# A Framework For Fully-Simulatable $h$-Out-Of-n Oblivious Transfer 

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#### Abstract

We present a framework for efficient, fully-simulatable $h$-out-of- $n$ oblivious transfer ( $O T_{h}^{n}$ ) with security against non-adaptive malicious adversaries. The number of communication round of the framework is six. Compared with existing fully-simulatable protocols for $O T_{h}^{n}$, our framework is round-efficient. Conditioning on no trusted common reference string is available, our DDH-based instantiation of the framework is the most efficient protocol for $O T_{h}^{n}$.

Our framework uses three abstract tools, i.e. perfectly binding commitment, perfectly hiding commitment and our new smooth projective hash. This allows a simple and intuitive understanding of its security.

We instantiate the new smooth projective hash under the lattice, decisional Diffie-Hellman, decisional N-th residuosity, decisional quadratic residuosity assumptions. This indeed shows that the folklore that it is technically difficult to instantiate the projective hash framework under the lattice assumption is not true. What's more, by using this lattice-based hash and Brassard's commitment scheme, we gain a concrete protocol for $O T_{h}^{n}$ which is secure against quantum algorithms.


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## 1 Introduction

### 1.1 Oblivious transfer

Oblivious transfer (OT), first introduced by [36] and later defined in another way with equivalent effect [9] by [11], is a fundamental primitive in cryptography and a concrete problem in the filed of secure multi-party computation. Considerable cryptographic protocols can be built from it. Most remarkable, [18, 21, 24, 42] proves that any secure multi-party computation can be based on a secure oblivious transfer protocol. In this paper, we concern a variant of OT, $h$-out-of- $n$ oblivious transfer $\left(O T_{h}^{n}\right)$. $O T_{h}^{n}$ deals with the following scenario. A sender holds $n$ private messages $m_{1}, m_{2}, \ldots, m_{n}$. A receiver holds $h$ private positive integers $i_{1}, i_{2}, \ldots, i_{h}$, where $i_{1}<i_{2}<\ldots<i_{h} \leqslant n$. The receiver expects to get the messages $m_{i_{1}}, m_{i_{2}}, \ldots, m_{i_{h}}$ without leaking any information about his private input, i.e. the $h$ positive integers he holds. The sender expects all new knowledge learned by the receiver from their interaction is at most $h$ messages. Obviously, the OT most literature refer to is $O T_{1}^{2}$ and can be viewed as a special case of $O T_{h}^{n}$.

Considering a variety of attack we have to confront in real environment, a protocol for $O T_{h}^{n}$ with security against malicious adversaries (a malicious adversary may act in any arbitrary malicious way to learn as much extra information as possible) is more desirable than the one with security against semi-honest adversaries (a semi-honest adversary, on one side, honestly does everything told by a prescribed protocol; on one side, records the messages he sees to deduce extra information which is not supposed to be known to he). Using Goldreich's compiler [16, 18], we can gain the former version from the corresponding latter version. However, the resulting protocol is prohibitive expensive for practical use, because it is embedded with so many invocations of zero-knowledge for NP. Thus, directly constructing the protocol based on specific intractable assumptions seems more feasible.

The first step in this direction is independently made by [31 and [1] which respectively presents a two-round efficient protocol for $O T_{1}^{2}$ based on the decisional Diffie-Hellman (DDH) assumption. Starting from these works and using the tool smooth projective hashing, [22] abstracts and generalizes the ideas of [1,31] to a framework for $O T_{1}^{2}$. Besides DDH assumption, the framework can be instantiated under the decisional $N$-th residuosity (DNR) assumption and decisional quadratic residuosity (DQR) assumption [22].

Unfortunately, these protocols (or frameworks) are only half-simulatable not fully-simulatable. By saying a protocol is fully-simulatable, we means that the protocol can be strictly proven its security under the real/ideal model simulation paradigm and without turning to a random oracle. The
paradigm requires that for any adversary in the real world, there exists a corresponding adversary, which can simulate him, in the ideal world. Thus, the real adversary can not do more harm than the corresponding ideal adversary does. Therefore the security level of the protocol is guaranteed not to be lower than that of the ideal world. Undesirably, a half-simulatable protocol for $O T_{1}^{2}$ only provides a simulator for the malicious sender.

