.3.6 [Shortest Independent Vectors Problem, SIVP ${ }_{\gamma}$ ] An instance of SIVP ${ }_{\gamma}$, for $\gamma=\gamma(n)>1$, is given by an $n$-dimensional lattice $\Lambda$. The goal is to output a set of $n$ linearly independent lattice vectors of length at most $\gamma(n) \cdot \lambda_{n}(\Lambda)$.

### 3.3.1 Sampling algorithms

The following algorithms are used to sample short vectors and random basis from specific lattices.

Algorithm ToBasis : Micciancio and Goldwasser [MG02] showed that a full-rank set $\boldsymbol{S}$ in a lattice $\Lambda$ can be converted into a basis $T$ for $\Lambda$ with an equally low Gram-Schmidt norm.

Lemma 3.3.7 ([MG02, Lemma 7.1]) Let $\Lambda$ be an $m$-dimensional lattice. There is a deterministic polynomial-time algorithm that, given an arbitrary basis of $\Lambda$ and a fullrank set $\boldsymbol{S}=s_{1}, \ldots, s_{m}$ in $\Lambda$, returns a basis $T$ of $\Lambda$ satisfying $\|\widetilde{\boldsymbol{T}}\| \leq\|\widetilde{\boldsymbol{S}}\|$ and $\|\boldsymbol{T}\| \leq\|\boldsymbol{S}\| \sqrt{m} / 2$

Algorithm TrapGen : Ajtai [Ajt96] and later Alwen and Peikert [AP09] showed how to sample an essentially uniform matrix $\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times m}$ along with a basis $\boldsymbol{S}$ of $\Lambda_{q}^{\perp}(\boldsymbol{A})$ with low GramSchmidt norm.

Theorem 3.3.8 ([AP09, Theorem 3.2] with $\delta=1 / 3$ ) Let $q, n, m$ be positive integers with $q \geq 2$ and $m \geq 6 n \lg q$. There is a probabilistic polynomial-time algorithm $\operatorname{TrapGen}(q, n, m)$ that outputs a pair $\left(\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times m}, \boldsymbol{S} \in \mathbb{Z}^{m \times m}\right)$ such that $\boldsymbol{A}$ is statistically close to uniform in $\mathbb{Z}_{q}^{n \times m}$ and $\boldsymbol{S}$ is a basis for $\Lambda_{q}^{\perp}(\boldsymbol{A})$, satisfying $\|\widetilde{\boldsymbol{S}}\| \leq O(\sqrt{n \log q})$ and $\|\boldsymbol{S}\| \leq O(n \log q)$ with overwhelming probability in $n$.

We let $\sigma_{\mathrm{TG}}=O(\sqrt{n \log q})$ denote the maximum with high probability Gram-Schmidt norm of a basis produced by TrapGen.

Algorithm ExtendBasis : Peikert et al. [CHKP10] shows how to construct a basis for $\Lambda_{q}^{\perp}(\boldsymbol{A}\|\boldsymbol{B}\| \boldsymbol{C})$ from a basis for $\Lambda_{q}^{\perp}(\boldsymbol{B})$.

Theorem 3.3.9 For $i=1,2,3$ let $\boldsymbol{A}_{i}$ be a matrix in $\mathbb{Z}_{q}^{n \times m_{i}}$ and let $\boldsymbol{A}:=\left(\boldsymbol{A}_{1}\left\|\boldsymbol{A}_{2}\right\| \boldsymbol{A}_{3}\right)$. Let $\boldsymbol{T}_{2}$ be a basis of $\Lambda_{q}^{\perp}\left(\boldsymbol{A}_{2}\right)$. There is deterministic polynomial time algorithm ExtendBasis $\left(\boldsymbol{A}_{1}\right.$, $\left.\boldsymbol{A}_{2}, \boldsymbol{A}_{3}, \boldsymbol{T}_{2}\right)$ that outputs a basis $T$ for $\Lambda_{q}^{\perp}(\boldsymbol{A})$ such that $\|\widetilde{\boldsymbol{T}}\|=\left\|\widetilde{\boldsymbol{T}_{2}}\right\|$

Algorithm SampleLeft : The algorithm takes as input a full rank matrix $\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times m}$, a short basis $\boldsymbol{T}_{\boldsymbol{A}}$ of $\Lambda_{q}^{\perp}(\boldsymbol{A})$, a matrix $\boldsymbol{B} \in \mathbb{Z}_{q}^{n \times m_{1}}$, a vector $\boldsymbol{u} \in \mathbb{Z}_{q}^{n}$, and a Gaussian parameter $\sigma$. Let $\boldsymbol{F}:=(\boldsymbol{A} \| \boldsymbol{B})$, then the algorithm outputs a vector $\boldsymbol{e} \in \mathbb{Z}^{m+m_{1}}$ in the coset $\Lambda_{q}^{\boldsymbol{u}}(\boldsymbol{F})$.

Theorem 3.3.10 ( $\overline{\mathrm{ABB} 10}$, Theorem 17], [CHKP10, Lemma 3.2]) Let $q>2, m>n$ and $\sigma>\left\|\boldsymbol{T}_{\boldsymbol{A}}\right\| \cdot \omega\left(\sqrt{\log \left(m+m_{1}\right)}\right)$. Then $\operatorname{SampleLeft}\left(\boldsymbol{A}, \boldsymbol{B}, \boldsymbol{T}_{\boldsymbol{A}}, \boldsymbol{u}, \sigma\right)$ outputs a vector $\boldsymbol{e} \in \mathbb{Z}^{m+m_{1}}$ statistically close to $\mathcal{D}_{\Lambda_{q}^{u}(\boldsymbol{F}), \sigma}$.

Algorithm SampleRight : The algorithm takes as input matrices $\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times k}$ and $\boldsymbol{R} \in \mathbb{Z}^{k \times m}$, a full rank matrix $\boldsymbol{B} \in \mathbb{Z}_{q}^{n \times m}$ and a short basis $\boldsymbol{T}_{\boldsymbol{B}}$ of $\Lambda_{q}^{\perp}(\boldsymbol{B})$, a vector $\boldsymbol{u} \in \mathbb{Z}_{q}^{n}$, and a Gaussian parameter $\sigma$. Let $\boldsymbol{F}:=(\boldsymbol{A} \| \boldsymbol{A} \boldsymbol{R}+\boldsymbol{B})$, then the algorithm outputs a vector $\boldsymbol{e} \in \mathbb{Z}^{k+m}$ in the coset $\Lambda_{q}^{u}(\boldsymbol{F})$.
Let $S^{m}$ be the $m$-sphere $\left\{\boldsymbol{x} \in \mathbb{R}^{m+1}:\|\boldsymbol{x}\|=1\right\}$. We define $s_{\boldsymbol{R}}:=\|\boldsymbol{R}\|=\sup _{\boldsymbol{x} \in S^{m-1}} \| \boldsymbol{R}$. $\boldsymbol{x} \|$.

Theorem 3.3.11 ([ABB10, Theorem 19]) Let $q>2, m>n$ and $\sigma>\left\|\boldsymbol{T}_{\boldsymbol{B}}\right\| \cdot s_{R}$. $\omega(\sqrt{\log (k+m)})$. Then SampleRight $\left(\boldsymbol{A}, \boldsymbol{B}, \boldsymbol{R}, \boldsymbol{T}_{\boldsymbol{B}}, \boldsymbol{u}, \sigma\right)$ outputs a vector $\boldsymbol{e} \in Z^{k+m}$ distributed statistically close to $\mathcal{D}_{\Lambda_{q}^{u}(\boldsymbol{F}), \sigma}$.

### 3.3.2 The LWE Problem

The Learning with Errors problem (LWE, for short) is the problem of determining a secret vector over $\mathbb{F}_{q}$ given a polynomial number of noisy inner products. The decision variant is to distinguish such samples from random. More formally, we define the (average-case) problem as follows:

Definition 3.3.12 ([|Reg05]) Let $n \geq 1$ and $q \geq 2$ be integers, and let $\chi$ be a probability distribution on $\mathbb{Z}_{q}$. For $\boldsymbol{r} \in \mathbb{Z}_{q}^{n}$, let $A_{\boldsymbol{r}, \chi}$ be the probability distribution on $\mathbb{Z}_{q}^{n} \times \mathbb{Z}_{q}$ obtained by choosing a vector $\boldsymbol{a} \in \mathbb{Z}_{q}^{n}$ uniformly at random, choosing $e \in \mathbb{Z}_{q}$ according to $\chi$, and outputting $(\boldsymbol{a},\langle\boldsymbol{a}, \boldsymbol{r}\rangle+e)$.
(a) The search- $\operatorname{LWE}_{q, n, \chi}$ problem is: for uniformly random $\boldsymbol{r} \in \mathbb{Z}_{q}^{n}$, given a poly $(n)$ number of samples from $A_{\boldsymbol{r}, \chi}$, output $\boldsymbol{r}$.
(b) The decision- $\operatorname{LWE}_{q, n, \chi}$ problem is: for uniformly random $\boldsymbol{r} \in \mathbb{Z}_{q}^{n}$, given a poly $(n)$ number of samples that are either (all) from $A_{r, \chi}$ or (all) uniformly random in $\mathbb{Z}_{q}^{n} \times Z_{q}$, output 0 if the former holds and 1 if the latter holds.

We say the decision- $\mathrm{LWE}_{q, n, \chi}$ problem is infeasible if for all polynomial-time algorithms $\mathcal{A}$, the probability that $\mathcal{A}$ solves the decision-LWE problem (over $r$ and $\mathcal{A}$ random coins) is negligibly close to $1 / 2$ as a function of $n$.

The power of the LWE problem comes from the fact that for certain noise distributions $\chi$, solving the search-LWE problem is as hard as finding approximate solutions to the shortest independent vectors problem (SIVP) and the decision version of the shortest vector problem (GapSVP) in the worst case. For polynomial size $q$ there is a quantum reduction due to Regev, while for exponential size $q$ there is a classical reduction due to Peikert. Furthermore, the search and decision versions of the problem are equivalent whenever $q$ is a product of small primes. These results are summarized in the following:

Definition 3.3.13 For $\alpha \in(0,1)$ and an integer $q>2$, let $\bar{\Psi}_{\alpha}$ denote the probability distribution over $\mathbb{Z}_{q}$ obtained by choosing $x \in \mathbb{R}$ according to the normal distribution with mean 0 and standard deviation $\alpha / \sqrt{2 \pi}$ and outputting $\lfloor q x\rceil$.

Theorem 3.3.14 ([|Reg05|) Let $n, q$ be integers and $\alpha \in(0,1)$ such that $q=\operatorname{poly}(n)$ and $\alpha q>$ $2 \sqrt{n}$. If there exists an efficient (possibly quantum) algorithm that solves decision-LWE ${ }_{q, n, \bar{\Psi}_{\alpha}}$, then there exists an efficient quantum algorithm that approximates SIVP and GapSVP to within $\widetilde{O}(n / \alpha)$ in the worst case.

Theorem 3.3.15 ([|Pei09]) Let $n, q$ be integers and $\alpha \in(0,1)$, and $q=\sum_{i} q_{i} \leq 2^{n / 2}$, where the $q_{i}$ are distinct primes satisfying $\omega(\log n) / \alpha \leq q_{i} \leq \operatorname{poly}(n)$. If there exists an efficient (classical) algorithm that solves decision-LWE $\underline{q}_{q, n, \bar{\Psi}_{\alpha}}$, then there exists an efficient (classical) algorithm that approximates GapSVP to within $\widetilde{O}(n / \alpha)$ in the worst case.

The following lemma will be used to show correctness of decryption.
Lemma 3.3.16 ([ABB10, Lemma 12]) Let $\boldsymbol{e}$ be some vector in $\mathbb{Z}^{m}$ and let $\boldsymbol{v} \leftarrow \bar{\Psi}_{\alpha}^{m}$. Then the quantity $|\langle\boldsymbol{e}, \boldsymbol{v}\rangle|$ when treated as an integer in $(-q / 2, q / 2]$ satisfies

$$
|\langle\boldsymbol{e}, \boldsymbol{v}\rangle| \leq\|\boldsymbol{e}\| \cdot(q \alpha \cdot \omega(\sqrt{\log m})+\sqrt{m} / 2)
$$

with overwhelming probability in $m$.

## Chapter 4

## Generalized Key Delegation for Wildcarded Identity-Based and Inner-Product Encryption

Even though the WIBE and WKD-IBE constructions in $\mathrm{ACD}^{+} 06$, AKN07] are very practical, they had two significant shortcomings. First, their security proofs only hold in cases where the maximum hierarchy depth $L$ is a constant due to the fact that they are not tight and lose a factor which is exponential in $L$. As a result, these schemes can only be used in scenarios where such a restriction is acceptable. In particular, when using WKD-IBE schemes to build identity-based broadcast encryption schemes, such a limitation on the maximum hierarchy depth will have a direct impact on the maximum size of the target group. Second, their solutions do not hide the pattern associated with the ciphertext. Hence, their schemes cannot be used in any application where the anonymity of the recipient needs to be preserved.

The main goal here is to overcome the limitations of existing WKD-IBE and WIBE schemes. To achieve the first of these goals, we show in Section 4.2 how to convert the WIBE scheme by Abdalla et al. [ACD ${ }^{+}$06] based on the Boneh-Boyen HIBE scheme [BB04] into a nonanonymous WW-IBE scheme which is fully secure even when the maximum hierarchy depth $L$ is a polynomial in the security parameter. Towards this goal, we make use of bilinear groups of composite order [BGN05], whose order is a product of three primes. Our non-anonymous fully secure WW-IBE scheme has three main ingredients in its design. The first one is the generalization of the secret key delegation mechanism already present in the underlying Boneh-Boyen HIBE scheme. This technique is similar to one used by Abdalla et al. in [AKN07]. The second one is the addition of a ciphertext delegation mechanism, using a technique similar to the one employed in [ $\left.\mathrm{ACD}^{+} 06\right]$. Finally, the third key ingredient is the use of the dual system technique of Waters Wat09, LW10] to achieve security with respect to adaptive adversaries. In a dual system, there are two kinds of keys and ciphertexts: normal and semi-functional. While normal keys can be used to decrypt both normal and semi-functional ciphertexts, semi-functional keys can only decrypt normal ciphertexts. When a semi-functional key is used to decrypt a semifunctional ciphertext, decryption will fail with all but negligible probability. In a dual system, a security proof usually employs a standard hybrid argument over a sequence of games. The first one is the real security game in which key derivation and encryption queries are answered with normal keys and ciphertexts. In the next game, the ciphertext given to the attacker is changed to semi-functional. Then, in each subsequent game, the secret keys used to answer key derivation queries are changed to semi-functional, one at a time. Finally, once everything the attacker receives is semi-functional, then it is straight-forward to prove security directly.

As we show in Section 4.2, these three ingredients fit nicely together and we are able to prove its security. However, this is not always the case. In particular, when considering WWIBE schemes based on the Boneh-Boyen-Goh HIBE [BBG05], we did not succeed extending the proof strategy by Lewko and Waters [LW10] to the WW-IBE case. This is due to the fact that, when a normal secret key is changed to semi-functional, the information leaked by the delegation mechanisms of the secret keys together with the presence of wildcards in the challenge ciphertext breaks down the information theoretical argument used to hide the nominality of semi-functional keys. Nevertheless, it is worth pointing out that recent techniques by Lewko and Waters [LW12] might be useful in overcoming these issues.

One of the key elements in designing a WW-IBE scheme, starting from the HIBE schemes in [BB04, BBG05], is to have some sort of delegation mechanism also for the ciphertext in order to replace the wildcards with the identities required by the secret key. So we have two delegation mechanisms, one for the secret keys and one for the ciphertexts. These two mechanisms can be symmetric in the sense that they are essentially the same. Let us think to the [BBG05] scheme: If we want a delegation mechanism for the ciphertext what we can do is essentially to reuse the secret keys delegation mechanism. The issue here is that we are introducing a stronger correlation between ciphertexts and secret keys. Thus when we proceed to prove the semantic security by means of the dual system encryption, during the secret key games the information theoretical argument necessary to finish the proof breaks down because the ciphertext now release too much information. Instead this issue doesn't arise in the [BB04] scheme because we are able to define two different delegation mechanisms avoiding in this way the stronger correlation between ciphertexts and secret keys. Unfortunately when we try to achieve the anonymity property then it seems that this requires a symmetric delegation mechanism and this blocks us to achieve even semantic security.

In order to achieve, and to avoid all the issues of the previous approach, our second goal of building WW-IBE, which are not only fully secure but also anonymous, we first introduce in Section 4.3 a generalization of the notion of hierarchical inner predicate encryption [OT09], which we call inner-product encryption with generalized key delegation (WKD-IPE, for short). Next, after introducing the new WKD-IPE notion, we present a generic transform which converts any WKD-IPE into a WW-IBE scheme. The reduction uses the same trick used in [KSW08] to build hidden vector encryption from inner product encryption. Specifically, zero entries are used to simulate wildcards.

Since the transform preserves anonymity and has a tight security reduction, the resulting WW-IBE scheme will be anonymous and fully secure as long as the underlying WKD-IPE is anonymous and fully secure. Finally, we show how to modify the hierarchical inner-product encryption scheme from $\left[\mathrm{LOS}^{+} 10\right]$ to allow for more general key delegation patterns. As for the previous scheme, we also use the dual system encryption methodology of Waters [Wat09] in the security proof.

It is worth noticing that, even though our constructions allow for wildcards in the key derivation algorithm, they do not take into account the limited delegation property through which one can prevent users from further deriving keys for identities below them. As remarked by Abdalla et al. in [AKN07], this is without loss of generality as the limited delegation property can be easily achieved in the case where the maximum depth is fixed by padding the identity vector with dummy strings at the unused levels. Hence, here we only focuses on wildcard key delegation.

### 4.1 Definitions

In this section, we introduce a new primitive, called identity-based encryption with wildcards and generalized key delegation, which combines the notions of wildcarded IBE $\mathrm{ACD}^{+} 06$ ] and

IBE with generalized key delegation AKN07]. In doing so, we adopt the same notation and definition style used in $\mathrm{ACD}^{+} 06$.

### 4.1.1 Identity-Based Encryption with Wildcards and Generalized Key Delegation

Like WIBE and WKD-IBE, Identity-Based Encryption with Wildcards and Generalized Key Delegation (WW-IBE, for short) allows for more general patterns either during the encryption or the key derivation processes. Such a pattern is described by a vector

$$
P=\left(P_{1}, \ldots, P_{\ell}\right) \in\left(\{0,1\}^{*} \cup\{*\}\right)^{\ell}
$$

where $\star$ is a special wildcard symbol. As in a WKD-IBE, a user in possession of the secret key for a given pattern $P$ can generate secret keys for any pattern $P^{\prime}$ that matches $P$. We say that a pattern $P^{\prime}=\left(P_{1}^{\prime}, \ldots, P_{\ell^{\prime}}^{\prime}\right)$ matches $P$, denoted $P^{\prime} \in_{\star} P$, if and only if $\ell^{\prime} \leq \ell ; \forall i=1 \ldots \ell^{\prime}$, $P_{i}^{\prime}=P_{i}$ or $P_{i}=\star$; and $\forall i=\ell^{\prime}+1 \ldots \ell, P_{i}=\star$. As in a WIBE, a user in possession of the secret key for a given pattern $P$ can decrypt ciphertexts for any pattern $P^{\prime}$ that matches $P$.

Formally, an WW-IBE scheme can be defined as a predicate encryption scheme with public index in the following way. Let $L$ be the hierarchy depth then, the keys space $K$ and the indices space $I$ consist of all the patterns $P=\left(P_{1}, \ldots, P_{\ell}\right) \in\left(\{0,1\}^{*} \cup\{*\}\right)^{\ell}$ for $1 \leq \ell \leq L$ and the WW-IBE predicate $P_{L}$ over $(K, I)$ is defined as follows:

$$
P_{L}\left(k=P^{\prime}, \mathfrak{i n d}=P\right)= \begin{cases}1 & \text { if } P^{\prime} \in_{\star} P \\ 0 & \text { otherwise }\end{cases}
$$

Definition 4.1.1 [WW-IBE] An Identity-Based Encryption with Wildcards and Generalized Key Delegation scheme is a tuple of 5 algorithms WW $-I B E=($ Setup, KeyGen, KeyDer, Enc, Dec) with the following syntax:

Setup $\left(1^{\lambda}, 1^{L}\right)$ outputs public and master secret keys ( $m p k, m s k$ ) for security parameter $\lambda$ and hierarchy depth $L$.

KeyGen $\left(m s k, P=\left(P_{1}, \ldots, P_{\ell}\right)\right)$ on input a master secret key $m s k$ and pattern $P$, outputs secret key $s k_{P}$;

KeyDer $\left(m p k, s k_{P}, P^{\prime}\right)$ on input a master public key $m p k$, a secret key $s k_{P}$ for pattern $P=$ $\left(P_{1}, \ldots, P_{\ell}\right)$ and pattern $P^{\prime}$ such that $P^{\prime} \in_{\star} P$, outputs secret key $s k_{P^{\prime}}$;
$\operatorname{Enc}(m p k, P, \mathfrak{m})$ on input public key $m p k$, pattern $P=\left(P_{1}, \ldots, P_{\ell}\right)$ and plaintext $\mathfrak{m}$, outputs ciphertext Ct ;
$\operatorname{Dec}\left(m p k, s k_{P^{\prime}}, C t\right)$ outputs a string $y$.

Correctness. We require that for all key pairs $(m p k, m s k)$ output by Setup, all messages $\mathfrak{m} \in\{0,1\}^{*}$, all $0 \leq \ell \leq L$, all patterns $P, P^{\prime} \in\left(\{0,1\}^{*} \cup\{*\}\right)^{\ell}$ such that $P^{\prime} \in_{*} P$, all $s k_{P^{\prime}}$ obtained through a chain of key derivation $s k_{P_{1}}=\operatorname{KeyGen}\left(m s k, P_{1}\right), s k_{P_{i+1}}=$ $\operatorname{KeyDer}\left(m p k, s k_{P_{i}}, P_{i+1}\right)$ such that $P_{1} \in_{\star} \ldots \epsilon_{\star} P_{n}=P^{\prime}$,

$$
\operatorname{Dec}\left(s k_{P^{\prime}}, \operatorname{Enc}(m p k, P, \mathfrak{m})\right)=\mathfrak{m}
$$

with probability one.

### 4.1.2 Security

As for Definition 2.2.1, IND-CPA security for WW-IBE schemes is formalized by means of the following game between an adversary $\mathcal{A}$ and a challenger $\mathcal{C}$. Specifically, we define the security of WW-IBE schemes through a game with an adversary which is different from the security games of HIBE schemes in two points: first, the adversary can query for the secret keys corresponding to arbitrary patterns, rather than specific identity vectors and second, the adversary chooses a challenge pattern instead of an identity to which the challenge ciphertext will be encrypted. Of course, the adversary is not allowed to query the key derivation oracle for any pattern matched by the challenge pattern.

We give complete form of the security definition following [SW08]. Specifically, game $\operatorname{IND}-\mathrm{CPA}_{\mathcal{A}}^{W W-I B E}\left(1^{\lambda}\right)$, where $\lambda$ is the security parameter, is defined as follows.

Setup: $\mathcal{C}$ generates master public key and secret key by invoking the setup algorithm on input the security parameter $\lambda$ and the hierarchy depth $L$ given in unary. Specifically, $(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}, 1^{L}\right)$. Then, $\mathcal{C}$ starts the interaction with $\mathcal{A}$ on input $m p k$.

We let $S$ denote the set of private keys that $\mathcal{C}$ has created but not yet given to the adversary. At this point, $S=\emptyset$.

Query Phase 1: $\mathcal{A}$ makes Create, Delegate, and Reveal key queries. Specifically:
To make a Create query, $\mathcal{A}$ specifies a pattern $P \in\left(\{0,1\}^{*} \cup\{*\}\right)^{\leq L}$. In response, $\mathcal{C}$ creates a key for this pattern by calling the key generation algorithm, $s k_{P}=$ KeyGen $(m s k, P)$, and places this key in the set $S$. $\mathcal{C}$ only gives $\mathcal{A}$ a reference to this key, not the key itself.
To make a Delegate query, $\mathcal{A}$ specifies a key $s k_{P}$ in the set $S$ and a pattern $P^{\prime}$. In response, $\mathcal{C}$ makes a key for this new vector by running the delegation algorithm on $s k_{P}$ and $P^{\prime} . \mathcal{C}$ adds this key to the set $S$ and again gives $\mathcal{A}$ only a reference to it, not the actual key.

To make a Reveal query, $\mathcal{A}$ specifies an element of the set $S$. $\mathcal{C}$ gives this key to $\mathcal{A}$ and removes it from the set $S$. Notice that $\mathcal{A}$ needs no longer make any delegation queries for this key because it can run delegation algorithm on the revealed key for itself.

Challenge: $\mathcal{A}$ submits two equal length messages $\mathfrak{m}_{0}, \mathfrak{m}_{1}$ and a challenge pattern $P=\left(P_{1}, \ldots, P_{\ell}\right)$
where $0 \leq \ell \leq L . \mathcal{C}$ chooses random $\eta \in\{0,1\}$ and generates ciphertext $\mathrm{Ct}^{\star}=$ $\operatorname{Enc}\left(m p k, P, \mathfrak{m}_{\eta}\right) . \mathcal{C}$ sends back $\mathrm{Ct}^{\star}$ to $\mathcal{A}$.

Query Phase 2: It is the same to Query Phase 1.
Guess: Eventually, $\mathcal{A}$ submits its guess $\eta^{\prime}$.
Winning Condition: $\mathcal{A}$ wins the game if the following conditions are satisfied:

1. $\eta=\eta^{\prime}$.
2. No secret key for any pattern obtained through delegation from any of the revealed secret key is such that it matches the target pattern (i.e., any $P^{\prime}$ such that $P^{\prime} \in \in_{\star} P$ ).

The advantage of $\mathcal{A}$ in the above game is defined as

$$
\operatorname{Adv}_{\text {IND-CPA }}^{W W-I B E, \mathcal{A}}(\lambda)=\operatorname{Prob}\left[\operatorname{IND}-\mathrm{CPA}_{\mathcal{A}}^{W W-I B E}\left(1^{\lambda}\right)=1\right]-1 / 2
$$

Definition 4.1.2 A WW-IBE scheme is IND-CPA-secure if any polynomial-time adversaries making at most a polynomial number of queries to the key derivation oracle only has a negligible advantage in the IND-CPA game described above, i.e., $\operatorname{Adv}_{\text {IND-CPA }}^{W W-I B E, \mathcal{A}}(\lambda)$ is negligible function of $\lambda$.

### 4.1.3 Anonymous WW-IBE

An important property we investigate is the anonymity of WW-IBE schemes. The objective is to hide the pattern in the ciphertext. Therefore, the adversary could also choose two patterns to be challenged on. More specifically, game ANO-CPA $\mathcal{A}^{W W-I B E}\left(1^{\lambda}\right)$, where $\lambda$ is the security parameter, is defined as follows.

Setup: $\mathcal{C}$ generates master public key and secret key by invoking the setup algorithm on input the security parameter $\lambda$ and the hierarchy depth $L$ given in unary. Specifically, $(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}, 1^{L}\right)$. Then, $\mathcal{C}$ starts the interaction with $\mathcal{A}$ on input $m p k$.
We let $S$ denote the set of private keys that $\mathcal{C}$ has created but not yet given to the adversary. At this point, $S=\emptyset$.

Query Phase 1: $\mathcal{A}$ makes Create, Delegate, and Reveal key queries. Specifically:
To make a Create query, $\mathcal{A}$ specifies a pattern $P \in\left(\{0,1\}^{*} \cup\{*\}\right)^{\leq L}$. In response, $\mathcal{C}$ creates a key for this pattern by calling the key generation algorithm, $s k_{P}=$ KeyGen $(m s k, P)$, and places this key in the set $S$. $\mathcal{C}$ only gives $\mathcal{A}$ a reference to this key, not the key itself.
To make a Delegate query, $\mathcal{A}$ specifies a key $s k_{P}$ in the set $S$ and a pattern $P^{\prime}$. In response, $\mathcal{C}$ makes a key for this new vector by running the delegation algorithm on $s k_{P}$ and $P^{\prime} . \mathcal{C}$ adds this key to the set $S$ and again gives $\mathcal{A}$ only a reference to it, not the actual key.
To make a Reveal query, $\mathcal{A}$ specifies an element of the set $S$. $\mathcal{C}$ gives this key to $\mathcal{A}$ and removes it from the set $S$. Notice that $\mathcal{A}$ needs no longer make any delegation queries for this key because it can run delegation algorithm on the revealed key for itself.

Challenge: $\mathcal{A}$ submits two equal length messages $\mathfrak{m}_{0}, \mathfrak{m}_{1}$ and two challenge patterns $\mathbf{P}_{0}=$ $\left(P_{1}^{(0)}, \ldots, P_{\ell}^{(0)}\right)$ and $\mathbf{P}_{1}=\left(P_{1}^{(1)}, \ldots, P_{\ell}^{(1)}\right)$ where $0 \leq \ell \leq L . \mathcal{C}$ chooses random $\eta \in\{0,1\}$ and generates ciphertext $\mathrm{Ct}^{\star}=\mathrm{Enc}\left(m p k, P_{\eta}, \mathfrak{m}_{\eta}\right) . \mathcal{C}$ sends back $\mathrm{Ct}^{\star}$ to $\mathcal{A}$.

Query Phase 2: It is the same as Query Phase 1.
Guess: Eventually, $\mathcal{A}$ submits its guess $\eta^{\prime}$.
Winning Condition: $\mathcal{A}$ wins the game if the following conditions are satisfied:

1. $\eta=\eta^{\prime}$.
2. No secret key for any pattern obtained through delegation from any of the revealed secret key is such that it matches one of the challenge pattern but not the other.

The advantage of $\mathcal{A}$ in the above game is defined as

$$
\operatorname{Adv}_{\mathrm{ANO}-\mathrm{CPA}}^{W W-I B E, \mathcal{A}}(\lambda)=\operatorname{Prob}\left[\operatorname{ANO}-\mathrm{CPA}_{\mathcal{A}}^{W W-I B E}\left(1^{\lambda}\right)=1\right]-1 / 2
$$

Definition 4.1.3 A WW-IBE scheme is ANO-CPA-secure if any polynomial-time adversaries making at most a polynomial number of queries to the key derivation oracle only has a negligible advantage in the ANO-CPA game described above, i.e., $\mathbf{A d v}_{\text {ANO-CPA }}^{W W-I B E, \mathcal{A}}(\lambda)$ is negligible function of $\lambda$.

### 4.2 A WW-IBE Construction

Our first construction is based on a slight variant of the Boneh-Boyen WIBE scheme given in $\left[\mathrm{ACD}^{+} 06\right]$. We extend it to obtain generalized key delegation and prove its security in the standard model, under static assumptions in composite order bilinear groups [BGN05], by using Waters' Dual Encryption System Wat09]. As in [LW10], we use groups $\mathbb{G}, \mathbb{G}_{\mathrm{T}}$ whose order $N=p_{1} p_{2} p_{3}$ is a product of three primes. We note that the actual computation occurs in $\mathbb{G}_{p_{1}}$, the subgroup of $\mathbb{G}$ of order $p_{1} \cdot \mathbb{G}_{p_{2}}$, the subgroup of $\mathbb{G}$ of order $p_{2}$, is only used to define the semi-functional space and is not used in the real scheme. Thus, secret keys and ciphertexts will be semi-functional when they include terms in $\mathbb{G}_{p_{2}}$. Finally, $\mathbb{G}_{p_{3}}$, the subgroup of $\mathbb{G}$ of order $p_{3}$, is used to re-randomize the secret keys.

### 4.2.1 Complexity Assumptions

For completeness we restate here the complexity assumptions we use, first introduced by Lewko and Waters in [LW10]. The first two are instances of the general subgroup decision assumption.

Assumption LW. 1 For a given prime-order bilinear setting generator BSGen, define the following distribution: Pick a random bilinear setting $\left(N=p_{1} p_{2} p_{3}, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{BSGen}\left(1^{\lambda}\right)$, and then pick

$$
g_{1}, T_{1} \leftarrow \mathbb{G}_{p_{1}}, g_{3} \leftarrow \mathbb{G}_{p_{3}}, T_{2} \leftarrow \mathbb{G}_{p_{1} p_{2}}
$$

and set $D=\left(N, g_{1}, g_{3}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking Assumption LW. 1 is

$$
\mathbf{A d v}_{L W .1}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, T_{1}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, T_{2}\right)=1\right]\right|
$$

Definition 4.2.1 We say that Assumption LW. 1 holds for generator BSGen if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \mathbf{A d v}_{L W .1}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

Assumption LW. 2 For a given prime-order bilinear setting generator BSGen, define the following distribution: Pick a random bilinear setting $\left(N=p_{1} p_{2} p_{3}, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{BSGen}\left(1^{\lambda}\right)$, and then pick

$$
g_{1}, X_{1} \leftarrow \mathbb{G}_{p_{1}}, X_{2}, Y_{2} \leftarrow \mathbb{G}_{p_{2}}, g_{3}, X_{3} \leftarrow \mathbb{G}_{p_{3}}, T_{1} \leftarrow \mathbb{G}_{p_{1} p_{3}}, T_{2} \leftarrow \mathbb{G}_{p_{1} p_{2} p_{3}}
$$

and set $D=\left(N, g_{1}, g_{3}, X_{1} X_{2}, Y_{2} X_{3}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking Assumption LW. 2 is

$$
\mathbf{A d v}_{L W .2}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, T_{1}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, T_{2}\right)=1\right]\right|
$$

Definition 4.2.2 We say that Assumption LW. 2 holds for generator BSGen if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \mathbf{A d v}_{L W .2}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

Assumption LW. 3 For a given prime-order bilinear setting generator BSGen, define the following distribution: Pick a random bilinear setting $\left(N=p_{1} p_{2} p_{3}, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{BSGen}\left(1^{\lambda}\right)$, and then pick

$$
\rho, s \leftarrow \mathbb{Z}_{N}, g_{1} \leftarrow \mathbb{G}_{p_{1}}, g_{2}, X_{2}, Y_{2} \leftarrow \mathbb{G}_{p_{2}}, g_{3} \leftarrow \mathbb{G}_{p_{3}}, T_{2} \leftarrow \mathbb{G}_{\mathrm{T}}
$$

and set $T_{1}=\hat{e}\left(g_{1}, g_{1}\right)^{\rho s}$ and $D=\left(N, g_{1}, g_{2}, g_{3}, g_{1}^{\rho} X_{2}, g_{1}^{s} Y_{2}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking Assumption LW. 3 is

$$
\operatorname{Adv}_{L W .3}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, T_{1}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, T_{2}\right)=1\right]\right|
$$

Definition 4.2.3 We say that Assumption LW. 3 holds for generator BSGen if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \mathbf{A d v}_{L W .3}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

### 4.2.2 Our Construction

We first introduce some additional notation. If $P=\left(P_{1}, \ldots, P_{\ell}\right)$ is a pattern, then let $|P|=\ell$ be the length of $P$, let $\mathrm{W}(P)$ be the set containing all wildcard indices in $P$, i.e. the indices $1 \leq i \leq \ell$ such that $P_{i}=\star$, and let $\overline{\mathrm{W}}(P)$ be the complementary set containing all nonwildcard indices. Clearly $\mathrm{W}(P) \bigcap \overline{\mathrm{W}}(P)=\emptyset$ and $\mathrm{W}(P) \bigcup \overline{\mathrm{W}}(P)=\{1, \ldots, \ell\}$.
$\operatorname{Setup}\left(1^{\lambda}, 1^{L}\right)$ : Choose a description of a composite order bilinear setting $\left(N=p_{1} p_{2} p_{3}, \mathbb{G}, g\right.$, $\left.\mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{CBSGen}\left(1^{\lambda}, 3\right)$ with known factorization and random $g_{1} \leftarrow \mathbb{G}_{p_{1}}, g_{3} \leftarrow \mathbb{G}_{p_{3}}$. Choose random $\rho \leftarrow \mathbb{Z}_{N}$ and random $\left(u_{i, j} \leftarrow \mathbb{G}_{p_{1}}\right)_{i \in[L], j \in\{0,1\}}$, and set $\Omega=\hat{e}\left(g_{1}, g_{1}\right)^{\rho}$. Then, the master public key is

$$
m p k \leftarrow\left[N, g_{1}, g_{3},\left(u_{i, j}\right)_{i \in[L], j \in\{0,1\}}, \Omega\right]
$$

and the master secret key is

$$
m s k \leftarrow[\rho]
$$

$\operatorname{KeyGen}\left(m s k, P=\left(P_{1}, \ldots, P_{\ell}\right)\right):$ Choose random $\left(r_{i} \leftarrow \mathbb{Z}_{N}\right)_{i \in \overline{\mathrm{~W}}(P)}, W_{0} \leftarrow \mathbb{G}_{p_{3}}$ and $\left(W_{i} \leftarrow\right.$ $\left.\mathbb{G}_{p_{3}}\right)_{i \in \overline{\mathrm{~W}}(P)}$. Then set

$$
K_{0}=g_{1}^{\rho} \cdot \prod_{i \in \overline{\mathrm{~W}}(P)}\left(u_{i, 0} \cdot u_{i, 1}^{P_{i}}\right)^{r_{i}} \cdot W_{0}
$$

and

$$
\left(D_{i}=g_{1}^{r_{i}} \cdot W_{i}\right)_{i \in \overline{\mathrm{~W}}(P)}
$$

The secret key for pattern $P$ is $s k_{P} \leftarrow\left[K_{0},\left(D_{i}\right)_{i \in \overline{\mathrm{~W}}(P)}\right]$.
$\operatorname{Key} \operatorname{Del}\left(m p k, s k_{P}, P^{\prime}=\left(P_{1}^{\prime}, \ldots, P_{\ell}^{\prime}\right)\right)$ : If $P^{\prime} \in_{\star} P$ then the key delegation algorithm chooses random $\left(r_{i}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{i \in \overline{\mathrm{~W}}\left(P^{\prime}\right)}$ and $W_{0}^{\prime} \leftarrow \mathbb{G}_{p_{3}}$ and $\left(W_{i}^{\prime} \leftarrow \mathbb{G}_{p_{3}}\right)_{i \in \overline{\mathrm{~W}}\left(P^{\prime}\right)}$. Then the algorithm sets

$$
K_{0}^{\prime}=K_{0} \cdot \prod_{i \in \overline{\mathrm{~W}}\left(P^{\prime}\right)}\left(u_{i, 0} \cdot u_{i, 1}^{P_{i}^{\prime}}\right)^{r_{i}^{\prime}} \cdot W_{0}^{\prime}
$$

and

$$
\left(D_{i}^{\prime}=D_{i} \cdot g_{1}^{r_{i}^{\prime}} \cdot W_{i}^{\prime}\right)_{i \in \overline{\mathrm{~W}}(P) \cap \overline{\mathrm{W}}\left(P^{\prime}\right)} \quad \text { and } \quad\left(D_{i}^{\prime}=g_{1}^{r_{i}^{\prime}} \cdot W_{i}^{\prime}\right)_{i \in \mathrm{~W}(P) \cap \overline{\mathrm{W}}\left(P^{\prime}\right)}
$$

The secret key for pattern $P^{\prime}$ is $s k_{P^{\prime}} \leftarrow\left[K_{0}^{\prime},\left(D_{i}^{\prime}\right)_{i \in \overline{\mathrm{~W}}\left(P^{\prime}\right)}\right]$.
$\operatorname{Enc}\left(m p k, P=\left(P_{1}, \ldots, P_{\ell}\right), \mathfrak{m}\right)$ : Choose random $s \leftarrow \mathbb{Z}_{N}$ and then set

$$
\begin{gathered}
C_{1}=g_{1}^{s} \\
\left(C_{2, i}=\left(u_{i, 0} \cdot u_{i, 1}^{P_{i}}\right)^{s}\right)_{i \in \overline{\mathrm{~W}}(P)} \\
C_{3}=\Omega^{s} \cdot \mathfrak{m} \\
\left(C_{4, i, j}=u_{i, j}^{s}\right)_{i \in \mathrm{~W}(P), j \in\{0,1\}}
\end{gathered}
$$

The ciphertext for pattern $P$ and message $\mathfrak{m}$ is $\mathrm{Ct}_{P} \leftarrow\left[C_{1},\left(C_{2, i}\right) C_{3},\left(C_{4, i, j}\right)\right]$.
$\operatorname{Dec}\left(m p k, \mathrm{Ct}_{P}, s k_{P^{\prime}}\right)$ : The decryption algorithms returns $\perp$ if $P \not \notin \star_{*} P^{\prime}$. If this is not the case then to decrypt it is enough to focus on the non-wildcard positions of the secret key. Then, each identity in these positions must be matched by a corresponding identity in the ciphertext. In particular, if position $i$ corresponds to a non-wildcard in the secret key and to a wildcard in the ciphertext (these positions are those in $\mathrm{W}(P) \cap \overline{\mathrm{W}}\left(P^{\prime}\right)$ ) then that wildcard needs to be delegated to the required identity $P_{i}^{\prime}$. Thus, to recover the message $\mathfrak{m}$ the decryption algorithms computes what follows:

$$
C_{2}=\prod_{i \in \overline{\mathrm{~W}}(P) \cap \overline{\mathrm{W}}\left(P^{\prime}\right)} \hat{e}\left(D_{i}, C_{2, i}\right) \cdot \prod_{i \in \mathrm{~W}(P) \cap \overline{\mathrm{W}}\left(P^{\prime}\right)} \hat{e}\left(D_{i}, C_{4, i, 0} \cdot C_{4, i, 1}^{P_{i}^{\prime}}\right)
$$

then the blinding factor is obtained as

$$
\frac{\hat{e}\left(K_{0}, C_{1}\right)}{C_{2}}=\frac{\Omega^{s} \cdot \prod_{i \in \overline{\mathrm{~W}}\left(P^{\prime}\right)} \hat{e}\left(\left(u_{i, 0} \cdot u_{i, 1}^{P_{i}}\right)^{r_{i}}, g_{1}^{s}\right)}{\prod_{i \in \overline{\mathrm{~W}}\left(P^{\prime}\right)} \hat{e}\left(g_{1}^{r_{i}},\left(u_{i, 0} \cdot u_{i, 1}^{P_{i}}\right)^{s}\right)}=\Omega^{s} .
$$

Remember that the $\mathbb{G}_{p_{3}}$ part cancels out by the orthogonality property.