Considering security, requiring a protocol to be fully-simulatable is necessary. Specifically, a fully-simulatable protocol provides security against all kinds of attacks, especially the future unknown attacks taken by any adversary whose the power is fixed when constructing the protocol (generally, the adversary is probabilistic polynomial-time) [5, 16], while a not fully-simulatable protocol doesn't. For example, the protocols proposed by [1,22, 31] suffer the selective-failure attacks, in which a malicious sender can induce transfer failures that are dependent on the messages that the receiver requests (32].

Constructing fully-simulatable protocols for OT with security against malicious adversaries naturally becomes the focus of the research community. [4] first presents such a fully-simulatable protocol. In detail, the OT is an adaptive h-out-n oblivious transfer (denoted by $O T_{h \times 1}^{n}$ in related literature) and based on $q$-Power Decisional Diffie-Hellman and $q$-Strong Diffie-Hellman assumptions. Unfortunately, these two assumptions are not standard assumptions used in cryptography and seem significantly stronger than DDH , DQR and so on. Motivated by basing OT on weaker complexity assumption, [19] presents a protocol for $O T_{h}^{n}$ using a blind identity-based encryption which is based on decisional bilinear Diffie-Hellman (DBDH) assumption. Using cut-choose technique, [25] later presents two efficient fully-simulatable protocols for $O T_{1}^{2}$ respectively based on DDH assumption and DNR assumption, which are weaker than DBDH . Since bilinear curves are considerably more expensive than regular Elliptic curves [12] and DDH is obtainable from Elliptic curves, [25]'s DDH-based protocol is the most efficient one among such fully-simulatable works.

The protocols mentioned above are proved their securities in the plain stand-alone model which not necessarily allows concurrent composition with other arbitrary malicious protocols. [35] overcomes this weakness and further the research by presenting a framework under common reference string (CRS) model for fully-simulatable, universally composable $O T_{1}^{2}$ and instantiating the framework respectively under $\mathrm{DDH}, \mathrm{DQR}$ and worst-case lattice assumption. It is notable that, the number of the communication rounds and the number of the public key operations of the framework, respectively, is two and one. Thus, conditioning on a trusted CRS is available, the DDH-based instantiation of the framework is the most efficient protocol for $O T_{1}^{2}$, no matter
seen from the number of communication rounds or the computational overhead. Recently, [13], using a novel compiler and somewhat non-committing encryption they present, convert 35 's instantiations based on DDH, DQR to the corresponding protocols with higher security level. In more detail, the resulting protocols for $O T_{1}^{2}$ are secure against adaptive malicious adversaries, which corrupts the parties dynamically based on his knowledge gathered by far. Note that, the fully-simulatable protocols for $O T_{1}^{2}$ mentioned by far except the one presented by [4] are only secure against non-adaptive malicious adversaries, which only corrupts the parties preset before the running of the protocol.

Though constructing protocols for fully-simulatable $O T_{1}^{2}$ with security against malicious adversaries has been studied well, constructing protocols for such $O T_{h}^{n}$ hasn't. We note that there are some works aiming to extend known cryptographic protocols to $O T_{h}^{n}$. [29] shows how to implementation $O T_{h}^{n}$ using $\log n$ invocation of $O T_{1}^{2}$ under half-simulation. A similar implementation for adaptive $O T_{h}^{n}$ can be seen in [30]. What's more, the same authors of [29, 30] propose a way to transform a singe-server privateinformation retrieval scheme (PIR) into an oblivious transfer scheme under half-simulation too [32]. With the help of a random oracle, [20] shows how to extend k oblivious transfers (for some security parameter k ) into many more, without much additional effort. However, the half-simulation and the random oracle are undesirable. To our best knowledge, only [4] and [19] respectively presents a fully-simulatable $O T_{h}^{n}$. However, as pointed out above, the assumptions the former uses are not standard and the latter uses is too expensive. Therefore, a well-motivated problem is to find a protocol or framework for efficient, fully-simulatable, secure against malicious adversaries $O T_{h}^{n}$ under weaker complexity assumptions.