### 4.2.3 Security

In this section we will prove the following main theorem:
Theorem 4.2.4 If Assumptions LW.1, LW.2, and LW. 3 hold, then our scheme is IND-CPA secure in the sense of Definition 4.1.2.

Following Lewko and Waters [LW10], we define two additional structures used in the proofs of security. Specifically:

Semi-Functional Ciphertext : Let $g_{2}$ denote a generator of $\mathbb{G}_{p_{2}}$. An semi-functional ciphertext is created as follows: first, we use the encryption algorithm to form a normal ciphertext $\mathrm{Ct}^{\prime} \leftarrow\left[C_{1}^{\prime},\left(C_{2, i}^{\prime}\right), C_{3}^{\prime},\left(C_{4, i, j}^{\prime}\right)\right]$. Then, we choose random exponents $x \leftarrow \mathbb{Z}_{N},\left(z_{c, i} \leftarrow\right.$ $\left.\mathbb{Z}_{N}\right)_{i \in \overline{\mathrm{~W}}(P)}$ and $\left(z_{c, i, j} \leftarrow \mathbb{Z}_{N}\right)_{i \in \mathrm{~W}(P), j \in\{0,1\}}$. Finally, we set

$$
\begin{gathered}
C_{1}=C_{1}^{\prime} \cdot g_{2}^{x}, \\
\left(C_{2, i}=C_{2, i}^{\prime} \cdot g_{2}^{x \cdot z_{c, i}}\right)_{i \in \overline{\mathrm{~W}}(P)}, \\
C_{3}=C_{3}^{\prime}, \\
\left(C_{4, i, j}=C_{4, i, j}^{\prime} \cdot g_{2}^{x \cdot z z_{c, i, j}}\right)_{i \in \mathrm{~W}(P), j \in\{0,1\}} .
\end{gathered}
$$

Semi-Functional Keys : To create a semi-functional key, we first create a normal key $s k_{P} \leftarrow$ $\left(K_{0}^{\prime},\left(D_{i}^{\prime}\right)\right)$ using the key generation algorithm. Then, we choose random $y, z_{k} \leftarrow \mathbb{Z}_{N}$ and random $\left(z_{k, i} \leftarrow \mathbb{Z}_{N}\right)_{i \in \overline{\mathrm{~W}}(P)}$. Finally, we set

$$
K_{0}=K_{0}^{\prime} \cdot g_{2}^{y \cdot z_{k}}
$$

and

$$
\left(D_{i}=D_{i}^{\prime} \cdot g_{2}^{y \cdot z_{k, i}} \cdot W_{i}\right)_{i \in \overline{\mathrm{~W}}(P)} .
$$

Remark 4.2.5 Notice that semi-functional ciphertexts can still be decrypted by normal secret keys and semi-functional secret keys can still be used to decrypt normal ciphertext. But an semifunctional ciphertext, for pattern $P^{\prime}$, and an semi-functional secret key, for pattern $P$, lead to a random decryption with high probability. In fact the decryption algorithm will compute the blinding factor multiplied by the additional term $\hat{e}\left(g_{2}, g_{2}\right)^{x y \cdot\left(z_{c}-z_{k}\right)}$ where

$$
z_{c}=\sum_{i \in \overline{\mathrm{~W}}(P) \cap \overline{\mathrm{W}}\left(P^{\prime}\right)} z_{k, i} \cdot z_{c, i}+\sum_{i \in \overline{\mathrm{~W}}(P) \cap \mathrm{W}\left(P^{\prime}\right)} z_{k, i} \cdot\left(z_{c, i, 0}+z_{c, i, 1}\right) .
$$

If $z_{c}=z_{k}$, decryption will still work. We say that such a ciphertexts and secret keys are nominally semi-functional. We will use this concept during the security proofs arguing that the nominality is hidden even to an unbounded adversary under the constraints of the security game.

For a probabilistic polynomial-time adversary $\mathcal{A}$ which makes $q$ key queries, our proof of security will consist of the following $q+5$ games between $\mathcal{A}$ and a challenger $\mathcal{C}$.

Game $_{\text {Real }}$ : It is the real IND-CPA security game.
Game $_{\text {WD }}$ : It is the same as Game $_{\text {Real }}$ except that all key queries will be answered by fresh calls to the key generation algorithm.

Game $_{\text {Restricted }}$ : It is the same as Game ${ }_{\text {WD }}$ except that $\mathcal{A}$ cannot ask for keys for patterns which are prefixes of one of the challenge patterns modulo $p_{2}$. We will retain this restriction in all subsequent games.

Game $_{\mathbf{k}}$ : For $k$ from 0 to $q$, we define $\mathbf{G a m e}_{\mathbf{k}}$ like $\mathbf{G a m e}_{\text {Restricted }}$ except that the ciphertext given to $\mathcal{A}$ is semi-functional and the first $k$ keys are semi-functional. The rest of the keys are normal.

Game $_{\text {Final }}$ : It is the same game as Game $_{\mathbf{q}}$, except that the challenge ciphertext has $C_{3}$ random in $\mathbb{G}_{\mathrm{T}}$ (thus the ciphertext is independent from the messages provided by $\mathcal{A}$ ). It is clear that no adversary can have advantage greater than 0 here.

We start by giving an overview of the proof of security.
Game $_{\text {Real }}=$ Game $_{\text {WD }}:$ The keys are identically distributed whether they are produced by key delegation from a previous key or from a fresh call to the key generation.

Game $_{\text {WD }} \approx_{\mathbf{c}}$ Game $_{\text {Restricted }}$ : Essentially, if the adversary is able to ask for keys for patterns which are prefixes of one of the challenge patterns modulo $p_{2}$, then this means that the adversary can find a non-trivial factor of $N$ and can be used to break either Assumption LW.2. This restriction is crucial to show that $\mathbf{G a m e}_{\mathbf{k}-\mathbf{1}} \approx \mathbf{G a m e}_{\mathbf{k}}$.

Game $_{\text {Restricted }} \approx_{\mathbf{c}}$ Game $_{0}$ : In Game ${ }_{0}$, the challenge ciphertext $\mathrm{Ct}^{\star}$ is semi-functional, while all keys are normal. Notice that Assumption LW. $1\left(N, g_{1}, g_{3}\right)$ is enough to generate $m p k$ and $m s k$. Then, $T$ can be used to generate $\mathrm{Ct}^{\star}$ and, depending on the nature of $T$, $\mathrm{Ct}^{\star}$ can be normal as in Game Restricted or semi-functional as in Game $\mathbf{0}_{\mathbf{0}}$.

Game $_{\mathbf{k}-\mathbf{1}} \approx_{\mathbf{c}} \mathbf{G a m e}_{\mathbf{k}}$ : Given an instance of Assumption LW. $2\left(N, g_{1}, g_{3}, X_{1} X_{2}, Y_{2} X_{3}\right.$ ), $m p k$ and $m s k$ can be generated by using $g_{1}$ and $g_{3}$. The first $k-1$ secret key queries, which are semi-functional, can be answered by employing $Y_{2} X_{3}$. The last $q-k$ queries can be answered by invoking KeyGen on input $m s k$. Finally, the challenge ciphertext can be generated by employing $X_{1} X_{2}$ and the $k$-th secret key by employing $T$. Now, if $T \in \mathbb{G}_{p_{1} p_{3}}$, then the $k$-th secret key is normal and the joint distribution of the $k$-th secret key and the challenge ciphertext is as in $\mathbf{G a m e}_{\mathbf{k}-\mathbf{1}}$. On the other hand, if $T \in \mathbb{G}_{p_{1} p_{2} p_{3}}$, then the $k$-th key is nominally semi-functional with the respect to the challenge ciphertext. Hence, the simulator cannot test by himself the nature of $T$. Moreover, the nominality of $k$-th key is hidden to the adversary under the restriction of Game Restricted and from the fact that the adversary cannot ask secret keys for patters matching the challenge pattern.

Game $_{\mathbf{q}} \approx_{\mathbf{c}}$ Game $_{\text {Final }}:$ In Game $_{\mathbf{q}}$, the challenge ciphertext and secret keys are semifunctional. It is easy to see that these two games are indistinguishable under Assumption LW.3.

Game $_{\text {Final }}$ gives no advantage: In Game ${ }_{\text {Final }}, \eta$ is information-theoretically hidden from the attacker. Hence the attacker can obtain no advantage in breaking the scheme.

We give now formal proofs of the above statements.
Lemma 4.2.6 Game $_{\text {Real }}=$ Game $_{\text {WD }}$.
Proof: We note that the keys are identically distributed whether they are produced by the key delegation algorithm from a previous key or from a fresh call to the key generation algorithm. Thus, in the attacker's view, there is no difference between these games.

Lemma 4.2.7 If Assumption LW. 2 holds, then for all probabilistic polynomial time adversaries $\mathcal{A}$, Game ${ }_{\text {WD }} \approx_{c}$ Game $_{\text {Restricted }}$.

Proof: Suppose that $\mathcal{A}$ has probability $\epsilon$ of producing a pattern $P=\left(P_{1}, \ldots, P_{k}\right)$, that is a prefix of one of the challenge pattern $P^{\star}=\left(P_{1}^{\star}, \ldots, P_{j}^{\star}\right)$ modulo $p_{2}$. That is, there exists $i$ such that that $P_{i} \neq P_{i}^{\star}$ modulo $N$ and both are different from $\star$, and that $p_{2}$ divides $P_{i}-P_{i}^{\star}$ and thus $a=\operatorname{gcd}\left(P_{i}-P_{i}^{\star}, N\right)$ is a nontrivial factor of $N$. We notice that $p_{2}$ divides $a$ and set $b=\frac{N}{a}$. The following two cases are exhaustive and at least one occurs with probability at least $\epsilon / 2$ :

1. $\operatorname{ord}\left(g_{1}\right) \mid b$. Suppose case 1 has probability at least $\epsilon / 2$. We describe algorithm $\mathcal{B}$ that breaks Assumption LW.2. $\mathcal{B}$ receives ( $N, g_{1}, g_{3}, X_{1} X_{2}, Y_{2} X_{3}$ ) and $T$ and constructs $m p k$ and $m s k$ by choosing $\rho \leftarrow \mathbb{Z}_{N}$ and $\left(u_{i, j} \leftarrow \mathbb{Z}_{N}\right)_{i \in[L], j \in\{0,1\}}$, and setting $\Omega=\hat{e}\left(g_{1}, g_{1}\right)^{\rho}$, $m p k \leftarrow\left[N, g_{1}, g_{3},\left(g_{1}^{u_{i, j}}\right)_{i \in[L], j \in\{0,1\}}, \Omega\right]$ and $m s k \leftarrow[\rho]$. Then $\mathcal{B}$ runs $\mathcal{A}$ on input $m p k$ and uses knowledge of $m s k$ to answer $\mathcal{A}$ 's queries. At the end of the game, for all pattern $P$ 's for which $\mathcal{A}$ has asked for the key and for the challenge pattern $P^{\star}, \mathcal{B}$ computes $a=\operatorname{gcd}\left(P_{i}-P_{i}^{\star}, N\right)$. Then, if $\hat{e}\left(\left(X_{1} X_{2}\right)^{a}, Y_{2} X_{3}\right)$ is the identity element of $\mathbb{G}_{T}$ then $\mathcal{B}$ tests if $\hat{e}\left(T^{b}, X_{1} X_{2}\right)$ is the identity element of $\mathbb{G}_{T}$. If this second test is successful, then $\mathcal{B}$ declares $T \in \mathbb{G}_{p_{1} p_{3}}$. If it is not, $\mathcal{B}$ declares $T \in \mathbb{G}_{p_{1} p_{2} p_{3}}$. It is easy to see that if $p_{2}$ divides $a$ and $p_{1}=\operatorname{ord}\left(g_{1}\right)$ divides $b$, then $\mathcal{B}$ 's output is correct.
2. $\operatorname{ord}\left(g_{1}\right) \nmid b$ and $\operatorname{ord}\left(g_{3}\right) \mid b$. In this case, $\mathcal{B}$ breaks Assumption LW. 2 in the same way except that $m p k$ by exchanging the roles of $g_{p_{1}}$ and $g_{p_{3}}$ and thus $p_{1}$ and $p_{3}$.

Lemma 4.2.8 If Assumption LW. 1 holds, then for all probabilistic polynomial time adversaries $\mathcal{A}$, Game $_{\text {Restricted }} \approx_{\mathrm{c}}$ Game $_{0}$.

Proof: We describe algorithm $\mathcal{B}$ that breaks Assumption LW.1. $\mathcal{B}$ receives $\left(N, g_{1}, g_{3}\right)$ and $T$ and simulates Game $_{\text {Restricted }}$ or Game $_{\mathbf{0}}$ with $\mathcal{A}$ depending on $T$.

Setup $: \mathcal{B}$ chooses random $\rho \leftarrow \mathbb{Z}_{N}$ and $\left(u_{i, j} \leftarrow \mathbb{Z}_{N}\right)_{i \in[L], j \in\{0,1\}}$, and set $\Omega=\hat{e}\left(g_{1}, g_{1}\right)^{\rho}$. Then, $\mathcal{B}$ set

$$
m p k \leftarrow\left[N, g_{1}, g_{3},\left(g_{1}^{u_{i, j}}\right)_{i \in[L], j \in\{0,1\}}, \Omega\right]
$$

as the public parameters and

$$
m s k \leftarrow[\rho]
$$

as the master secret key. $\mathcal{B}$ starts the interaction with $\mathcal{A}$ on input $m p k$.
Secret Keys : Notice that $\mathcal{B}$ knows $m s k$ and thus can answer all $\mathcal{A}$ 's queries.
Challenge : $\mathcal{A}$ sends $\mathcal{B}$ challenge messages $\mathfrak{m}_{0}, \mathfrak{m}_{1} \in\{0,1\}^{*}$ and a challenge pattern $P=$ $\left(P_{1}, \ldots, P_{\ell}\right)$ where $0 \leq \ell \leq L . \mathcal{B}$ chooses random $\eta \leftarrow\{0,1\}$ and computes the challenge ciphertext $\mathrm{Ct}^{\star}$ as follows:

$$
\begin{gathered}
C_{1}=T \\
\left(C_{2, i}=T^{u_{i, 0}+u_{i, 1} \cdot P_{i}}\right)_{i \in \overline{\mathrm{~W}}(P)} \\
C_{3}=\hat{e}\left(T, g_{1}\right)^{\rho} \cdot \mathfrak{m} \\
\left(C_{4, i, j}=T^{u_{i, j}}\right)_{i \in \mathrm{~W}(P), j \in\{0,1\}}
\end{gathered}
$$

Notice that if $T \in \mathbb{G}_{p_{1}}$, then $T$ can be written as $g_{1}^{s}$ and $\mathrm{Ct}^{\star}$ is a normal ciphertext with randomness $s$. Instead, if $T \in \mathbb{G}_{p_{1} p_{2}}$, then $T$ can be written as $g_{1}^{s} g_{2}^{x}$ and $\mathrm{Ct}^{\star}$ is semi-functional with randomness $s, x, z_{c, i}=u_{i, 0}+u_{i, 1} \cdot P_{i}, z_{c, i, j}=u_{i, j}$.

Lemma 4.2.9 If Assumption LW. 2 holds, then for all probabilistic polynomial time adversaries $\mathcal{A}$, Game $_{\mathrm{k}-1} \approx_{\mathbf{c}}$ Game $_{\mathrm{k}}$.

Proof: We describe algorithm $\mathcal{B}$ that breaks Assumption LW.2. $\mathcal{B}$ receives $\left(N, g_{1}, g_{3}, X_{1} X_{2}, Y_{2} X_{3}\right)$ and $T$ and simulates $\mathbf{G a m e}_{\mathbf{k}-\mathbf{1}}$ or $\mathbf{G a m e}_{\mathbf{k}}$ with $\mathcal{A}$ depending on $T$.

Setup $: \mathcal{B}$ chooses random $\rho \leftarrow \mathbb{Z}_{N}$ and $\left(u_{i, j} \leftarrow \mathbb{Z}_{N}\right)_{i \in[L], j \in\{0,1\}}$, and set $\Omega=\hat{e}\left(g_{1}, g_{1}\right)^{\rho}$. Then $\mathcal{B}$ set

$$
m p k \leftarrow\left[N, g_{1}, g_{3},\left(g_{1}^{u_{i, j}}\right)_{i \in[L], j \in\{0,1\}}, \Omega\right]
$$

as the public parameters and

$$
m s k \leftarrow[\rho]
$$

as the master secret key. $\mathcal{B}$ starts the interaction with $\mathcal{A}$ on input mpk.

Secret Keys: Let us now explain how $\mathcal{B}$ answers the $i$-th key query for pattern $P=\left(P_{1}, \ldots, P_{\ell}\right)$ where $0 \leq \ell \leq L$.
$i<k: \mathcal{B}$ creates an semi-functional secret key as follows: Choose random $z_{k}^{\prime} \leftarrow \mathbb{Z}_{N}$, $\left(r_{i}, z_{k, i}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{i \in \overline{\mathrm{~W}}(P)}, W_{0} \leftarrow \mathbb{G}_{p_{3}}$ and $\left(W_{i} \leftarrow \mathbb{G}_{p_{3}}\right)_{i \in \overline{\mathrm{~W}}(P)}$. Then set

$$
K_{0}=g_{1}^{\rho} \cdot \prod_{i \in \overline{\mathrm{~W}}(P)}\left(u_{i, 0} \cdot u_{i, 1}^{P_{i}}\right)^{r_{i}} \cdot\left(Y_{2} X_{3}\right)^{z_{k}^{\prime}} \cdot W_{0}
$$

and

$$
\left(D_{i}=g_{1}^{r_{i}} \cdot\left(Y_{2} X_{3}\right)^{z_{k, i}^{\prime}} \cdot W_{i}\right)_{i \in \overline{\mathrm{~W}}(P)}
$$

By writing $Y_{2}$ as $g_{2}^{y}$, we have that this is a properly distributed semi-functional key with randomness $y, z_{k}=z_{k}^{\prime}$ and $z_{k, i}=z_{k, i}^{\prime}$.
$i=k:$ To answer the $k$-th key query $\mathcal{B}$ chooses random $\left(r_{i}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{i \in \overline{\mathrm{~W}}(P)}, W_{0} \leftarrow \mathbb{G}_{p_{3}}$ and $\left(W_{i} \leftarrow \mathbb{G}_{p_{3}}\right)_{i \in \overline{\mathrm{~W}}(P)}$, and sets

$$
K_{0}=g_{1}^{\rho} \cdot \prod_{i \in \overline{\mathrm{~W}}(P)}\left(T^{u_{i, 0}+u_{i, 1} \cdot P_{i}}\right)^{r_{i}^{\prime}} \cdot W_{0}
$$

and

$$
\left(D_{i}=T^{r_{i}^{\prime}} \cdot W_{i}\right)_{i \in \overline{\mathrm{~W}}(P)}
$$

Notice that, if $T \in \mathbb{G}_{p_{1} p_{3}}$, then $T$ can be written as $g_{1}^{r} g_{3}^{w}$ and the $k$-th secret key is a normal key with randomness $r_{i}=r \cdot r_{i}^{\prime}$, If $T \in \mathbb{G}_{p_{1} p_{2} p_{3}}$, then $T$ can be written as $g_{1}^{r} g_{2}^{y} g_{3}^{w}$ and the $k$-th secret key is semi-functional with randomness $r_{i}=r \cdot r_{i}^{\prime}, y$, $z_{k}=\sum_{i \in \overline{\mathrm{~W}}(P)} r_{i}^{\prime} \cdot\left(u_{i, 0}+u_{i, 1} \cdot P_{i}\right)$ and $z_{k, i}=r_{i}^{\prime}$.
$i>k: \mathcal{B}$ runs KeyGen using msk.
Challenge : $\mathcal{A}$ sends $\mathcal{B}$ challenge messages $\mathfrak{m}_{0}, \mathfrak{m}_{1} \in\{0,1\}^{*}$ and a challenge pattern $P=$ $\left(P_{1}, \ldots, P_{\ell}\right)$ where $0 \leq \ell \leq L . \mathcal{B}$ chooses random $\eta \leftarrow\{0,1\}$ and computes the challenge ciphertext $\mathrm{Ct}^{\star}$ as follows:

$$
\begin{gathered}
C_{1}=X_{1} X_{2} \\
\left(C_{2, i}=\left(X_{1} X_{2}\right)^{u_{i, 0}+u_{i, 1} \cdot P_{i}}\right)_{i \in \overline{\mathrm{~W}}(P)} \\
C_{3}=\hat{e}\left(X_{1} X_{2}, g_{1}\right)^{\rho} \cdot \mathfrak{m} \\
\left(C_{4, i, j}=\left(X_{1} X_{2}\right)^{u_{i, j}}\right)_{i \in \mathrm{~W}(P), j \in\{0,1\}}
\end{gathered}
$$

Notice that $X_{1} X_{2}$ can be written as $g_{1}^{s} g_{2}^{x}$ for some random $s, x \in \mathbb{Z}_{N}$, thus the challenge ciphertext is semi-functional with randomness $s, x, z_{c, i}=u_{i, 0}+u_{i, 1} \cdot P_{i}, z_{c, i, j}=u_{i, j}$.
Since the $k$-th secret key pattern is not a prefix of the challenge pattern modulo $p_{2}$ we have that $z_{k}$ and $z_{c}$ are independent and randomly distributed. We observe that, if $\mathcal{B}$ attempts to test whether the $k$-th key is semi-functional by using the above procedure to create an semi-functional ciphertext for $k$-th secret key pattern then we will have that $z_{k}=z_{c}$ and thus decryption always works (independently of $T$ ).

Lemma 4.2.10 If Assumption LW. 3 holds, then for all probabilistic polynomial time adversaries $\mathcal{A}$, Game $_{\mathbf{q}} \approx_{\mathbf{c}}$ Game $_{\text {Final }}$.

Proof: We describe algorithm $\mathcal{B}$ that breaks Assumption LW.3. $\mathcal{B}$ receives $\left(N, g_{1}, g_{2}, g_{3}, g_{1}^{\rho} X_{2}, g_{1}^{s} Y_{2}\right)$ and $T$ and simulates Game $_{\mathbf{q}}$ or Game Final with $\mathcal{A}$ depending on $T$.

Setup : $\mathcal{B}$ chooses random $\left(u_{i, j} \leftarrow \mathbb{Z}_{N}\right)_{i \in[L], j \in\{0,1\}}$, and set $\Omega=\hat{e}\left(g_{1}^{\rho} X_{2}, g_{1}\right)$. Then $\mathcal{B}$ set

$$
m p k \leftarrow\left[N, g_{1}, g_{3},\left(g_{1}^{u_{i, j}}\right)_{i \in[L], j \in\{0,1\}}, \Omega\right]
$$

as the public parameters. $\mathcal{B}$ starts the interaction with $\mathcal{A}$ on input mpk.
Secret Keys : For pattern $P=\left(P_{1}, \ldots, P_{\ell}\right)$ where $0 \leq \ell \leq L, \mathcal{B}$ creates a semi-functional secret key as follows: Choose random $z_{k}^{\prime} \leftarrow \mathbb{Z}_{N},\left(r_{i}, z_{k, i}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{i \in \bar{W}(P)}, W_{0} \leftarrow \mathbb{G}_{p_{3}}$ and $\left(W_{i} \leftarrow \mathbb{G}_{p_{3}}\right)_{i \in \overline{\mathrm{~W}}(P)}$. Then set

$$
K_{0}=\left(g_{1}^{\rho} X_{2}\right) \cdot \prod_{i \in \overline{\bar{W}}(P)} g_{1}^{r_{i} \cdot\left(u_{i, 0}+u_{i, 1} \cdot P_{i}\right)} \cdot g_{2}^{z_{k}^{\prime}} \cdot W_{0}
$$

and

$$
\left(D_{i}=g_{1}^{r_{i}} \cdot g_{2}^{z_{k, i}^{\prime}} \cdot W_{i}\right)_{i \in \overline{\mathrm{~W}}(P)}
$$

Challenge : $\mathcal{A}$ sends $\mathcal{B}$ challenge messages $\mathfrak{m}_{0}, \mathfrak{m}_{1} \in\{0,1\}^{*}$ and a challenge pattern $P=$ $\left(P_{1}, \ldots, P_{\ell}\right)$ where $0 \leq \ell \leq L . \mathcal{B}$ chooses random $\eta \leftarrow\{0,1\}$ and computes the challenge ciphertext $C$ as follows:

$$
\begin{gathered}
C_{1}=\left(g_{1}^{s} Y_{2}\right) \\
\left(C_{2, i}=\left(g_{1}^{s} Y_{2}\right)^{u_{i, 0}+u_{i, 1} \cdot P_{i}}\right)_{i \in \overline{\mathrm{~W}}(P)} \\
C_{3}=T \cdot \mathfrak{m} \\
\left(C_{4, i, j}=\left(g_{1}^{s} Y_{2}\right)^{u_{i, j}}\right)_{i \in \mathrm{~W}(P), j \in\{0,1\}}
\end{gathered}
$$

This sets $z_{c, i}=u_{i, 0}+u_{i, 1} \cdot P_{i}$ and $z_{c, i, j}=u_{i, j}$. We note that $g_{1}^{u_{i, j}}$ are elements of $\mathbb{G}_{p_{1}}$, so when $u_{i, j}$ are randomly chosen from $\mathbb{Z}_{N}$, their value modulo $p_{1}$ and modulo $p_{2}$ are random and independent.

We finish by observing that, if $T=\hat{e}\left(g_{1}, g_{1}\right)^{\rho s}$, then the challenge ciphertext is a properly distributed semi-functional with message $\mathfrak{m}_{\eta}$. If $T \leftarrow \mathbb{G}_{\mathrm{T}}$, then the ciphertext is an semifunctional ciphertext with a random message.

Wrapping up. We have proved by the previous lemmata that the real security game is indistinguishable from Game Final , where $\eta$ is information-theoretically hidden from the attacker. Thus under assumptions LW.1-3, Theorem 4.2.4 holds.

### 4.3 WKD-IPE and Its application to Anonymous WW-IBE

In this section, we first introduce inner-product encryption with generalized key delegation (WKD-IPE, for short) and show how to generically reduce Anonymous WW-IBE to it in a way that preserves full security. The reduction is based on an observation by [KSW08] that zero entries can be used to simulate wildcards. The generalized key delegation capability of the WKD-IPE scheme will provide the final ingredient of the reduction. Finally, we give a new WKD-IPE construction based on a slight modification of the Okamoto-Takashima hierarchical inner-product predicate encryption scheme $\left[\mathrm{LOS}^{+} 10\right]$, which allows for more general key delegation patterns.

### 4.3.1 Definition of Inner-Product Encryption with Generalized Key Delegation

In this section, following [OT09, $\left.\mathrm{LOS}^{+} 10\right]$, we define the WKD-IPE primitive and its security. We start by extending Definition 2.1.2 of hierarchical format to more general patterns. The main difference with Definition 2.1.2 is that we do not require any more that each $\Sigma_{i}$ does not contain the all zero vector. Specifically:

Definition 4.3.1 [Generalized Hierarchical Format] Let $\boldsymbol{\mu}$ be a tuple of positive integers $\boldsymbol{\mu}=$ $\left(n, d ; \mu_{1}, \ldots, \mu_{d}\right)$ such that $\mu_{0}=0<\mu_{1}<\cdots<\mu_{d}=n$. For $i=1, \ldots, d$, let $\Sigma_{i}=\mathbb{F}_{p}^{\mu_{i}-\mu_{i-1}}$ the set of attributes. Let $\Sigma$ be the hierarchical attributes $\Sigma=\cup_{i=1}^{d}\left(\Sigma_{1} \times \ldots \times \Sigma_{i}\right) \backslash\left\{\mathbf{0}_{n}\right\}$, where the union is a disjoint union. Them, we call $\boldsymbol{\mu}$ an generalized hierarchical format of depth $d$ for the attribute space $\Sigma$.

Delegation. Given $\boldsymbol{y}, \boldsymbol{y}^{\prime} \in \Sigma$, we say that $\boldsymbol{y}^{\prime}$ is a delegation of $\boldsymbol{y}$, in symbols $\boldsymbol{y}^{\prime} \prec \boldsymbol{y}$, if and only if for each $i \in[d]$ we have $\boldsymbol{y}_{i}=\mathbf{0}$ or $\boldsymbol{y}_{i}=\boldsymbol{y}^{\prime}{ }_{i}$. Again as opposed to [OT09, LOS ${ }^{+}$10] any $\boldsymbol{y}_{i}=\mathbf{0}$ can be replaced by a non-zero vector, for any $i \in[d]$.

Definition 4.3.2 [WKD-IPE] A inner-product encryption with generalized key delegation with hierarchical format $\boldsymbol{\mu}$ is defined by the following tuple of algorithms: WKD - IPE $=$ (Setup, Enc, KeyGen, Dec, KeyDel) with the following syntax:
$\operatorname{Setup}\left(1^{\lambda}, \boldsymbol{\mu}\right)$ takes as input security parameter $\lambda$ and generalized hierarchical format $\boldsymbol{\mu}$ and outputs public parameters $m p k$ and master secret key msk.
$\operatorname{KeyGen}\left(m s k, \boldsymbol{y}=\left(\boldsymbol{y}_{1}, \ldots, \boldsymbol{y}_{d}\right) \in \Sigma\right)$ takes as input $m s k$, an attribute vector $\boldsymbol{y}$ following hierarchy $\boldsymbol{\mu}$ and outputs a private key $s k_{\boldsymbol{y}}$.
$\operatorname{KeyDel}\left(m p k, s k_{\boldsymbol{y}}, \boldsymbol{y}^{\prime} \in \Sigma\right)$ takes as input $m p k$, secret key for attribute vectors $\boldsymbol{y}=\left(\boldsymbol{y}_{1}, \ldots, \boldsymbol{y}_{d}\right)$ and $\boldsymbol{y}^{\prime}=\left(\boldsymbol{y}^{\prime}{ }_{1}, \ldots, \boldsymbol{y}^{\prime}{ }_{d}\right)$ such that $\boldsymbol{y}^{\prime} \prec \boldsymbol{y}$, and outputs a secret key for attribute vector $\left(\boldsymbol{y}^{\prime}{ }_{1}, \ldots, \boldsymbol{y}^{\prime}{ }_{d}\right)$.
$\operatorname{Enc}(m p k, \mathfrak{m}, \boldsymbol{x})$ takes as input $m p k, \mathfrak{m}$ in some associated message space and an attribute vector $\boldsymbol{x}$, and outputs a ciphertext Ct .
$\operatorname{Dec}\left(m p k, \mathrm{Ct}, s k_{\boldsymbol{y}}\right)$ takes as input $m p k$, ciphertext Ct and secret key $s k_{\boldsymbol{y}}$ and outputs the message $\mathfrak{m}$.

Correctness. We make the following consistency requirement. Suppose ciphertext Ct is obtained by running Enc on input $m p k$, message $\mathfrak{m}$ and attribute vector $\boldsymbol{x}$ and that $s k_{\boldsymbol{y}}$ is a secret key for attribute vector $\boldsymbol{y}$ obtained through a sequence of KeyGen and KeyDel calls using the same $m p k$. Then Dec, on input $m p k, \mathrm{Ct}$ and $s k_{\boldsymbol{y}}$, returns $\mathfrak{m}$, except with negligible probability, if and only if $f_{\boldsymbol{y}}(\boldsymbol{x})=1$.

### 4.3.2 Security Definition

We give complete form of the security definition following [SW08]. Our security definition captures semantic security and ciphertext anonymity by means of the following game between an adversary $\mathcal{A}$ and a challenger $\mathcal{C}$. More specifically, game $\operatorname{ANO}^{-C P A} \mathcal{A}_{\mathcal{A}}^{W K D-I P E}\left(1^{\lambda}\right)$, where $\lambda$ is the security parameter, is defined as follows.

Setup: $\mathcal{C}$ generates master public key and secret key by invoking the setup algorithm on input the security parameter $\lambda$ and the generalized hierarchical format $\boldsymbol{\mu}$. Specifically, $(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}, \boldsymbol{\mu}\right)$. Then, $\mathcal{C}$ starts the interaction with $\mathcal{A}$ on input $m p k$.
We let $S$ denote the set of private keys that $\mathcal{C}$ has created but not yet given to the adversary. At this point, $S=\emptyset$.

Query Phase 1: $\mathcal{A}$ makes Create, Delegate, and Reveal key queries. Specifically:
To make a Create query, $\mathcal{A}$ specifies an attribute vector $\boldsymbol{y}$. In response, $\mathcal{C}$ creates a key for this vector by calling the key generation algorithm, and places this key in the set $S . \mathcal{C}$ only gives $\mathcal{A}$ a reference to this key, not the key itself.
To make a Delegate query, $\mathcal{A}$ specifies a key $s k_{\boldsymbol{y}}$ in the set $S$ and $\boldsymbol{y}^{\prime}$. In response, $\mathcal{C}$ makes a key for this new vector by running the delegation algorithm on $s k_{\boldsymbol{y}}$ and $\boldsymbol{y}^{\prime}$. $\mathcal{C}$ adds this key to the set $S$ and again gives $\mathcal{A}$ only a reference to it, not the actual key.
To make a Reveal query, $\mathcal{A}$ specifies an element of the set $S$. $\mathcal{C}$ gives this key to $\mathcal{A}$ and removes it from the set $S$. Notice that $\mathcal{A}$ needs no longer make any delegation queries for this key because it can run delegation algorithm on the revealed key for itself

Challenge: $\mathcal{A}$ gives two pairs of message and identity $\left(\mathfrak{m}_{0}, \boldsymbol{y}_{0}\right)$ and $\left(\mathfrak{m}_{1}, \boldsymbol{y}_{1}\right)$ to $\mathcal{C}$. Then $\mathcal{C}$ chooses random $\eta \leftarrow\{0,1\}$, encrypts $\mathfrak{m}_{\eta}$ under $\boldsymbol{y}_{\eta}$ and sends the resulting ciphertext to $\mathcal{A}$.

Query Phase 2: This is the same as Phase 1.
Guess: Eventually, $\mathcal{A}$ submits its guess $\eta^{\prime}$.
Winning Condition: $\mathcal{A}$ wins the game if the following conditions are satisfied:

1. $\eta=\eta^{\prime}$.
2. No secret key for any vector $\boldsymbol{w}$ obtained through delegation from any of the revealed secret key is such that $f_{\boldsymbol{w}}\left(\boldsymbol{x}_{0}\right) \neq f_{\boldsymbol{w}}\left(\boldsymbol{x}_{1}\right)$.

The advantage of $\mathcal{A}$ in the above game is defined as

$$
\mathbf{A d v}_{\mathrm{ANO}-\mathrm{CPA}}^{W K D-I P E, \mathcal{A}}(\lambda)=\operatorname{Prob}\left[\mathrm{ANO}-\mathrm{CPA}_{\mathcal{A}}^{W K D-I P E}\left(1^{\lambda}\right)=1\right]-1 / 2
$$

Definition 4.3.3 A inner-product encryption with generalized key delegation scheme is ANO-CPA-secure if any polynomial-time adversaries making at most a polynomial number of queries to the key derivation oracle only has a negligible advantage in the ANO-CPA game described above, meaning that $\mathbf{A d v}_{\text {ANO-CPA }}^{W K D-I P E, \mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

In the Section 4.4, we will describe a construction of inner-product encryption with generalized key delegation that achieves a weaker security definition called wAH-ANO-CPA and defined in Section 2.2.1. Specifically, game wAH-ANO-CPA is like ANO-CPA but with a relaxed winning condition. Specifically, wAH-ANO-CPA only requires that $f_{\boldsymbol{w}}\left(\boldsymbol{x}_{0}\right)=f_{\boldsymbol{w}}\left(\boldsymbol{x}_{1}\right)=0$ for any vector $\boldsymbol{w}$ for which the relative secret key has been obtained through delegation from one of the revealed secret key.

Definition 4.3.4 A inner-product encryption with generalized key delegation scheme is wAH-ANO-CPA-secure if any polynomial-time adversaries making at most a polynomial number of queries to the key derivation oracle only has a negligible advantage in the wAH-ANO-CPA game, meaning that $\mathbf{A d v} \mathbf{w A H - A N O - C P A}_{W K D-I P E, \mathcal{A}}^{W}(\lambda)$ is a negligible function of $\lambda$.

### 4.3.3 Reducing Anonymous WW-IBE to WKD-IPE

Let WKD - IPE $=\left(\right.$ Setup $_{W K D}$, KeyGen $_{\text {WKD }}$, KeyDel $_{\text {WKD }}$, EncwKD, DecwKd $)$ be a Innerproduct encryption with generalized key delegation. We can construct the scheme WW - IBE = (Setup $_{w w}$, KeyGen $_{w w}$, KeyDel $_{w w}$, Encww $_{w}$, Dec $_{w w}$ ) as follows:
$\operatorname{Setup}_{\mathrm{ww}}\left(1^{\lambda}, 1^{L}\right)$ : The algorithm returns the output of $\operatorname{Setup}_{\mathrm{WKD}}\left(1^{\lambda}, \boldsymbol{\mu}=\left(2 L, L ;\left(\mu_{i}=\right.\right.\right.$ $\left.2 i)_{i \in[L]}\right)$ ). So the hierarchy $\boldsymbol{\mu}$ is of depth $L$ and each level has dimension 2 .
KeyGen ${ }_{\mathrm{WW}}(m s k, P)$ : For a pattern $P=\left(P_{1}, \ldots, P_{\ell}\right)$, the key generation algorithm constructs vector $\boldsymbol{y} \in \Sigma$ by setting, for each $i \in[\ell]$,

$$
\boldsymbol{y}_{i}=\left\{\begin{array}{ll}
\left(1, P_{i}\right) & \text { if } P_{i} \neq \star \\
(0,0) & \text { otherwise }
\end{array} .\right.
$$

We denote this transformation by

$$
\boldsymbol{y}=\operatorname{KEncode}(P) .
$$

Then the key generation algorithm returns

$$
s k_{P}=\operatorname{KeyGen}_{\mathrm{WKD}}(m s k, \boldsymbol{y}) .
$$

$\operatorname{KeyDel}_{\mathrm{WW}}\left(m p k, s k_{P}, P^{\prime}\right)$ : The algorithm returns the output of

$$
s k_{P^{\prime}}=\operatorname{KeyDel}_{\mathrm{WKD}}\left(m p k, s k_{P}, \operatorname{KEncode}\left(P^{\prime}\right)\right) .
$$

Encww $(m p k, P)$ : The algorithm constructs vector $\boldsymbol{x} \in \Sigma$ in the following way: For each $i \in[\ell]$ the algorithms sets

$$
\boldsymbol{x}_{i}=\left\{\begin{array}{ll}
\left(-r_{i} \cdot P_{i}, r_{i}\right) & \text { if } P_{i} \neq \star \\
(0,0) & \text { otherwise }
\end{array} .\right.
$$

We denote this transformation by

$$
\boldsymbol{x}=\mathrm{CEncode}(P) .
$$

Then the encryption algorithm returns

$$
\mathrm{Ct}=\operatorname{Enc}_{\mathrm{WKD}}(m p k, \boldsymbol{x}) .
$$

$\operatorname{Dec}_{\mathrm{Ww}}\left(s k_{P^{\prime}}, \mathrm{Ct}\right)$ : Returns $\operatorname{Dec}_{\mathrm{WKD}}\left(s k_{P^{\prime}}, \mathrm{Ct}\right)$.
Correctness. It follows from the observation that for patterns $P$ and $P^{\prime}$, we have that $f_{\mathrm{KEncode}\left(P^{\prime}\right)}(\operatorname{CEncode}(P))=1$ if and only if $P \in_{\star} P^{\prime}$.

Theorem 4.3.5 If WKD - IPE is ANO-CPA secure then WW - IBE is so.

Proof: Let $\mathcal{A}$ be an adversary that breaks the ANO-CPA security of the WW - IBE scheme for an hierarchy of depth $\ell$ and consider the following adversary $\mathcal{B}$ that tries to break the ANO-CPA security of the WKD - IPE scheme for hierarchy format $\boldsymbol{\mu}=\left(2 \ell, \ell ;\left(\mu_{i}=2 i\right)_{i \in[\ell]}\right)$ that uses $\mathcal{A}$ as a subroutine and interact with the challenger $\mathcal{C}$.

Then, $\mathcal{B}$ receives from $\mathcal{C}$ a master public key $m p k$ for the WKD - IPE scheme and passes it to $\mathcal{A}$. Whenever $\mathcal{A}$ asks for the key for pattern $P, \mathcal{B}$ constructs $\boldsymbol{y}=\operatorname{KEncode}(P)$ and asks $\mathcal{C}$ for a key $s k_{\boldsymbol{y}}$ for $\boldsymbol{y}$ and returns it to $\mathcal{A}$. When $\mathcal{A}$ asks for a challenge ciphertext by providing $\left(\mathfrak{m}_{0}, P_{0}^{\star}\right)$ and $\left(\mathfrak{m}_{1}, P_{1}^{\star}\right), \mathcal{B}$ simply computes $\boldsymbol{x}_{0}=\mathrm{CEncode}\left(P_{0}^{\star}\right)$ and $\boldsymbol{x}_{1}=\mathrm{CEncode}\left(P_{1}^{\star}\right)$ and gives the pair $\left(\mathfrak{m}_{0}, \boldsymbol{x}_{0}\right),\left(\mathfrak{m}_{1}, \boldsymbol{x}_{1}\right)$ to $\mathcal{C}$. $\mathcal{B}$ then forwards the challenge ciphertext to $\mathcal{A}$. Finally, $\mathcal{B}$ outputs $\mathcal{A}$ 's guess.
First, $\mathcal{B}$ 's simulation is perfect. Indeed, we have that if for all $\mathcal{A}$ 's queries satisfy the game constraints, then all $\mathcal{B}$ 's queries have the same property. Thus $\mathcal{B}$ 's advantage is the same as $\mathcal{A}$ 's.

## I

### 4.4 Constructing WKD-IPE

In this section, we show our new WKD-IPE construction and prove its security. Our construction is based on a slight modification of the Okamoto-Takashima hierarchical predicate encryption scheme for the class of the inner-product predicates [ $\left.\mathrm{LOS}^{+} 10\right]$. As the HPE scheme in [ $\left.\mathrm{LOS}^{+} 10\right]$, our scheme will be proved to be only weakly attribute-hiding secure.

### 4.4.1 Complexity Assumptions

For completeness we restate here the complexity assumptions used, first introduced by Okamoto and Takashima in [ $\left.\mathrm{LOS}^{+} 10\right]$. All these assumptions can be seen as variants of the decisional subspace assumption.

Assumption OT. 1 For a given prime-order dual-pairing vector spaces generator DPVSGen, define the following distribution: Pick a random bilinear setting $\left(\right.$ param $\left._{\mathbb{V}}, \mathbb{B}, \mathbb{B}^{\star}\right) \leftarrow \operatorname{DPVSGen}\left(1^{\lambda}\right.$, $2 n+3)$, and then pick

$$
\delta \leftarrow \mathbb{Z}_{N}, \boldsymbol{\delta} \leftarrow \mathbb{Z}_{N}^{n}, \rho \leftarrow \mathbb{Z}_{N}, \boldsymbol{U}=\left(\boldsymbol{u}_{i}\right)_{i \in[n]} \leftarrow \mathrm{GL}\left(n, \mathbb{Z}_{N}\right)
$$

and set

$$
\begin{gathered}
\hat{\mathbb{B}}=\left(\boldsymbol{b}_{1}, \ldots, \boldsymbol{b}_{n}, \boldsymbol{b}_{2 n+1}, \boldsymbol{b}_{2 n+3}\right), \quad \hat{\mathbb{B}}^{\star}=\left(\boldsymbol{b}_{1}^{\star}, \ldots, \boldsymbol{b}_{n}^{\star}, \boldsymbol{b}_{2 n+1}^{\star}, \boldsymbol{b}_{2 n+2}^{\star}\right) \\
\boldsymbol{e}_{0}=\left(\delta \cdot \boldsymbol{b}_{i}+\delta_{i} \cdot \boldsymbol{b}_{2 n+3}\right)_{i \in[n]}, \quad \boldsymbol{e}_{1}=\left(\delta \cdot \boldsymbol{b}_{i}+\rho \cdot \sum_{j \in[n]} u_{i, j} \cdot \boldsymbol{b}_{n+j}+\delta_{i} \cdot \boldsymbol{b}_{2 n+3}\right)_{i \in[n]}
\end{gathered}
$$

and $\mathcal{D}=\left(\right.$ param $\left._{\mathbb{V}}, \hat{\mathbb{B}}, \hat{\mathbb{B}}^{\star}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking Assumption OT. 1 is

$$
\mathbf{A d v}_{O T .1}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, \boldsymbol{e}_{0}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, \boldsymbol{e}_{1}\right)=1\right]\right|
$$

Definition 4.4.1 We say that Assumption OT. 1 holds for generator DPVSGen if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \mathbf{A d} \mathbf{v}_{O T .1}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

Assumption OT. 2 For a given prime-order dual-pairing vector spaces generator DPVSGen, define the following distribution: Pick a random bilinear setting $\left(\right.$ param $\left._{\mathbb{V}}, \mathbb{B}, \mathbb{B}^{\star}\right) \leftarrow \operatorname{DPVSGen}\left(1^{\lambda}\right.$, $2 n+3$ ), and then pick

$$
\omega, \gamma, \rho, \tau \leftarrow \mathbb{Z}_{N}, \gamma \leftarrow \mathbb{Z}_{N}^{n}, \boldsymbol{U}=\left(\boldsymbol{u}_{i}\right)_{i \in[n]} \leftarrow \mathrm{GL}\left(n, \mathbb{Z}_{N}\right)
$$

and set

$$
\hat{\mathbb{B}}=\left(\boldsymbol{b}_{1}, \ldots, \boldsymbol{b}_{n}, \boldsymbol{b}_{2 n+1}, \boldsymbol{b}_{2 n+3}\right), \quad \hat{\mathbb{B}}^{\star}=\left(\boldsymbol{b}_{1}^{\star}, \ldots, \boldsymbol{b}_{2 n+2}^{\star}\right),
$$

$$
\begin{gathered}
\boldsymbol{Z}=\left(\boldsymbol{z}_{i}\right)_{i \in[n]}=\left(\boldsymbol{U}^{\top}\right)^{-1}, \quad \boldsymbol{e}=\left(\delta \cdot \boldsymbol{b}_{i}+\rho \cdot \sum_{j \in[n]} u_{i, j} \cdot \boldsymbol{b}_{n+j}\right)_{i \in[n]} \\
\boldsymbol{h}_{0}=\left(\omega \cdot \boldsymbol{b}_{i}^{\star}+\gamma_{i} \cdot \boldsymbol{b}_{2 n+2}^{\star}\right)_{i \in[n]}, \quad \boldsymbol{h}_{1}=\left(\omega \cdot \boldsymbol{b}_{i}^{\star}+\tau \cdot \sum_{j \in[n]} z_{i, j} \cdot \boldsymbol{b}_{n+j}^{\star}+\gamma_{i} \cdot \boldsymbol{b}_{2 n+2}\right)_{i \in[n]}
\end{gathered}
$$

and $\mathcal{D}=\left(\operatorname{param}_{\mathbb{V}}, \hat{\mathbb{B}}, \hat{\mathbb{B}}^{\star}, \boldsymbol{e}\right)$.
Then, the advantage of a PPT adversary $\mathcal{A}$ in breaking Assumption OT. 2 is

$$
\mathbf{A d v}_{O T .2}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, \boldsymbol{h}_{0}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, \boldsymbol{h}_{1}\right)=1\right]\right|
$$

Definition 4.4.2 We say that Assumption OT. 2 holds for generator DPVSGen if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \mathbf{A d v} \mathbf{v}_{O T .2}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

Remark 4.4.3 In [LOS $\left.{ }^{+} 10\right]$, Okamoto and Takashima proved that Assumptions OT. 1 and OT. 2 can be reduced to the $n$-Extended Decisional Diffie-Hellman Assumption. More recently in [OT10] Okamoto and Takashima proved that these kind of decisional subspace problems can be reduced to the Decision Linear Assumption [ $\overline{\mathrm{BBS} 04}$ ].

### 4.4.2 Our Construction

The algorithms of our WKD - IPE are defined as follows:
$\operatorname{Setup}\left(1^{\lambda}, \mu=\left(n, d ; \mu_{1}, \ldots, \mu_{d}\right)\right)$ : Choose a random DPVS instance

$$
\left(\operatorname{param}_{\mathbb{V}}, \mathbb{B}, \mathbb{B}^{\star}\right) \leftarrow \operatorname{DPVSGen}\left(1^{\lambda}, 2 n+3\right)
$$

Then set $\hat{\mathbb{B}}=\left(\boldsymbol{b}_{1}, \ldots, \boldsymbol{b}_{n}, \boldsymbol{b}_{2 n+1}, \boldsymbol{b}_{2 n+3}\right), \Omega=\hat{e}(g, g)$. The master public key is defined as

$$
m p k \leftarrow\left[\operatorname{param}_{\mathbb{V}}, \hat{\mathbb{B}}, \Omega\right]
$$

and the master secret key

$$
m s k \leftarrow\left[\mathbb{B}^{\star}\right]
$$

$\operatorname{KeyGen}\left(m s k, \boldsymbol{y}=\left(\boldsymbol{y}_{1}, \ldots, \boldsymbol{y}_{d}\right)\right)$ : Given $\boldsymbol{y}$ according to hierarchy $\mu$, generate the following sub-keys:

Decryption Key : First apply a random translation to $\boldsymbol{y}$ choosing $\left(r_{i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[d]}$ so that

$$
\hat{\boldsymbol{y}}=\left(\hat{\boldsymbol{y}}_{1}, \ldots, \hat{\boldsymbol{y}}_{d}\right)=\left(r_{1} \cdot \boldsymbol{y}_{1}, \ldots, r_{d} \cdot \boldsymbol{y}_{d}\right)
$$

Then choose $r \leftarrow \mathbb{Z}_{N}$ and computes the decryption key:

$$
s k^{d e c}=\left(\hat{\boldsymbol{y}}_{1}, \ldots, \hat{\boldsymbol{y}}_{d}, \mathbf{0}_{n}, 1, r, 0\right)_{\mathbb{B}^{\star}}
$$

Randomization Keys : Let $S_{\boldsymbol{y}}=\left\{k \in[d]: \boldsymbol{y}_{k} \neq \mathbf{0}\right\}$ and $T_{\boldsymbol{y}}=S_{\boldsymbol{y}} \bigcup\{d+1\}$. For each $k \in T_{\boldsymbol{y}}$ apply a random translation to $\boldsymbol{y}$ choosing $\left(s_{k, i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[d]}$ so that

$$
\hat{\boldsymbol{y}}_{k}^{r a n}=\left(\hat{\boldsymbol{y}}_{k, 1}^{r a n}, \ldots, \hat{\boldsymbol{y}}_{k, d}^{r a n}\right)=\left(s_{k, 1} \cdot \boldsymbol{y}_{1}, \ldots, s_{k, d} \cdot \boldsymbol{y}_{d}\right) .
$$

Then choose $s_{k} \leftarrow \mathbb{Z}_{N}$ and compute the randomization keys:

$$
s k_{k}^{r a n}=\left(\hat{\boldsymbol{y}}_{k, 1}^{r a n}, \ldots, \hat{\boldsymbol{y}}_{k, d}^{r a n}, \mathbf{o}_{n}, 0, s_{k}, 0\right)_{\mathbb{B}^{\star}} .
$$

Delegation Keys : Let $D_{\boldsymbol{y}}=\left\{j \in[n]: \exists k \in[d]\right.$ s.t. $\boldsymbol{y}_{k}=\mathbf{0}$ and $\left.\mu_{k-1}+1 \leq j \leq \mu_{k}\right\}$. Choose $t \leftarrow \mathbb{Z}_{N}$. Then for each $j \in D_{y}$, let $k$ be the index such that $\boldsymbol{y}_{k}=\mathbf{0}$ and $\mu_{k-1}+1 \leq j \leq \mu_{k}$, apply a random translation to $\boldsymbol{y}$ choosing $\left(t_{j, i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[d]}$ so that

$$
\hat{\boldsymbol{y}}_{j}^{\text {del }}=\left(\hat{\boldsymbol{y}}_{j, 1}^{d e l}, \ldots, \hat{\boldsymbol{y}}_{j, d}^{d e l}\right)=\left(t_{k, 1} \cdot \boldsymbol{y}_{j, 1}^{\text {del }}, \ldots, t_{k, d} \cdot \boldsymbol{y}_{j, d}^{d e l}\right)
$$

where $\hat{\boldsymbol{y}}_{j, i}^{d e l}=t \cdot \boldsymbol{e}_{j}$ if $j=i_{k}$, otherwise $\hat{\boldsymbol{y}}_{j, i}^{\text {del }}=t_{j, i} \cdot \boldsymbol{y}_{i}$ where $\boldsymbol{e}_{j}$ is the zero vector with 1 in the $j$-th coordinate.
Then choose $t_{j} \leftarrow \mathbb{Z}_{N}$ and compute the delegation keys:

$$
s k_{j}^{d e l}=\left(\hat{\boldsymbol{y}}_{j, 1}^{d e l}, \ldots, \hat{\boldsymbol{y}}_{j, d}^{d e l}, \mathbf{0}_{n}, 0, t_{j}, 0\right)_{\mathbb{B}^{\star}}
$$

The secret key for $\boldsymbol{y}$ is $s k_{\boldsymbol{y}} \leftarrow\left[s k^{\text {dec }},\left(s k_{k}^{r a n}\right)_{k \in T_{\boldsymbol{y}}},\left(s k_{j}^{d e l}\right)_{j \in D_{\boldsymbol{y}}}\right]$. Notice that $s k_{\boldsymbol{y}}$ contains at most $n+1$ sub-keys.
$\operatorname{Key} \operatorname{Del}\left(m s k, d_{\boldsymbol{y}}, \boldsymbol{y}^{\prime}=\left(\boldsymbol{y}_{1}^{\prime}, \ldots, \boldsymbol{y}_{d}^{\prime}\right)\right)$ : Given $\boldsymbol{y}^{\prime}$ according to the hierarchy $\mu$, Let $\overline{S_{\boldsymbol{y}}}=\{i \in$ $\left.[d]: \boldsymbol{y}_{i}=\mathbf{0}\right\}$ and let $E_{k}=\left\{i \in[n]: \mu_{k-1}+1 \leq i \leq \mu_{k}\right\}$. Then the key delegation algorithm does the following:

Decryption Key : Chooses $r^{\prime} \leftarrow \mathbb{Z}_{N},\left(r_{j}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{j \in[d]}$ and computes

$$
s k^{\prime d e c}=s k^{d e c}+\sum_{j \in T_{y}} r_{j}^{\prime} \cdot s k_{j}^{r a n}+\sum_{j \in \overline{S_{y}}}\left[r_{j}^{\prime} \cdot \sum_{i \in E_{j}} y_{i}^{\prime} \cdot s k_{i}^{d e l}\right] .
$$

Randomization Keys : For each $k \in T_{\boldsymbol{y}^{\prime}}$ chooses $s_{k}^{\prime} \leftarrow \mathbb{Z}_{N}$ and $\left(s_{k, j}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{j \in[d]}$. Then computes

$$
s k_{k}^{\text {lran }}=\sum_{j \in T_{y}} s_{k, j}^{\prime} \cdot s k_{j}^{r a n}+\sum_{j \in \overline{S_{y}}}\left[s_{k, j}^{\prime} \cdot \sum_{i \in E_{j}} y_{i}^{\prime} \cdot s k_{i}^{d e l}\right] .
$$

Delegation Keys : Chooses $t^{\prime} \leftarrow \mathbb{Z}_{N}$ and for each $j \in D_{y^{\prime}}$, chooses $\left(t_{j, i}^{\prime} \leftarrow \mathbb{Z}_{N}\right)_{i \in[n]}$, $t_{j}^{\prime}, \leftarrow \mathbb{Z}_{N}$. Then computes

$$
s k_{j}^{\prime d e l}=\sum_{i \in T_{y}} t_{j, i}^{\prime} \cdot s k_{i}^{r a n}+\sum_{i \in \overline{S_{y}}}\left[t_{j, i}^{\prime} \cdot \sum_{k \in E_{j}} y_{k}^{\prime} \cdot s k_{k}^{\text {del }}\right]+t^{\prime} \cdot s k_{j}^{\text {del }}
$$

Thus $s k_{\boldsymbol{y}^{\prime}}=\left[s k^{\prime \text { dec }},\left(s k_{k}^{\prime \text { ran }}\right)_{k \in T_{y^{\prime}}},\left(s k_{j}^{\prime \text { del }}\right)_{j \in D_{y^{\prime}}}\right]$.
$\operatorname{Enc}\left(m p k, \boldsymbol{x}=\left(\boldsymbol{x}_{1}, \ldots, \boldsymbol{x}_{\ell}\right), \mathfrak{m}\right):$ Choose $s \leftarrow \mathbb{Z}_{N},\left(s_{i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[d]}$ and pads $\boldsymbol{x}$ to the maximum length $d$ with random vectors following the hierarchy $\mu$. Then apply a random translation to $\boldsymbol{x}$ so that $\hat{\boldsymbol{x}}=\left(\hat{\boldsymbol{x}}_{1}, \ldots, \hat{\boldsymbol{x}}_{d}\right)=\left(s_{1} \cdot \boldsymbol{x}_{1}, \ldots, s_{d} \cdot \boldsymbol{x}_{d}\right)$, and compute

$$
C_{1}=\left(\hat{\boldsymbol{x}}_{1}, \ldots, \hat{\boldsymbol{x}}_{d}, \mathbf{0}_{n}, \alpha, 0, s\right)_{\mathbb{B}} \text { and } C_{2}=\Omega^{\alpha} \cdot \mathfrak{m}
$$

The ciphertext for vector $x$ and message $\mathfrak{m}$ is $\mathrm{Ct} \leftarrow\left[C_{1}, C_{2}\right]$.
$\operatorname{Dec}\left(m p k, \mathrm{Ct}, s k_{y}\right)$ : The decryption algorithms computes the blinding

$$
\hat{e}\left(C_{1}, s k^{d e c}\right)=\hat{e}(g, g)^{\alpha+\sum_{i \in[d]}\left\langle\hat{\boldsymbol{x}}_{i}, \hat{\boldsymbol{y}}_{i}\right\rangle}
$$

Remark 4.4.4 Notice that we need $\left|T_{\boldsymbol{y}}\right|$ randomization sub-keys because in order to re-randomize the decryption key it is necessary to generate $\left|T_{\boldsymbol{y}}\right|$ random coefficients, one for each $i \in S_{\boldsymbol{y}}$ plus $r$. By taking a random linear combination of the randomization sub-keys we achieve the task.

### 4.4.3 Security

We will prove the following main theorem:
Theorem 4.4.5 If Assumptions OT. 1 and OT. 2 hold then our WKD-IPE scheme is secure in the sense of Definition 4.3.4

Following [LW10], we define two additional structures that will be used in the proof of security. Specifically:

Semi-Functional Ciphertext : An semi-functional ciphertext has additional dimensions. More specifically, choose $\boldsymbol{r} \leftarrow \mathbb{Z}_{N}^{n}$ and set

$$
C_{1}=\left(\hat{\boldsymbol{x}}_{1}, \ldots, \hat{\boldsymbol{x}}_{d}, \boldsymbol{r}, \alpha, 0, s\right)_{\mathbb{B}} .
$$

Semi-Functional Keys : A secret key is semi-functional if all its sub-keys are semi-functional. More specifically:

## Decryption Key :

$$
s k^{s f . d e c}=\left(\hat{\boldsymbol{y}}_{1}, \ldots, \hat{\boldsymbol{y}}_{d}, \boldsymbol{r}^{d e c}, 1, r, 0\right)_{\mathbb{B}^{\star}},
$$

## Randomization Keys :

$$
s k_{k}^{s f . r a n}=\left(\hat{\boldsymbol{y}}_{k, 1}^{r a n}, \ldots, \hat{\boldsymbol{y}}_{k, d}^{r a n}, \boldsymbol{r}_{k}^{r a n}, 0, s_{k}, 0\right)_{\mathbb{B}^{\star}},
$$

## Delegation Keys :

$$
s k_{j}^{s f . d e l}=\left(\hat{\boldsymbol{y}}_{j, 1}^{d e l}, \ldots, \hat{\boldsymbol{y}}_{j, d}^{d e l}, \boldsymbol{r}_{j}^{d e l}, 0, t_{j}, 0\right)_{\mathbb{B}^{\star}},
$$

where $\boldsymbol{r}^{\text {dec }}, \boldsymbol{r}_{k}^{r a n}, \boldsymbol{r}_{j}^{\text {del }} \leftarrow \mathbb{Z}_{N}^{n}$.
Remark 4.4.6 Notice that semi-functional ciphertexts and semi-functional secret keys lead to a random decryption with high probability. In fact with high probability the $\left\langle\boldsymbol{r}, \boldsymbol{r}^{d e c}\right\rangle \neq 0$ given that $r$ and $\boldsymbol{r}^{d e c}$ are uniformly and independently distributed. If $\boldsymbol{r}$ and $\boldsymbol{r}^{d e c}$ were correlated such that

$$
\boldsymbol{r}=\rho \cdot \hat{\boldsymbol{x}} \cdot \boldsymbol{U} \quad \text { and } \quad \boldsymbol{r}^{d e c}=\tau \cdot \hat{\boldsymbol{y}} \cdot \boldsymbol{Z},
$$

where $\rho, \tau \leftarrow \mathbb{Z}_{N}, \boldsymbol{U}=\left(u_{i, j} \leftarrow \mathbb{Z}_{N}\right)_{i \in[n], j \in[n]}$ and $\boldsymbol{Z}=\left(\boldsymbol{U}^{\boldsymbol{\top}}\right)^{-1}$, then the decryption would succeed for all vectors $\hat{\boldsymbol{x}}$ and $\hat{\boldsymbol{y}}$ such that their inner-product is 0 .

For a PPT adversary $\mathcal{A}$ which makes $q$ key queries, our proof of security will consist of the following $q \cdot(n+1)+3$ games between $\mathcal{A}$ and a challenger $\mathcal{C}$.

Game $_{\text {Real }}$ : It is the real Inner-product encryption with generalized key delegation security game.

Game $_{\text {WD }}$ : It it the same as Game Real except that all key queries will be answered by fresh calls to the key generation.

Game $_{0}$ : It is the same as Game $_{\text {WD }}$ except that the challenge ciphertext given to $\mathcal{A}$ is semifunctional.

Game $_{\mathbf{i}, \mathbf{j}}$ : For $i \in[q]$ and $0 \leq j \leq n+1$, is the same as Game $_{0}$ except that the first $i-1$ keys are semi-functional and the remaining $q-i$ are normal. For the $i$-th key, only the first $j$ sub-keys are semi-functional and the rest of the sub-keys are normal. Thus Game $_{\mathbf{1}, \mathbf{0}}=$ Game $_{\mathbf{0}}$ and Game $_{\mathbf{i}, \mathbf{n}+\mathbf{1}}=$ Game $_{\mathbf{i}+\mathbf{1}, \mathbf{0}}$.

Game $_{\text {Final }}$ : It is the same as $\mathbf{G a m e}_{\mathbf{q}, \mathbf{n}+\mathbf{1}}$ expect that the challenge ciphertext is for a random message and random attribute vector. Notice that in Game Final the value of $\eta$ is $^{\text {sen }}$ information-theoretically hidden from the attacker. Hence the attacker has no advantage in breaking the scheme.

For completeness we restate some useful lemmas whose proofs can be found in [LOS $\left.{ }^{+} 10\right]$.
Lemma 4.4.7 Let $C=\{(\boldsymbol{x}, \boldsymbol{y}) \mid\langle\boldsymbol{x}, \boldsymbol{y}\rangle \neq 0\} \subset V \times V^{\star}$ where $V$ is an $n$-dimensional vector space over final field $\mathbb{F}_{N}$ and $V^{\star}$ its dual. Then for all $(\boldsymbol{x}, \boldsymbol{y}) \in C$ and for all $(\boldsymbol{r}, \boldsymbol{w}) \in C$ we have that $\operatorname{Pr}_{\boldsymbol{Z} \leftarrow G L\left(n, \mathbb{F}_{N}\right), \rho, \tau \leftarrow \mathbb{F}_{N}^{\times}}[\boldsymbol{x}(\rho \boldsymbol{U})=\boldsymbol{r}$ and $\boldsymbol{v}(\tau \boldsymbol{Z})=\boldsymbol{w}]=\frac{1}{|C|}$ where $\boldsymbol{U}=\left(\boldsymbol{Z}^{\boldsymbol{\top}}\right)^{-1}$.

Lemma 4.4.8 Let $N>2$ and $\Delta=\{\boldsymbol{M} \mid \operatorname{det} \boldsymbol{M}=0\} \subset \mathbb{F}_{N}^{n \times n}$. Then $|\Delta| / N^{n^{2}}<2 / N$
We give now formal proofs of the above statements.

Proof: We show that the distribution of $s k_{y^{\prime}} \leftarrow \operatorname{KeyGen}\left(m s k, \boldsymbol{y}^{\prime}\right)$ is equivalent to that of $\operatorname{Key} \operatorname{Del}\left(m p k, s k_{\boldsymbol{y}}, \boldsymbol{y}^{\prime}\right)$, for $s k_{\boldsymbol{y}}=\operatorname{KeyGen}(m s k, \boldsymbol{y})$ such that $\boldsymbol{y}^{\prime}$ can be delegated from $\boldsymbol{y}$, except with probability $3 / N$. Notice that the distribution of the secret key $s k_{y^{\prime}}$ is represented by the coefficients of the square matrix $\boldsymbol{S}=\left(\left(s_{k . i}\right)_{i \in S_{y^{\prime}}}, s_{k}\right)_{k \in T_{y^{\prime}}}$ (which is a $\left|T_{\boldsymbol{y}^{\prime}}\right| \times\left|T_{\boldsymbol{y}^{\prime}}\right|$ matrix in $\mathbb{Z}_{N}$ ) and by $t$ used in the delegation sub-keys, which are all uniformly and independently distributed. By Lemma 4.4.8 the span of matrix $\boldsymbol{S}_{y^{\prime}}$ is $\mathbb{Z}_{N}^{\left|T_{y^{\prime}}\right|}$ except with probability $2 / N$ and the coefficient $t$ is different from 0 except with probability $1 / N$. Thus the distribution of $\operatorname{Key} \operatorname{Del}\left(m p k, s k_{\boldsymbol{y}}, \boldsymbol{y}^{\prime}\right)$ is equivalent to that of $s k_{\boldsymbol{y}^{\prime}}$ except with probability $2 / N+1 / N=3 / N$. The Shoup's difference lemma gives us the upper bound.
 Then we can build an algorithm $\mathcal{B}$ with advantage $\epsilon$ in breaking Assumption OT.1.

Proof: We describe algorithm $\mathcal{B}$ that breaks Assumption OT.1. $\mathcal{B}$ receives (param $\left., \hat{\mathbb{B}}, \hat{\mathbb{B}}^{\star}\right)$ and $\boldsymbol{e}_{\beta}$. Depending on the value of $\beta, \mathcal{B}$ simulates Game ${ }_{\mathbf{W D}}$ or Game $_{\mathbf{0}}$.

Setup : $\mathcal{B}$ sets $\Omega=\hat{e}(g, g)$ and gives $\mathcal{A} m p k \leftarrow\left[\operatorname{param}_{\mathbb{V}}, \hat{\mathbb{B}}, \Omega\right]$ and keeps $m s k \leftarrow\left[\hat{\mathbb{B}}^{\star}\right]$ secret.
Secret Keys : $\mathcal{B}$ uses the $m s k$ to generate secret keys.
Challenge : $\mathcal{A}$ sends $\mathcal{B}$ the two challenges $\left(\left(\mathfrak{m}^{(0)}, \boldsymbol{x}^{(0)}\right),\left(\mathfrak{m}^{(1)}, \boldsymbol{x}^{(1)}\right)\right) . \mathcal{B}$ chooses $\eta \leftarrow\{0,1\}$ and then encrypts $\left(\mathfrak{m}^{(\eta)}, x^{(\eta)}\right)$ as follows: $\mathcal{B}$ chooses $\alpha \leftarrow \mathbb{Z}_{N},\left(s_{i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[d]}$ and pads $\boldsymbol{x}^{\eta}$ to the maximum length $d$ with random vectors following the hierarchy $\boldsymbol{\mu}$. Then $\mathcal{B}$ applies a random translation to $\boldsymbol{x}^{\eta}$ so that

$$
\begin{aligned}
\hat{\boldsymbol{x}}^{(\eta)} & =\left(\hat{\boldsymbol{x}}_{1}^{(\eta)}, \ldots, \boldsymbol{x}_{d}^{(\eta)}\right) \\
& =\left(s_{1} \cdot \boldsymbol{x}_{1}^{(\eta)}, \ldots, s_{d} \cdot \boldsymbol{x}_{d}^{(\eta)}\right)
\end{aligned}
$$

and computes

$$
C_{1}=\sum_{i \in[n]} \hat{x}_{i}^{(\eta)} \cdot \boldsymbol{e}_{\beta, i}+\alpha \cdot \boldsymbol{b}_{2 n+1} \text { and } C_{2}=\Omega^{\alpha} \cdot \mathfrak{m}^{(\eta)} .
$$

Now, if $\beta=0$ then

$$
C_{1}=\left(\overline{\boldsymbol{x}}_{1}, \ldots, \overline{\boldsymbol{x}}_{d}, \mathbf{0}_{n}, \alpha, 0, s\right)_{\mathbb{B}}
$$

where $\overline{\boldsymbol{x}}_{i}=\delta \cdot s_{i} \cdot \boldsymbol{x}_{i}^{(\eta)}$ and $s=\left\langle\hat{\boldsymbol{x}}^{(\eta)}, \boldsymbol{\delta}\right\rangle$, and the challenge ciphertext is distributed as in Game $_{\text {WD }}$. Instead if $\beta=1$ then

$$
C_{1}=\left(\overline{\boldsymbol{x}}_{1}, \ldots, \overline{\boldsymbol{x}}_{d}, \boldsymbol{r}, \alpha, 0, s\right)_{\mathbb{B}}
$$

where $\overline{\boldsymbol{x}}_{i}=\delta \cdot s_{i} \cdot \boldsymbol{x}_{i}^{(\eta)}, \boldsymbol{r}=\left(r_{1}, \ldots, r_{n}\right)$ with $r_{i}=\rho \cdot\left\langle\hat{\boldsymbol{x}}^{(\eta)}, \boldsymbol{u}_{i}\right\rangle$, and $s=\left\langle\hat{\boldsymbol{x}}^{(\eta)}, \boldsymbol{\delta}\right\rangle$. Because $\hat{\boldsymbol{x}}^{(\eta)} \neq \mathbf{0}_{n}$ and that $\left(\boldsymbol{u}_{i}\right)_{i \in[n]}$ and $\boldsymbol{\delta}$ are uniformly and independently distributed then the challenge ciphertext is properly distributed as in Game ${ }_{\mathbf{0}}$. Hence, $\mathcal{B}$ can use the output of $\mathcal{A}$ to distinguish the value of $\beta$.

Lemma 4.4.11 Suppose there exists an algorithm $\mathcal{A}$ such that $\mathbf{A d v}_{\mathbf{G a m e}_{\mathbf{i}, \mathbf{j}-\mathbf{1}}}^{\mathcal{A}}-\mathbf{A d v}_{\mathbf{G a m e}_{\mathbf{i}, \mathbf{j}}}^{\mathcal{A}}=$ $\epsilon$. Then we can build an algorithm $\mathcal{B}$ with advantage $\epsilon+1 / N$ in breaking Assumption OT.2.

Proof: We describe algorithm $\mathcal{B}$ that breaks Assumption OT.2. $\mathcal{B}$ receives (param ${ }_{\mathbb{V}}, \hat{\mathbb{B}}, \hat{\mathbb{B}}^{\star}, \boldsymbol{e}$ ) and $\boldsymbol{h}_{\beta}$. Depending on the value of $\beta, \mathcal{B}$ simulates $\mathbf{G a m e}_{\mathbf{i}, \mathbf{j}-\mathbf{1}}$ or $\mathbf{G a m e}_{\mathbf{i}, \mathbf{j}}$.

Setup $: \mathcal{B}$ sets $\Omega=\hat{e}(g, g)$ and gives $\mathcal{A} m p k \leftarrow\left[\operatorname{param}_{\mathbb{V}}, \hat{\mathbb{B}}, \Omega\right]$ and sets $m s k \leftarrow\left[\hat{\mathbb{B}}^{\star}\right]$.
Secret Keys : For the first $i-1$ secret key queries, $\mathcal{B}$ creates semi-functional secret keys in the following way: $\mathcal{B}$ invokes the key generation algorithm on input the master secret key $m s k$ and the given vector $\boldsymbol{y}$ to obtain the secret key $s k_{\boldsymbol{y}} \leftarrow\left[s k^{d e c},\left(s k_{k}^{r a n}\right)_{k \in T_{\boldsymbol{y}}},\left(s k_{k}^{d e l}\right)_{k \in D_{\boldsymbol{y}}}\right] \leftarrow$ $\operatorname{KeyGen}(m s k, \boldsymbol{y})$. Then $\mathcal{B}$ introduces the additional dimensions to create semi-functional secret key $s k_{\boldsymbol{y}}^{s f} \leftarrow\left[s k^{s f . d e c},\left(s k_{k}^{s f . r a n}\right)_{k \in T_{\boldsymbol{y}}},\left(s k_{k}^{s f . d e l}\right)_{k \in D_{\boldsymbol{y}}}\right]$ by using $\hat{\mathbb{B}}^{\star}$ as:

## Decryption Key :

$$
s k^{s f . d e c}=s k^{d e c}+\sum_{i \in[n]} r_{i}^{d e c} \cdot \boldsymbol{b}_{n+i}^{\star}
$$

## Randomization Keys :

$$
s k_{k}^{s f . r a n}=s k_{k}^{r a n}+\sum_{i \in[n]} r_{k, i}^{r a n} \cdot \boldsymbol{b}_{n+i}^{\star}
$$

## Delegation Key :

$$
s k_{j}^{s f . d e l}=s k_{j}^{d e l}+\sum_{i \in[n]} r_{j, i}^{d e l} \cdot \boldsymbol{b}_{n+i}^{\star}
$$

where $r_{i}^{d e c}, r_{k, i}^{r a n}, r_{j, i}^{d e l} \leftarrow \mathbb{Z}_{N}^{n}$.
Then, for the $i$-th secret key query on vector $\boldsymbol{y}$, the first $j-1$ sub-keys are generated semi-functional as before. The remaining $n-j$ sub-keys are normal, so without the semifunctional part. For the $j$-th sub-key of the $i$-th key, $\mathcal{B}$ does the following: Depending on the type of the $j$-th sub-key (decryption, randomization, delegation), $\mathcal{B}$ applies the necessary translation to the vector $\boldsymbol{y}$ to obtain the vector $\hat{\boldsymbol{y}}$. Then it encodes $\hat{\boldsymbol{y}}$ as:

$$
s k_{j}=\sum_{i \in[n]} \hat{y}_{i} \cdot \boldsymbol{h}_{\beta, i}+\boldsymbol{b}_{2 n+1}^{\star} .
$$

The remaining $q-i$ secret key queries are answered by generating normal secret keys using $m s k$.