### 1.2 Our contribution

In this paper, we present a framework for efficient, fully-simulatable, secure against non-adaptive malicious adversaries $O T_{h}^{n}$ whose security is proven under stand model (i.e. without turning to a random oracle). To our best knowledge, this is the first framework for such $O T_{h}^{n}$. The framework have the following features,

1. Fully-simulatable and secure against malicious adversaries without using a CRS. [22]'s framework for $O T_{1}^{2}$ is half-simulatable. Thought [35]'s framework for $O T_{1}^{2}$ is fully-simulatable, it doesn't work without a CRS. What is more, how to provide a trusted CRS before the protocol run still is a problem to be solved. The existing possible solutions, such as
natural process suggested by [35] , are only conjectures without formal proofs. The same problem remains in its adaptive version presented by [13]. Therefore, considering practical use, our framework are better.
2. Efficient. The number of communication rounds of our framework is six. Compared with the existing fully-simulatable $O T_{h}^{n}$ protocols, i.e. the protocols respectively presented by [4], [19] with round number $4+4 h, a+h \cdot b$ respectively ( $a, b \geq 2$ respectively is the round number of two zero-knowledge proof used in the protocol), our framework is round-efficient.

The computational overhead of our framework, in the worst case, consists of $K \cdot n$ public key encryption operations and $K \cdot h$ public key decryption operations, where $K$ is a value such that the probability of our simulator failing is at most $1 / 2^{K-1}$. The computational overhead in the average case is half of that in the worst case. Setting $K$ to be 40 is secure enough to be used in practice. Our framework covers [25]'s DDH-based protocol (with some straightforward modification) as a specific case. Since the price of DDH-based operations are lower than that of the operations [4] and [19] use, our DDH-based instantiation are the most efficient protocol for $O T_{h}^{n}$, in the sense of defending non-adaptive malicious adversaries without using a CRS.

We also admit that, in the context of a trusted CRS is available and only $O T_{1}^{2}$ is needed, [35]'s DDH-based instantiation is most efficient no matter seen from the number of communication rounds or the computational overhead.
3. Abstract and modular. The framework is described using just three high-level cryptographic tools, i.e. perfectly binding commitment (PBC), perfectly hiding commitment (PHC) and our new smooth projective hash (denoted by $S P W H_{h, t}$ ). This allows a simple and intuitive understanding of its security.
4. Generally realizable. The high-level cryptographic tools PBC, PHC and $S P W H_{h, t}$ are realizable from a variety of known specific assumption, even future assumption maybe. This makes our framework generally realizable. In particular, we instantiate $S P W H_{h, t}$ from DDH, DNR, DQR and lattice assumption. Instantiating PBC or PHC under specific assumptions is beyond the scope of this paper. Please see $[15,17$ for such examples. Generally realizability is vital to make framework live long, considering the future progress in breaking a specific intractable problem. If this case happen, replacing the instan-
tiation based on the broken problem with that based on a unbroken problem suffices.

What is more, we fix a folklore 25 that it appears technically difficult to instantiate the projective hash under lattice assumption by presenting a lattice-based $S P W H_{h, t}$ instantiation. It is notable that we gain an $O T_{h}^{n}$ instantiation which is secure against quantum algorithms, using this latticebased $S P W H_{h, t}$ instantiation and [3]'s commitment scheme. Considering that factoring integers and finding discrete logarithms are easy for quantum algorithms [38-40], this is an example showing the benefits from the generally realizability of the framework.

As an independent contribution, we present several propositions related to the indistinguishability of probability ensembles defined by sampling polynomial instances. Such propositions simplify our security proof very much. We believe that they are as useful in other security proof as in this paper.

### 1.3 Our Approach

We note that the smooth projective hash is a good abstract tool. Using this tool, [22] in fact presents a framework for half-simulatable $O T_{1}^{2}$, [14] present a framework for password-based authenticated key exchange protocols. We also note that the cut-and-choose is a good technique to make protocol fullysimulatable. Using this tool, [25] present several fully-simulatable protocol for $O T_{1}^{2},[26]$ presents a general fully-simulatable protocol for two-party computation. Indeed, we are inspired by such works. Our basic ideal is to use cut-and-choose technique and smooth projective hash to get a fully-simulatable framework.

We define a new smooth projective hash called $h$-smooth $t$-projective hash family with witnesses and hard subset membership ( $S P W H_{h, t}$ ). Loosely speaking, a smooth projective hash (SPH) is a set of operations defined over two languages $\dot{L}$ and $\ddot{L}$, where $\dot{L} \cap \ddot{L}=\emptyset$. For any projective instance $\dot{x} \in \dot{L}$, there are two ways to obtain its hash value, i.e. the way using its hash key or the way using its projective key and its witness $\dot{w}$. For any smooth instance $\ddot{x} \in \ddot{R}$, there is only one way to obtain its hash value, i.e. the way using its hash key.