Challenge : $\mathcal{A}$ sends $\mathcal{B}$ the tuple $\left(\left(\mathfrak{m}^{(0)}, \boldsymbol{x}^{(0)}\right),\left(\mathfrak{m}^{(1)}, \boldsymbol{x}^{(1)}\right)\right)$. $\mathcal{B}$ chooses $\eta \leftarrow\{0,1\}$ and encrypts $\left(\mathfrak{m}^{(\eta)}, \boldsymbol{x}^{(\eta)}\right)$ as follows: $\mathcal{B}$ chooses random $\alpha, s \leftarrow \mathbb{Z}_{N},\left(s_{i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[d]}$ and pads $\boldsymbol{x}^{(\eta)}$ to the maximum length $d$ with random vectors following the hierarchy $\mu$. Then $\mathcal{B}$ applies a random translation to $\boldsymbol{x}^{(\eta)}$ so that

$$
\begin{aligned}
\hat{\boldsymbol{x}}^{(\eta)} & =\left(\hat{\boldsymbol{x}}_{1}^{(\eta)}, \ldots, \hat{\boldsymbol{x}}_{d}^{(\eta)}\right) \\
& =\left(s_{1} \cdot \boldsymbol{x}_{1}^{(\eta)}, \ldots, s_{d} \cdot \boldsymbol{x}_{d}^{(\eta)}\right)
\end{aligned}
$$

$\mathcal{B}$ computes

$$
\begin{aligned}
C_{1} & =\sum_{i \in[n]} \hat{x}_{i}^{(\eta)} \cdot \boldsymbol{e}_{i}+\alpha \cdot \boldsymbol{b}_{2 n+1}+s \cdot \boldsymbol{b}_{2 n+3}, \\
& =\left(\overline{\boldsymbol{x}}_{1}, \ldots, \overline{\boldsymbol{x}}_{d}, \boldsymbol{t}, \alpha, 0, s\right)_{\mathbb{B}}
\end{aligned}
$$

and

$$
C_{2}=\Omega^{\alpha} \cdot \mathfrak{m}^{(\eta)}
$$

where $\overline{\boldsymbol{x}}_{i}=\delta \cdot s_{i} \cdot \boldsymbol{x}_{i}^{(\eta)}, \boldsymbol{t}=\left(t_{1}, \ldots, t_{n}\right)$ with $\left(t_{i}=\rho \cdot\left\langle\hat{\boldsymbol{x}}^{(\eta)}, \boldsymbol{u}_{i}\right\rangle\right)_{i \in[n]}$.
Now, if $\beta=0$ then the $j$-th sub-key can be written as

$$
s k_{j}=\left(\overline{\boldsymbol{y}}_{1}, \ldots, \overline{\boldsymbol{y}}_{d}, \mathbf{0}_{n}, 1, r, 0\right)_{\mathbb{B}^{\star}},
$$

where $\overline{\boldsymbol{y}}_{i}=\omega \cdot \hat{\boldsymbol{y}}_{i}$ and $r=\langle\hat{\boldsymbol{y}}, \gamma\rangle$. Thus $s k_{j}$ is a normal sub-key and the joint distribution of the challenge ciphertext and $s k_{j}$ is as in $\mathbf{G a m e}_{\mathbf{i} \mathbf{j} \mathbf{j} \mathbf{1}}$. Instead if $\beta=1$ then the $j$-th sub-key can be written

$$
s k_{j}=\left(\overline{\boldsymbol{y}}_{1}, \ldots, \overline{\boldsymbol{y}}_{d}, \boldsymbol{r}, 1, r, 0\right)_{\mathbb{B}^{\star}}
$$

where $\overline{\boldsymbol{y}}_{i}=\omega \cdot \hat{\boldsymbol{y}}_{i}, \boldsymbol{r}=\left(r_{1}, \ldots, r_{n}\right)$ with $\left(r_{i}=\tau \cdot\left\langle\hat{\boldsymbol{y}}, \boldsymbol{z}_{i}\right\rangle\right)_{i \in[n]}, r=\langle\hat{\boldsymbol{y}}, \boldsymbol{\gamma}\rangle$. Since $\langle\hat{\boldsymbol{y}}, \hat{\boldsymbol{x}}\rangle \neq 0$, by the constraints of the security game and from Lemma 4.4.7, the coefficients $\left(t_{i}, r_{i}\right)_{i \in[n]}$ are pairwise-independently and uniformly distributed under the condition that $\langle\boldsymbol{t}, \boldsymbol{r}\rangle \neq 0$. Thus the joint distribution of the challenge ciphertext and $s k_{j}$ is as in Game $_{\mathbf{i}, \mathbf{j}}$ expect with probability $1 / N$ that is the probability of the event $\langle\boldsymbol{t}, \boldsymbol{r}\rangle=0$. I

Lemma 4.4.12 For any algorithm $\mathcal{A}$, Game $_{\mathbf{q}, \mathbf{n}+\mathbf{1}}=$ Game $_{\text {Final }}$.
Proof: We show that the view of $\mathcal{A}$ in $\mathbf{G a m e}_{\mathbf{q}, \mathbf{n + 1}}$ is identical to the view of $\mathcal{A}$ in Game $_{\text {Final }}$. Given the dual orthonormal bases $\left(\mathbb{B}, \mathbb{B}^{\star}\right)$, used in $\mathrm{Game}_{\mathbf{q}, \mathbf{n}+\mathbf{1}}$, of $\mathbb{V}$, we will construct new dual orthonormal bases $\mathbb{D}, \mathbb{D}^{\star}$ of $\mathbb{V}$ such that:

1. $\left(\mathbb{D}, \mathbb{D}^{\star}\right)$ is distributed like $\left(\mathbb{B}, \mathbb{B}^{\star}\right)$,
2. $\left(\mathbb{D}, \mathbb{D}^{\star}\right)$ is consistent with the public key seen by $\mathcal{B}$ in both games $\boldsymbol{G a m e}_{\mathbf{q}, \mathbf{n + 1}}$ and Game $_{\text {Final }}$, and
3. the secret keys and the challenge ciphertext in Game Final can be written in those bases.

Thus, given

$$
\mathbb{B}=\left(\boldsymbol{b}_{1}, \ldots, \boldsymbol{b}_{n}, \boldsymbol{b}_{n+1}, \ldots, \boldsymbol{b}_{2 n}, \boldsymbol{b}_{2 n+1}, \boldsymbol{b}_{2 n+2}, \boldsymbol{b}_{2 n+3}\right)
$$

and

$$
\mathbb{B}^{\star}=\left(\boldsymbol{b}_{1}^{\star}, \ldots, \boldsymbol{b}_{n}^{\star}, \boldsymbol{b}_{n+1}^{\star}, \ldots, \boldsymbol{b}_{2 n}^{\star}, \boldsymbol{b}_{2 n+1}^{\star}, \boldsymbol{b}_{2 n+2}^{\star}, \boldsymbol{b}_{2 n+3}^{\star}\right),
$$

we define new dual orthonormal bases $\mathbb{D}, \mathbb{D}^{\star}$ of $\mathbb{V}$ as follow: First choose $\boldsymbol{U}=\left(\boldsymbol{u}_{i} \leftarrow \mathbb{Z}_{N}^{n}\right)_{i \in[n]}$ and $\boldsymbol{v}=\left(v_{i} \leftarrow \mathbb{Z}_{N}\right)_{i \in[n]}$, then define

$$
\mathbb{D}=\left(\boldsymbol{b}_{1}, \ldots, \boldsymbol{b}_{n}, \boldsymbol{d}_{n+1}, \ldots, \boldsymbol{d}_{2 n}, \boldsymbol{b}_{2 n+1}, \boldsymbol{b}_{2 n+2}, \boldsymbol{b}_{2 n+3}\right),
$$

and

$$
\mathbb{D}^{\star}=\left(\boldsymbol{d}_{1}^{\star}, \ldots, \boldsymbol{d}_{n}^{\star}, \boldsymbol{b}_{n+1}^{\star}, \ldots, \boldsymbol{b}_{2 n}^{\star}, \boldsymbol{d}_{2 n+1}^{\star}, \boldsymbol{b}_{2 n+2}^{\star}, \boldsymbol{b}_{2 n+3}^{\star}\right),
$$

where $\left(\boldsymbol{d}_{i}^{\star}=\boldsymbol{b}_{i}^{\star}+\sum_{j \in[n]} u_{i, j} \cdot \boldsymbol{b}_{n+j}^{\star}\right)_{i \in[n]}$ and $\left(\boldsymbol{d}_{n+i}=\boldsymbol{b}_{n+i}-\sum_{j \in[n]} u_{i, j} \cdot \boldsymbol{b}_{j}-v_{i} \cdot \boldsymbol{b}_{2 n+1}\right)_{i \in[n]}$ and $\boldsymbol{d}_{2 n+1}^{\star}=\boldsymbol{b}_{2 n+1}^{\star}+\sum_{j \in[n]} v_{i} \cdot \boldsymbol{b}_{n+j}^{\star}$.
It is easy to see that $\left(\mathbb{D}, \mathbb{D}^{\star}\right)$ are dual orthonormal because the transformation applied preserve this property, distributed like $\left(\mathbb{B}, \mathbb{B}^{\star}\right)$ and consistent with public key in both games (notice that $\boldsymbol{b}_{n+1}, \ldots, \boldsymbol{b}_{2 n}$ are secret to the adversary). Now in $\operatorname{Game}_{\mathbf{q}, \mathbf{n + 1}}$ each sub-key on vector $\boldsymbol{y}$ can be written as the encoding of a randomly translated vector $\hat{\boldsymbol{y}}$ (the translation depends on the sub-key type) in base $\mathbb{B}^{\star}$. This sub-key can be rewritten in base $\mathbb{D}^{\star}$, affecting only the semi-functional part, as follows:

$$
\begin{aligned}
s k & =\left(\hat{\boldsymbol{y}}_{1}, \ldots, \hat{\boldsymbol{y}}_{d}, \boldsymbol{r}, 1, r, 0\right)_{\mathbb{B}^{\star}} \\
& =\sum_{i \in[n]} \hat{y}_{i} \cdot \boldsymbol{b}_{i}^{\star}+\sum_{i \in[n]} r_{i} \cdot \boldsymbol{b}_{n+i}^{\star}+\boldsymbol{b}_{2 n+1}^{\star}+r \cdot \boldsymbol{b}_{2 n+2}^{\star} \\
& =\sum_{i \in[n]} \hat{y}_{i} \cdot\left(\boldsymbol{d}_{i}^{\star}-\sum_{j \in[n]} u_{i, j} \cdot \boldsymbol{b}_{n+j}^{\star}\right)+\sum_{i \in[n]} r_{i} \cdot \boldsymbol{b}_{n+i}^{\star}+\left(\boldsymbol{d}_{2 n+1}^{\star}-\sum_{j \in[n]} v_{i} \cdot \boldsymbol{b}_{n+j}^{\star}\right)+r \cdot \boldsymbol{b}_{2 n+2}^{\star}, \\
& =\sum_{i \in[n]} \hat{y}_{i} \cdot \boldsymbol{d}_{i}^{\star}+\sum_{i \in[n]}\left(r_{i}-\left\langle\hat{\boldsymbol{y}}_{i}, \boldsymbol{u}_{i}\right\rangle-v_{i}\right) \cdot \boldsymbol{b}_{n+i}^{\star}+\boldsymbol{d}_{2 n+1}^{\star}+r \cdot \boldsymbol{b}_{2 n+2}^{\star} \\
& =\left(\hat{\boldsymbol{y}}_{1}, \ldots, \hat{\boldsymbol{y}}_{d}, \boldsymbol{r}^{\prime}, 1, r, 0\right)_{\mathbb{D}^{\star}}
\end{aligned}
$$

where $\boldsymbol{r}^{\prime}=\left(r_{1}^{\prime}, \ldots, r_{n}^{\prime}\right)$ with $r_{i}^{\prime}=r_{i}-\left\langle\hat{\boldsymbol{y}}_{i}, \boldsymbol{u}_{i}\right\rangle-v_{i}$. Notice that $\boldsymbol{r}^{\prime}$ is uniformly distributed.
Moreover in $\operatorname{Game}_{\mathbf{q}, \mathbf{n}+\mathbf{1}}$ the challenge ciphertext on vector $\boldsymbol{x}=\boldsymbol{x}^{(\eta)}$ can be written as the encoding of a randomly translated vector $\hat{\boldsymbol{x}}$ in base $\mathbb{B}$. This challenge ciphertext can be rewritten in base $\mathbb{D}$ as follows:

$$
\begin{aligned}
C_{1} & =\left(\hat{\boldsymbol{x}}_{1}, \ldots, \hat{\boldsymbol{x}}_{d}, \boldsymbol{t}, \alpha, 0, t\right)_{\mathbb{B}} \\
& =\sum_{i \in[n]} \hat{x}_{i} \cdot \boldsymbol{b}_{i}+\sum_{i \in[n]} t_{i} \cdot \boldsymbol{b}_{n+i}+\alpha \cdot \boldsymbol{b}_{2 n+1}+t \cdot \boldsymbol{b}_{2 n+3} \\
& =\sum_{i \in[n]} \hat{x}_{i} \cdot \boldsymbol{b}_{i}+\sum_{i \in[n]} t_{i} \cdot\left(\boldsymbol{d}_{n+i}+\sum_{j \in[n]} u_{i, j} \cdot \boldsymbol{b}_{j}+v_{i} \cdot \boldsymbol{b}_{2 n+1}\right)+\alpha \cdot \boldsymbol{b}_{2 n+1}+t \cdot \boldsymbol{b}_{2 n+3}, \\
& =\sum_{i \in[n]}\left(\hat{x}_{i}+\left\langle\boldsymbol{t}, \boldsymbol{u}_{i}\right\rangle\right) \cdot \boldsymbol{b}_{i}+\sum_{i \in[n]} t_{i} \cdot \boldsymbol{b}_{n+i}+(\alpha \cdot\langle\boldsymbol{t}, \boldsymbol{u}\rangle) \cdot \boldsymbol{b}_{2 n+1}+t \cdot \boldsymbol{b}_{2 n+3} \\
& =\left(\overline{\boldsymbol{x}}_{1}, \ldots, \overline{\boldsymbol{x}}_{d}, \boldsymbol{t}, \alpha^{\prime}, 0, t\right)_{\mathbb{D}}
\end{aligned}
$$

affecting only $\hat{\boldsymbol{x}}$ and $\alpha$ and where $\bar{x}_{i}=\hat{x}_{i}+\left\langle\boldsymbol{t}, \boldsymbol{u}_{i}\right\rangle$ and $\alpha^{\prime}=\alpha \cdot\langle\boldsymbol{t}, \boldsymbol{u}\rangle$. Notice that $\overline{\boldsymbol{x}}$ and $\alpha^{\prime}$ are independently and uniformly distributed as requested in Game Final $^{\text {. Thus, the view of } \mathcal{A}}$ in Game $_{\mathbf{q}, \mathbf{n + 1}}$ and Game ${ }_{\text {Final }}$ is the same.

Wrapping up. We have proved by the previous lemmata that the real security game is indistinguishable from Game Final , where $\eta$ is information-theoretically hidden from the attacker. Thus under assumptions OT. 1 and OT.2, Theorem 4.4.5holds.

## Chapter 5

## Lattice-based Hierarchical Inner Product Encryption

In this chapter we consider the problem of constructing hierarchical inner-product encryption scheme based on lattices assumptions. To achieve this goal, we extend the lattice-based IPE scheme by Agrawal et al. [AFV11] to the hierarchical setting by employing basis delegation technics by Peikert et al. [CHKP10] and by Agrawal et al. [ABB10]. As the underlying IPE scheme, our new scheme is shown to be selective weakly attribute-hiding secure based on the difficulty of the learning with errors problem in the standard model, as long as the total number of levels in the hierarchy is a constant.

### 5.1 Definitions

In this section, we introduce the hierarchical inner-product encryption primitive and its security. In doing so, we adopt the same notation and definition style used in [AFV11].

### 5.1.1 Hierarchical inner-product encryption

Definition 5.1.1 An hierarchical inner-product encryption for hierarchical format $\boldsymbol{\mu}$ (Definition 2.1.2 is defined by the following tuple of algorithms HIPE $=($ Setup, Derive, Enc, Dec) with the following syntax:
$\operatorname{Setup}\left(1^{\lambda}, \boldsymbol{\mu}\right)$ Takes as input security parameter $\lambda$ and hierarchical format $\boldsymbol{\mu}$ and outputs public parameters $m p k$ and master secret key $m s k$.

Derive $\left(m p k, s k_{\boldsymbol{v}}, \boldsymbol{v}_{t}\right)$ Takes as input the master public key $m p k$, the secret key for the vector $\boldsymbol{v}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t-1}\right) \in \Sigma_{\mid t-1}$, and a vector $\boldsymbol{v}_{t} \in \Sigma_{t}$, and outputs a secret key $s k_{\boldsymbol{v}^{\prime}}$ for the vector $\boldsymbol{v}^{\prime}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t-1}, \boldsymbol{v}_{t}\right)$. If $\boldsymbol{v}$ is the empty vector then $s k_{\boldsymbol{v}}$ is the master secret key $m s k$.
$\operatorname{Enc}\left(m p k, \mathfrak{m}, \boldsymbol{w}=\left(\boldsymbol{w}_{1}, \ldots, \boldsymbol{w}_{t}\right) \in \Sigma_{\mid t}\right)$ Takes as input public parameters $\mathfrak{m}$ in some associated message space, public parameters $m p k$ and an attribute vector $\boldsymbol{w}$ and outputs a ciphertext Ct.
$\operatorname{Dec}\left(m p k, C t, s k_{\boldsymbol{v}}\right)$ Takes as input public parameters $m p k$, ciphertext Ct and secret key $s k_{\boldsymbol{v}}$ and outputs the message $\mathfrak{m}$.

Correctness. We make the following consistency requirement. Suppose ciphertext Ct is obtained by running Enc on input $m p k$, message $\mathfrak{m}$ and attribute vector $\boldsymbol{w}$ and that $s k_{v}$ is a secret key for attribute vector $\boldsymbol{v}$ obtained through a sequence of Derive calls using the same $m p k$. Then Dec, on input $m p k, \mathrm{Ct}$ and $s k_{\boldsymbol{v}}$, returns $\mathfrak{m}$, except with negligible probability, if and only if $f_{\boldsymbol{v}}(\boldsymbol{w})=1$.

### 5.1.2 Security Definition

The notion we achieve is the selective weakly attribute-hiding indistinguishability under-chosen message attack (wAH-sIND-CPA, for short) and can be formalized by means of the following game between an adversary $\mathcal{A}$ and a challenger $\mathcal{C}$. Game wAH-sIND-CPA ${ }_{\mathcal{A}}^{\mathrm{HIPE}}\left(1^{\lambda}\right)$, where $\lambda$ is the security parameter, is defined as follows.

Init: $\mathcal{A}$ is given hierarchical format $\boldsymbol{\mu}$ of depth $d$ and outputs challenge vectors $\boldsymbol{w}_{0}, \boldsymbol{w}_{1} \in \Sigma_{\mid h}$.
Setup: $\mathcal{C}$ generates master public key and secret key by invoking the setup algorithm on input the security parameter $\lambda$ given in unary. Specifically, $(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}\right)$. Then, $\mathcal{C}$ starts the interaction with $\mathcal{A}$ on input mpk.

Query Phase 1: $\mathcal{A}$ is given oracle access to Derive $(m p k, m s k, \cdot)$. Then, $\mathcal{A}$ can delegate secret keys directly by invoking the Derive algorithm.

Challenge: $\mathcal{A}$ gives a pair of message $\left(\mathfrak{m}_{0}, \mathfrak{m}_{1}\right)$ to $\mathcal{C}$. Then $\mathcal{C}$ chooses random $\eta \leftarrow\{0,1\}$, encrypts $\mathfrak{m}_{\eta}$ under $\boldsymbol{w}_{\eta}$ and sends the resulting ciphertext to $\mathcal{A}$.

Query Phase 2: It is the same as Query Phase 1.
Guess: Eventually, $\mathcal{A}$ submits its guess $\eta^{\prime}$.
Winning Condition: $\mathcal{A}$ wins the game if the following conditions are satisfied:

1. $\eta=\eta^{\prime}$.
2. for any vector $\boldsymbol{v}$ for which the relative secret key has been obtained through derivation from one of the revealed secret key, it holds that $f_{\boldsymbol{v}}\left(\boldsymbol{w}_{0}\right)=f_{\boldsymbol{v}}\left(\boldsymbol{x}_{1}\right)=0$.

The advantage of $\mathcal{A}$ in the above game is defined as

$$
\operatorname{Adv}_{\mathrm{wAH}-\mathrm{SIND}-\mathrm{CPA}}^{\mathrm{HIPE}}(\lambda)=\operatorname{Prob}\left[\mathrm{wAH}-\mathrm{sIND}-\mathrm{CPA}_{\mathcal{A}}^{\mathrm{HIPE}}\left(1^{\lambda}\right)=1\right]-1 / 2 .
$$

Definition 5.1.2 An hierarchical inner-product encryption scheme HIPE is wAH-sIND-CPAsecure if any polynomial-time adversaries making at most a polynomial number of queries to the key derivation oracle only has a negligible advantage in the wAH-sIND-CPA game described above, meaning that $\operatorname{Adv}_{\text {wAH-sind-CPA }}^{\mathrm{HIPE}, \mathcal{A}}(\lambda)$ is negligible function of $\lambda$.

### 5.2 An HIPE Construction

We start by giving some intuitions on how we have extended the scheme by Agrawal et al. [AFV11] to the hierarchical setting.

Public Parameters. For hierarchical format $\boldsymbol{\mu}=\left(\ell, d ; \mu_{1}, \ldots, \mu_{d}\right)$, the public parameters will contain random matrices $\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\}\right)$ in $\mathbb{Z}_{q}^{n \times m}$. The master secret key is a trapdoor $\boldsymbol{T}_{\boldsymbol{A}}$ for $\boldsymbol{A}$. To generate a secret key for vector $\boldsymbol{v}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$ at depth $t \leq d$ we use the matrix:

$$
\begin{equation*}
F_{\boldsymbol{v}}=\left(\boldsymbol{A}\left\|\sum_{j \in\left[\mu_{1}\right]} \sum_{\gamma=0}^{k} v_{1, j, \gamma} \cdot \boldsymbol{A}_{1, j, \gamma}\right\| \ldots \| \sum_{j \in\left[\mu_{t}\right]} \sum_{\gamma=0}^{k} v_{t, j, \gamma} \cdot \boldsymbol{A}_{t, j, \gamma}\right) \in \mathbb{Z}_{q}^{n \times(t+1) m} \tag{5.1}
\end{equation*}
$$

where each $v_{i, j}$ is $r$-decomposed for a certain fixed $r$ and $k=\left\lfloor\log _{r} q\right\rfloor$.

Secret Keys. Then the secret key for $\boldsymbol{v}$ is a short basis for the lattice $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$. By using the short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$, it is possible to generate a random short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v} \| \boldsymbol{v}_{t+1}}\right)$. This provides the delegation mechanism.

Simulation. For challenge vector $\boldsymbol{w}^{\star}=\left(\boldsymbol{w}_{1}^{\star}, \ldots, \boldsymbol{w}_{t^{\star}}^{\star}\right)$, the simulator chooses the matrices $\boldsymbol{A}$ and $\boldsymbol{B}$ uniformly at random in $\mathbb{Z}_{q}^{n \times m}$ and construct the matrices $\boldsymbol{A}_{i, j, \gamma}$ as follows:

$$
\boldsymbol{A}_{i, j, \gamma}=\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}
$$

where $\boldsymbol{R}_{i, j, \gamma} \in\{-1,1\}^{m \times m}$. Since the matrices $\boldsymbol{A},\left\{\boldsymbol{R}_{i, j, \gamma}\right\}$ are uniform and independent in $\mathbb{Z}_{q}^{n \times m}$, we have that the $\boldsymbol{A}_{i, j, \gamma}$ 's are uniform in $\mathbb{Z}_{q}^{n \times m}$ as in the real system. Moreover the simulator has a trapdoor $\boldsymbol{T}_{\boldsymbol{B}}$ for $\Lambda_{q}^{\perp}(\boldsymbol{B})$ but no trapdoor for $\Lambda_{q}^{\perp}(\boldsymbol{A})$. To generate a secret key for vector $\boldsymbol{v}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$, the simulator must produce a short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$ where

$$
F_{\boldsymbol{v}}=\left(\boldsymbol{A}\left\|\boldsymbol{A} \overline{\boldsymbol{R}_{1}}-\left\langle\boldsymbol{v}_{1}, \boldsymbol{w}_{1}^{\star}\right\rangle \boldsymbol{B}\right\| \ldots \| \boldsymbol{A} \overline{\boldsymbol{R}_{t}}-\left\langle\boldsymbol{v}_{t}, \boldsymbol{w}_{t}^{\star}\right\rangle \boldsymbol{B}\right)
$$

Then let

$$
\begin{aligned}
\overline{\boldsymbol{R}_{i}} & =\sum_{j \in\left[\mu_{i}\right]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \cdot \boldsymbol{R}_{i, j, \gamma} \quad \in \mathbb{Z}_{q}^{m \times m} \\
\overline{\boldsymbol{R}} & =\left[\overline{\boldsymbol{R}_{1}}\|\ldots\| \overline{\boldsymbol{R}_{t}}\right] \quad \in \mathbb{Z}_{q}^{m \times t \cdot m} \\
\boldsymbol{B}_{\boldsymbol{v}} & =\left[-\left\langle\boldsymbol{v}_{1}, \boldsymbol{w}_{1}^{\star}\right\rangle \boldsymbol{B}\|\ldots\|-\left\langle\boldsymbol{v}_{t}, \boldsymbol{w}_{t}^{\star}\right\rangle \boldsymbol{B}\right] \quad \in \mathbb{Z}_{q}^{n \times t \cdot m}
\end{aligned}
$$

Thus $F_{v}$ can be written as:

$$
\begin{equation*}
F_{\boldsymbol{v}}=\left(\boldsymbol{A} \| \boldsymbol{A} \overline{\boldsymbol{R}}+\boldsymbol{B}_{\boldsymbol{v}}\right) \quad \in \mathbb{Z}_{q}^{n \times(t+1) m} \tag{5.2}
\end{equation*}
$$

When $\boldsymbol{v}$ is not a prefix of $\boldsymbol{w}$ meaning that there exists an index $i$ such that $\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}^{\star}\right\rangle \neq 0$, the simulator can then extend $\boldsymbol{T}_{\boldsymbol{B}}$ to a short basis for the entire lattice $\Lambda_{q}^{\perp}\left(\boldsymbol{B}_{\boldsymbol{v}}\right)$. The simulator can now generate short vectors in $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$ using algorithm SampleRight, which is sufficient for constructing a short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$, as required. When $\boldsymbol{v}$ is a prefix of $\boldsymbol{w}$, meaning that for each $i=1, \ldots, t,\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}^{\star}\right\rangle=0$, then the matrix $F_{\boldsymbol{v}}$ no longer depends on $\boldsymbol{B}$ and the simulator's trapdoor disappears. Consequently, the simulator can generate secret keys for all vectors other than prefixes of $\boldsymbol{w}^{\star}$. As we will see, for $\boldsymbol{w}^{\star}$ the simulator can produce a challenge ciphertext that helps it solve the given LWE challenge.

### 5.2.1 Sampling a random basis

To realize the above delegation mechanism and for the simulation we will need the following algorithms. Following, almost verbatim, ABB10, CHKP10], let $\Lambda$ be an $m$-dimensional lattice and let $\mathcal{O}(\Lambda, \sigma)$ be an algorithm that generates independent samples from a distribution statistically close to $\mathcal{D}_{\Lambda, \sigma}$. The following algorithm called SampleBasis ${ }^{\mathcal{O}}(\Lambda, \sigma)$ uses $\mathcal{O}$ to generate a basis T of $\Lambda$ in the following way:

1. For $i=1, \ldots, m$, generate $v \leftarrow \mathcal{O}(\Lambda, \sigma)$, if $v$ is independent of $\left\{v_{1}, \ldots, v_{i-1}\right\}$, set $v_{i} \leftarrow v$, if not, repeat.
2. Convert the set of independent vectors $v_{1}, \ldots, v_{m}$ to a basis $T$ using Lemma 3.3.8 (and using some canonical basis of $\Lambda$ ) and output $T$.

The following theorem summarizes properties of this algorithm.
Lemma 5.2.1 For $\sigma>\tilde{b} l(\Lambda) \omega(\sqrt{\log m})$ algorithm SampleBasis ${ }^{\mathcal{O}}(\Lambda, \sigma)$ satisfies the following properties:

1. Step 1 requires at most $O(m \log m)$ w.h.p and $2 m$ samples in expectation.
2. With overwhelming probability $\|\widetilde{T}\| \leq\|T\| \leq \sigma \sqrt{m}$.
3. Up to a statistical distance, the distribution of T does not depend on the implementation of $\mathcal{O}$. That is, the random variable $\operatorname{SampleBasis}^{\mathcal{O}}(\Lambda, \sigma)$ is statistically close to SampleBasis $\mathcal{O}^{\prime}(\Lambda, \sigma)$ for any algorithm $\mathcal{O}^{\prime}$ that samples from a distribution statistically close to $\mathcal{D}_{\Lambda, \sigma}$.

By using the above SampleBasis algorithm, we define the following two algorithms. The SampleBasisLeft algorithm will be used in the real scheme, instead the SampleBasisRight will be used in our proof of security. Specifically:

Algorithm SampleBasisLeft. Given the lattice $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$ where $F_{\boldsymbol{v}}$ is defined as in Equation 5.1 for $\boldsymbol{v}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$, write $F_{\boldsymbol{v}}=(\boldsymbol{A} \mid \boldsymbol{M})$ for some matrices $\boldsymbol{A}$ and $\boldsymbol{M}$.
Then given a short basis $\boldsymbol{T}_{\boldsymbol{A}}$ for $\Lambda_{q}^{\perp}(\boldsymbol{A})$ we can implement algorithm $\mathcal{O}\left(F_{\boldsymbol{v}}, \sigma\right)$ as

$$
\mathcal{O}\left(F_{\boldsymbol{v}}, \sigma\right):=\text { SampleLeft }\left(\boldsymbol{A}, \boldsymbol{M}, \boldsymbol{T}_{\boldsymbol{A}}, 0, \sigma\right)
$$

When $\sigma>\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{A}}}\right\| \cdot \omega(\sqrt{\log ((t+1) m)})$, Theorem 3.3.10 shows that the resulting vector is distributed statistically close to $\mathcal{D}_{\Lambda_{q}\left(F_{v}\right), \sigma}$ as required for SampleBasis. Using the above algorithm in algorithm SampleBasis leads to an algorithm to sample a random basis of $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$ given a short basis of $\boldsymbol{A}$. We refer to this algorithm as SampleBasisLeft and summarize its properties in the following corollary.

Corollary 5.2.2 Algorithm SampleBasisLeft $\left(\boldsymbol{A}, \boldsymbol{M}, \boldsymbol{T}_{\boldsymbol{A}}, \sigma\right)$ outputs a basis of $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$ satisfying the three properties in Lemma 5.2 .1 provided that $\boldsymbol{A}$ is rank $n$ and $\sigma>\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{A}}}\right\|$. $\omega(\sqrt{\log ((t+1) m)})$.

Algorithm SampleBasisRight. In the simulation, the matrix $F_{\boldsymbol{v}}$ is defined as in Equation 5.2 In this case, given a short basis $\boldsymbol{T}_{\boldsymbol{B}}$ for $\Lambda_{q}^{\perp}(\boldsymbol{B})$ we can implement algorithm $\mathcal{O}\left(F_{\boldsymbol{v}}, \sigma\right)$ as follows:

1. Using Theorem 3.3.9, extend basis $\boldsymbol{T}_{\boldsymbol{B}}$ for $\Lambda_{q}^{\perp}(\boldsymbol{B})$ to a basis $\boldsymbol{T}_{\boldsymbol{B}_{\boldsymbol{v}}}$ for $\Lambda_{q}^{\perp}\left(\boldsymbol{B}_{\boldsymbol{v}}\right)$ such $\left\|\widetilde{\boldsymbol{T}_{B_{v}}}\right\|=\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{B}}}\right\|$.
2. Then run SampleRight $\left(\boldsymbol{A}, \boldsymbol{B}_{\boldsymbol{v}}, \overline{\boldsymbol{R}}, \boldsymbol{T}_{\boldsymbol{B}_{\boldsymbol{v}}}, 0, \sigma\right)$ and output the result. When $\boldsymbol{B}_{\boldsymbol{v}}$ is rank $n$ and $\boldsymbol{v}$ is not a prefix of $\boldsymbol{w}^{\star}$ the matrix $\boldsymbol{B}_{\boldsymbol{v}}$ is rank $n$ as required for SampleRight.

Let $s_{\overline{\boldsymbol{R}}}:=\|\overline{\boldsymbol{R}}\|$ be the norm of the matrix $\overline{\boldsymbol{R}}$. Then, when $\sigma>\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{B}}}\right\| \cdot s_{\overline{\boldsymbol{R}}} \cdot \omega(\sqrt{\log ((t+1) m)})$, Theorem 3.3 .11 shows that the resulting vector is distributed statistically close to $\mathcal{D}_{\Lambda_{q}^{\perp}\left(F_{v}\right), \sigma}$ as required for SampleBasis. Using the above algorithm in algorithm SampleBasis leads to an algorithm to sample a random basis of $\Lambda_{q}^{\perp}\left(F_{v}\right)$ for $F_{\boldsymbol{v}}$ defined in 5.2 given a short basis of $\boldsymbol{B}$. We refer to this algorithm as SampleBasisRight and summarize its properties in the following corollary.

Corollary 5.2.3 Algorithm SampleBasisRight $\left(\boldsymbol{A}, \boldsymbol{B}_{\boldsymbol{v}}, \overline{\boldsymbol{R}}, \boldsymbol{T}_{\boldsymbol{B}}, \sigma\right)$ outputs a basis of $\mathcal{D}_{\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right), \sigma}$ satisfying the three properties in Lemma 5.2.1 provided that $\boldsymbol{B}$ is rank $n$, that $\boldsymbol{v}$ is not a prefix of $\boldsymbol{w}^{\star}$ and $\sigma>\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{B}}}\right\| \cdot s_{\overline{\boldsymbol{R}}} \cdot \omega(\sqrt{\log ((t+1) m)})$.

### 5.2.2 Our Construction

Let $n>0$ be the security parameter and $\boldsymbol{\mu}=\left(\ell, d ; \mu_{1}, \ldots, \mu_{d}\right)$ the hierarchical format. Let $q=q(n, \boldsymbol{\mu})$ and $m=m(n, \boldsymbol{\mu})$ be positive integers. Let $r=r(n, \boldsymbol{\mu}) \geq 2$ be and integer and define $k=k(n, \boldsymbol{\mu}):=\left\lfloor\log _{r} q\right\rfloor$. Our hierarchical inner-product encryption for hierarchical format $\boldsymbol{\mu}$ consists of the following algorithms.
$\operatorname{Setup}\left(1^{n}, \boldsymbol{\mu}\right)$ : On input a security parameter $n$, and an hierarchical format of depth $d \boldsymbol{\mu}=$ $\left(\ell, d ; \mu_{1}, \ldots, \mu_{d}\right)$, the algorithm generates public and secret parameters as follows: Use algorithm $\operatorname{TrapGen}(q, n, m)$ to select a uniformly random $n \times m$-matrix $\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times m}$ with a basis $\boldsymbol{T}_{\boldsymbol{A}} \in \mathbb{Z}^{m \times m}$ for $\Lambda_{q}^{\perp}(\boldsymbol{A})$ such that $\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{A}}}\right\| \leq O(\sqrt{n \log q})$. For $i \in[d], j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$, choose uniformly random matrices $\boldsymbol{A}_{i, j, \gamma} \in \mathbb{Z}_{q}^{n \times m}$. Select a uniformly random vector $\boldsymbol{u} \in \mathbb{Z}_{q}^{n}$. Output

$$
m p k=\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\}, \boldsymbol{u}\right)
$$

and

$$
m s k=\boldsymbol{T}_{\boldsymbol{A}}
$$

Derive $\left(m p k, s k_{\boldsymbol{v}}, \boldsymbol{v}_{t}\right)$ : On input the master public key $m p k$, the secret key for the vector $\boldsymbol{v}=$ $\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t-1}\right)$, and the vector $\boldsymbol{v}_{t}$, the algorithm generates a secret key for the vector $\boldsymbol{v}^{\prime}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$ as follows: Construct short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}^{\prime}}\right)$ by invoking

$$
\boldsymbol{S} \leftarrow \text { SampleBasisLeft }\left(F_{\boldsymbol{v}}, \sum_{j \in\left[\mu_{i}\right]} \sum_{\gamma \in k} v_{t, j, \gamma} \cdot \boldsymbol{A}_{t, j, \gamma}, s k_{\boldsymbol{v}}, \sigma_{t}\right)
$$

where

$$
F_{\boldsymbol{v}^{\prime}}=\left[F_{\boldsymbol{v}} \| \sum_{j \in\left[\mu_{i}\right]} \sum_{\gamma=0}^{k} v_{t, j, \gamma} \cdot \boldsymbol{A}_{t, j, \gamma}\right]
$$

and $s k_{\boldsymbol{v}}$ is a short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$. Then, output $s k_{\boldsymbol{v}^{\prime}}=\boldsymbol{S}$.
By Corollary 5.2.2, when $\sigma_{t}>\left\|\widetilde{s k_{\boldsymbol{v}}}\right\| \cdot \omega(\sqrt{\log ((t+1) m)})$ then $\left\|\widetilde{s k_{\boldsymbol{v}^{\prime}}}\right\| \leq\left\|s k_{\boldsymbol{v}^{\prime}}\right\| \leq$ $\sigma_{t} \cdot \sqrt{(t+1) m}$.
Notice that, for the special case of the first level secret keys when $v$ is the empty vector $\epsilon$, we define $F_{\epsilon}:=\boldsymbol{A}$ and $s k_{\boldsymbol{v}}=m s k$.
$\operatorname{Enc}(m p k, \boldsymbol{w}, \mathfrak{m})$ : On input the master public key $m p k$, the vector $\boldsymbol{w}=\left(\boldsymbol{w}_{1}, \ldots, \boldsymbol{w}_{t}\right)$, and the message $\mathfrak{m} \in\{0,1\}$, the algorithm generates a ciphertext Ct as follows: The algorithm chooses a uniformly random matrix $\boldsymbol{B} \leftarrow \mathbb{Z}_{q}^{n \times m}$ and $s \leftarrow \mathbb{Z}_{q}^{n}$, a noise vector $\boldsymbol{x} \leftarrow \bar{\Psi}_{\alpha_{t}}^{m}$ and a noise term $x \leftarrow \bar{\Psi}_{\alpha_{t}}$. Then, the algorithm computes

$$
\boldsymbol{c}_{0}=\boldsymbol{A}^{\top} \boldsymbol{s}+\boldsymbol{x} \in \mathbb{Z}_{q}^{m} .
$$

For $i \in[t], j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$, the algorithm chooses a random matrix $\boldsymbol{R}_{i, j, \gamma} \in$ $\{-1,1\}^{m \times m}$ and computes

$$
\boldsymbol{c}_{i, j, \gamma}=\left(\boldsymbol{A}_{i, j, \gamma}+r^{\gamma} w_{i, j} \boldsymbol{B}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\top} \boldsymbol{x} \in \mathbb{Z}_{q}^{m} .
$$

Finally, the algorithm computes

$$
c_{1}=\boldsymbol{u}^{\top} \boldsymbol{s}+x+\mathfrak{m} \cdot\lfloor q / 2\rceil \in \mathbb{Z}_{q},
$$

and outputs $\mathrm{Ct}=\left(\boldsymbol{c}_{0},\left\{\boldsymbol{c}_{i, j, \gamma}\right\}, c_{1}\right)$.
$\operatorname{Dec}\left(m p k, s k_{\boldsymbol{v}}, C\right)$ : On input the master public key $m p k$, the secret key for the vector $\boldsymbol{v}=$ $\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$, and a ciphertext $\mathrm{Ct}=\left(\boldsymbol{c}_{0},\left\{\boldsymbol{c}_{i, j, \gamma}\right\}, c_{1}\right)$, the algorithm does the following: For $i \in[t]$, define the $r$-ary expansion of the vector $\boldsymbol{v}_{i}$, then the algoirthm computes

$$
\boldsymbol{c}_{\boldsymbol{v}_{i}}=\sum_{j \in\left[\mu_{i}\right]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \cdot \boldsymbol{c}_{i, j, \gamma} .
$$

Let $\boldsymbol{c}=\left[\boldsymbol{c}_{0}\left\|\boldsymbol{c}_{\boldsymbol{v}_{1}}\right\| \ldots \| \boldsymbol{c}_{\boldsymbol{v}_{t}}\right]$, then the algorithm sets $\tau_{t}:=\sigma_{t} \cdot \sqrt{(t+1) m} \cdot \omega(\sqrt{(t+1) m})$ and $\tau_{t} \geq\left\|s k_{\boldsymbol{v}}\right\| \cdot \omega(\sqrt{(t+1) m})$, and computes

$$
e_{\boldsymbol{v}}=\operatorname{SamplePre}\left(F_{\boldsymbol{v}}, s k_{\boldsymbol{v}}, \boldsymbol{u}, \tau_{t}\right)
$$

Finally, the algorithm computes

$$
z=c_{1}-\boldsymbol{e}_{\boldsymbol{v}}^{\top} \cdot \boldsymbol{c},
$$

and interpreters $z$ as in integer in $(-q / 2, q / 2]$. Then the algorithm outputs 0 if $|z|<q / 4$ and 1 otherwise.

With the following lemma, we show the correctness of our scheme.
Lemma 5.2.4 For hierarchical format $\boldsymbol{\mu}=\left(\ell, d ; \mu_{1}, \ldots, \mu_{d}\right)$ of depth $d$, suppose the parameters $q$ and $\alpha_{t}$, for each $t \in[d]$, are such that
$q / \log q=\Omega\left(\sigma_{t} \cdot \mu \cdot \frac{r}{\log r} \cdot m^{3 / 2}\right)$ and $\alpha_{t} \leq\left(t \cdot \log q \cdot \sigma_{t} \cdot \mu \cdot \frac{r}{\log r} \cdot m \cdot \omega(\sqrt{\log m})\right)^{-1}$,
where $\mu=\max _{i \in[d]} \mu_{i}$. Moreover, for vectors $\boldsymbol{v}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$ and $\boldsymbol{w}=\left(\boldsymbol{w}_{1}, \ldots, \boldsymbol{w}_{h}\right)$, let $s k_{v}$ be the basis of the lattice $F_{v}$ obtained through a sequence of calls to the Derive algorithm, let $\mathrm{Ct} \leftarrow \operatorname{Enc}(m p k, \boldsymbol{w}, \mathfrak{m})$ and $\mathfrak{m}^{\prime} \leftarrow \operatorname{Dec}\left(m p k, s k_{\boldsymbol{v}}, C\right)$.

Then, if $f_{\boldsymbol{v}}(\boldsymbol{w})=1$, namely $\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}\right\rangle=0(\bmod q)$ for each $i \in[t]$, then with overwhelming probability we have $\mathfrak{m}^{\prime}=\mathfrak{m}$.

Proof: Just for notational convenience, let

$$
\widetilde{\boldsymbol{A}_{i}}=\left(\sum_{j \in[\mu]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \boldsymbol{A}_{i, j, \gamma}\right) \quad \text { and } \quad \widetilde{\boldsymbol{R}_{i}}=\sum_{j \in[\mu]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \boldsymbol{R}_{i, j, \gamma}^{\top} \boldsymbol{x} .
$$

Then, for each $i \in[t]$, the decryption algorithm computes:

$$
\begin{aligned}
\boldsymbol{c}_{\boldsymbol{v}_{i}} & =\sum_{j \in[\mu]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \cdot \boldsymbol{c}_{i, j, \gamma} \\
& =\sum_{j \in[\mu]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \cdot\left[\left(\boldsymbol{A}_{i, j, \gamma}+r^{\gamma} w_{i, j} \boldsymbol{B}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\top} \boldsymbol{x}\right] \\
& =\widetilde{\boldsymbol{A}}_{i}^{\top} \boldsymbol{s}+\underbrace{\left(\sum_{j \in[\mu]} \sum_{\gamma=0}^{k} r^{\gamma} v_{i, j, \gamma} w_{i, j}\right)}_{\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}\right\rangle} \boldsymbol{B}^{\top} \boldsymbol{s}+\widetilde{\boldsymbol{R}}_{i}
\end{aligned}
$$

If $\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}\right\rangle=0$ then we have:

$$
\boldsymbol{c}_{\boldsymbol{v}_{i}}=\widetilde{\boldsymbol{A}}_{i}^{\top} \boldsymbol{s}+\widetilde{\boldsymbol{R}}_{i} \quad(\bmod q) .
$$

Thus, if for each $i \in[t],\left\langle\boldsymbol{v}_{i}, \boldsymbol{w}_{i}\right\rangle=0$ then $\boldsymbol{c}$ can be written as:

$$
\begin{aligned}
\boldsymbol{c} & =\left[\boldsymbol{c}_{0}\left\|\boldsymbol{c}_{\boldsymbol{v}_{1}}\right\| \ldots \| \boldsymbol{c}_{\boldsymbol{v}_{t}}\right] \\
& =\left[\boldsymbol{A}\left\|\widetilde{\boldsymbol{A}_{1}}\right\| \ldots \| \widetilde{\boldsymbol{A}_{t}}\right]^{\top} s+\left[\boldsymbol{x}\left\|\widetilde{\boldsymbol{R}_{1}}\right\| \ldots \| \widetilde{\boldsymbol{R}}_{t}\right]^{\top}(\bmod q) \\
& =F_{\boldsymbol{v}}^{\top} \cdot \boldsymbol{s}+\left[\boldsymbol{x}\left\|\widetilde{\boldsymbol{R}_{1}}\right\| \ldots \| \widetilde{\boldsymbol{R}_{t}}\right]^{\top}(\bmod q)
\end{aligned}
$$

At this point, the short vector $\boldsymbol{e}_{\boldsymbol{v}}=\operatorname{SampleLeft}\left(F_{\boldsymbol{v}}, s k_{\boldsymbol{v}}, \boldsymbol{u}, \tau_{t}\right)$ is computed by the decryption algorithm such that by Theorem 3.3.10, $F_{\boldsymbol{v}} \cdot \boldsymbol{e}_{\boldsymbol{v}}=\boldsymbol{u}(\bmod q)$. It follows that

$$
\boldsymbol{e}_{\boldsymbol{v}}^{\top} \boldsymbol{c}=\boldsymbol{u}^{\top} \boldsymbol{s}+\boldsymbol{e}_{\boldsymbol{v}}^{\top}\left[\boldsymbol{x}\left\|\widetilde{\boldsymbol{R}}_{1}\right\| \ldots \| \widetilde{\boldsymbol{R}}_{t}\right]^{\top}(\bmod q)
$$

Finally, the decryption algorithm computes:

$$
\begin{aligned}
z & =c_{1}-\boldsymbol{e}_{\boldsymbol{v}}^{\top} \boldsymbol{c} \quad(\bmod q) \\
& =\left(\boldsymbol{u}^{\top} \boldsymbol{s}+x+\mathfrak{m} \cdot\lfloor q / 2\rceil\right)-\boldsymbol{u}^{\top} \boldsymbol{s}-\boldsymbol{e}_{\boldsymbol{v}}^{\top}\left[\boldsymbol{x}\left\|\widetilde{\boldsymbol{R}_{1}}\right\| \ldots \| \widetilde{\boldsymbol{R}}_{t}\right]^{\top}(\bmod q) \\
& =\mathfrak{m} \cdot\lfloor q / 2\rceil+\underbrace{\left(x-\boldsymbol{e}_{\boldsymbol{v}}^{\top}\left[\boldsymbol{x}\left\|\widetilde{\boldsymbol{R}_{1}}\right\| \ldots \| \widetilde{\boldsymbol{R}_{t}}\right]^{\top}\right)}_{\text {noise term }}(\bmod q) .
\end{aligned}
$$

Thus, to have a successful decryption, it suffices to set the parameters so that with overwhelming probability,

$$
\left|x-\boldsymbol{e}_{\boldsymbol{v}}^{\top}\left[\boldsymbol{x}\left\|\widetilde{\boldsymbol{R}_{1}}\right\| \ldots \| \widetilde{\boldsymbol{R}_{t}}\right]^{\top}\right|<\frac{q}{4}
$$

Let us write $\boldsymbol{e}_{\boldsymbol{v}}=\left[e_{\boldsymbol{v}, 0}\left\|\boldsymbol{e}_{\boldsymbol{v}, 1}\right\| \ldots \| e_{\boldsymbol{v}, t}\right]$ with $\boldsymbol{e}_{\boldsymbol{v}, i} \in \mathbb{Z}^{m}$ for $i=0, \ldots, t$. Then the noise term can be rewritten as

$$
x-\left(\boldsymbol{e}_{\boldsymbol{v}, 0}+\sum_{i \in[t]} \sum_{j \in[\mu]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \boldsymbol{R}_{i, j, \gamma} \boldsymbol{e}_{\boldsymbol{v}, i}\right)^{\top} \boldsymbol{x} .
$$

By Lemma 3.3.1, we have $\left\|e_{\boldsymbol{v}}\right\|<\tau_{t} \sqrt{(t+1) m}$ with overwhelming probability. Moreover by Lemma 3.3.2, we have $\left\|\boldsymbol{R}_{i, j, \gamma} \cdot \boldsymbol{e}_{\boldsymbol{v}, i}\right\| \leq 12 \sqrt{2 m} \cdot\left\|\boldsymbol{e}_{\boldsymbol{v}, i}\right\|$ with overwhelming probability, and since $v_{i, j, \gamma} \in[0, r-1]$ it follows that

$$
\left\|\boldsymbol{e}_{\boldsymbol{v}, 0}+\sum_{i \in[t]} \sum_{j \in[\mu]} \sum_{\gamma=0}^{k} v_{i, j, \gamma} \boldsymbol{R}_{i, j, \gamma} \boldsymbol{e}_{\boldsymbol{v}, i}\right\|=O\left(t \cdot \mu \cdot k \cdot r \cdot \sigma_{t} \cdot m\right) .
$$

Finally, by Lemma 3.3.16, the error term has absolute value at most:

$$
\left(q \alpha_{t} \cdot \omega(\sqrt{\log m})+\sqrt{m} / 2\right) \cdot O\left(t \cdot \mu \cdot k \cdot r \cdot \sigma_{t} \cdot m\right)
$$

## I

### 5.3 Proofs of Security

In this section we prove the following theorem.
Theorem 5.3.1 If the decision- $\mathrm{LWE}_{q, n, \chi}$ problem is infeasible (Definition 3.3.12), then our HIPE scheme is wAH-sIND-CPA-secure.

Following [ABB10, AFV11], we define additional algorithms. These will not be used in the real scheme, but we need them in our proofs.
$\operatorname{Sim} . \operatorname{Setup}\left(1^{n}, \boldsymbol{\mu}, \boldsymbol{w}^{\star}\right):$ The main difference with the normal setup algorithm is that now the simulator does not generate any more a trapdoor for the matrix $\boldsymbol{A}$. Instead, a trapdoor for a matrix $\boldsymbol{B}$, that will be embedded into the $\boldsymbol{A}_{i, j, \gamma}$ matrices, is generated. Furthermore, the simulator embeds the challenge vector $\boldsymbol{w}^{\star}$ in the public parameters in such a way it is still possible to generate a ciphertext for $\boldsymbol{w}^{\star}$ but no secret key can be generate for vectors $\boldsymbol{v}$ such that $f_{v}\left(\boldsymbol{w}^{\star}\right)=1$.
Specifically, the algorithm chooses random $\boldsymbol{A}, \boldsymbol{R}_{i, j, k}^{\star}$ and $\boldsymbol{u}$, it uses TrapGen to generate $\boldsymbol{B}^{\star}$ and defines

$$
\boldsymbol{A}_{i, j, \gamma}=\boldsymbol{A} \boldsymbol{R}_{i, j \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star} .
$$

Specifically, on input a security parameter $n$, an hierarchical format $\boldsymbol{\mu}=\left(\ell, d ; \mu_{1}, \ldots\right.$, $\left.\mu_{d}\right)$ of depth $d$, and a challenge vector $\boldsymbol{w}^{\star}=\left(\boldsymbol{w}_{1}^{\star}, \ldots, \boldsymbol{w}_{d}^{\star}\right)$, the algorithm generates public and secret parameters as follows: Choose random matrix $\boldsymbol{A} \in \mathbb{Z}_{q}^{n \times m}$. For $i \in[d]$, $j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$, choose uniformly random matrices $\boldsymbol{R}_{i, j, \gamma}^{\star} \in \mathbb{Z}_{q}^{n \times m}$. Select a uniformly random vector $\boldsymbol{u} \in \mathbb{Z}_{q}^{n}$. Use algorithm $\operatorname{TrapGen}(q, n, m)$ to select a uniformly random $n \times m$-matrix $\boldsymbol{B}^{\star} \in \mathbb{Z}_{q}^{n \times m}$ with a basis $\boldsymbol{T}_{\boldsymbol{B}^{\star}} \in \mathbb{Z}^{m \times m}$ for $\Lambda_{q}^{\perp}\left(\boldsymbol{B}^{\star}\right)$ such that $\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{B}^{\star}}}\right\| \leq O(\sqrt{n \log q})$. For $i \in[d], j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$, set

$$
\boldsymbol{A}_{i, j, \gamma}=\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star} .
$$

Output $m p k=\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\}, \boldsymbol{u}\right)$ and $m s k=\left(\left\{\boldsymbol{R}_{i, j, \gamma}^{\star}\right\}, \boldsymbol{B}^{\star}, \boldsymbol{T}_{\boldsymbol{B}^{\star}}\right)$.

Sim.Derive $\left(m p k, s k_{\boldsymbol{v}}, \boldsymbol{v}_{t}\right)$ : Secret keys are now created by using the trapdoor $\boldsymbol{T}_{\boldsymbol{B}^{\star}}$, sampled by Sim.Setup, and the SampleBasisRight algorithm.

Specifically, on input the master public key $m p k$, the secret key for the vector $\boldsymbol{v}=$ $\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t-1}\right)$, and the vector $\boldsymbol{v}_{t}$, the algorithm generates a secret key for the vector $\boldsymbol{v}^{\prime}=\left(\boldsymbol{v}_{1}, \ldots, \boldsymbol{v}_{t}\right)$ by constructing a short basis for $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}^{\prime}}\right)$, as defined by Equation 5.2 , by invoking

$$
\boldsymbol{S} \leftarrow \text { SampleBasisRight }\left(\boldsymbol{A}, \boldsymbol{B}_{\boldsymbol{v}^{\prime}}^{\star}, \overline{\boldsymbol{R}^{\star}}, \boldsymbol{T}_{\boldsymbol{B}^{\star}}, \sigma_{t}\right)
$$

Output $s k_{\boldsymbol{v}^{\prime}}=\boldsymbol{S}$.
Notice that if the simulator tries to generate a secret key for a vector $\boldsymbol{v}$ such that $f_{\boldsymbol{v}}\left(\boldsymbol{w}^{\star}\right)=$ 1 then the simulator's trapdoor disappears meaning that the simulator cannot generate such a secret key.

Sim.Enc $(m p k, \boldsymbol{w}, \mathfrak{m})$ : The algorithm differs from Enc in the sense that it uses matrices $\boldsymbol{R}_{i, j, \gamma}^{\star}$ and $\boldsymbol{B}^{\star}$ instead of matrices $\boldsymbol{R}_{i, j, \gamma}$ and $\boldsymbol{B}$.
Specifically, on input master public key $m p k$, vector $\boldsymbol{w}=\left(\boldsymbol{w}_{1}, \ldots, \boldsymbol{w}_{t}\right)$, and message $\mathfrak{m} \in\{0,1\}$, the algorithm generates a ciphertext $C t$ as follows: The algorithm chooses a uniformly random vector $s \leftarrow \mathbb{Z}_{q}^{n}$, a noise vector $\boldsymbol{x} \leftarrow \bar{\Psi}_{\alpha_{t}}^{m}$ and a noise term $x \leftarrow \bar{\Psi}_{\alpha_{t}}$. Then, the algorithm computes

$$
\boldsymbol{c}_{0}=\boldsymbol{A}^{\top} \boldsymbol{s}+\boldsymbol{x} \in \mathbb{Z}_{q}^{m}
$$

and, for $i \in[t], j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$,

$$
\boldsymbol{c}_{i, j, \gamma}=\left(\boldsymbol{A}_{i, j, \gamma}+r^{\gamma} w_{i, j} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x} \in \mathbb{Z}_{q}^{m}
$$

Finally, the algorithm computes

$$
c_{1}=\boldsymbol{u}^{\top} \boldsymbol{s}+x+\mathfrak{m} \cdot\lfloor q / 2\rceil \in \mathbb{Z}_{q}
$$

and outputs $\mathrm{Ct}=\left(\boldsymbol{c}_{0},\left\{\boldsymbol{c}_{i, j, \gamma}\right\}, c_{1}\right)$.
We now turn on the proof of security that follows the same line of that of Agrawal et al. AFV11 adapted to new hierarchical setting.

Specifically, for a probabilistic polynomial-time adversary $\mathcal{A}$, our proof of security will consist of the following sequence of 6 games between $\mathcal{A}$ and $\mathcal{C}$ showing that the encryption of the pair $\left(\boldsymbol{w}_{0}, \mathfrak{m}_{0}\right)$ is computationally indistinguishable from the encryption of the pair $\left(\boldsymbol{w}_{1}, \mathfrak{m}_{1}\right)$.

The six games are defined as follows:
Game $_{0}: \mathcal{C}$ runs the Setup algorithm, answers $\mathcal{A}$ 's secret key queries using the Derive algorithm, and generates the challenge ciphertext for vector $\boldsymbol{w}_{0}$ and message $\mathfrak{m}_{0}$ by using the Enc algorithm.

Game $_{1}$ : In this game $\mathcal{C}$ uses the simulation algorithms. Specifically, $\mathcal{C}$ runs the Sim.Setup algorithm with $\boldsymbol{w}^{\star}=\boldsymbol{w}_{0}$, answers $\mathcal{A}$ 's secret key queries using the Sim.Derive algorithm, and generates the challenge ciphertext for vector $\boldsymbol{w}_{0}$ and message $\mathfrak{m}_{0}$ by using the Sim.Enc algorithm .

Game $_{2}$ : It is the same as the Game Gexcept that the challenge ciphertext is randomly chosen $^{\text {en }}$ from the ciphertext space.

Game $_{3}:$ It is the same as the Game $_{2}$ except that $\mathcal{C}$ runs the Sim.Setup algorithm with $\boldsymbol{w}^{\star}=$ $\boldsymbol{w}_{1}$.

Game $_{4}$ : It is the same as the Game $_{\mathbf{3}}$ except that $\mathcal{C}$ generates the challenge ciphertext for vector $\boldsymbol{w}_{1}$ and message $\mathfrak{m}_{1}$ by using the Sim. Enc algorithm

Game $_{5}: \mathcal{C}$ runs the Setup algorithm, answers $\mathcal{A}$ 's secret key queries using the Derive algorithm, and generates the challenge ciphertext for vector $\boldsymbol{w}_{1}$ and message $\mathfrak{m}_{1}$ by using the Enc algorithm.

We give now formal proofs of the above statements.
Lemma 5.3.2 The view of the adversary $\mathcal{A}$ in $\mathbf{G a m e}_{\mathbf{0}}$ (resp. Game ${ }_{4}$ ) is statistically close to the view of $\mathcal{A}$ in Game $_{1}$ (resp. Game ${ }_{5}$ ).

Proof: We will show that Game $_{0}$ is statically close to Game $_{1}$. The same proof also applies to Game $_{4}$ and Game ${ }_{5}$.

Setup: In Game $_{\mathbf{0}}$, matrix $\boldsymbol{A}$ is generated by TrapGen and, for each $i \in[d], j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$, matrix $\boldsymbol{A}_{i, j, \gamma}$ is uniformly random in $\mathbb{Z}_{q}^{n \times m}$.
Instead, in Game $_{\mathbf{1}} \boldsymbol{A}$ is chosen uniformly at random and, for each $i \in[d], j \in\left[\mu_{i}\right]$ and $\gamma=0, \ldots, k$, we have $\boldsymbol{A}_{i, j, \gamma}=\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}$, where $\boldsymbol{B}^{\star}$ is generated by TrapGen and the matrices $\boldsymbol{R}_{i, j, \gamma}^{\star}$ are uniformly and independently chosen at random in $\{-1,1\}^{m \times m}$.
Moreover, in both the games the vector $\boldsymbol{u}$ is chosen at random in $\mathbb{Z}_{q}^{n}$. Notice that, by Theorem 3.3.8, since $m \geq 6 n \log q$, the matrixes $\boldsymbol{A}$ and $\boldsymbol{B}^{\star}$ output by TrapGen are statistically indistinguishable from a uniformly random matrix.

Secret keys: Assuming, for each level $t \in[d]$, a sufficiently large $\sigma_{t}$, then in Game ${ }_{\mathbf{0}}$, the secret key for vector $\boldsymbol{v}, s k_{\boldsymbol{v}}$, is a basis of $\Lambda_{q}^{\perp}\left(F_{\boldsymbol{v}}\right)$, sampled by using the SampleBasisLeft algorithm, satisfying the properties in Lemma 5.2.1. The same happens in Game ${ }_{1}$ by using SampleBasisRight algorithm this time. Moreover, Lemma 5.2.1, guarantees the independence from the derivation path. Thus, the secret keys have the same distribution in both games.

Challenge: In Game ${ }_{0}$, the challenge ciphertext components $\boldsymbol{c}_{i, j, \gamma}$ are computed as follow:

$$
\boldsymbol{c}_{i, j, \gamma}=\left(\boldsymbol{A}_{i, j, \gamma}+r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star}{ }^{\top} \boldsymbol{x} \in \mathbb{Z}_{q}^{m},
$$

where $\boldsymbol{B}^{\star}$ is uniformly random in $\mathbb{Z}_{q}^{n \times m}$ and the matrices $\boldsymbol{R}_{i, j, \gamma}^{\star}$ are uniformly and independently chosen at random in $\{-1,1\}^{m \times m}$.
On the other side, in Game ${ }_{1}$, we have:

$$
\begin{aligned}
\boldsymbol{c}_{i, j, \gamma} & =\left(\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}+r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star}{ }^{\top} \boldsymbol{x} \\
& =\boldsymbol{R}_{i, j, \gamma}^{\star}{ }^{\top}\left(\boldsymbol{A}^{\top} \boldsymbol{s}+\boldsymbol{x}\right)
\end{aligned}
$$

where, this time, the matrices $\boldsymbol{R}_{i, j, \gamma}^{\star}$ are the same used to construct the public parameters.
Let us now analyze the join distribution of the public parameters and the challenge ciphertext in $\boldsymbol{G a m e}_{\mathbf{0}}$ and $\mathbf{G a m e}_{\mathbf{1}}$. We will show that the distributions of $\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\},\left\{\boldsymbol{c}_{i, j, \gamma}\right\}\right)$ in $\mathbf{G a m e}_{\mathbf{0}}$ and in Game ${ }_{1}$ are statistical indistinguishable.

First notice that by ABB 10 , Lemma 13] we have that the following two distributions are statistically indistinguishable for every fixed matrix $\boldsymbol{B}^{\star}$, every $\boldsymbol{w}^{\star}$ and every vector $\boldsymbol{x} \in \mathbb{Z}_{q}^{m}$.

$$
\left(\boldsymbol{A}, \boldsymbol{A}_{i, j, \gamma}, \boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x}\right) \approx_{s}\left(\boldsymbol{A}, \boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}, \boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x}\right)
$$

Since each $\boldsymbol{R}_{i, j, \gamma}^{\star}$ is chosen independently for every $i, j$ and $\gamma$, then the joint distribution of them are statistically close:

$$
\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\},\left\{\boldsymbol{R}_{i, j, \gamma}^{\star \top} \boldsymbol{x}\right\}\right) \approx_{s}\left(\boldsymbol{A},\left\{\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right\},\left\{\boldsymbol{R}_{i, j, \gamma}^{\star \top} \boldsymbol{x}\right\}\right)
$$

Since $\left(\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}$ is statistically close to $\boldsymbol{A}_{i, j, \gamma}^{\top} \boldsymbol{s}$, it is possible to add each term to one side of the equation:

$$
\begin{array}{r}
\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\},\left\{\boldsymbol{A}_{i, j, \gamma}^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x}\right\}\right) \approx_{s} \\
\left(\boldsymbol{A},\left\{\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right\},\left\{\left(\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x}\right\}\right)
\end{array}
$$

Finally, we add $\left(r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}$ to both sides:

$$
\begin{aligned}
&\left(\boldsymbol{A},\left\{\boldsymbol{A}_{i, j, \gamma}\right\},\left\{\left(\boldsymbol{A}_{i, j, \gamma}+r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x}\right\}\right) \approx_{s} \\
&\left(\boldsymbol{A},\left\{\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right\},\left\{\left(\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x}\right\}\right)
\end{aligned}
$$

To conclude, observe that the distribution on the left hand side is that of the public parameters and the challenge ciphertext in $\mathrm{Game}_{\mathbf{0}}$, while that on the right hand side is the distribution in Game $_{1}$.

Lemma 5.3.3 The view of the adversary $\mathcal{A}$ in Game $_{\mathbf{1}}$ (resp. Game ${ }_{3}$ ) is computationally indistinguishable to the view of $\mathcal{A}$ in Game $_{2}$ (resp. Game ${ }_{4}$ ) under the decision-LWE problem.

Proof: We will show that Game $_{1}$ is computationally indistinguishable from Game ${ }_{2}$ by giving a reduction from the decision-LWE problem. The same proof applies also to Game $\mathbf{H}_{3}$ and Game $_{4}$.

Suppose $\mathcal{A}$ can distinguish between Game $_{1}$ and Game $_{2}$ with non-negligible advantage. Then, it is possible to use $\mathcal{A}$ to build an algorithm $\mathcal{B}$ to solve the decision-LWE problem.

Init: $\mathcal{B}$ is given $m+1 \operatorname{LWE}$ challenge pairs $\left(\boldsymbol{a}_{j}, y_{j}\right) \in \mathbb{Z}_{q}^{n} \times \mathbb{Z}_{q}$, where either $y_{j}=\left\langle\boldsymbol{a}_{j}, s\right\rangle+x_{j}$ for a random $s \in \mathbb{Z}_{q}^{n}$ and a noise term $x_{j} \leftarrow \Psi_{\alpha}$ or $y_{j}$ is uniformly random in $\mathbb{Z}_{q}$.

Setup: The public parameters are constructed using the vectors of the pairs $\left(\boldsymbol{a}_{j}, y_{j}\right)$. The $i$-th column of matrix $\boldsymbol{A}$ will be the vector $\boldsymbol{a}_{i}, 1 \leq i \leq m$ and vector $\boldsymbol{u}$ will be $\boldsymbol{a}_{0}$. The matrices $\boldsymbol{A}_{i, j, \gamma}$ are still computed as in Sim.Setup on input $\boldsymbol{w}^{\star}=\boldsymbol{w}_{0}$, i.e.,

$$
\boldsymbol{A}_{i, j, \gamma}=\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}
$$

Secret keys: All private-key extraction queries are answered using Sim.Derive.

Challenge: The ciphertext $\mathrm{Ct}=\left(\boldsymbol{c}_{0},\left\{c_{i, j, \gamma}\right\}, c_{1}\right)$ is constructed using the LWE challenge pairs ( $a_{j}, y_{j}$ ), as follows:

$$
\begin{gathered}
\boldsymbol{c}_{0}=\left(y_{1}, \ldots, y_{m}\right), \\
\boldsymbol{c}_{i, j, \gamma}=\boldsymbol{R}_{i, j, \gamma}^{\star \top} \boldsymbol{c}_{0},
\end{gathered}
$$

and

$$
c_{1}=y_{0}+\mathfrak{m} \cdot\lfloor q / 2\rceil .
$$

Then, If $y_{j}=\left\langle\boldsymbol{a}_{j}, s\right\rangle+x_{j}$, we have that

$$
\begin{aligned}
\boldsymbol{c}_{0} & =\left(y_{1}, \ldots, y_{m}\right) \\
& =\left(\left\langle\boldsymbol{a}_{1}, s\right\rangle+x_{1}, \ldots,\left\langle\boldsymbol{a}_{m}, s\right\rangle+x_{m}\right)=\boldsymbol{A}^{\top} \boldsymbol{s}+\boldsymbol{x} \\
\boldsymbol{c}_{i, j, \gamma} & =\boldsymbol{R}_{i, j, \gamma}^{\star \top} \boldsymbol{c}_{0} \\
& =\boldsymbol{R}_{i, j, \gamma}^{\star \top}\left(\boldsymbol{A}^{\top} \boldsymbol{s}+\boldsymbol{x}\right) \\
& =\left(\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star \top} \boldsymbol{x} \\
& =\left(\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}+r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}^{\star}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star} \boldsymbol{x} \\
& =\left(\boldsymbol{A}_{i, j, \gamma}+r^{\gamma} w_{i, j}^{\star} \boldsymbol{B}\right)^{\top} \boldsymbol{s}+\boldsymbol{R}_{i, j, \gamma}^{\star \top} \boldsymbol{x}
\end{aligned}
$$

and

$$
\begin{aligned}
c_{1} & =y_{0}+\mathfrak{m}\lfloor q / 2\rceil \\
& =\boldsymbol{u}^{\top} \boldsymbol{s}+x+\mathfrak{m} \cdot\lfloor q / 2\rceil
\end{aligned} .
$$

Therefore the challenge ciphertext is distributed exactly as in Game $\mathbf{1}_{\mathbf{1}}$.
Instead, if $y_{j}$ is uniformly random in $\mathbb{Z}_{q}$ then the challenge ciphertext is

$$
\left(\boldsymbol{c}_{0}, \overline{\boldsymbol{R}}^{\star \top} \boldsymbol{c}_{0}, c_{1}\right),
$$

where $\overline{\boldsymbol{R}}^{\star \top}$ is the concatenation of all the $\boldsymbol{R}_{i, j, \gamma}^{\star}$ matrices, and by the leftover hash lemma [Sho08, Theorem 8.38], $\boldsymbol{A} \overline{\boldsymbol{R}}^{\star}$ and $\overline{\boldsymbol{R}}^{\star \top} \boldsymbol{c}_{0}$ are independent uniformly random samples. Therefore, the ciphertext is uniformly random, as the ciphertext generated by Game ${ }_{2}$.

At the end $\mathcal{A}$ must guess if it is interacting with $\mathrm{Game}_{1}$ or $\mathrm{Game}_{2}$. Notice that, the answer to this guess is also the answer to the LWE challenge. We showed that if $y_{j}$ is uniformly random in $\mathbb{Z}_{q}$, than $\mathcal{A}$ 's view is the same as in $\mathbf{G a m e}_{\mathbf{2}}$ and if $y_{j}=\left\langle\boldsymbol{a}_{j}, s\right\rangle+x_{j}$, than $\mathcal{A}$ 's view is the same as in Game. Therefore, the $\mathcal{B}$ 's advantage in solving LWE is the same as $\mathcal{A}$ 's advantage in distinguish the views of Game ${ }_{1}$ and Game ${ }_{2}$.

Them, it remains to show that $\mathbf{G a m e}_{2}$ is indistinguishability from $\mathbf{G a m e}_{3}$.
Lemma 5.3.4 The view of the adversary $\mathcal{A}$ in $\mathrm{Game}_{2}$ is statistically indistinguishability to the view of $\mathcal{A}$ in $\mathrm{Game}_{3}$

Proof: We will show that $\mathbf{G a m e}_{2}$ is statically close to Game $_{3}$. The main difference between the two games is in the way the Sim.Setup algorithm is invoked. Specifically, in Game ${ }_{2}$, Sim.Setup is invoked on input $\boldsymbol{w}^{\star}=\boldsymbol{w}_{0}$. In Game $\boldsymbol{a}_{\boldsymbol{3}}, \boldsymbol{w}^{\star}=\boldsymbol{w}_{1}$ is used instead.

Notice that, all public parameters are randomly generated in both games, except for the matrices $\boldsymbol{A}_{i, j, \gamma}$. Therefore, the indistinguishability of Game $\mathbf{2}_{2}$ and $\mathbf{G a m e}_{3}$ only depends on the indistinguishability of the $\boldsymbol{A}_{i, j, \gamma}$ matrices that in both games they are generated as

$$
\boldsymbol{A}_{i, j, \gamma}=\boldsymbol{A} \boldsymbol{R}_{i, j, \gamma}^{\star}-r^{\gamma} \boldsymbol{w}_{i, j}^{\star} \boldsymbol{B}^{\star} .
$$

By Lemma 3.3.4, $\left(\boldsymbol{A}, \boldsymbol{A} \overline{\boldsymbol{R}}^{\star}\right)$ is statistically indistinguishable from $\left(\boldsymbol{A}, \boldsymbol{A}^{\prime}\right)$ where $\overline{\boldsymbol{R}}^{\star}$ is the concatenation of all the $\boldsymbol{R}_{i, j, \gamma}^{\star}$ matrices. Moreover, for any fixed $\boldsymbol{X}$ and uniformly random $\boldsymbol{C}$, the variable $\boldsymbol{C}-\boldsymbol{X}$ is also uniformly random. Therefore the distributions of $\boldsymbol{A}_{i, j, \gamma}$ in both games are statistically indistinguishable.

Wrapping up. We have proved by the previous lemmata that our hierarchical inner-product encryption is weak attribute hiding-selective attribute secure assuming decision- $\mathrm{LWE}_{q, n, \chi}$ problem is infeasible. To finalize we extract the parameters required for correctness and security of the system. Specifically:

- We need to ensure that for each $t \in[d]$, correctness holds. Specifically, Lemma 5.2.4 requires

$$
q / \log q=\Omega\left(\sigma_{t} \cdot \mu \cdot \frac{r}{\log r} \cdot m^{3 / 2}\right)
$$

and

$$
\alpha_{t} \leq\left(t \cdot \log q \cdot \sigma_{t} \cdot \mu \cdot \frac{r}{\log r} \cdot m \cdot \omega(\sqrt{\log m})\right)^{-1} .
$$

- By Theorem 3.3.8, algorithm TrapGen requires

$$
q>2 \text { and } m>6 n \lg q .
$$

- By Corollary 5.2.2, to have algorithm SampleBasisLeft working correctly in the Derive algorithm, we need for each $t \in[d]$,

$$
\sigma_{t}>\left\|\widetilde{s k_{\boldsymbol{v}}}\right\| \cdot \omega(\sqrt{\log ((t+1) m)})
$$

Thus, we have

$$
\sigma_{t}>\sigma_{\mathrm{TG}} \cdot \omega\left((\log m)^{t / 2}\right) .
$$

- By Corollary 5.2.3, to have algorithm SampleBasisRight working correctly in the Sim. Derive algorithm, we need for each $t \in[d]$,

$$
\sigma_{t}>\left\|\widetilde{\boldsymbol{T}_{\boldsymbol{B}}}\right\| \cdot s_{\overline{\boldsymbol{R}}} \cdot \omega(\sqrt{\log ((t+1) m)}) .
$$

Thus, by Theorem 3.3.8,

$$
\left\|\widetilde{\boldsymbol{T}_{B}}\right\|<\sigma_{\mathrm{TG}}
$$

and, by Lemma 3.3.2,

$$
s_{\overline{\boldsymbol{R}}}=\|\overline{\boldsymbol{R}}\|=O\left(t \cdot \mu \cdot\left(\log _{r} q+1\right) \cdot \sqrt{(t+1) m}\right)
$$

due the particular structure of $\overline{\boldsymbol{R}}$, where $\mu=\max _{i \in[d]} \mu_{i}$. Thus,

$$
\sigma_{t}>O(\sqrt{n \log q}) \cdot O\left(\mu \cdot\left(\log _{r} q+1\right) \cdot \sqrt{(t+1) m}\right) \cdot \omega(\sqrt{(t+1) m}) .
$$

- Finally, Regev's reduction must apply. Thus

$$
q>2 \sqrt{n} / \alpha_{t}
$$

## Chapter 6

## Adaptive Simulation-Based Secure Constructions for Functional <br> Encryption

In this chapter we consider the problem of designing simulation-based secure functional encryption schemes.

Specifically, we give a general transformation that takes a game-based secure functional encryption scheme for all polynomial-size circuits and constructs a functional encryption scheme for the same functionality that is simulation-based secure against adaptive adversaries that can ask for a bounded number of tokens and can see one ciphertext. Our transformation is both blackbox and for the standard model and is inspired by the work of Feige, Lapidot and Shamir [FLS90]. We remark that by the recent impossibility results of [AGVW12], the restriction to bounded number of tokens is necessary.

We then investigate functional encryption schemes for Hidden-Vector Encryption [BW07] which is readily seen to be a generalization of (Anonymous) Identity-Based Encryption. Therefore the impossibility result for IBE of [BSW11] applies also to HVE. The current state of the art [BSW11, O'N10, AGVW12] tells us that in general for functional encryption the gamebased and simulation-based definitions of security are not equivalent. Despite this, the work of O'Neill [O’N10] provides us with an avenue for constructing simulation-based secure functional encryption schemes from game-based secure ones for non-adaptive adversaries (these are adversaries that cannot ask token queries after having seen the ciphertext). Specifically, for a pre-image sampleable functionality (see later for a definition) any game-based secure functional encryption scheme is also non-adaptive simulation-based secure. We show that if HVE is preimage sampleable then it is possible to decide $\mathbb{N P}$ in probabilistic polynomial time. Therefore it is unlikely that pre-image samplability can be used to construct simulation-based secure HVE.

Motivated by this result, we then show how to construct a HVE scheme whose simulationbased security can be proved under standard assumptions in the bilinear pairing setting in the standard model. Our construction is shown secure against adaptive adversaries obtaining one ciphertext and asking an unbounded number of tokens. Again, this is the best one can hope for in the standard model, given the impossibility results of [BSW11, BO12] for IBE. The only previous simulation-based construction for IBE secure against adaptive adversary was given in [BSW11] and is in the programmable random oracle model and imposes no bound on the number of tokens and ciphertexts obtained by the adversary.

We remark that a functional encryption scheme for HVE can be derived from the general result of [GVW12]. However, our construction allows the adversary to obtain an unbounded
number of tokens (whereas the construction of [GVW12] imposes an upper bound on the number of tokens that can be seen by an adversary) and is more efficient (this is due to the fact that our construction is tailored for HVE whereas the one of [GVW12] is for a general class). Finally, we remark that the impossibility result of Agrawal et al. [AGVW12] does not apply to HVE. Indeed, the impossibility results uses the fact that polynomial size circuits can compute incompressible functionalities (i.e., weak pseudo-random functions). It is easy to see that HVE is not incompressible.

### 6.1 A General Transformation

In this section we show how to transform any game-based secure functional encryption scheme for the Circuit functionality (see Definition 2.0.2 into one for the same functionality that is simulation-based secure against adaptive adversaries that can ask a bounded number of tokens and can see one ciphertext. The transformation has some similarities with the FLS paradigm introduced by Feige, Lapidot and Shamir [FLS90] to obtain zero-knowledge proof systems from witness indistinguishable proof systems.

Notice, that by the impossibility result of [AGVW12], in the full model, there exists no functional encryption scheme for the functionality of all circuits that is simulation-based secure against adversaries that ask an unbounded number of tokens (even if the adversary is nonadaptive).

Let us start by showing how to transform a given $n$-input Boolean circuit $C$ in a new randomized circuit over $n \cdot(q+1)$ Boolean inputs $C^{r}$ such that $C^{r}$ evaluates the same function of $C$ and provides additional slots in the message that will be used in a critical way by our simulator. Later we will set $q$ as the bound on the number of tokens the adversary can ask for, and $r$ as a random $n$-bit Boolean string.

Definition 6.1.1 [The transformed circuit $C^{r}$ ] Fix the security parameter $\lambda$ and let $n=\operatorname{poly}(\lambda)$. For any string $r \in\{0,1\}^{n}$ and any $n$-bit input circuit $C$, we define $(n \cdot(q+1))$-input Boolean circuit $C^{r}$ as follows:

$$
C^{r}\left(m, m_{1}, \ldots, m_{q}\right)= \begin{cases}1, & \text { if } m_{i}=r \text { for some } i \in[q] \\ 0, & \text { if } m=\perp \text { and } m_{i} \neq r, \text { for all } i \in[q] \\ C(m), & \text { otherwise }\end{cases}
$$

Moreover, we let $\perp$ be a special message not contained in $X_{n}$.
Now, we are ready to show how to transform a functional encryption scheme IndFE for the Circuit functionality in an another functional encryption scheme SimFE for the same functionality by using $C^{r}$. We call this transformation Trans $_{\text {Bound }}$.

Definition 6.1.2 [The transformation Trans Bound ] Let IndFE be a functional encryption scheme for the functionality Circuit defined by the algorithm = (IndFE.Setup, IndFE.Enc, IndFE.KeyGen, IndFE.Eval), and let $q>0$ be a fixed integer. We define a new functional encryption scheme SimFE $=$ (Setup, KeyGen, Enc, Eval) for Circuit as follows.