For simplicity, we only compare our $S P W H_{h, t}$ with the version of SPH presented by [22] and denoted by $V S P H$, since on the one hand, $V S P H$ is the most complete version among previous works. On the other hand, the aim, constructing a framework for half-simulatable $O T_{1}^{2}$, of [22] is the closest to ours, constructing a framework for fully-simulatable $O T_{h}^{n}$.
$S P W H_{h, t}$ can be viewed as a generalized version of $V S P H$ to deal with
$O T_{h}^{n}$. $V S P H$ resembles $S P W H_{1,1}$ very much and can be converted into $S P W H_{1,1}$ through some straightforward modifications. $S P W H_{h, t}$ extends $V S P H$ in the following way.

1. Extends the instance-sampler algorithm to generate $h \dot{x}$ s and $t \ddot{x}$ s in a invocation. What is more, we not only require each $\dot{x}$ to hold a witness $\dot{w}$ as previous work do, but also require each $\ddot{x}$ to hold a witness $\ddot{w}$.
2. Discards the instance test (IT) algorithm and provide a new verification (VF) algorithm which is more useful for applying cut-and-choose technique.

We observe that the VSPH indeed is easy to be extended to deal with $O T_{1}^{n}$, but seems difficult to be extended to deal with the more general $O T_{h}^{n}$. The reason is that, on one hand, the $\ddot{x}$ lacks a direct witness, which result in $\dot{x}$ and $\ddot{x}$ being generated in a dependent way. This makes designing IT for $O T_{h}^{n}$ difficult without leaking information which is conductive to distinguish such $\dot{x}$ s and $\ddot{x} \mathrm{~s}$. Thus, even constructing a framework for $O T_{h}^{n}$ which is half-simulatable as [22] seems difficult. On the other hand, to use the technique cut-and-choose, a direct witness for $\ddot{x}$ indeed is needed. Because the simulator have to use such witness to extract the receiver's real input which is encoded as a permutation of $\dot{x}$ s and $\ddot{x} s$. The difficulties mentioned above can be overcome by requiring each $\ddot{x}$ to hold a direct witness too. What is more, the implementation of VF is easier than that of its predecessor IT. Because the operated object essentially is a pair of the form $(x, w)$ which is simpler than $\left(x_{1}, \ldots, x_{h+t}, w_{1}, \ldots, w_{h}\right)$ which is the general form of operated object of IT.
3. Extends key generation (KG) algorithm such that there is more information available for it. In more details, the information about the instance (i.e. $\dot{x}$ or $\ddot{x}$ ) is available to KG to generate hash key and projective key. This makes constructing hash system easier. In indeed, this makes lattice-based hash system come true which is thought difficult by 25].
4. Extends "smoothness" to guarantee that, loosely speaking, for any $\overrightarrow{\vec{x}} \stackrel{\text { def }}{=}$ $\left(\ddot{x}_{1}, \ldots, \ddot{x}_{t}\right)$ generated by invocations of the operations of hash system in some given way, even if given the corresponding projective keys, $\left(c_{1}, c_{2}, \ldots, c_{t}\right)$ and $\left(r_{1}, r_{2}, \ldots, r_{t}\right)$ are computationally indistinguishable, where $c_{i}$ is the hash value of $\ddot{x}_{i}$ and $r_{i}$ is uniformly chosen from all possible hash values.
5. Extends "hard subset membership" to guarantee that, loosely speaking, for any permutation $\pi, \overrightarrow{x y}, \pi(\overrightarrow{x y})$ and $\vec{x}$ are computationally indistinguishable, where $\overrightarrow{x y} \stackrel{\text { def }}{=}\left(\dot{x}_{1}, \ldots, \dot{x}_{h}, \ddot{x}_{h+1}, \ldots, \ddot{x}_{h+t}\right), \vec{x} \stackrel{\text { def }}{=}\left(\dot{x}_{1}, \ldots, \dot{x}_{h+t}\right)$ are generated by invocations of the operations of hash system in some given way.

Using $S P W H_{h, t}$ we construct the framework described with high-level as follows .