- $\operatorname{Setup}\left(1^{\lambda}, 1^{n}\right)$ : the algorithm returns the output of $\operatorname{IndFE}$. Setup $\left(1^{\lambda}, 1^{n \cdot(q+1)}\right)$ as its own output.
- Enc $(m p k, m)$ : on input $m p k$ and $m \in\{0,1\}^{n}$, the algorithm chooses $q$ random $n$-bit Boolean strings $m_{1}, \ldots, m_{q} \in\{0,1\}^{n}$ and sets $m^{\prime}=\left(m, m_{1}, \ldots, m_{q}\right)$ and returns the output of IndFE.Enc $\left(m p k, m^{\prime}\right)$ as its own output.
- KeyGen $(m s k, C)$ : on input $m s k$ and a $n$-input Boolean circuit $C$, the algorithm chooses random $r \in\{0,1\}^{n}$ and returns the pair $(r, s k)$ where $s k=\operatorname{IndFE}$.KeyGen $\left(m s k, C^{r}\right)$.
- Eval $(m p k, \mathrm{Ct}, s k)$ : on input $m p k, \mathrm{Ct}$ and $s k$, the algorithm returns the output IndFE.Eval( $m p k$, $\mathrm{Ct}, s k)$ as its own output.

Correctness. Suppose that Eval is invoked on a ciphertext $C t$ for $m \in\{0,1\}^{n}$ and $s k_{C}$ for circuit $C$. Then if $m_{i} \neq r$ for all $i \in[q]$, Eval returns the correct evaluation, namely $C(m)$. Thus, the probability that Eval fails is at most $q / 2^{n}$ that is negligible in $n$.

Before proving the $q$-SIM-Security of SimFE, we mention that SimFE is also IND-Secure. Notice that it is not implied by $q$-SIM-Security. The lemma follows from a simple reduction to the IND-Security of IndFE and we state it without proof.

Lemma 6.1.3 If IndFE is IND-Secure then SimFE is IND-Secure.

We now prove the main result of this section, namely, that SimFE is $q$-SIM-Secure.

Theorem 6.1.4 If there exists an IND-Secure functional encryption scheme for circuits, then, for all constant $q$, there exists an $q$-SIM-Secure functional encryption scheme for the same functionality.

Proof: Let IndFE $=$ (IndFE.Setup, IndFE.Enc, IndFE. KeyGen, IndFE.Eval) be an IND-Secure functional encryption scheme for the functionality Circuit, and let SimFE = (Setup, KeyGen, Enc, Eval) be the functional encryption scheme for the same functionality obtained applying Trans Bound . Then, we show that $\operatorname{SimFE}$ is $q$-SIM-Secure by constructing a simulator $\operatorname{Sim}=\left(\operatorname{Sim}_{0}, \operatorname{Sim}_{1}, \operatorname{Sim}_{2}\right)$ such that for any p.p.t. adversaries $\mathcal{A}=\left(\mathcal{A}_{0}, \mathcal{A}_{1}\right)$, RealExp $\mathrm{SimFE}, \mathcal{A}^{\operatorname{Sind}}$ andeaIExp $\mathrm{SimFE}, \mathcal{A}$ are computationally indistinguishable, and reducing the $q$-SIM security to the IND security of IndFE.
Specifically, the simulator works as follows:

- $\operatorname{Sim}_{0}$ on input the security parameters invokes the IndFE.Setup algorithm to generate public parameters and master secret key.
- $\operatorname{Sim}_{1}$ on input the public parameter $m p k$, token queries $C_{1}, \ldots, C_{q_{1}}$ made by $\mathcal{A}_{0}$, the Boolean values $z_{1}=C_{1}(m), \ldots, z_{q_{1}}=C_{q_{1}}(m)$ where $m$ is the challenge message output by the adversary, and the tokens $s k_{1}, \ldots, s k_{q_{1}}$ generated by KeyGen to answers $A_{0}$ 's token queries, does the following:
$\operatorname{Sim}_{0}$ chooses values $\tilde{m}_{1}, \ldots, \tilde{m}_{q} \in\{0,1\}^{n}$ as follows. For $i=1, \ldots, q_{1}$, if $z_{i}=1, \operatorname{Sim}_{0}$ sets $\tilde{m}_{i}=r_{i}$; otherwise, if $z_{i}=0, \operatorname{Sim}_{0}$ picks $\tilde{m}_{i}$ at random in $\{0,1\}^{n}$. Furthermore for $i=q_{1}+1, \ldots, q, \operatorname{Sim}_{0}$ picks $\tilde{m}_{i}$ at random in $\{0,1\}^{n}$.
Finally, $\operatorname{Sim}_{0}$ sets $\tilde{m}=\left(\perp, \tilde{m}_{1}, \ldots, \tilde{m}_{q}\right)$ and computes $\tilde{C} t$ as $\tilde{C} t=\operatorname{IndFE} . \operatorname{Enc}(m p k, \tilde{m})$ and stores $\tilde{m}_{1}, \ldots, \tilde{m}_{q}$ in the state.
- $\operatorname{Sim}_{2}$ on input the master secret key $m s k$, Boolean circuit $C_{i}$ and $z_{i}=C_{i}(m)$, for $q_{1}<$ $i \leq q$, does the following:
If $z_{i}=0$ then $\operatorname{Sim}_{2}$ picks random $r_{i} \in\{0,1\}^{n}$; otherwise, if $z_{i}=1$, $\operatorname{Sim} \operatorname{Sets}_{1}$ set $r_{i}$. $\operatorname{Sim}_{2}$ then sets $s k_{i}$ as the pair $\left(r_{i}\right.$, IndFE.KeyGen $\left.\left(m s k, C_{i}^{r_{i}}\right)\right)$.

We now prove that $\operatorname{Sim}$ is a good simulator meaning that for all PPT adversaries $\mathcal{A}=\left(\mathcal{A}_{0}, \mathcal{A}_{1}\right)$, RealExp ${ }^{\operatorname{SimFE}, \mathcal{A}}$ and IdealExp $\mathrm{SimFE}_{\mathrm{Sim}}^{\operatorname{Sim}}$ are computationally indistinguishable.
For sake of contradiction, suppose there exists a distinguisher $\mathcal{D}$ that distinguishes with nonnegligible probability the output distribution of a certain $\mathcal{A}$ in experiment RealExp ${ }^{\operatorname{SimFE}, \mathcal{A}}$ and in experiment IdealExp $\operatorname{SimFE}_{\mathrm{Sim}}^{\operatorname{Sim}}$. Then, $\mathcal{A}$ and $\mathcal{D}$ can be used to construct a successful IND adversary $\mathcal{B}$ for IndFE. Specifically, $\mathcal{B}=\left(\mathcal{B}_{0}, \mathcal{B}_{1}\right)$ does the following.

- $\mathcal{B}_{0}$ on input the public parameters $m p k$ generated by IndFE.Setup, runs $\mathcal{A}_{0}$ on input $m p k$. Then, $\mathcal{B}_{0}$ answers $\mathcal{A}_{0}$ 's query $C_{i}$ for $i \in\left[q_{1}\right]$ by using its own oracle IndFE.KeyGen $(m s k, \cdot)$ as follows: $\mathcal{B}_{0}$ chooses a random $r_{i} \in\{0,1\}^{n}$ and outputs the pair $\left(r_{i}, s k_{i}\right)$ where $s k_{i}=\operatorname{IndFE} . \operatorname{KeyGen}\left(m s k, C_{i}^{r_{i}}\right)$.
Eventually, $\mathcal{A}_{0}$ outputs a message $m$ and the state aux. $\mathcal{B}_{0}$ sets $x=\left(m, m_{1}, \ldots, m_{q}\right)$ and $x^{\prime}=\left(\perp, m_{1}^{\prime}, \ldots, m_{q}^{\prime}\right)$ where the values $m_{1}, \ldots, m_{q}$ are chosen at random in $\{0,1\}^{n}$ and the values $m_{1}^{\prime}, \ldots, m_{q}^{\prime}$ are defined as follows. For $i=1$ to $q_{1}$ :
- If $C_{i}(m)=0$, picks $m_{i}^{\prime}$ at random in $\{0,1\}^{n}$,
- if $C_{i}(m)=1$, sets $m_{i}^{\prime}=r_{i}$.

Furthermore for $q_{1}<i \leq q, \operatorname{Sim}_{0}$ pick $m_{i}^{\prime}$ at random in $\{0,1\}^{n}$.
Finally, $\mathcal{B}_{0}$ outputs ( $x, x^{\prime}$, st); i.e., $x$ and $x^{\prime}$ are the two challenge plaintexts and st is $\mathcal{B}$ 's state.

- $\mathcal{B}_{1}$ on input the public parameters $m p k$, a ciphertext Ct and the state st, runs $\mathcal{A}_{1}$ on the same inputs.
Then, $\mathcal{B}_{1}$ answers $\mathcal{A}_{1}$ 's queries by using its own oracle IndFE.KeyGen $(m s k, \cdot)$ as follows: Given a query $C_{i}$, for $i=q_{1}+1, \ldots, q$,
- If $C_{i}(m)=0, \mathcal{B}_{1}$ picks random $r_{i} \in\{0,1\}^{n}$,
- if $C_{i}(m)=1$, sets $r_{i}=m_{i}^{\prime}$.

Then, $\mathcal{B}_{1}$ outputs $\left(r_{i}, s k_{i}\right)$ where $s k_{i}=\operatorname{IndFE} . \operatorname{KeyGen}\left(m s k, C_{i}^{r_{i}}\right)$.
Eventually, $\mathcal{A}_{1}$ outputs $\alpha$, then $\mathcal{B}_{1}$ invokes $\mathcal{D}$ on input ( $m p k, x, \alpha$ ) and returns $\mathcal{D}$ 's guess as its own guess.

We now show that, except with negligible probability, $x=\left(m, m_{1}, \ldots, m_{q}\right)$ and $x^{\prime}=(\perp$ , $m_{1}^{\prime}, \ldots, m_{q}^{\prime}$ ) are such that, for all $i=1, \ldots, q, C_{i}^{r_{1}}(x)=C_{i}^{r_{1}}\left(x^{\prime}\right)$ and thus $\mathcal{B}$ is a valid IND adversary.
Recall that $C_{i}^{r_{i}}\left(m, m_{1}, \ldots, m_{q}\right)=1$ iff $r_{i}=m_{j}$ or $C_{i}(m)=1$ for some $j \in[q]$, and that $m \neq \perp$. For each $i \in[q]$, we distinguish two mutually exclusive cases.

- $C_{i}^{r_{i}}(x)=0$. In this case, we know that $C_{i}(m)=0$ since $m \neq \perp$. On the other side, since $m_{1}^{\prime}, \ldots, m_{q}^{\prime}$ are randomly chosen and independent from $r_{i}$, from the definition of $C_{i}^{r_{i}}$, it follows that, except with negligible probability, $C_{i}^{r_{i}}\left(x^{\prime}\right)=0$.
- $C_{i}^{r_{i}}(x)=1$. Suppose toward a contradiction that with non-negligible probability $C_{i}(m)=$ 0 . Thus $C_{i}^{r_{i}}(x)=1$ because $r_{i}=m_{j}$ for some $j \in[q]$ given that $m \neq \perp$. But when $C_{i}(m)=0$ then $r_{i}$ is randomly and independently chosen. Thus, with negligible probability, $r_{i} \neq m_{j}$ for all $j \in[q]$. Thus, $C_{i}^{r_{i}}(x)=0$. Contradiction. Then, it must be that $C_{i}(m)=1$. Thus, it holds that $r_{i}=m_{j}^{\prime}$ for some $j \in[q]$, so that $C_{i}^{r_{i}}\left(x^{\prime}\right)=1$ too.

Now notice that if the challenger of the IND-Security game encrypts $x$ then $\mathcal{B}$ is simulating $\mathcal{A}$ 's output distribution in the RealExp ${ }^{\operatorname{SimFE}, \mathcal{A}}$ experiment. Instead if $x^{\prime}$ is encrypted then $\mathcal{B}$ is simulating Sim's output distribution in the $\operatorname{Ideal} \operatorname{Exp}_{\mathrm{Sim}}^{\operatorname{SimFE}, \mathcal{A}}$ experiment.
Finally, for $\mathcal{A}$ to be successful it is important that $x \neq x^{\prime}$, which holds with non-negligible probability. By the hypothesis on $\mathcal{D}$, this contradicts our initial assumption that IndFE is INDSecure.

Comparison with the construction of Gorbunov, Vaikuntanathan, and Wee [GVW12]. Gorbunov, Vaikuntanathan, and Wee [GVW12], building on the construction of Sahai and Seyalioglu [SS10], gave a functional encryption scheme for all polynomial size circuits that is simulationbased secure against adaptive adversaries that have access to a bounded number of tokens. We stress that they assume a (one-query) SIM-Secure functional encryption scheme whereas we rely only on the existence of IND-Secure functional encryption schemes. Furthermore, their construction results in a scheme with a ciphertext of size super linear (at least quadratic even in the case of NC1 circuits) in $q$ whereas our construction has ciphertext of size linear in $q$.

### 6.2 Adaptive Security for HVE

In this section, we show how to construct a HVE scheme whose semantic security can be proved under static assumptions in the bilinear pairing setting in the standard model. We show our construction secure against adaptive adversaries obtaining one ciphertext and asking an unbounded number of tokens.

One possible avenue for obtaining (non-adaptively) simulation-based secure functional encryption scheme for HVE could be via the notion of pre-image samplability introduced by O'Neill [O'N10]. In Section 6.2.1] we prove that HVE is unlikely to be pre-image sampleable. We then embark into the construction of an adaptively secure simulation-based functional encryption scheme for HVE.

### 6.2.1 Pre-Image Samplability

In this section we prove that if HVE is pre-image sampleable then we can decide 3SAT in polynomial time. We start by reviewing the notion of a pre-image sampleable functionality by O'Neill [O’N10].

Definition 6.2.1 [O’N10] Functionality $F: K \times M \rightarrow\{0,1\}$ is pre-image sampleable $(\mathrm{PS}$, for short) if there exists a sampler algorithm Sam such that for all PPT adversaries $\mathcal{A}$,

$$
\begin{aligned}
& \operatorname{Prob}\left[\left(m,\left(k_{i}\right)_{i=1}^{\ell=\operatorname{poly}(\lambda)}\right) \leftarrow \mathcal{A}\left(1^{\lambda}\right) ; m^{\prime} \leftarrow \operatorname{Sam}\left(1^{\lambda},|m|,\left(k_{i}, F\left(k_{i}, m\right)\right)_{i=1}^{\ell}\right):\right. \\
& \left.F\left(k_{i}, m\right)=F\left(k_{i}, m^{\prime}\right) \text { for } i=1, \ldots, \ell\right]=1-\operatorname{negl}(\lambda)
\end{aligned}
$$

For $q>0$, we say that $F$ is $q$-pre-image sampleable ( $q-\mathrm{PS}$, for short) if the sampler algorithm Sam is guaranteed to work only with respect to adversaries $\mathcal{A}$ that output at most $q$ pairs $\left(k_{i}, b_{i}\right)$.

The following theorem is from [O’N10].
Theorem 6.2.2 If functionality $F$ is PS then any IND-Secure functional encryption scheme $\mathcal{F E}$ for $F$ is also NA-SIM-Secure.

Next we prove that the fact that the HVE functionality is PS has unexpected complexitytheoretic consequences.

Theorem 6.2.3 If HVE is PS then SAT can be decided in probabilistic polynomial time.
Proof: Let Sam be a sampler algorithm for HVE and consider the following algorithm $\mathcal{B}$ that, on input a Boolean formula $\Phi=\phi_{1} \wedge \ldots \wedge \phi_{c}$ in CNF with $c$ clauses and $\ell$ variables $x_{1}, \ldots, x_{\ell}$, performs the following computation. In the description of $\mathcal{B}$, we will identify $\ell$-bit strings with truth assignment to variables $x_{1}, \ldots, x_{\ell}$.
For each clause $\phi_{i}$ of $\Phi, \mathcal{B}$ computes key $k_{i}=\left(k_{1, i}, \ldots, k_{\ell, i}\right) \in\{0,1, \star\}^{\ell}$ in the following way. Let $h \in\{0,1\}^{\ell}$ be a truth assignment that does not satisfy the $i$-th clause $\phi_{i}$. For $j=1, \ldots, \ell$, $\mathcal{B}$ sets $k_{j, i}$ in the following way

$$
k_{j, i}= \begin{cases}h_{j}, & \text { if } x_{j} \text { appears in } \phi_{i} ; \\ 1-h_{j}, & \text { if } \bar{x}_{j} \text { appears in } \phi_{i} ; \\ \star, & \text { otherwise } .\end{cases}
$$

Keys $k_{1}, \ldots, k_{c}$ enjoy the following (easy to verify) property. Let $m \in\{0,1\}^{\ell}$ be a truth assignment over the variables $x_{1}, \ldots, x_{\ell}$. Then, $m$ satisfies $\Phi$ if and only if $\operatorname{HVE}\left(m, k_{i}\right)=0$ for $i=1, \ldots, c$.
$\mathcal{B}$ then runs algorithm Sam on input $\left(\left(k_{1}, 0\right), \ldots,\left(k_{c}, 0\right)\right)$ and let $m$ be Sam's output. If $m$ is a satisfying truth assignment for $\Phi$ then $\mathcal{B}$ decides that $\Phi$ is satisfiable. Otherwise, $\mathcal{B}$ decides that $\Phi$ is not satisfiable.
Let us now prove that $\mathcal{B}$ 's output is correct with high probability. Notice that Sam is only guaranteed to work if it is given in input a sequence $\left(k_{i}, b_{i}\right)$ for which there exists an $m \in\{0,1\}^{\ell}$ such that $b_{i}=\operatorname{HVE}\left(m, k_{i}\right)$. So we distinguish two cases. In the first case, we assume that $\Phi$ is satisfiable. Then the input of Sam is exactly as required by Definition 6.2.1 and thus, except with negligible probability, Sam outputs $m$ such that $\operatorname{HVE}\left(m, k_{i}\right)=0$ for $i=1, \ldots, c$. By our previous observation such an $m$ is a satisfying assignment for $\Phi$ and $\mathcal{B}$ is correct.
Suppose instead that $\Phi$ is not satisfiable. Then Sam will not output a satisfying assignment and $\mathcal{B}$ is correct.

### 6.2.2 Complexity Assumptions

In our construction we will make hardness assumptions for bilinear settings whose order $N$ is product of five distinct primes each of length $\Theta(\lambda)$.

Assumption 1 (General Diffie-Hellman for composite order groups). The assumption is a kind of Diffie-Hellman assumption in the bilinear setting of composite order. Specifically, for a given composite-order bilinear setting generator CBSGen, define the following distribution: Pick a random bilinear setting $\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{CBSGen}\left(1^{\lambda}, 5\right)$ and then pick

$$
\begin{gathered}
A_{1} \leftarrow \mathbb{G}_{p_{1}}, A_{2} \leftarrow \mathbb{G}_{p_{2}}, A_{3} \leftarrow \mathbb{G}_{p_{3}}, A_{4}, B_{4}, C_{4}, D_{4} \leftarrow \mathbb{G}_{p_{4}}, A_{5} \leftarrow \mathbb{G}_{p_{5}}, \alpha, \beta \leftarrow \mathbb{Z}_{N}, \\
T_{0}=A_{1}^{\alpha \beta} \cdot D_{4}, T_{1} \leftarrow \mathbb{G}_{p_{1} p_{4}},
\end{gathered}
$$

and set $D=\left(N, A_{1}, A_{2}, A_{3}, A_{4}, A_{5}, A_{1}^{\alpha} \cdot B_{4}, A_{1}^{\beta} \cdot C_{4}\right)$. We define the advantage of any $\mathcal{A}$ in breaking Assumption 1 to be

$$
\operatorname{Adv}_{1}^{\mathcal{A}}(\lambda)=\left|\operatorname{Prob}\left[\mathcal{A}\left(D, T_{0}\right)=1\right]-\operatorname{Prob}\left[\mathcal{A}\left(D, T_{1}\right)=1\right]\right| .
$$

Definition 6.2.4 We say that the Assumption 1 holds for generator CBSGen if, for all probabilistic polynomial-time algorithms $\mathcal{A}, \operatorname{Adv}_{1}^{\mathcal{A}}(\lambda)$ is a negligible function of $\lambda$.

### 6.2.3 The Scheme

In this section we describe our HVE scheme. To make our description and proofs simpler, we add to all vectors $\vec{x}$ and $\vec{y}$ two dummy components and set both of them equal to 0 . We can thus assume that all vectors have at least two non-star positions.
$\operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right)$ : The setup algorithm randomly chooses a description of composite-order bilinear setting $\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{CBSGen}\left(1^{\lambda}, 5\right)$ with known factorization, and random $g_{1} \in$ $\mathbb{G}_{p_{1}}, g_{2} \in \mathbb{G}_{p_{2}}, g_{3} \in \mathbb{G}_{p_{3}}, g_{4} \in \mathbb{G}_{p_{4}}$, and, for $i \in[\ell]$ and $b \in\{0,1\}$, random $t_{i, b} \in Z_{N}$ and random $R_{i, b} \in \mathbb{G}_{p_{3}}$ and sets

$$
T_{i, b}=g_{1}^{t_{i, b}} \cdot R_{i, b}
$$

The public key is $m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]$, and the master secret key is $m s k=$ $\left[g_{12}, g_{4},\left(t_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]$, where $g_{12}=g_{1} \cdot g_{2}$. The algorithm returns ( $m p k, m s k$ ).

KeyGen $(m s k, \vec{y})$ : Let $S_{\vec{y}}$ be the set of indices $i$ such that $y_{i} \neq \star$. The key generation algorithm chooses random $a_{i} \in \mathbb{Z}_{N}$ for $i \in S_{\vec{y}}$ under the constraint that $\sum_{i \in S_{\vec{y}}} a_{i}=0$. For $i \in S_{\vec{y}}$, the algorithm chooses random $W_{i} \in \mathbb{G}_{p_{4}}$ and sets

$$
Y_{i}=g_{12}^{a_{i} / t_{i, y_{i}}} \cdot W_{i}
$$

The algorithm returns ciphertext $\mathrm{Ct}=\left(Y_{i}\right)_{i \in S_{\vec{y}}}$.
Here we use the fact that $S_{\vec{y}}$ has size at least 2 .
$\operatorname{Enc}(m p k, \vec{x})$ : The encryption algorithm chooses random $s \in \mathbb{Z}_{N}$. For $i \in[\ell]$, the algorithm chooses random $Z_{i} \in \mathbb{G}_{p_{3}}$ and sets

$$
X_{i}=T_{i, x_{i}}^{s} \cdot Z_{i}
$$

and returns the token $s k_{\vec{y}}=\left(X_{i}\right)_{i \in[\ell]}$.
$\operatorname{Eval}\left(m p k, \mathrm{Ct}, s k_{\vec{y}}\right)$ : The test algorithm computes

$$
T=\prod_{i \in S_{\vec{y}}} \hat{e}\left(X_{i}, Y_{i}\right)
$$

It returns TRUE if $T=1$, FALSE otherwise.

Correctness. It easy to see that the scheme is correct.

Remark 6.2.5 In our construction the computation of the HVE predicate is performed in the $\mathbb{G}_{p_{1}}$ subgroup. All the other subgroups are used during the proof of security and let us able to change the distribution of the real experiment to one that is computationally indistinguishable and easy to simulate. Notice that at this stage the $\mathbb{G}_{p_{2}}$ subgroup can be removed from our construction at the expense of introducing another game in the security proof. To simplify the proof we have decided to include that subgroup directly.

### 6.2.4 Proof of Security

In this section we prove the following theorem.
Theorem 6.2.6 Under the General Subgroup Decision Assumption and Assumption 1, the HVE scheme described in Section 6.2.3 is SIM-Secure in the sense of Definition 2.2.2

We next give an overview of the proof of the theorem and break down the proof into 4 lemmata.

Our simulator Sim consists of three algorithms: Sim.Setup, Sim.Enc, and Sim.KeyGen that share a common state. Informally,

Sim. Setup produces two public keys $m p k$ and $m p k^{\prime} . m p k$ is given to the adversary (that may use it to generate his own ciphertexts) and $m p k^{\prime}$ instead is used to generate the simulated ciphertext.

Sim. Enc is used to compute the simulated ciphertext and takes as input the public key $m p k^{\prime}$ and the sequence $\left(\vec{y}_{k}, z_{k}\right)_{k=1}^{q_{1}}$, where $z_{k}=\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right), \vec{x}$ is the plaintext output by the adversary and $q_{1}$ is the number of queries asked by the adversary in the first stage.

Sim.KeyGen instead is used to answer the adaptive queries asked by the adversary after seeing the simulated ciphertext. When the adversary asks to see a token for $\vec{y}, \operatorname{Sim}$. KeyGen is invoked with the master secret key $m s k$, the value $\vec{y}$ and the value $\operatorname{HVE}(\vec{x}, \vec{y})$, where $\vec{x}$ is the challenge ciphertext output by the adversary. We stress that the simulator does not have access to $\vec{x}$. Thus we have to prove indistinguishability of the following two experiments.

Therefore, to prove Theorem 6.2.6 it is enough to show that the following two experiments are computational indistinguishable.

```
RealExp \({ }^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right)\)
Output: \((m p k, \vec{x}, \alpha)\)
```

$(m p k, m s k) \leftarrow \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ; \quad\left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} . \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ;$
$(\vec{x}$, aux $) \leftarrow \mathcal{A}_{0}^{\text {KeyGen }(m s k,)}(m p k) ; \quad(\vec{x}$, aux $) \leftarrow \mathcal{A}_{0}^{\text {Key } \operatorname{Gen}(m s k, \cdot)}(m p k)$;
$\mathrm{Ct} \leftarrow \operatorname{Enc}(m p k, \vec{x}) ; \quad \operatorname{Ct} \leftarrow \operatorname{Sim} \cdot \operatorname{Enc}\left(m p k^{\prime},\left(\vec{y}_{k}, \operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)\right)_{k=1}^{q_{1}}\right)$;
$\alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Key} \operatorname{Gen}(m s k, \cdot)}(m p k, \mathrm{Ct}$, aux $) ; \quad \alpha \leftarrow \mathcal{A}_{1}^{\text {Sim. } \operatorname{Key} \operatorname{Gen}(m s k, ; \operatorname{HVE}(\vec{x}, \cdot))}(m p k, \mathrm{Ct}$, aux $) ;$
Ideal $\operatorname{Exp}_{\operatorname{Sim}}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right)$
Output: $(m p k, \vec{x}, \alpha)$

More formally, we next describe the three algorithms used by Sim.
Algorithm Sim.Setup constructs two public keys $m p k$ (that is given to the adversary) and $m p k^{\prime}$ (that is used to produce the simulated ciphertext) and a master secret key msk (that is used to answer the queries of the adversary). In public key $m p k$ the $t_{i, b}$ 's are encoded in the $\mathbb{G}_{p_{2}}$ part of the $T_{i, b}$ 's (instead of the $\mathbb{G}_{p_{1}}$ part as in normal public key) whereas $m p k^{\prime}$ is generated correctly. Then, notice that the simulated ciphertext is independent from $m p k$, and this gives us enough freedom to manipulate the $\mathbb{G}_{p_{1}}$ part of the simulated ciphertext and tokens the adversary sees in the second stage. The subgroup $\mathbb{G}_{p_{3}}$ will help us to hide this transition.
Specifically, Sim.Setup randomly chooses a description of a composite-order bilinear $\operatorname{group}\left(N, \mathbb{G}, g, \mathbb{G}_{T}, \hat{e}\right) \leftarrow \operatorname{CBSGen}\left(1^{\lambda}, 5\right)$ with known factorization, and random $g_{1} \in$
$\mathbb{G}_{p_{1}}, g_{2} \in \mathbb{G}_{p_{2}}, g_{3} \in \mathbb{G}_{p_{3}}, g_{4} \in \mathbb{G}_{p_{4}}, g_{5} \in \mathbb{G}_{p_{5}}$. Then, for $i \in[\ell]$ and $b \in\{0,1\}$, Sim.Setup picks random $t_{i, b} \in Z_{N}, R_{i, b}, R_{i, b}^{\prime} \in \mathbb{G}_{p_{3}}$ and sets

$$
T_{i, b}=g_{2}^{t_{i, b}} \cdot R_{i, b}, \quad T_{i, b}^{\prime}=g_{1}^{t_{i, b}} \cdot R_{i, b}^{\prime}
$$

The public keys are

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}]}\right] \quad \text { and } \quad m p k^{\prime}=\left[N, g_{3}, g_{5},\left(T_{i, b}^{\prime}\right)_{i \in[\ell], b \in\{0,1\}}\right],
$$

and the master secret key is

$$
m s k=\left[g_{12}, g_{4}, g_{5},\left(t_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

where $g_{12}=g_{1} \cdot g_{2}$. Sim. Setup returns $\left(m p k, m p k^{\prime}, m s k\right)$.
Algorithm Sim.Enc simulates the encryption of the challenge plaintext $\vec{x}$ provided by the adversary. It receives on input the queries $\vec{y}_{k}, k=1, \ldots, q_{1}$, asked by the adversary in the first stage (the non-adaptive one) and the values $z_{k}=\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)$, for $k=1, \ldots, q_{1}$.

Sim.Enc does not have access to $\vec{x}$ and by Theorem 6.2 .3 the pre-image samplability cannot be used to sample an $\vec{x}^{\prime}$ such that $\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)=\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}^{\prime}\right)$ for all $k$. But still some information about $\vec{x}$ can be derived from the inputs. Indeed, observe that if $\operatorname{HVE}(\vec{y}, \vec{x})=1$, then $\vec{y}$ and $\vec{x}$ coincide in all non-star entries of $\vec{y}$. For these positions, Sim. Enc computes the ciphertext Ct just like the encryption algorithm; for all remaining positions, the ciphertext output by Sim. Enc has a random $\mathbb{G}_{p_{1}}$ part.

Notice that such a simulated ciphertext is then compatible with all the tokens the adversary sees in first stage in the sense that $\operatorname{Eval}\left(m p k, \mathrm{Ct}, s k_{\vec{y}_{k}}\right)=\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)$, for $k=1, \ldots, q_{1}$. Unfortunately the simulation is not perfect. In the second stage, the adversary could ask to see a token $s k_{\vec{y}}$ for a vector $\vec{y}$ such that $\operatorname{HVE}(\vec{y}, \vec{x})=1$ but $\operatorname{Eval}\left(m p k, \mathrm{Ct}, s k_{\vec{y}}\right)=0$. This is because the $\mathbb{G}_{p_{1}}$ part of simulated ciphertext Ct and the token $s k_{\vec{y}}$ does not cancel out correctly upon decryption. A possible solution is to remove the $\mathbb{G}_{p_{1}}$ part from the matching tokens the adversary sees in the second stage. This is where $\mathbb{G}_{p_{5}}$ comes to help us. $\mathbb{G}_{p_{5}}$ will be introduced in the simulated ciphertext and matching second stage tokens in such a way it will cancel out upon decryption and will provide us enough entropy to remove the $\mathbb{G}_{p_{1}}$ part from the adaptive tokens. Thus, each component of the ciphertext computed by Sim.Enc contains also a random $\mathbb{G}_{p_{5}}$ part that will be used for constructing the answers to the adaptive queries of the adversary.

More formally, for a sequence $Q=\left(\vec{y}_{k}, z_{k}\right)_{k=1}^{q_{1}}$ with $z_{k}=\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)$, for $k=1, \ldots, q_{1}$, we let MPos denote the set of indices $1 \leq i \leq \ell$ for which there exists $k \in\left\{1, \ldots, q_{1}\right\}$ such that $z_{k}=1$ and $y_{k, i} \neq \star$. Notice that, by the observation above, for all $i \in \operatorname{MPos}$, the value $x_{i}$ can be derived from the sequence $Q$. Then, $\operatorname{Sim}$.Enc $\left(m p k^{\prime},\left(\vec{y}_{k}, z_{k}\right)_{k=1}^{q_{1}}\right)$ can be described as follows. The algorithm parses $m p k^{\prime}$ as $m p k^{\prime}=\left[N, g_{3}, g_{5},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]$ and, for all $i \in$ MPos, derives the value $x_{i}$ from the input sequence. Then, the algorithm chooses random $s \in \mathbb{Z}_{N}$ and, for each $i \in[\ell]$, random $Z_{i} \in \mathbb{G}_{p_{3}}$, and, for each $i \in[\ell]$ :

- if $i \notin$ MPos, randomly select $r_{i} \in \mathbb{Z}_{N}$ and set $X_{i}=T_{i, 0}^{r_{i}} \cdot g_{5}^{s_{i}} \cdot Z_{i}$.
- if $i \in$ MPos, set $X_{i}=T_{i, x_{i}}^{s} \cdot g_{5}^{s_{i}} \cdot Z_{i}$.

Sim.Enc returns one output, the challenge ciphertext $\mathrm{Ct}=\left(X_{i}\right)_{i \in[\ell]}$, and keeps the vector $\left(s_{i}\right)_{i \in[\ell]}$ in the state so that it can be used by algorithm Sim.KeyGen.

| Real | $H_{1}$ | $H_{2}$ | $H_{3}$ | Ideal |
| :--- | :--- | :--- | :--- | :--- |
| Setup | Sim.Setup | Sim.Setup | Sim.Setup | Sim.Setup |
| Enc | Enc | Sim.A.Enc | Sim.A.Enc | Sim.Enc |
| KeyGen | KeyGen | KeyGen | Sim.KeyGen | Sim.KeyGen |

Figure 6.1: The hybrids used in the proof of security.


#### Abstract

Algorithm Sim.KeyGen simulates the answer to the second stage queries of the adversary. It receives as input the master secret key $m s k$, the vector $\vec{y}$ for which the token has to be simulated and the value $z=\operatorname{HVE}(\vec{y}, \vec{x})$, where $\vec{x}$ is the challenge plaintext.


For each $j \in S_{\vec{y}}$, Sim. KeyGen selects random $W_{j} \in \mathbb{G}_{p_{4}}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$. Then, it distinguishes two cases.

- $z=0$ : In this case, the algorithm generates a token with a random $\mathbb{G}_{p_{1}}$ part. Specifically, for each $j \in S_{\vec{y}}$, Sim. KeyGen chooses random $c_{j} \in \mathbb{Z}_{N}$ and sets

$$
Y_{j}=g_{1}^{c_{j}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

- $z=1$ : In this case, the algorithm generates a token without the $\mathbb{G}_{p_{1}}$ part and with a $\mathbb{G}_{p_{5}}$ part that will cancel out against the simulated ciphertext upon decryption. Specifically, for each $j \in S_{\vec{y}}$, Sim. KeyGen sets

$$
Y_{j}=g_{2}^{a_{j} / t_{j, y_{j}}} \cdot g_{5}^{a_{j} / s_{j}} \cdot W_{j}
$$

We remind the reader that the vector $\left(s_{j}\right)_{j \in[\ell]}$ is the same vector used by Sim.Enc (and is stored in the state of the simulator). Also, note that the key output by Sim.KeyGen behaves correctly when matched against the ciphertext output by Sim.Enc.

To prove that the real and the ideal experiments are indistinguishable we use three intermediate hybrids that we call $H_{1}, H_{2}$ and $H_{3}$. Then, the proof consists of four main steps (see Figure 6.2.4 for a quick reference).

The first step of our proof consists in constructing the public key by means of Sim.Setup. This has the effect of projecting the public key (and thus the ciphertexts the adversary constructs by himself) to a different subgroup from the one of the challenge ciphertext. Specifically:

$$
\begin{aligned}
& H_{1}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} \cdot \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) \\
& (\vec{x}, \text { aux }) \leftarrow \mathcal{A}_{0}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k) ; \\
& \mathrm{Ct} \leftarrow \operatorname{Enc}\left(m p k^{\prime}, \vec{x}\right) ; \\
& \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k, \mathrm{Ct}, \text { aux }) ; \\
& \text { Output: }(m p k, \vec{x}, \alpha)
\end{aligned}
$$

The first step of the proof consists in proving that the outputs of ReaIExp ${ }^{\mathcal{A}}$ and $H_{1}^{\mathcal{A}}$ are indistinguishable for all PPT adversaries $\mathcal{A}$.

The second step modifies the simulated ciphertext by adding a $\mathbb{G}_{p_{5}}$ part. This will be used in successive experiments to answer to the adversary's adaptive token queries. Specifically:

$$
\begin{aligned}
& H_{2}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} \cdot \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ; \\
& (\vec{x}, \text { aux }) \leftarrow \mathcal{A}_{0}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k) ; \\
& \mathrm{Ct} \leftarrow \operatorname{Sim} . \operatorname{A.Enc}\left(m p k^{\prime}, \vec{x}\right) ; \\
& \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k, \mathrm{Ct}, \text { aux }) ; \\
& \text { Output: }(m p k, m, \alpha)
\end{aligned}
$$

where algorithm Sim.A.Enc proceeds as follows.
Algorithm Sim.A.Enc on input public key $m p k^{\prime}=\left[N, g_{3}, g_{5},\left(T_{i, b}^{\prime}\right)_{i \in[\ell], b \in\{0,1\}}\right]$, the algorithm chooses random $s \in \mathbb{Z}_{N}$. Then, for $i \in[\ell]$, the algorithm chooses random $Z_{i} \in \mathbb{G}_{p_{3}}$ and random $s_{i} \in \mathbb{Z}_{N}$, sets

$$
X_{i}=T_{i, x_{i}}^{\prime}{ }^{s} \cdot g_{5}^{s_{i}} \cdot Z_{i}
$$

returns the tuple $\left(X_{i}\right)_{i \in[\ell]}$ and stores the vector $\left(s_{i}\right)_{i \in[\ell]}$ in the state so that it can be used by Sim.KeyGen.

The second step of the proof consists in proving that the outputs of $H_{1}^{\mathcal{A}}$ and $H_{2}^{\mathcal{A}}$ are indistinguishable for all PPT adversaries $\mathcal{A}$.

The third step computes the simulated answers using the Sim.KeyGen algorithm that adds a $\mathbb{G}_{p_{5}}$ part that cancels out with the one added in the simulated ciphertext.

$$
\begin{aligned}
& H_{3}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} \cdot \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) \\
& (\vec{x}, \text { aux }) \leftarrow \mathcal{A}_{0}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k) \\
& \mathrm{Ct} \leftarrow \operatorname{Sim} . \operatorname{A.Enc}\left(m p k^{\prime}, \vec{x}\right) \\
& \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Sim} . \operatorname{KeyGen}(m s k, \cdot, \operatorname{HV}(\vec{x}, \cdot))}(m p k, \mathrm{Ct}, \text { aux }) \\
& \text { Output: }(m p k, \vec{x}, \alpha)
\end{aligned}
$$

Thus the third step of the proof consists in the defining hybrid experiment $H_{3}^{\mathcal{A}}$ and in proving that it is indistinguishable from $H_{2}^{\mathcal{A}}$ for all PPT adversaries $\mathcal{A}$.

The fourth step consists in proving that hybrid $H_{3}^{\mathcal{A}}$ is indistinguishable from the ideal experiment for all PPT adversaries $\mathcal{A}$.

This concludes the proof of Theorem6.2.6. In the next sections we give formal details for each step of the proof.

### 6.2.5 The first step of the proof

In this section we prove that RealGame and PKGame are indistinguishable.
Lemma 6.2.7 If the GSD Assumption holds, then for all PPT adversaries $\mathcal{A}$, RealGame ${ }^{\mathcal{A}} \approx_{c}$ PKGame ${ }^{\mathcal{A}}$.

Proof. Suppose there exists an adversary $\mathcal{A}$ for which RealGame ${ }^{\mathcal{A}}$ and PKGame ${ }^{\mathcal{A}}$ are distinguishable. Then, we show a PPT algorithm $\mathcal{B}$ which receives the following instance of the GSD Assumption, $\left(N, T,\left\{g_{3}, g_{4}, g_{13}, g_{12}\right\}\right) \leftarrow \operatorname{GSDGen}\left(1^{\lambda},\{1,3\},\{2,3\},\{3\},\{4\},\{1,3\},\{1,2\}\right)$ and, depending on the nature of $T$, simulates RealGame ${ }^{\mathcal{A}}$ or PKGame $^{\mathcal{A}}$. This suffices to prove the Lemma.

Setup: $\mathcal{B}$, for each $i \in[\ell]$ and $b \in\{0,1\}, \mathcal{B}$ chooses random $t_{i, b} \in \mathbb{Z}_{N}$ and sets

$$
T_{i, b}=T^{t_{i, b}} \quad \text { and } \quad T_{i, b}^{\prime}=g_{13}^{t_{i, b}}
$$

Then $\mathcal{B}$ sets

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right] \quad \text { and } \quad m p k^{\prime}=\left[N, g_{3},\left(T_{i, b}^{\prime}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and

$$
m s k=\left[g_{12}, g_{4},\left(t_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and starts the interaction with $\mathcal{A}$ on input $m p k$.
Token Queries: $\mathcal{B}$ uses $m s k$ to answer the token queries of $\mathcal{A}$.
Ciphertext: $\mathcal{B}$ computes $\mathrm{Ct}=\operatorname{Enc}\left(m p k^{\prime}, \vec{x}\right)$.
This concludes the description of algorithm $\mathcal{B}$.
The view of $\mathcal{A}$ consists of $m p k, \mathrm{Ct}$ and the answer of the queries. The answers of the queries are distributed, independently from the nature of the challenge $T$, as in RealGame ${ }^{\mathcal{A}}$ and as in PKGame ${ }^{\mathcal{A}}$.

If $T \in \mathbb{G}_{p_{1} p_{3}}$ then $m p k$ is constructed exactly as in RealGame ${ }^{\mathcal{A}}$. Moreover, the ciphertext Ct , even though is constructed using $m p k^{\prime}$, has the same distribution of a ciphertext constructed using $m p k$ and thus the view of $\mathcal{A}$ is the same as in RealGame ${ }^{\mathcal{A}}$.

On the other hand, when $T \in \mathbb{G}_{p_{2} p_{3}}$ then $m p k$ and $C t$ are distributed as in PKGame ${ }^{\mathcal{A}}$.

### 6.2.6 The second step of the proof

In this section we prove that PKGame is indistinguishable from TypeACtGame.
Lemma 6.2.8 If GSD Assumption holds then, for all PPT adversaries $\mathcal{A}, \mathrm{PKGame}{ }^{\mathcal{A}} \approx_{c}$ TypeACtGame ${ }^{\mathcal{A}}$.
Proof. Suppose there exists an adversary $\mathcal{A}$ for which PKGame ${ }^{\mathcal{A}}$ and TypeACtGame ${ }^{\mathcal{A}}$ are not indistinguishable. Then, we show a PPT algorithm $\mathcal{B}$ which receives the following instance of the GSD Assumption, $\left(N, T,\left\{g_{1}, g_{2}, g_{3}, g_{4}\right\}\right) \leftarrow \operatorname{GSDGen}\left(1^{\lambda},\{3\},\{3,5\},\{1\},\{2\},\{3\},\{4\}\right)$ and, depending on the nature of $T$, simulates PKGame ${ }^{\mathcal{A}}$ or TypeACtGame ${ }^{\mathcal{A}}$. This suffices to prove the Lemma.

Setup: $\mathcal{B}$, for each $i \in[\ell]$ and $b \in\{0,1\}, \mathcal{B}$ chooses random $t_{i, b} \in \mathbb{Z}_{N}, R_{i, b}, R_{i, b}^{\prime} \in \mathbb{G}_{p_{3}}$ and sets $T_{i, b}=g_{2}^{t_{i, b}} \cdot R_{i, b}$. Then $\mathcal{B}$ sets

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and

$$
m s k=\left[g_{12}, g_{4},\left(t_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and starts the interaction with $\mathcal{A}$ on input mpk.
Token Queries: $\mathcal{B}$ runs algorithm KeyGen on input $m s k$ to answer token queries.
Ciphertext: $\mathcal{B}$ generates the simulated ciphertext for $\vec{x}$ as follows. $\mathcal{B}$ chooses, for $i \in[\ell]$, random $r_{i} \in \mathbb{Z}_{N}$ and random $\mathbb{Z}_{i} \in \mathbb{G}_{p_{3}}$, and sets

$$
X_{i}=g_{1}^{t_{i, x_{i}}} \cdot T^{r_{i}} \cdot Z_{i}
$$

This concludes the description of algorithm $\mathcal{B}$.
Finally it is easy to see that if $T \in \mathbb{G}_{p_{3}}$ then Ct is distributed as in PKGame ${ }^{\mathcal{A}}$ and if $T \in$ $\mathbb{G}_{p_{3} p_{5}}$ then Ct is distributed as in TypeACtGame ${ }^{\mathcal{A}}$.

### 6.2.7 The third step of the proof

In this section we prove that $H_{2}$ is indistinguishable from $H_{3}$.
Lemma 6.2.9 Under the GSD Asummption and Assumption 5, for all PPT adversaries $\mathcal{A}$, $H_{2}^{\mathcal{A}} \approx_{c} H_{3}^{\mathcal{A}}$.

The proof of the lemma above consists of three sub-steps.
First Substep. In the first substep, we define a new experiment called TypeAKeysGame which differs from TypeACtGame in the way the token queries of the second stage are answered. More specifically, the token for vector $\vec{y}$ such that $\operatorname{HVE}(\vec{x}, \vec{y})=1$ contains a $\mathbb{G}_{p_{5}}$ part that is related with the one of the simulated ciphertext. The remaining tokens are like those in TypeACtGame. We will prove that, under Assumption 3, TypeACtGame $\approx_{c}$ TypeAKeysGame.

Second Substep. In the second substep, we define a new experiment called TypeBKeysGame which differs from TypeAKeysGame in the way the token for matching queries of the second stage are generated. More specifically, if the adversary asks for the token for a vector $\vec{y}$ such that $\operatorname{HVE}(\vec{x}, \vec{y})=1$, then the answer does not contain a $\mathbb{G}_{p_{1}}$ part. The tokens for $\vec{y}$ such that $\operatorname{HVE}(\vec{x}, \vec{y})=0$ instead are computed as in TypeAKeysGame. We will prove that, under Assumption 4, TypeAKeysGame $\approx_{c}$ TypeBKeysGame.

Third Substep. Let $q_{2}$ be the number of token queries made by the adversary in the second stage. In the third substep we define, for $k=0, \ldots, q_{2}$, a new experiment called $H_{3, k}$ which differs from TypeBKeysGame in the way the the second stage token queries are answered. More specifically, the first $k$ tokens asked by the adversary are modified in the following way. The token for $\vec{y}$ such that $\operatorname{HVE}(\vec{x}, \vec{y})=0$ contains a random $\mathbb{G}_{p_{1}}$ part. Instead if $\operatorname{HVE}(\vec{x}, \vec{y})=1$ then the token is computed as in TypeBKeysGame. The tokens for the remaining $q_{2}-k$ queries are computed like in experiment TypeBKeysGame. We observe that TypeBKeysGame $=H_{3,0}$ and that $H_{3, q_{2}}=H_{3}$.
We will prove that, under Assumption 5, $H_{3, k-1} \approx_{c} H_{3, k}$.

## The first substep

Let us start by formally defining experiment TypeAKeysGame in terms of the algorithm Sim.A.KeyGen used to answer the token queries of the second stage.

Algorithm Sim.A.KeyGen receives in input $m s k, \vec{y}$ and $z=\operatorname{HVE}(\vec{y}, \vec{x})$ where $\vec{x}$ is the challenge plaintext provided by the adversary and distinguishes two cases.

- $z=0$ : The answer to the query is computed by running KeyGen on input $m s k$ and $\vec{y}$.
- $z=1$ : For each $j \in S_{\vec{y}}$, the algorithm chooses random $W_{j} \in \mathbb{G}_{p_{4}}$, random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}_{i}}} a_{j}=0$, and sets

$$
Y_{j}=g_{12}^{a_{j} / t_{j, y_{j}}} \cdot g_{5}^{a_{j} / s_{j}} \cdot W_{j}
$$

Notice that the vector $\left(s_{j}\right)_{j_{\in}[\ell]}$ is the vector stored in the state by Sim.A.Enc.
We define hybrid experiment TypeAKeysGame ${ }^{\mathcal{A}}$ as follows.

```
TypeAKeysGame \({ }^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right)\)
\(\left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} . \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ;\)
\((\vec{x}\), aux \() \leftarrow \mathcal{A}_{0}^{\text {KeyGen }(m s k, \cdot)}(m p k) ;\)
\(\mathrm{Ct} \leftarrow \operatorname{Sim} . \mathrm{A} . \operatorname{Enc}\left(m p k^{\prime}, \vec{x}\right) ;\)
\(\alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Sim} . \mathrm{A} . \operatorname{KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x} \cdot \cdot))}(m p k, \mathrm{Ct}\), aux \() ;\)
Output: \((m p k, m, \alpha)\)
```

Lemma 6.2.10 If the GSD Assumption holds then, for all PPT $\mathcal{A}, H_{2}^{\mathcal{A}} \approx_{c}$ TypeAKeysGame ${ }^{\mathcal{A}}$.
Proof. Suppose there exists an adversary $\mathcal{A}$ for which TypeACtGame ${ }^{\mathcal{A}}$ and TypeAKeysGame ${ }^{\mathcal{A}}$ are distinguishable. Then, we show a PPT algorithm $\mathcal{B}$ that receives the following instance of the GSD Assumption, $\left(N, T,\left\{g_{1}, g_{2}, g_{3}, g_{4}, g_{15}\right\}\right) \leftarrow \operatorname{GSDGen}\left(1^{\lambda},\{1\},\{1,5\},\{1\},\{2\},\{3\},\{4\}\right.$, $\{1,5\}$ ) and, depending on the nature of $T$, simulates TypeACtGame ${ }^{\mathcal{A}}$ or TypeAKeysGame ${ }^{\mathcal{A}}$. This suffices to prove the Lemma.

Setup: $\mathcal{B}$, for each $i \in[\ell]$ and $b \in\{0,1\}, \mathcal{B}$ chooses random $t_{i, b} \in \mathbb{Z}_{N}, R_{i, b}, R_{i, b}^{\prime} \in \mathbb{G}_{p_{3}}$ and sets $T_{i, b}=g_{2}^{t_{i, b}} \cdot R_{i, b}$. Then $\mathcal{B}$ sets

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and

$$
m s k=\left[g_{12}, g_{4},\left(t_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and starts the interaction with $\mathcal{A}$ on input $m p k$.
First Stage Token Queries: $\mathcal{B}$ computes the token for $\vec{y}$ by running KeyGen on input $m s k$ and $\vec{y}$.

Ciphertext: $\mathcal{B}$ generates the simulated ciphertext for vector $\vec{x}$ as follows. $\mathcal{B}$ chooses random $s \in \mathbb{Z}_{N}$ and, for $i \in[\ell]$, random $\mathbb{Z}_{i} \in \mathbb{G}_{p_{3}}$, and sets

$$
X_{i}=g_{15}^{s \cdot t_{i, x_{i}}} \cdot Z_{i}
$$

Second Stage Token Queries: $\mathcal{B}$ generates the token for $\vec{y}$ in the following way.
If $\operatorname{HVE}(\vec{x}, \vec{y})=0$, the token is computed by running KeyGen on input $m s k$ and $\vec{y}$.
If $\operatorname{HVE}(\vec{x}, \vec{y})=1$, the token is computed as follows. For each $j \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$. Then $\mathcal{B}$, for each $j \in S_{\vec{y}}$, sets:

$$
Y_{j}=T^{a_{j} / t_{j, y_{j}}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

This concludes the description of algorithm $\mathcal{B}$.
Now we observe that the output of $\mathcal{B}$ 's setup is distributed like the output of algorithm Sim.Setup and thus like in $H_{2}$ and TypeAKeysGame. Similarly, the simulated ciphertext is distributed like the output of A.Enc on input $m p k^{\prime}$ and $\vec{x}$ and thus exactly as in $H_{2}$ and TypeAKeysGame. Finally, let us consider the answers to the second stage queries. If $T \in \mathbb{G}_{p_{1}}$ then the answer to the query for $\vec{y}$ has the same distribution as the output of KeyGen on input $m s k$ and $\vec{y}$ and thus it is distributed as in $H_{2}$. On the other hand, if $T \in \mathbb{G}_{p_{1} p_{5}}$ then the answer to the query for $\vec{y}$ has the same distribution as the output of Sim.A.KeyGen on input $m s k, \vec{y}$ and $\operatorname{HVE}(\vec{x}, \vec{y})$ just like in TypeAKeysGame.

## The second substep

Let us start by formally defining experiment TypeBKeysGame in terms of the algorithm Sim.B.KeyGen that is used to answer the token queries of the second stage.

Algorithm Sim.B.KeyGen receives in input $m s k, \vec{y}$ and $z=\operatorname{HVE}(\vec{y}, \vec{x})$, where $\vec{x}$ is the challenge plaintext provided by the adversary and distinguishes two cases.

- $z=0$ : The answer to the query is computed by running KeyGen on input $m s k$ and $\vec{y}$.
- $z=1$ : For each $j \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$ and sets

$$
Y_{j}=g_{2}^{a_{j} / t_{j, y_{j}}} \cdot g_{5}^{a_{j} / s_{j}} \cdot W_{j}
$$

Notice that vector $\left(s_{j}\right)_{j \in[\ell]}$ is the vector stored in the state by Sim.A.Enc.
We define hybrid experiment TypeBKeysGame ${ }^{\mathcal{A}}$ as follows.

$$
\begin{aligned}
& \text { TypeBKeysGame } \mathcal{A}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} . \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) \\
& (\vec{x}, \text { aux }) \leftarrow \mathcal{A}_{0}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k) \\
& \operatorname{Ct} \leftarrow \operatorname{Sim} . \operatorname{A.Enc}\left(m p k^{\prime}, \vec{x}\right) ; \\
& \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Sim} . \operatorname{B.KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot))}(m p k, \mathrm{Ct}, \text { aux }) ; \\
& \text { Output: }(m p k, m, \alpha)
\end{aligned}
$$

Lemma 6.2.11 If the GSD Assumption holds then, for all PPT $\mathcal{A}$, TypeAKeysGame ${ }^{\mathcal{A}} \approx_{c}$ TypeBKeysGame ${ }^{\mathcal{A}}$.

Proof. Suppose there exists an adversary $\mathcal{A}$ for which TypeAKeysGame ${ }^{\mathcal{A}}$ and TypeBKeysGame ${ }^{\mathcal{A}}$ are distinguishable. Then, we show a PPT algorithm $\mathcal{B}$ that receives the following instance of the GSD Assumption, $\left(N, T,\left\{g_{2}, g_{3}, g_{4}, g_{14}, g_{15}\right\}\right) \leftarrow \operatorname{GSDGen}\left(1^{\lambda},\{1,4,5\},\{4,5\},\{2\},\{3\},\{4\}\right.$, $\{1,4\},\{1,5\}$ ) and, depending on the nature of $T$, simulates TypeAKeysGame ${ }^{\mathcal{A}}$ or TypeBKeysGame ${ }^{\mathcal{A}}$. This suffices to prove the Lemma.

Setup: $\mathcal{B}$, for each $i \in[\ell]$ and $b \in\{0,1\}, \mathcal{B}$ chooses random $t_{i, b} \in \mathbb{Z}_{N}, R_{i, b}, R_{i, b}^{\prime} \in \mathbb{G}_{p_{3}}$ and sets $T_{i, b}=g_{2}^{t_{i, b}} \cdot R_{i, b}$. Then $\mathcal{B}$ sets

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and

$$
m s k=\left[\perp, g_{4},\left(t_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and starts the interaction with $\mathcal{A}$ on input $m p k$. Notice that $\mathcal{B}$ is unable to compute a complete $m s k$ as it does not have access to an element from $\mathbb{G}_{p_{1} p_{2}}$.

First Stage Token Queries: $\mathcal{B}$ generates the token for $\vec{y}$ in the following way. For each $i \in S_{\vec{y}}$, $\mathcal{B}$ chooses random $W_{i} \in \mathbb{G}_{p_{4}}$, random $a_{i} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{i \in S_{\vec{y}}} a_{i}=0$ and then sets

$$
Y_{i}=g_{14}^{a_{i} / t_{i, y_{i}}} \cdot g_{2}^{a_{i} / t_{i, y_{i}}} \cdot W_{i}
$$

Ciphertext: $\mathcal{B}$ generates the simulated ciphertext for vector $\vec{x}$ as follows. $\mathcal{B}$ chooses random $s \in \mathbb{Z}_{N}$ and, for $i \in[\ell]$, random $\mathbb{Z}_{i} \in \mathbb{G}_{p_{3}}$, and sets

$$
X_{i}=g_{15}^{s \cdot t_{i, x_{i}}} \cdot Z_{i}
$$

Second Stage Token Queries: On input $m s k, \vec{y}$ and $z=\operatorname{HVE}(\vec{x}, \vec{y}), \mathcal{B}$ generates the token for vector $\vec{y}$ in the following way.
For each $j \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{y}} a_{j}=0$. Then $\mathcal{B}$ distinguishes two cases.
$z=0:$ For each $j \in S_{\vec{y}}, \mathcal{B}$ sets

$$
Y_{j}=g_{14}^{a_{j} / t_{j, y_{j}}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

$z=1:$ For each $j \in S_{\vec{y}}, \mathcal{B}$ sets

$$
Y_{j}=T^{a_{j} / t_{j, y_{j}}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

This concludes the description of algorithm $\mathcal{B}$.
Let us now look at the answers to the token queries of the second stage. It is easy to see that queries for $\vec{y}$ such that $\operatorname{HVE}(\vec{x}, \vec{y})=0$ are distributed as the output of KeyGen. Suppose that $T \in$ $\mathbb{G}_{p_{1 p_{4} p_{5}}}$. Then the answers to matching queries are distributed like in the output of algorithm Sim.A.KeyGen and thus like in TypeAKeysGame. On the other hand, if $T \in \mathbb{G}_{p_{4} p_{5}}$ then the answers to matching queries are distributed like in the output of algorithm Sim.B.KeyGen and thus like in TypeBKeysGame.

## The third substep

In this section we prove that, for any PPT adversary $\mathcal{A}$, TypeBKeysGame ${ }^{\mathcal{A}}$ and $H_{3}^{\mathcal{A}}$ are indistinguishable, under Assumption 5.

For any adversary $\mathcal{A}$ that makes $q_{2}$ queries in the second stage, we define a sequence of hybrid experiments $H_{3, k}^{\mathcal{A}}$, for $k=0, \ldots, q_{2}$, such that

$$
H_{3,0}^{\mathcal{A}}=\text { TypeBKeysGame }{ }^{\mathcal{A}} \quad \text { and } \quad H_{3, q_{2}}^{\mathcal{A}}=H_{3}^{\mathcal{A}} .
$$

Then if TypeBKeysGame ${ }^{\mathcal{A}}$ and $H_{3}^{\mathcal{A}}$ can be distinguished then there must exist $k \in\left\{1, \ldots, q_{2}\right\}$ such that hybrid experiments $H_{3, k-1}^{\mathcal{A}}$ and $H_{3, k}^{\mathcal{A}}$ are indistinguishable. Moreover, from the definition of experiment $H_{3, k}$, it is clear that if $k$-th token query of the second stage is a matching query (that is, $\operatorname{HVE}(\vec{x}, \vec{y})=1$ ) then $H_{3, k-1}^{\mathcal{A}}$ and $H_{3, k}^{\mathcal{A}}$ coincide. Therefore, if $\mathcal{A}$ is such that TypeBKeysGame ${ }^{\mathcal{A}}$ and $H_{3}^{\mathcal{A}}$ can be distinguished there must exist $k$ such that $H_{3, k-1}^{\mathcal{A}}$ and $H_{3, k}^{\mathcal{A}}$ can be distinguished and the $k$-th token query of $\mathcal{A}$ in the second stage is with non-negligible probability a non-matching query (that is, $\operatorname{HVE}(\vec{x}, \vec{y})=0$ ).

Let us now define experiment $H_{3, k}$. In $H_{3, k}$ second stage token queries are answered by running a parametrized version of algorithm Sim.KeyGen that, with a slight abuse of notation, we also call Sim.KeyGen.

Algorithm Sim.KeyGen receives as input the master secret key $m s k$, the vector $\vec{y}$ for which the token must be computed and the value $z=\operatorname{HVE}(\vec{x}, \vec{y})$. In addition, Sim.KeyGen receives the number $i$ of the query and the value $k$.
For each $j \in S_{\vec{y}}$, Sim. KeyGen selects random $W_{j} \in \mathbb{G}_{p_{4}}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$. Then, it distinguishes two cases.

- $z=0$ : The depending on the query index $i, \mathcal{B}$ does the following.
- If $i \leq k$ : For each $j \in S_{\vec{y}}$, Sim.KeyGen chooses random $c_{j} \in \mathbb{Z}_{N}$ and sets

$$
Y_{j}=g_{1}^{c_{j}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

- If $i>k$ : For each $j \in S_{\vec{y}}$, Sim.KeyGen sets

$$
Y_{j}=g_{12}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

- $z=1$ : For each $j \in S_{\vec{y}}$, Sim.KeyGen sets

$$
Y_{j}=g_{2}^{a_{j} / t_{j, y_{j}}} \cdot g_{5}^{a_{j} / s_{j}} \cdot W_{j} .
$$

We remind the reader that the vector $\left(s_{j}\right)_{j \in[\ell]}$ is the vector stored in the state of the simulator.

We observe that for algorithm Sim.B.KeyGen is the parametrized version of Sim.KeyGen with $k=0$. On the other hand, algorithm Sim.KeyGen is the parametrized version of Sim.KeyGen with $k=q_{2}$.

Next we define hybrid experiment $H_{3, k}^{\mathcal{A}}$, for an adversary $\mathcal{A}$ that ask $q_{2}$ token queries in the second stage and for $k=0, \ldots, q_{2}$,

$$
\begin{aligned}
& H_{3, k}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} \cdot \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) \\
& (\vec{x}, \text { aux }) \leftarrow \mathcal{A}_{0}^{\operatorname{KeyGen}(m s k, \cdot)}(m p k) \\
& \mathrm{Ct} \leftarrow \operatorname{Sim} . \operatorname{A.Enc}\left(m p k^{\prime}, \vec{x}\right) ; \\
& \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Sim} \cdot \operatorname{KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k)}(m p k, \mathrm{Ct}, \text { aux }) ; \\
& \text { Output: }(m p k, \vec{x}, \alpha)
\end{aligned}
$$

Notice that $H_{3,0}=$ TypeBKeysGame and $H_{3, q_{2}}=H_{3}$.
Lemma 6.2.12 If Assumption 1 holds then, for any PPT adversary $\mathcal{A}$, TypeBKeysGame ${ }^{\mathcal{A}} \approx_{c}$ $H_{3}^{\mathcal{A}}$

Proof. Suppose there exists an adversary $\mathcal{A}$ such that TypeBKeysGame and $H_{3}$ are distinguishable. We show a PPT algorithm $\mathcal{B}$ that receives an instance of Assumption 1, consisting of $\left(N, A_{1}, A_{2}, A_{3}, A_{4}, A_{5}, A_{1}^{\alpha} \cdot B_{4}, A_{1}^{\beta} \cdot C_{4}\right)$ and challenge $T$, and, depending on the nature of $T$, simulates $H_{3, k-1}^{\mathcal{A}}$ or $H_{3, k}^{\mathcal{A}}$ with some non-negligible probability for a random $k \in\left\{1, \ldots, q_{2}\right\}$. This suffices to prove the Lemma.

Setup: $\mathcal{B}$ randomly chooses position $\mathrm{j} \in[\ell]$ and $b_{\mathrm{j}} \in\{0,1\}$.
$\mathcal{B}$ sets $g_{1}=A_{1}, g_{2}=A_{2}, g_{12}=A_{1} \cdot A_{2}, g_{3}=A_{3}, g_{4}=A_{4}, g_{5}=A_{5}$ and, for each $i \in[\ell]$ and $b \in\{0,1\}, \mathcal{B}$ chooses random $t_{i, b} \in \mathbb{Z}_{N}, R_{i, b}, R_{i, b}^{\prime} \in \mathbb{G}_{p_{3}}$ and sets $T_{i, b}=g_{2}^{t_{i, b}} \cdot R_{i, b}$. Finally, $\mathcal{B}$ sets

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[\ell], b \in\{0,1\}}\right]
$$

and

$$
m s k=\left[g_{12}, g_{4},,\left(t_{i, b}\right)_{(i, b) \in([\ell] \times\{0,1\}) \backslash\left\{\left(\mathrm{j}, b_{\mathrm{j}}\right)\right\}}\right]
$$

and starts the interaction with $\mathcal{A}$ on input $m p k$.

First Stage Token Queries: In the first stage, $\mathcal{B}$ generates token for $\vec{y}$ in the following way. For each $i \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{i} \in \mathbb{G}_{p_{4}}$, random $a_{i} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{i \in S_{\vec{y}}} a_{i}=0$ and then sets

$$
Y_{i}=g_{12}^{a_{i} / t_{i, y_{i}}} \cdot W_{i}
$$

However, if $y_{\mathrm{j}}=b_{\mathrm{j}}$ then $\mathcal{B}$ sets

$$
Y_{\mathrm{j}}=\left(A_{1}^{\beta} \cdot C_{4}\right)^{a_{i}} \cdot g_{2}^{a_{\mathrm{j}} / \mathrm{t}_{\mathrm{j}, y_{\mathrm{j}}}} \cdot W_{\mathrm{j}} .
$$

This last setting has the effect of implicitly setting $t_{\mathrm{j}, y_{\mathrm{j}}} \equiv 1 / \beta\left(\bmod p_{1}\right)$.
Ciphertext: $\mathcal{B}$ generate the challenge ciphertext for vector $\vec{x}$ in the following way. If $x_{\mathrm{j}}=b_{\mathrm{j}}$ then $\mathcal{B}$ aborts. Otherwise $\mathcal{B}$ chooses random $s \in \mathbb{Z}_{N}$ and, for $i \in[\ell]$, random $\mathbb{Z}_{i} \in \mathbb{G}_{p_{3}}$, and sets

$$
X_{i}=g_{1}^{s \cdot t_{i}, x_{i}} \cdot g_{5}^{s \cdot t_{i, x_{i}}} \cdot Z_{i}
$$

Second Stage Token Queries: At the start of the second stage of token queries, $\mathcal{B}$ picks a ran$\operatorname{dom} k \in\left\{1, \ldots, q_{2}\right\}$.
The $i$-th token query of the second stage for vector $\vec{y}$ is answered by $\mathcal{B}$ in the following way by distinguishing the following two cases.
$\operatorname{HVE}(\vec{x}, \vec{y})=0$ : We distinguish between the following three cases
$i<k$ : For each $j \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}, c_{j} \in \mathbb{Z}_{N}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$ and sets

$$
Y_{j}=g_{1}^{c_{j}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

$i=k: \quad$ If $y_{\mathrm{j}} \neq b_{\mathrm{j}}$ then $\mathcal{B}$ aborts.
Otherwise let $h \in[\ell]$ be such that $h \neq \mathrm{j}$ and $y_{h} \neq \star$ and set $S=S_{\vec{y}} \backslash\{\mathrm{j}, h\}$, Notice that such an $h$ always exists since we assumed that each query contains at least two non-» entries.
Then, for each $j \in S, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}$ and random $a_{j} \in \mathbb{Z}_{N}$ and sets

$$
Y_{j}=g_{12}^{a_{j} / t_{j, y_{j}}} \cdot W_{j} .
$$

Then, for position $\mathrm{j}, \mathcal{B}$ chooses random $W_{\mathrm{j}} \in \mathbb{G}_{p_{4}}$ and random $a_{\mathrm{j}} \in \mathbb{Z}_{N}$ and sets

$$
Y_{\mathrm{j}}=T \cdot g_{2}^{a_{\mathrm{j}} / t_{\mathrm{j}, y_{\mathrm{j}}}} \cdot W_{\mathrm{j}} .
$$

Finally, for position $h, \mathcal{B}$ sets

$$
Y_{h}=\left(A_{1}^{\alpha} B_{4}\right)^{-1 / t_{h, y_{h}}} \cdot g_{1}^{-s / t_{h, y_{h}}} \cdot g_{2}^{-\left(s+a_{\mathrm{j}}\right) / t_{h, y_{h}}} \cdot W_{h},
$$

where $s=\sum_{j \in S} a_{j}$.
$i>k$ : For each $j \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}$ and $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$ and sets

$$
Y_{j}=g_{12}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

In the special case that $y_{\mathrm{j}}=b_{\mathrm{j}}$, then $\mathcal{B}$ sets

$$
Y_{\mathrm{j}}=\left(A_{1}^{\beta} \cdot C_{4}\right)^{a_{\mathrm{j}}} \cdot g_{2}^{a_{\mathrm{j}} / \mathrm{t}_{\mathrm{j}, y_{\mathrm{j}}}} \cdot W_{\mathrm{j}}
$$

$\operatorname{HVE}\left(\vec{x}, \vec{y}_{i}\right)=1:$ For each $j \in S_{\vec{y}}, \mathcal{B}$ chooses random $W_{j} \in \mathbb{G}_{p_{4}}$ and $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\vec{y}}} a_{j}=0$ and sets

$$
Y_{j}=g_{2}^{a_{j} / t_{j, y_{j}}} \cdot g_{5}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

This concludes the description of algorithm $\mathcal{B}$.
Let us now prove that the probability that $\mathcal{B}$ does not abort while interacting with $\mathcal{A}$ is non-negligible. First observe that the probability that $\mathcal{B}$ aborts while constructing the simulated ciphertext is $1 / 2$. Indeed, the view of $\mathcal{A}$ up to this point is independent from j and $b_{\mathrm{j}}$ and thus, since $b_{\mathrm{j}}$ is chosen at random by $\mathcal{B}$, the probability that $x_{\mathrm{j}}=b_{\mathrm{j}}$ is $1 / 2$. Let us now look at the probability that $\mathcal{B}$ aborts while answering the $k$-th token query of the second stage of $\mathcal{A}$, given that it has not aborted in the construction of the simulated ciphertext. In this case the view of $\mathcal{A}$ is independent from j and also remember that the probability that the $k$-th query of $\mathcal{A}$ is non-matching is non-negligible. If this is the case then there must exist one position $j$ such that $y_{j} \neq \star$ and $y_{j} \neq x_{j}$. If $j=\mathrm{j}$ (which happens with non-negligible probability since $\mathcal{A}$ 's view is independent from the value of $j$ ) then $\mathcal{B}$ does not abort. Indeed, in this case we would have $y_{\mathrm{j}} \neq x_{j} \neq b_{j}$ which implies that $y_{\mathrm{j}}=b_{j}$.

Let us now look at the view of $\mathcal{A}$ while interacting with $\mathcal{B}$. We observe that $m p k$ has the same distribution of the corresponding output of Sim.Setup. Even though $\mathcal{B}$ does not have a complete $m s k$ as it misses $t_{\mathrm{j}, b_{\mathrm{j}}}$, the answer of the first stage queries are distributed as the output of algorithm KeyGen on input $m s k$ in which $t_{\mathrm{j}, b_{\mathrm{j}}} \equiv 1 / \beta\left(\bmod p_{1}\right)$. Given that $\mathcal{B}$ does not abort, it is straightforward to see that Ct constructed by $\mathcal{B}$ has the same distribution as the output of $\mathrm{Ct} \leftarrow$ Sim.A.Enc on input $m p k^{\prime}$ and $\vec{x}$. Let us now look at the answers of the second stage queries. The matching queries and the first $k-1$ non-matching queries have the same distribution of the output of algorithms Sim.KeyGen $(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k)$ and $\operatorname{Sim} . \operatorname{KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k-$ 1). Notice that, for both types of queries, the missing element of $m s k$ plays no role. Similarly, the last $q_{2}-k$ non-matching queries of the second stage are distributed as the output $\operatorname{Sim} . \operatorname{KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k)$ and $\operatorname{Sim} . \operatorname{KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k-1)$ in which the missing element of $m s k$ is set equal to $t_{\mathrm{j}, b_{\mathrm{j}}} \equiv 1 / \beta\left(\bmod p_{1}\right)$. Let us finally look at the answer to the $k$-th query. If $T=A_{1}^{\alpha \beta} \cdot D_{4}$ then the answer of the query is distributed according to the output of Sim.KeyGen $(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k-1)$. Notice that if $B$ does not abort the missing element of $m s k$ plays no role. On the other hand, if $T$ is random $\mathbb{G}_{p_{1} p_{4}}$ then the answer of the query is distributed according Sim.KeyGen $(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot), \cdot, k)$.

### 6.2.8 The fourth step of the proof

In this section we prove that $H_{3}$ is indistinguishable from IdealExp.
Lemma 6.2.13 If Assumption 1 holds then, for all PPT adversaries $\mathcal{A}, H_{3}^{\mathcal{A}} \approx_{c}$ IdealExp $^{\mathcal{A}}$.
We start by defining $\ell+1$ intermediate hybrid experiments $I_{1}, \ldots, I_{\ell+1}$ such that $I_{1}=H_{3}$ and $I_{\ell+1}=$ IdealExp and show that, for $f=1, \ldots, \ell+1, I_{f}$ and $I_{f+1}$ are indistinguishable, under Assumption 1.

To define $I_{f}$, we introduce a parametrized version of algorithm Sim.Enc that, with a slight abuse of notation, we also call Sim.Enc.

The parametrized version of $\operatorname{Sim}$. Enc takes as input public key $m p k^{\prime}=\left[N, g_{3}, g_{5},\left(T_{i, b}^{\prime}\right)_{i \in[\ell], b \in\{0,1\}}\right]$, the challenge ciphertext $\vec{x}$, the sequence $\left(\vec{y}_{k}, z_{k}\right)_{k=1}^{q_{1}}$ of the $q_{1}$ queries asked by the adversary in the first stage along with $z_{k}=\operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)$. In addition we let Sim.Enc take parameter $0 \leq f \leq \ell$. Sim. Enc returns a simulated ciphertext in which the $\mathbb{G}_{p_{1}}$ part of
positions $i \leq f$ that do not belong to MPos is random. The remaining positions are well formed.
More formally, Sim. Enc chooses random $s \in \mathbb{Z}_{N}$, and, for each $i \in[\ell]$, random $Z_{i} \in \mathbb{G}_{p_{3}}$ and random $s_{i} \in \mathbb{Z}_{N}$. Then for each $i \in[\ell]$, Sim. Enc distinguishes the following cases.

- if $i<f$ and $i \notin$ MPos, Sim. Enc randomly selects $r_{i} \in \mathbb{Z}_{N}$ and sets

$$
X_{i}=T_{i, 0}^{\prime r_{i}} \cdot g_{5}^{s_{i}} \cdot Z_{i}
$$

- if $i<f$ and $i \in$ MPos, Sim. Enc sets

$$
X_{i}=T_{i, x_{i}}^{\prime \prime} \cdot g_{5}^{s_{i}} \cdot Z_{i}
$$

- if $i \geq f$ Sim.Enc sets

$$
X_{i}=T_{i, x_{i}}^{\prime} \cdot g_{5}^{s_{i}} \cdot Z_{i}
$$

Sim. Enc returns the simulated ciphertext $\mathrm{Ct}=\left(X_{i}\right)_{i \in[\ell]}$ and stores the vector $\left(s_{i}\right)_{i \in[\ell]}$ in the state. Notice that if $f=\ell+1$ the input $\vec{x}$ is not used by the algorithm and we obtain algorithm Sim.Enc. On the other hand, if $f=1$, we obtain algorithm Sim.A.Enc.

Next, we define, for $1 \leq f \leq \ell+1$, experiment $I_{f}^{\mathcal{A}}$ as follows.

$$
\begin{aligned}
& I_{f}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} . \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ; \\
& (\vec{x}, \operatorname{aux}) \leftarrow \mathcal{A}_{0}^{\operatorname{KeyGen}(m s k,)}(m p k) ; \\
& \mathrm{Ct} \leftarrow \operatorname{Sim} \cdot \operatorname{Enc}\left(m p k^{\prime}, \vec{x},\left(\vec{y}_{k}, \operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)\right){ }_{k=1}^{q_{1}}, f\right) ; \\
& \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Sim} \cdot \operatorname{KeyGen}(m s k, \cdot, \operatorname{HVE}(\vec{x}, \cdot))}(m p k, \mathrm{Ct}, \text { aux }) ; \\
& \text { Output: }(m p k, \vec{x}, \alpha)
\end{aligned}
$$

Clearly, for all PPT adversaries $\mathcal{A}, I_{1}^{\mathcal{A}}=H_{3}$ and $I_{\ell+1}^{\mathcal{A}}=I$ dealExp ${ }^{\mathcal{A}}$. Therefore to prove Lemma 6.2.13, it is enough to prove the following lemma.

Lemma 6.2.14 If Assumption 1 holds, then for all PPT adversaries $\mathcal{A}$, and for $f=1, \ldots, \ell$, $I_{f}^{\mathcal{A}} \approx_{c} I_{f+1}^{A}$.

To prove the above lemma, we introduce another sequence of intermediate games and make the following observation.

Observation 6.2.15 If $f \in$ MPos, the output distributions of $\operatorname{Sim} . \operatorname{Enc}\left(m p k, \vec{x},\left(\vec{y}_{k}, H V E\left(\vec{y}_{k}, \vec{x}\right)_{k=1}^{q}\right.\right.$, $f)$ and $\operatorname{Sim} . \operatorname{Enc}\left(m p k, \vec{x},\left(\vec{y}_{k}, H V E\left(\vec{y}_{k}, \vec{x}\right)_{k=1}^{q}, f+1\right)\right.$ coincide.

Therefore if $I_{f}^{\mathcal{A}}$ and $I_{f+1}^{\mathcal{A}}$ are distinguishable then it must be the case that $\mathcal{A}$ has a nonnegligible probability of outputting a challenge plaintext $\vec{x}$ such that $f \notin$ MPos. For an adversary $\mathcal{A}$ that makes $q_{1}$ first stage token queries we introduce $2 \cdot(q+1)$ intermediate hybrid experiments $L_{f, 0}^{\mathcal{A}}, \ldots, L_{f, q_{1}}^{\mathcal{A}}$ and $M_{f, q_{1}}^{\mathcal{A}}, \ldots, M_{f, 0}^{\mathcal{A}}$ which differ in the way in which first stage token queries are answered. Specifically, first stage queries are answered by running the following algorithm.

Algorithm Sim.C.KeyGen takes as input the master secret key $m s k$, the query $\vec{y}$ for which a token has to be computed, the number $1 \leq i \leq q_{1}$ of the query, integer $1 \leq f \leq \ell+1$ and integer $0 \leq k \leq q_{1}$. The algorithm distinguishes the following cases.

- $i \leq k$ and $y_{f} \neq \star$ :

The $\mathbb{G}_{p_{1}}$ part of the token is random.
For each $j \in S_{\vec{y}}$, the algorithm chooses random $W_{j} \in \mathbb{G}_{p_{4}}$, random $c_{j} \in \mathbb{Z}_{N}$ and random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{y}} a_{j}=0$ and sets

$$
Y_{j}=g_{1}^{c_{j}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

- $i \leq k$ and $y_{f}=\star$ : the algorithm returns the output of $\operatorname{KeyGen}(m s k, \vec{y})$.
- $i>k$ : the algorithm returns the output of $\operatorname{KeyGen}(m s k, \vec{y})$.