1. The receiver generates hash parameters and instance vectors, then sends them to the sender after disorders each vector.
2. The receiver and the sender cooperate to toss coin to decide which vector to be opened.
3. The receiver opens the chosen instances and encodes his private input by reordering each unchosen vector.
4. The sender checks that the chosen vectors are generated in the legal way which guarantees that the receiver learns at most $h$ message. If the check pass, the sender encrypts his private input (i.e. the $n$ messages he holds) using the hash system, and sends the ciphertext together with some information (i.e. the projective hash keys) to the receiver which is conductive to decrypt some ciphertext.
5. The receiver decrypts the ciphertext and gains the messages he expects.

Intuitively speaking, the receiver's security is implied by the hard subset membership, which guarantee that it is difficult for a malicious sender to distinguish $\dot{x}$ s and $\ddot{x}$ s. The sender's security is implied by the cut-and-choose technique, which guarantees that a malicious receiver's cheating is caught with probability nearly 1 . Formally speaking, in case the sender is corrupted, a simulator can extract the malicious sender's real input by sending cheatingly generated instance vectors while avoid to be caught by rewinding the malicious sender to get an appropriate result of tossing coin. In case the receiver is corrupted, a simulator can extract the malicious receiver's real input by rewinding the malicious receiver to open the instance vectors two times.

Motivated by making instantiating $S P W H_{h, t}$ easier and making use of the existing works as much as possible, we proves some propositions which essentially guarantees that the existing $S P H$ instantiations can be converted into $S P W H_{h, t}$ instantiations with some modification. The key idea is generating $\dot{x}$ s and $\ddot{x}$ s in independent ways.

Our lattice-based $S P W H_{h, t}$ instantiation is builded on the lattice-based cryptosystem presented by [25]. It is noticeable that it appears difficult to get lattice-based instantiation for $S P H$ [25]. Our solution is to let the instance $x(x \in \dot{L} \cup \ddot{L})$ be available to KG.

### 1.4 Organization

In Section 2, we describe the notations used in this paper, the security definition of $O T_{h}^{n}$, the definition of commitment scheme. In Section 3, we define our new hash system, i.e. $S P W H_{h, t}$. In Section 4, we construct our framework. In Section 5, we prove the security of the framework. In Section 6, we instantiate $S P W H_{h, t}$ under the lattice, DDH, DNR, DQR assumptions, respectively.

## 2 Preliminaries

Most notations and concepts mentioned in this section originate from [5, 15 , 16 which are basic literature in the filed of secure multi-party computation (SMPC). We tailor them to the need of dealing with $O T_{h}^{n}$.

### 2.1 Basic Notations

We denote an unspecified positive polynomial by poly(.). We denote the set consists of all natural numbers by $\mathbb{N}$. For any $i \in \mathbb{N},[i] \stackrel{\text { def }}{=}\{1,2, \ldots, i\}$. We denote the set consists of all prime numbers by $\mathbb{P}$.

We denote security parameter used to measure security and complexity by $k$. A function $\mu($.$) is negligible in k$, if there exists a positive constant integer $n_{0}$, for any poly(.) and any $k$ which is greater than $n_{0}$ (for simplicity, we later call such $k$ sufficiently large $k$ ), it holds that $\mu(k)<1 / \operatorname{poly}(k)$. A probability ensemble $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ is an infinite sequence of random variables indexed by $(k, a)$, where $a$ represents various types of inputs used to sample the instances according to the distribution of the random variable $X\left(1^{k}, a\right)$. Probability ensemble $X$ is polynomial-time constructible, if there exists a probabilistic polynomial-time (PPT) sample algorithm $S_{X}($. such that for any $a$, any $k$, the random variables $S_{X}\left(1^{k}, a\right)$ and $X\left(1^{k}, a\right)$ are identically distributed. We denote sampling an instance according to $X\left(1^{k}, a\right)$ by $\alpha \leftarrow S_{X}\left(1^{k}, a\right)$.

Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \stackrel{\text { def }}{=}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two probability ensembles. They are computationally indistinguishable, denoted
$X \stackrel{c}{=} Y$, if for any non-uniform PPT algorithm $D$ with an infinite auxiliary information sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ (where each $z_{k} \in\{0,1\}^{*}$ ), there exists a negligible function $\mu($.$) such that for any sufficiently large k$, any $a$, it holds that

$$
\left|\operatorname{Pr}\left(D\left(1^{k}, X\left(1^{k}, a\right), a, z_{k}\right)=1\right)-\quad \operatorname{Pr}\left(D\left(1^{k}, Y\left(1^{k}, a\right), a, z_{k}\right)=1\right)\right| \leqslant \mu(k)
$$

They are the same, denoted $X=Y$, if for any sufficiently large $k$, any $a$, $X\left(1^{k}, a\right)$ and $Y\left(1^{k}, a\right)$ are defined in the same way. They are equal, denoted $X \equiv Y$, if for any sufficiently large $k$, any $a$, the distributions of $X\left(1^{k}, a\right)$ and $Y\left(1^{k}, a\right)$ are identical. Obviously, if $X=Y$ then $X \equiv Y$; If $X \equiv Y$ then $X \stackrel{c}{=} Y$.