We are now ready to describe experiments $L_{f, k}^{\mathcal{A}}$ and $M_{f, k}^{\mathcal{A}}$ for $f=1, \ldots, \ell+1$ and $k=$ $0, \ldots, q_{1}$.

$$
\begin{aligned}
& L_{f, k}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} . \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ; \\
& (\vec{x}, \text { aux }) \leftarrow \mathcal{A}_{0}^{\text {Sim.C.KeyGen }(m s k, \cdot, \cdot, f, k)}(m p k) ; \\
& \mathrm{Ct} \leftarrow \operatorname{Sim} . \operatorname{Enc}\left(m p k^{\prime}, \vec{x},\left(\vec{y}_{k}, \operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)\right)_{k=1}^{q}, f-1\right) ; \mathrm{Ct} \leftarrow \operatorname{Sim} \operatorname{Enc}\left(m p k^{\prime}, \vec{x},\left(\vec{y}_{k}, \operatorname{HVE}\left(\vec{y}_{k}, \vec{x}\right)\right)_{k=1}^{q}, f\right) ; \\
& \alpha \leftarrow \mathcal{A}_{1}^{\text {Sim.KeyGen }(m s k,, \cdot \operatorname{HVE}(\vec{x}, \cdot))}(m p k, \mathrm{Ct}, \text { aux }) ; \quad \alpha \leftarrow \mathcal{A}_{1}^{\operatorname{Sim} \cdot \operatorname{KeyGen}(m s k,, \operatorname{HVE}(\vec{x}, \cdot))}(m p k, \mathrm{Ct}, \text { aux }) ; \\
& \text { Output: }(m p k, m, \alpha) \\
& M_{f, k}^{\mathcal{A}}\left(1^{\lambda}, 1^{\ell}\right) \\
& \left(m p k, m p k^{\prime}, m s k\right) \leftarrow \operatorname{Sim} \cdot \operatorname{Setup}\left(1^{\lambda}, 1^{\ell}\right) ; \\
& (\vec{x}, \mathrm{aux}) \leftarrow \mathcal{A}_{0}^{\text {Sim.C.KeyGen }(m s k, \cdot,, f, k)}(m p k) ; \\
& \text { Output: }(m p k, m, \alpha)
\end{aligned}
$$

Before continuing, we observe the following:

Observation 6.2.16 For all PPT adversaries $\mathcal{A}, I_{f}^{\mathcal{A}}=L_{f, 0}$.
Directly from the definition of the experiments.
Observation 6.2.17 For all PPT adversaries $\mathcal{A}, L_{f, q_{1}}^{\mathcal{A}}=M_{f, q_{1}}^{\mathcal{A}}$ for $f=1, \ldots, \ell$, where $q_{1}$ is the number of first stage queries made by $\mathcal{A}$.
From the definitions of the two experiments, it is clear that all the token queries are answered in the same way in both the experiments and all components $X_{i}$ for $i \neq f$ of the challenge ciphertext are computed in the same way. Let us now look at $X_{f}$ and more precisely to its $\mathbb{G}_{p_{1}}$ part. In $L_{f, q_{1}}$, the $\mathbb{G}_{p_{1}}$ part of $X_{f}$ is computed as $T_{f, x_{f}}^{\prime s}$ which is exactly how it is computed in $M_{f, q_{1}}$ when $f \in$ MPos. On the other hand, when $f \notin$ MPos, the $\mathbb{G}_{p_{1}}$ part of $X_{f}$ is chosen at random. However, observe that exponents $t_{f, 0} \bmod p_{1}$ and $t_{f, 1} \bmod p_{1}$ have not appeared in the answers to key queries since every query has either $\mathrm{a} \star$ in position $f$ (in which case position $f$ of the answer is empty) or a non- $\star$ value in position $f$ (in which case the $\mathbb{G}_{p_{1}}$ part of the element in position $f$ of the answer is random). Therefore, we can conclude that the $\mathbb{G}_{p_{1}}$ part of the component $X_{f}$ of the answer to the challenge query is also random in $\mathbb{G}_{p_{1}}$.

Observation 6.2.18 For all PPT adversaries $\mathcal{A}$ and for $f=1, \ldots, \ell-1, M_{f, 0}^{\mathcal{A}}=L_{f+1,0}^{\mathcal{A}}$. Indeed, in both experiments all key queries are answered correctly, and the challenge query in $M_{f, 0}$ is by definition answered in the same way as in $L_{f+1,0}$.

By the above observations, for proving Lemma 6.2 .14 it suffices to prove that $L_{f, k-1}$ and $L_{f, k}$ are indistinguishable and that $M_{f, k-1}$ and $M_{f, k}$ are indistinguishable, for $k=1, \ldots, q_{1}$. In the next section we prove that, under Assumption 1, $L_{f, k-1}$ and $L_{f, k}$ are indistinguishable. The proof that $M_{f, k-1}$ and $M_{f, k}$ are indistinguishable is similar and omitted.

Indistinguishability of $L_{f, k}$ and $L_{f, k-1}$
We start by describing an algorithm B that takes as input $f$ and $k$ and an instance of Assumption 1 and interacts with an adversary $\mathcal{A}$. Then, provided that $\mathcal{A}$ outputs a challenge such that $f \notin$ MPos, B simulates with some non-negligible probability $L_{f, k}$ or $L_{f, k-1}$ depending on the nature of the challenge. This suffices to prove that the two hybrids are indistinguishable.

## Description of algorithm $B$

Input: Integers $1 \leq f \leq \ell+1$ and $0 \leq k \leq q$, and a randomly chosen instance of Assumption 1 consisting of $D=\left(N, A_{1}, A_{2}, A_{3}, A_{4}, A_{5}, A_{1}^{\alpha} B_{4}, A_{1}^{\beta} C_{4}\right)$ and challenge $T$ which is either $T=A_{1}^{\alpha \beta} D_{4}$ or random $\mathbb{G}_{p_{1} p_{4}}$.

Setup: B starts by constructing public parameters mpk. B sets $g_{1}=A_{1}, g_{2}=A_{2}, g_{12}=$ $A_{1} \cdot A_{2}, g_{3}=A_{3}, g_{4}=A_{4}, g_{5}=A_{5}$ and, for each $i \in[\ell]$ and $b \in\{0,1\}, \mathrm{B}$ chooses random $t_{i, b} \in \mathbb{Z}_{N}, R_{i, b} \in \mathbb{G}_{p_{3}}$ and sets $T_{i, b}=g_{2}^{t_{i, b}} \cdot R_{i, b}$. Then B sets

$$
m p k=\left[N, g_{3},\left(T_{i, b}\right)_{i \in[l], b \in\{0,1\}}\right],
$$

and starts the interaction with $\mathcal{A}$ on input mpk.
Now, B guesses a position $\mathrm{j} \in[\ell]$ and a bit $b_{\mathrm{j}} \in\{0,1\}$ and chooses the following randomness that will be used in the $\mathbb{G}_{p_{1}}$ subgroup. Specifically, B chooses, for each $i \in[\ell]$ and $i \neq \mathrm{j}$, and $b \in\{0,1\}$, random $t_{i, j}^{\prime} \in \mathbb{Z}_{N}$. Moreover, B chooses random $t_{\mathrm{j}, c_{\mathrm{j}}}^{\prime} \in \mathbb{Z}_{N}$ where $c_{\mathrm{j}}=1-b_{\mathrm{j}}$. Notice that the value $t_{\mathrm{j}, b_{\mathrm{j}}}^{\prime}$ is unknown to B. It will be provided by the assumption as $\beta$ at the exponent.

First Stage Secret Tokens: To simulate the output of Sim.C.KeyGen $\left(\mathrm{Sk}, \vec{y}_{i}, f, k\right)$, B does the following way:
$i \leq k$ : We have the following mutually exclusive cases.
Case A.1: $y_{f} \neq \star$. In this case, B outputs a key whose $\mathbb{G}_{p_{1}}$ part is random.
Specifically, for each $j \in S_{\vec{y}_{i}}$, B chooses random $a_{j}$ such that $\sum_{j \in S_{\vec{y}_{j}}} a_{i}=0$, random $r_{j} \in \mathbb{Z}_{N}$, and random $W_{j} \in \mathbb{G}_{p_{4}}$. Then, for each $j \in S_{\overrightarrow{y_{i}}}$, B sets

$$
Y_{j}=g_{1}^{r_{j}} \cdot g_{2}^{a_{j} / v_{i, y_{j}}} \cdot W_{j} .
$$

Case A.2: $y_{f}=\star$. In this case, B outputs a key with a well-formed $\mathbb{G}_{p_{1}}$ part.
Specifically, B, for each $j \in S_{\overrightarrow{y_{i}}}$, chooses random $W_{j} \in \mathbb{G}_{p_{4}}$, random $a_{j} \in \mathbb{Z}_{N}$ under the constraint that $\sum_{j \in S_{\overrightarrow{y_{i}}}} a_{j}=0$ and then sets

$$
Y_{j}=g_{1}^{a_{j} / t_{j, y_{j}}^{\prime}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j} .
$$

In the special case that $y_{\mathrm{j}}=b_{\mathrm{j}}$, then B sets

$$
Y_{\mathrm{j}}=\left(A_{1}^{\beta} \cdot C_{4}\right)^{a_{i}} \cdot g_{2}^{g_{\mathrm{j}} / /_{\mathrm{j}, y_{\mathrm{j}}}} \cdot W_{\mathrm{j}}
$$

$i=k$ : We distinguish between the following cases:
Case B.1: $y_{f}=\star$. B performs the same steps of Case A.2.
Case B.2: $y_{f} \neq \star$ and $y_{\mathrm{j}} \neq b_{\mathrm{j}}$. In this case, B aborts.

Case B.3: $y_{f} \neq \star$ and $y_{\mathrm{j}}=b_{\mathrm{j}}$. B mounts $T$, the challenge of the assumption, in position j .
Specifically, let $S=S_{\vec{y}_{i}} \backslash\{\mathrm{j}, h\}$, where $h$ is an index such that $y_{k, h} \neq \star$. Such an index $h$ always exists since we assumed that each query contains at least two non- $\star$ entries. Then, for each $j \in S, \mathrm{~B}$ sets

$$
Y_{j}=g_{12}^{a_{j} / t_{j, y_{j}}^{\prime}} \cdot g_{2}^{a_{j} / t_{j, y_{j}}} \cdot W_{j}
$$

Then, for position $\mathrm{j}, \mathrm{B}$ sets

$$
Y_{\mathrm{j}}=T \cdot g_{2}^{a_{\mathrm{j}} / /_{\mathrm{j}, y_{\mathrm{j}}}} \cdot W_{\mathrm{j}}
$$

and for position $h$, B sets

$$
Y_{h}=\left(A_{1}^{\alpha} B_{4}\right)^{-1 / t_{h, y_{h}}^{\prime}} \cdot g_{1}^{-s / t_{h, y_{h}}^{\prime}} \cdot g_{2}^{-\left(s+a_{j}\right) / t_{h, y_{h}}} \cdot W_{h}
$$

where $s=\sum_{j \in S} a_{j}$.
$i>k:$ B handles these queries as in Case A.2, independently from whether $y_{f}=\star$ or $y_{f} \neq \star$.

Ciphertext: B receives $\vec{x}$ and has to compute a ciphertext. We distinguishes the following two cases:

Case C.1: $x_{\mathrm{j}}=b_{\mathrm{j}}$. In this case, $\mathbf{B}$ aborts.
Case C.2: $x_{\mathrm{j}} \neq b_{\mathrm{j}}$. In this case, B computes sequence $\left(\vec{y}_{i}, \operatorname{HVE}\left(\overrightarrow{y_{i}}, \vec{x}\right)\right)_{i=1}^{q_{1}}$ for all queries $\vec{y}_{i}$ that it has received from $\mathcal{A}$ to induce the set MPos.
Then, B chooses random $s \in \mathbb{Z}_{N}$, and, for each $i \in[\ell]$, random $Z_{i} \in \mathbb{G}_{p_{3}}$ and random $s_{i} \in \mathbb{Z}_{N}$. Then, for each $i \in[\ell]$, if $i<f$ and $i \notin$ MPos, the algorithm randomly select $r_{i} \in \mathbb{Z}_{N}$ and sets

$$
X_{i}=g_{1}^{r_{i}} \cdot g_{5}^{s_{i}} \cdot Z_{i}
$$

otherwise, the algorithm sets

$$
X_{i}=g_{1}^{s \cdot t_{i, x_{i}}^{\prime}} \cdot g_{5}^{s_{i}} \cdot Z_{i}
$$

Second Stage Secret Tokens: To simulate the output of Sim. $\operatorname{KeyGen}\left(m s k, \vec{y}_{i}, \operatorname{HVE}\left(\vec{x}, \vec{y}_{i}\right)\right)$, B does the following way:
$\operatorname{HVE}\left(\vec{x}, \vec{y}_{i}\right)=0: \mathrm{B}$ generates a secret key with a random $\mathbb{G}_{p_{1}}$ part and without the $\mathbb{G}_{p_{5}}$ part.
Specifically, for each $j \in S_{\vec{y}_{i}}$, B chooses random $a_{j}$ such that $\sum_{j \in S_{\vec{y}_{j}}} a_{i}=0$, random $r_{j} \in \mathbb{Z}_{N}$, and random $W_{j} \in \mathbb{G}_{p_{4}}$. Then, for each $j \in S_{\vec{y}_{i}}$, B sets

$$
Y_{j}=g_{1}^{r_{j}} \cdot g_{2}^{a_{j} / t_{i, y_{j}}} \cdot W_{j} .
$$

$\operatorname{HVE}\left(\vec{x}, \vec{y}_{i}\right)=1:$ B generates a secret key without the $\mathbb{G}_{p_{1}}$ part and with a $\mathbb{G}_{p_{5}}$ part correlated with that of the ciphertext.
Specifically, for each $j \in S_{\overrightarrow{y_{i}}}$, B chooses random $a_{j}$ such that $\sum_{j \in S_{\vec{y}_{j}}} a_{i}=0$, and random $W_{j} \in \mathbb{G}_{p_{4}}$. Then, for each $j \in S_{\overrightarrow{y_{i}}}$, B sets

$$
Y_{j}=g_{2}^{a_{j} / t_{i, y_{j}}} \cdot g_{5}^{a_{j} / s_{j}} \cdot W_{j} .
$$

This ends the description of B.
The algorithm B will be used to prove properties of experiments $L$. We can modify B so that, on input $f$ and $k$, the challenge ciphertext is constructed by randomizing the $\mathbb{G}_{p_{1}}$ part also of the $f$-th component. The so modified algorithm, that we call $\mathrm{B}_{2}$, closely simulates the work of experiments $M$ and will be used to prove properties of these experiments.

Let us define the following two events.
$\operatorname{NotAbort}{ }_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)$ : denotes the event that B does not abort while computing the answer to the $k$-th query in an interaction with $\mathcal{A}$ on input $f$ and $k$. This is equivalent to the event that $y_{k, f}=\star$ or $y_{k, \mathrm{j}}=b_{\mathrm{j}}$.
$\operatorname{NotAbort}{ }_{2, \mathrm{~B}}^{\mathcal{A}}(f, k)$ : denotes the event that B does not abort while computing the ciphertext in an interaction with $\mathcal{A}$ on input $f$ and $k$. This is equivalent to the event that adversary $\mathcal{A}$ outputs vector $\vec{x}$ such that $x_{\mathrm{j}}=c_{\mathrm{j}}=1-b_{\mathrm{j}}$.

We can modify, experiments PKGame, $L(f, k)$ and $M(f, k)$ so that j and $b_{\mathrm{j}}$ are chosen just like B does. This modification makes the definitions of events NotAbort ${ }_{1, \operatorname{Exp}}^{\mathcal{A}}$ and NotAbort ${ }_{2, \mathrm{Exp}}^{\mathcal{A}}$ meaningful also for these experiments. We write NotAbort ${ }_{2}^{\mathcal{A}}$ as a shorthand for NotAbort ${ }_{2}^{\mathcal{A}} \mathrm{PKGame}$.

Lemma 6.2.19 For all $f, k$ and $\mathcal{A}, \operatorname{Prob}\left[\operatorname{NotAbort}{ }_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)\right] \geq \frac{1}{\ell}$.
Proof. The probability of $\operatorname{NotAbort}_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)$ is at least the probability that $y_{k, \mathrm{j}}=b_{\mathrm{j}}$. Moreover, the view of $\mathcal{A}$ up to the $k$-th key query is independent from $b_{\mathrm{j}}$ and j . Now observe that the $\vec{y}_{k}$ has at least two non-star entry and, provided that j is one of these (which happens with probability at least $2 / \ell)$, the probability that $y_{k, \mathrm{j}}=b_{\mathrm{j}}$ is $1 / 2$.

Lemma 6.2.20 For all $f, k$ and $\mathcal{A}, \operatorname{Prob}\left[\operatorname{NotAbort}{ }_{2, \operatorname{Exp}}^{\mathcal{A}}(f, k)\right] \geq \frac{1}{2 \ell}$ for $\operatorname{Exp}=L(f, k)$.
Proof. NotAbort ${ }_{2, \mathrm{Exp}}^{\mathcal{A}}(f, k)$ is the event that $y_{\vec{k}, j} \neq x_{\mathrm{j}}$ in the experiment $\operatorname{Exp}$. It is easy to see that the probability that j and $b_{\mathrm{j}}$ are correctly guessed such that $x_{\mathrm{j}}=c_{\mathrm{j}}=1-b_{\mathrm{j}}$ is at least $1 /(2 \ell)$, independently from the view of $\mathcal{A}$.

Lemma 6.2.21 Suppose event $\operatorname{NotAbort}_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)$ occurs. If $T=T_{1}$ then $\mathcal{A}$ 's view up to the challenge ciphertext in the interaction with B running on input $(f, k)$ is the same as in $L_{f, k-1}$. If instead $T=T_{2}$ then $\mathcal{A}$ 's view up to the challenge ciphertext in the interaction with B running on input $(f, k)$ is the same as in $L_{f, k}$.

Moreover, suppose events $\operatorname{NotAbort}_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)$ and $\operatorname{NotAbort}_{2, \mathrm{~B}}^{\mathcal{A}}(f, k)$ occur. If $T=T_{1}$ then $\mathcal{A}$ 's total view in the interaction with B running on input $(f, k)$ is the same as in $L_{f, k-1}$. If instead $T=T_{2}$ then $\mathcal{A}$ 's total view in the interaction with $\mathbf{B}$ running on input $(f, k)$ is the same as in $L_{f, k}$.

Proof. First observe that $m p k$ has the same distribution as the public parameters seen by $\mathcal{A}$ in both experiments. The same holds for the answers to the first $(k-1)$ key queries and to the last $\left(q_{1}-k\right)$. Let us now focus on the answer to the $k$-th key query. We have two cases:

Case 1: $y_{f}=\star$. Then the view of $\mathcal{A}$ in the interaction with B is independent from $T$ (see Case B.1) and, on the other hand, by definition, the two experiments coincide. Therefore the lemma holds in this case.

Case 2: $y_{f} \neq \star$. Suppose $T=T_{1}=A_{1}^{\alpha \beta} \cdot D_{4}$ and that $\operatorname{NotAbort}_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)$ occurs. Therefore, $y_{\mathrm{j}}=b_{\mathrm{j}}$ and B's answer to the $k$-th key query has the same distributions as in $L(f, k-1)$. On the other hand if $T$ is random in $\mathbb{G}_{p_{1} p_{4}}$ and $\operatorname{NotAbort}{ }_{1, \mathrm{~B}}^{\mathcal{A}}(f, k)$ occurs, the $\mathbb{G}_{p_{1}}$ parts of the $Y_{j}$ 's are random and thus the answer to the $k$-th query of $\mathcal{A}$ is distributed as in $L_{f, k}$.

For the second part of the lemma, we observe that the challenge ciphertext has the same distribution in both experiments and that, if $\operatorname{Not}_{\mathrm{Abort}}^{2, \mathrm{~B}} \mathcal{A}(f, k)$ occurs, B properly constructs the challenge ciphertext.

Let us now analyze the probability that $\mathcal{A}$ does output a challenge such that $f \notin \mathrm{MPos}$, this is crucial for B to successfully simulate with some non-negligible probability $L(f, k)$ or $L(f, k-1)$.

We start by introducing some notation.
$E_{f, \operatorname{Exp}}^{\mathcal{A}}$ is defined as the event that in experiment $\operatorname{Exp}$ the adversary $\mathcal{A}$ declare a challenge vector such that $f \notin \mathrm{MPos}$. When the adversary $\mathcal{A}$ is clear from the context we will simply write $E_{f, \text { Exp }}$.
$E_{f}^{\mathcal{A}}$ : is defined as the event that in experiment PKGame, the adversary $\mathcal{A}$ declares a challenge vector such that $f \notin \mathrm{MPos}$. When the adversary $\mathcal{A}$ is clear from the context we will simply write $E_{f}$.
$E_{f, \mathrm{~B}}^{\mathcal{A}}\left(f^{\prime}, k\right)$ : we extend the definition of $E_{f, \mathbf{E x p}}$ to include the experiment played by $\mathcal{A}$ against the algorithm B . Thus we denote by the event that in the interaction between $\mathcal{A}$ and B on input $f^{\prime}$ and $k$, B does not abort and $\mathcal{A}$ declares a challenge vector such that $f \notin \mathrm{MPos}$. If $\mathcal{A}, f^{\prime}$ and $k$ are clear from the context, we will simply write $E_{f, \mathrm{~B}}$.
$E_{f, f^{\prime}}^{\mathcal{A}}$ : is defined as the event that during the execution of $I_{f^{\prime}}$ adversary $\mathcal{A}$ outputs a challenge vector such that $f \notin$ MPos.

Observation 6.2.22 For all PPT adversaries $\mathcal{A}$ and distinguisher $\mathcal{D}$ and all $1 \leq f \leq \ell$, we have that $\operatorname{Prob}\left[\mathcal{D}\left(I(f)^{\mathcal{A}}\right)=1 \mid \neg E_{f, f}\right]=\operatorname{Prob}\left[\mathcal{D}\left(I(f+1)^{\mathcal{A}}\right)=1 \mid \neg E_{f, f+1}\right]$.

Proof. By definition of $I$, if the challenge vector is such that $f \in \mathrm{MPos}$, then $\mathcal{A}$ 's view in $I_{f}$ and $I_{f+1}$ is the same.

Observation 6.2.23 For all PPT adversaries $\mathcal{A}$ and all $1 \leq f \leq \ell$, we have that $\operatorname{Prob}\left[E_{f, f}^{\mathcal{A}}\right]=$ $\operatorname{Prob}\left[E_{f, f+1}^{\mathcal{A}}\right]$.

Proof. The view of $\mathcal{A}$ in $I_{f}$ up to the challenge ciphertext is independent from $f$.
Therefore we can set $\operatorname{Prob}\left[E_{f, f}^{\mathcal{A}}\right]=\operatorname{Prob}\left[E_{f, 1}^{\mathcal{A}}\right]=\operatorname{Prob}\left[E_{f}^{\mathcal{A}}\right]$.
Lemma 6.2.24 If Assumption 1 holds, then for $k=1, \ldots, q$ and $f=1, \ldots, \ell$, and for all PPT adversaries $\mathcal{A},\left|\operatorname{Prob}\left[E_{f, G}^{\mathcal{A}}\right]-\operatorname{Prob}\left[E_{f, H}^{\mathcal{A}}\right]\right| \quad$ and $\quad\left|\operatorname{Prob}\left[\operatorname{NotAbort}{ }_{2, G}^{\mathcal{A}}\right]-\operatorname{Prob}\left[\operatorname{Not} A b o r t_{2, H}^{\mathcal{A}}\right]\right|$ are negligible functions of $\lambda$, for experiments $G=L(f, k-1)$ and $H=L(f, k)$.

Proof. We prove the lemma for $E_{f, G}$ and $E_{f, H}$. A similar reasoning holds for NotAbort ${ }_{2, G}^{\mathcal{A}}$ and $\operatorname{NotAbort}{ }_{2, H}^{\mathcal{A}}$. For the sake of contradiction, suppose that $\operatorname{Prob}\left[E_{f, G}^{\mathcal{A}}\right] \geq \operatorname{Prob}\left[E_{f, H}^{\mathcal{A}}\right]+\epsilon$ for some non-negligible $\epsilon$. Then we can modify algorithm B into algorithm $\mathcal{B}$ with a nonnegligible advantage in breaking Assumption 1. Algorithm $\mathcal{B}$ simply execute B 's code. By

Lemma 6.2.19 event NotAbort ${ }_{1, \mathrm{~B}}$ occurs with probability at least $1 / \ell$ and in this case $\mathcal{B}$ can continue the execution of B 's code and receive the challenge vector from $\mathcal{A}$. At this point, $\mathcal{B}$ checks whether $f \notin$ MPos. If it is the case, $\mathcal{B}$ outputs 1 ; else $\mathcal{B}$ outputs 0 . It is easy to see that, by Lemma 6.2.21, the above algorithm has a non-negligible advantage in breaking Assumption 1.

The proof of the following corollary is straightforward from Lemma 6.2.24 and Observations $6 . 2 . 1 7 \longdiv { 6 . 2 . 1 8 }$

Corollary 6.2.25 For all $f=1, \ldots, \ell+1$ and $k=0, \ldots, q$, and all PPT adversaries $\mathcal{A}$, we have that, for $H=L_{f, k}\left|\operatorname{Prob}\left[E_{f, H}^{\mathcal{A}}\right]-\operatorname{Prob}\left[E_{f}^{\mathcal{A}}\right]\right|$ and $\left|\operatorname{Prob}\left[\operatorname{NotAbort}_{2, H}^{\mathcal{A}}\right]-\operatorname{Prob}\left[\operatorname{NotAbort}{ }_{2}^{\mathcal{A}}\right]\right|$ are negligible.

We are now ready to prove that $L_{f, k-1} \approx_{c} L_{f, k}$. To do this let us define the event $\operatorname{Succ}^{\mathcal{A}}(f, k)$ as

$$
\begin{equation*}
\operatorname{Succ}^{\mathcal{A}}(f, k):=\operatorname{NotAbort}_{1, \mathrm{~B}}^{\mathcal{A}}(f, k) \wedge \operatorname{NotAbort}_{2, \mathrm{~B}}^{\mathcal{A}}(f, k) \wedge E_{f, \mathrm{~B}}^{\mathcal{A}}(f, k) . \tag{6.1}
\end{equation*}
$$

When $\mathcal{A}$ is clear from the context we use the $\operatorname{shortcut} \operatorname{Succ}(f, k)$.
We are now ready to prove Lemma 6.2.26
Lemma 6.2.26 Suppose there exists an adversary $\mathcal{A}$, a distinguisher $\mathcal{D}$ and integers $1 \leq f \leq$ $\ell$ and $1 \leq k \leq q$ such that $\left|\operatorname{Prob}\left[\mathcal{D}\left(G^{\mathcal{A}}\right)=1\right]-\operatorname{Prob}\left[\mathcal{D}\left(H^{\mathcal{A}}\right)=1\right]\right| \geq \epsilon$, where $G=$ $L(f, k-1), H=L(f, k)$ and $\epsilon>0$. Then, there exists a PPT algorithm $\mathcal{B}$ with $\mathbf{A d v}_{1}^{\mathcal{B}} \geq$ $\operatorname{Prob}\left[E_{f}\right] \cdot \epsilon /\left(2 \cdot \ell^{2}\right)-\nu(\lambda)$, for a negligible function $\nu$.
Proof. Assume without loss of generality that $\operatorname{Prob}\left[\mathcal{D}\left(G^{\mathcal{A}}\right)=1\right] \geq \operatorname{Prob}\left[\mathcal{D}\left(H^{\mathcal{A}}\right)=1\right]+\epsilon$ and consider the following algorithm $\mathcal{B}$. $\mathcal{B}$ uses algorithm B as a subroutine and interacts with $\mathcal{A}$ on input integers $f$ and $k$ for which the above inequality holds, and an instance $(D, T)$ of Assumption 1. If event $\operatorname{Succ}(f, k)$ does not occur, $\mathcal{B}$ outputs $\perp$. Otherwise, $\mathcal{B}$ outputs $\mathcal{D}$ 's output. Therefore we have

$$
\begin{gather*}
\operatorname{Prob}\left[\mathcal{B} \text { outputs } 1 \mid T=T_{1}\right]=\operatorname{Prob}\left[\mathcal{B} \text { outputs } 1 \mid T=T_{1} \wedge \operatorname{Succ}(f, k)\right] .  \tag{6.2}\\
\operatorname{Prob}\left[\operatorname{Succ}(f, k) \mid T=T_{1}\right]
\end{gather*}
$$

By definition of $\operatorname{Succ}(f, k)$ we have

$$
\begin{aligned}
\operatorname{Prob}\left[\operatorname{Succ}(f, k) \mid T=T_{1}\right]= & \operatorname{Prob}\left[E_{f, \mathrm{~B}} \wedge \operatorname{NotAbort}_{1, \mathrm{~B}} \wedge \operatorname{NotAbort}_{2, \mathrm{~B}} \mid T=T_{1}\right] \\
= & \operatorname{Prob}\left[\operatorname{Not} A b o r t_{1, \mathrm{~B}} \mid T=T_{1}\right] . \\
& \operatorname{Prob}\left[E_{f, \mathrm{~B}} \wedge \operatorname{Not}^{2} \text { bbort }_{2, \mathrm{~B}} \mid \operatorname{Not}^{2} \text { Abort }_{1, \mathrm{~B}} \wedge T=T_{1}\right] .
\end{aligned}
$$

Now observe that event NotAbort $t_{1, \mathrm{~B}}$ is determined before B uses $T$ and thus

$$
\operatorname{Prob}\left[\operatorname{Not} A \operatorname{bort}_{1, \mathrm{~B}} \mid T=T_{1}\right]=\operatorname{Prob}\left[\operatorname{NotAbort}_{1, \mathrm{~B}}\right] .
$$

Moreover, by Lemma 6.2.21, if event NotAbort ${ }_{1, \mathrm{~B}}$ occurs and $T=T_{1}$, the view of $\mathcal{A}$ up to Challenge Query is equal to the view of $\mathcal{A}$ in experiment $G$ and thus

$$
\operatorname{Prob}\left[E_{f, \mathrm{~B}} \wedge \operatorname{Not}^{\text {Abort }}{ }_{2, \mathrm{~B}} \mid \operatorname{NotAbort}_{1, \mathrm{~B}} \wedge T=T_{1}\right]=\operatorname{Prob}\left[E_{f, G} \wedge \operatorname{NotAbort} 2_{2, G}\right]
$$

whence

$$
\begin{aligned}
\operatorname{Prob}\left[\operatorname{Succ}^{\mathcal{A}}(f, k) \mid T=T_{1}\right] & =\operatorname{Prob}\left[\text { NotAbort }_{1, \mathrm{~B}}\right] \cdot \operatorname{Prob}\left[\operatorname{NotAbort}_{2, G} \wedge E_{f, G}\right] \\
& =\operatorname{Prob}\left[\operatorname{NotAbort}_{1, \mathrm{~B}}\right] \cdot \operatorname{Prob}\left[\operatorname{NotAbort}_{2, G}\right] \cdot \operatorname{Prob}\left[E_{f, G}\right]
\end{aligned},
$$

where $\operatorname{NotAbort}{ }_{2, G}$ and $E_{f, G}$ are independent. Finally, if $T=T_{1}$ and $\operatorname{Succ}^{\mathcal{A}}(f, k)$ occurs, then, by Lemma 6.2.21, $\mathcal{A}$ 's view is exactly as in experiment $G$, and thus the probability that $\mathcal{B}$ outputs 1 is equal to the probability that $\mathcal{D}$ output 1 . We can thus rewrite Eq. 6.2 as

$$
\begin{aligned}
& \operatorname{Prob}\left[\mathcal{B} \text { outputs } 1 \mid T=T_{1}\right]=\operatorname{Prob}\left[\mathcal{D}\left(G^{\mathcal{A}}\right)=1\right] \\
& \operatorname{Prob}\left[\operatorname{NotAbort}_{1, \mathrm{~B}}\right] \cdot \operatorname{Prob}\left[\operatorname{NotAbort}{ }_{2, G}\right] \cdot \operatorname{Prob}\left[E_{f, G}\right]
\end{aligned}
$$

A similar reasoning yields

$$
\begin{aligned}
& \operatorname{Prob}\left[\mathcal{B} \text { outputs } 1 \mid T=T_{2}\right]=\operatorname{Prob}\left[\mathcal{D}\left(H^{\mathcal{A}}\right)=1\right] \\
& \operatorname{Prob}\left[\operatorname{NotAbort}_{1, \mathrm{~B}}\right] \cdot \operatorname{Prob}\left[\operatorname{NotAbort}_{2, H}\right] \cdot \operatorname{Prob}\left[E_{f, H}\right]
\end{aligned}
$$

By using Corollary 6.2.25, Lemma 6.2.19 and Lemma 6.2.20, we can conclude that there exists a negligible function $\nu$ such that we have

$$
\left.\begin{array}{rl}
\mathbf{A d v}_{1}^{\mathcal{B}}= & \operatorname{Prob}\left[\operatorname{Not}^{2 b o r t}\right. \\
1, \mathrm{~B}
\end{array}\right] \cdot \operatorname{Prob}\left[\operatorname{Not} A \text { bort }_{2}\right] \cdot \operatorname{Prob}\left[E_{f}\right] . ~\left(\operatorname{Prob}\left[\mathcal{D}\left(G^{\mathcal{A}}\right)=1\right]-\operatorname{Prob}\left[\mathcal{D}\left(H^{\mathcal{A}}\right)=1\right]\right)-\nu(\lambda) .
$$

The following Lemma can be proved by referring to algorithm $\mathrm{B}_{2}$. We omit further details since the proof is essentially the same as the one of Lemma 6.2.26.

Lemma 6.2.27 Suppose there exists an adversary $\mathcal{A}$, a distinguisher $\mathcal{D}$ and integers $1 \leq f \leq \ell+$ 1 and $1 \leq k \leq q$ such that $\left|\operatorname{Prob}\left[\mathcal{D}\left(G^{\mathcal{A}}\right)=1\right]-\operatorname{Prob}\left[\mathcal{D}\left(H^{\mathcal{A}}\right)=1\right]\right| \geq \epsilon$, where $G=M_{f, k-1}$, $H=M_{f, k}$ and $\epsilon>0$. Then, there exists a PPT algorithm $\mathcal{B}$ with $\mathbf{A d v}_{1}^{\mathcal{B}} \geq \operatorname{Prob}\left[E_{f}\right] \cdot \epsilon /(2$. $\left.\ell^{2}\right)-\nu(\lambda)$, for a negligible function $\nu$.

We are finally ready to prove Lemma 6.2.14.
Lemma 6.2.14. If Assumption 1 holds, $I_{f} \approx_{c} I_{f+1}$.
Proof. Suppose that for some adversary $\mathcal{A}$, distinguisher $\mathcal{D}$ and $f \in[\ell]$

$$
\begin{equation*}
\left|\operatorname{Prob}\left[\mathcal{D}\left(I_{f}^{\mathcal{A}}\right)=1\right]-\operatorname{Prob}\left[\mathcal{D}\left(I_{f+1}^{\mathcal{A}}\right)=1\right]\right| \geq \epsilon \tag{6.3}
\end{equation*}
$$

Now recall that $I_{f}=L_{f, 0}$ and $I_{f+1}=M_{f, 0}$. Thus, there exists $1 \leq k \leq q$ such that:

$$
\left|\operatorname{Prob}\left[\mathcal{D}\left(G^{\mathcal{A}}\right)=1\right]-\operatorname{Prob}\left[\mathcal{D}\left(H^{\mathcal{A}}\right)=1\right]\right| \geq \epsilon /(2 q)
$$

where $G=L_{f, k}$ and $H=L_{f, k-1}$ or where $G=M_{f, k}$ and $H=M_{f, k-1}$. Then by Lemma6.2.26, in the former case, and by Lemma 6.2.27 in the latter, we can construct an adversary $\mathcal{B}$ against Assumption 1, such that

$$
\mathbf{A d v}_{1}^{\mathcal{B}} \geq \frac{\epsilon}{4 q \ell^{2}} \cdot \operatorname{Prob}\left[E_{f}\right]-\nu(\lambda)
$$

Now it remains to estimate $\operatorname{Prob}\left[E_{f}\right]$. Notice that we can write

$$
\begin{aligned}
\operatorname{Prob}\left[\mathcal{D}\left(I_{f}^{\mathcal{A}}\right)=1\right]=\operatorname{Prob}\left[E_{f, f}\right] \cdot \operatorname{Prob}\left[\mathcal{D}\left(I_{f}^{\mathcal{A}}\right)=1 \mid E_{f, f}\right]+ \\
\operatorname{Prob}\left[\neg E_{f, f}\right] \cdot \operatorname{Prob}\left[\mathcal{D}\left(I_{f}^{\mathcal{A}}\right)=1 \mid \neg E_{f, f}\right]
\end{aligned}
$$

and

$$
\begin{array}{r}
\operatorname{Prob}\left[\mathcal{D}\left(I_{f+1}^{\mathcal{A}}\right)=1\right]=\operatorname{Prob}\left[E_{f, f+1}\right] \cdot \operatorname{Prob}\left[\mathcal{D}\left(I_{f+1}^{\mathcal{A}}\right)=1 \mid E_{f, f+1}\right]+ \\
\operatorname{Prob}\left[\neg E_{f, f+1}\right] \cdot \operatorname{Prob}\left[\mathcal{D}\left(I_{f}^{\mathcal{A}}\right)=1 \mid \neg E_{f, f+1}\right]
\end{array}
$$

and by combining Equation 6.3 and Observations 6.2 .22 and 6.2.23, we obtain

$$
\operatorname{Prob}\left[E_{f}\right] \cdot\left|\operatorname{Prob}\left[\mathcal{D}\left(I_{f}^{\mathcal{A}}\right)=1 \mid E_{f, f}\right]-\operatorname{Prob}\left[\mathcal{D}\left(I_{f+1}^{\mathcal{A}}\right)=1 \mid E_{f, f+1}\right]\right| \geq \epsilon
$$

Thus, we can conclude that

$$
\operatorname{Prob}\left[E_{f}\right] \geq \epsilon
$$

and thus $\mathcal{B}$ has advantage

$$
\mathbf{A d v}_{1}^{\mathcal{B}} \geq \frac{\epsilon^{2}}{4 q \ell^{2}}-\nu(\lambda)
$$

## Chapter 7

## Summary of Contributions

We briefly summarize our main contributions.
Generalized Key Delegation for Wildcarded Identity-Based and Inner-Product Encryption. Even though the WIBE and WKD-IBE constructions in ACD $^{+} 06$, AKN07] are very practical, they had two significant shortcomings. First, their security proofs only hold in cases where the maximum hierarchy depth $L$ is a constant due to the fact that they are not tight and lose a factor which is exponential in $L$. As a result, these schemes can only be used in scenarios where such a restriction is acceptable. In particular, when using WKD-IBE schemes to build identity-based broadcast encryption schemes, such a limitation on the maximum hierarchy depth will have a direct impact on the maximum size of the target group. Second, their solutions do not hide the pattern associated with the ciphertext. Hence, their schemes cannot be used in any application where the anonymity of the recipient needs to be preserved.

In Chapter[4 we showed how to overcome the limitations of existing WKD-IBE and WIBE schemes to obtain the first fully secure anonymous WW-IBE scheme.

Lattice-based Hierarchical Inner Product Encryption. In Chapter 5, we considered the problem of constructing hierarchical inner-product encryption scheme based on lattices assumptions. To achieve this goal, we extended the lattice-based IPE scheme by Agrawal et al. [AFV11] to the hierarchical setting by employing basis delegation technics by Peikert et al. [CHKP10] and by Agrawal et al. [ABB10].

Adaptive Simulation-Based Secure Constructions for Functional Encryption. In Chapter 6, we considered the problem of designing simulation-based secure functional encryption schemes. We looked both at general functionalities corresponding to the class of poly-sized circuits and at a more specialized functionality, specifically HVE.

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