Let $\vec{x}$ be a vector (note that arbitrary binary string can be viewed as a vector). We denote its $i$-th element by $\vec{x}\langle i\rangle$, denote its dimensionality by $\# \vec{x}$, denote its length in bits by $|\vec{x}|$. For any positive integers set $I$, any vector $\vec{x}$, $\vec{x}\langle I\rangle \stackrel{\text { def }}{=}(\vec{x}\langle i\rangle)_{i \in I, i \leq \# \vec{x}}$.

Let $M$ be a probabilistic (interactive) Turing machine. By $M_{r}($.$) we$ denote $M$ 's output generated at the end of an execution using randomness $r$.

Let $f: D \rightarrow R$. Let $D^{\prime} \subseteq\{0,1\}^{*}$. Then $f\left(D^{\prime}\right) \stackrel{\text { def }}{=}\left\{f(x) \mid x \in D^{\prime} \cap D\right\}$, Range $(f) \stackrel{\text { def }}{=} f(D)$.

Let $x \in_{\chi} Y$ denotes sampling an instance $x$ from domain $Y$ according to the distribution law (or probability density function ) $\chi$. Specifically, let $x \in_{U} Y$ denotes uniformly sampling an instance $x$ from domain $Y$.

### 2.2 Security Definition Of A Protocol For $O T_{h}^{n}$

### 2.2.1 Functionality Of $O T_{h}^{n}$

$O T_{h}^{n}$ involves two parties, party $P_{1}$ (i.e. the sender) and party $P_{2}$ (i.e. the receiver). $O T_{h}^{n}$ 's functionality is formally defined as follows

$$
\begin{aligned}
f: \mathbb{N} \times\{0,1\}^{*} \times\{0,1\}^{*} & \rightarrow\{0,1\}^{*} \times\{0,1\}^{*} \\
f\left(1^{k}, \vec{m}, H\right) & =(\lambda, \vec{m}\langle H\rangle)
\end{aligned}
$$

where

- $k$ is the public security parameter.
- $\vec{m} \in\left(\{0,1\}^{*}\right)^{n}$ is $P_{1}$ 's private input, and each $|\vec{m}\langle i\rangle|$ is the same.
- $H \in \Psi \stackrel{\text { def }}{=}\{B \mid B \subseteq[n], \# B=h\}$ is $P_{2}$ 's private input.
- $\lambda$ denotes a empty string and is supposed to be got by $P_{1}$. That is, $P_{1}$ is supposed to get nothing.
- $\vec{m}\langle H\rangle$ is supposed to be got by $P_{2}$.

Note that, the length of all parties' private input have to be identical in SMPC (please see [16] for the reason and related discussion). This means that $|\vec{m}|=|H|$ is required. Without loss of generality, in this paper, we assume $|\vec{m}|=|H|$ always holds, because padding can be easily used to meet such requirement.

Intuitively speaking, the security of $O T_{h}^{n}$ requires that $P_{1}$ can't learn any new knowledge - typically, $P_{2}$ 's private input, from the interaction at all, and $P_{2}$ can't learn more than $h$ messages held by $P_{1}$. To capture the security in a formal way, the concepts such as adversary, trusted third party, ideal world, real world were introduced. Note that the security target in this paper is to be secure against non-adaptive malicious adversaries, so only concepts related to this case is referred to in the following.

### 2.2.2 Non-Adaptive Malicious Adversary

Before running $O T_{h}^{n}$, the adversary $A$ has to corrupt all parties listed in $I \subseteq[2]$. In case $U \in\left\{P_{1}, P_{2}\right\}$ is not corrupted, he will strictly follow the prescribed protocol as an honest party. In case party $U$ is corrupted, $U$ will be fully controlled by $A$ as a cheating party. In this case, $U$ will have to pass all his knowledge to $A$ before the protocol runs and follows $A$ 's instructions from then on - so there is a probability that $U$ arbitrarily deviates from prescribed protocol. In fact, after $A$ finishes corrupting, $A$ and all cheating parties have formed a coalition led by $A$ to learn as much extra knowledge, e.g. the honest parties' private inputs, as possible. From then on, they share knowledge with each other and coordinate their behavior. Without loss of generality, we can view this coalition as follows. All cheating parties are dummy. $A$ receives messages addressed to the members of the coalition and sends messages on behalf of the members.

Loosely speaking, we say $O T_{h}^{n}$ is secure, if and only if, for any malicious adversary $A$, the knowledge $A$ learns in the real world is not more than that he learns in the ideal world. In other words, if and only if, for any malicious adversary $A$, what harm $A$ can do in real world is not more than what harm he can do in the ideal world. In the ideal world, there is an incorruptible trusted third party (TTP). All parties hand their private inputs to TTP. TTP
computes $f$ and sends back $f().\langle i\rangle$ to $P_{i}$. In the real world, there is no TTP, and the computation of $f($.$) is finished by A$ and all parties's interaction.

### 2.2.3 $O T_{h}^{n}$ In The Ideal World

In the ideal world, an execution of $O T_{h}^{n}$ proceeds as follows.
Initial Inputs. All entities know the public security parameter k. $P_{1}$ holds a private input $\vec{m} \in\left(\{0,1\}^{*}\right)^{n}$. Party $P_{2}$ holds a private input $H \in \Psi$. Adversary $A$ holds a name list $I \subseteq[2]$, a randomness $r_{A} \in\{0,1\}^{*}$ and an infinite auxiliary input sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$, where $z_{k} \in\{0,1\}^{*}$. Before proceeds to next stage, $A$ corrupts parties listed in $I$ and learns $\vec{x}\langle I\rangle$, where $\vec{x} \stackrel{\text { def }}{=}(\vec{m}, H)$.

Submitting inputs to TTP. Each honest party $P_{i}$ always submits its private input $\vec{x}\langle i\rangle$ unchanged to TTP. $A$ submits arbitrary string based on his knowledge to TTP for cheating parties. The string TTP receives is a twodimensional vector $\vec{y}$ which is formally described in the following way.

$$
\vec{y}\langle i\rangle= \begin{cases}\vec{x}\langle i\rangle & \text { if } i \in I, \\ \alpha & \text { if } i \notin I\end{cases}
$$

where $\alpha \in\left\{\vec{x}\langle i\rangle\right.$, abort $\left._{i}\right\} \cup\{0,1\}^{\mid \vec{x}\langle i\rangle}$ and $\alpha$ is generated in the way of $\alpha \leftarrow$ $A\left(1^{k}, I, r_{A}, z_{k}, \vec{x}\langle I\rangle\right)$. Obviously, there is a probability that $\vec{x} \neq \vec{y}$.

TTP computing $f$. TTP checks that $\vec{y}$ is a valid input to $f$, i.e. no entry of $\vec{y}$ is abort ${ }_{i}$. If $\vec{y}$ passes the check, then TTP computes $f$ and sets $\vec{w}$ to be $f\left(1^{k}, \vec{y}\right)$. Otherwise, TTP sets $\vec{w}$ to be (abort $\left.i_{i}, a b o r t_{i}\right)$. Finally, TTP hands $\vec{w}\langle i\rangle$ to $P_{i}$ respectively and halts.

Outputs. Each honest party $P_{i}$ always outputs the message $\vec{w}\langle i\rangle$ it obtains from the TTP. Each cheating party $P_{i}$ outputs nothing (i.e. $\lambda$ ). The adversary outputs something generated by executing arbitrary function of the information he gathers by far. Without loss of generality, this can be assumed to be a string consisting of $1^{k}, I, r_{A}, z_{k}, \vec{x}\langle I\rangle, \vec{w}\langle I\rangle$.

The output, denoted ddeal $_{f, I, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, S\right)$, of the protocol for $O T_{h}^{n}$ in the ideal world is a three-dimensional vector orderly consisting of the outputs of $A, P_{1}, P_{2}$. Obviously, Ideal $l_{f, I, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, S\right)$ is a random variable whose randomness is $r_{A}$.

### 2.2.4 $O T_{h}^{n}$ In The Real World

In the real world, there is no TTP. A execution of $O T_{h}^{n}$ proceeds as follows.
Initial Inputs. Initial input each entity holds in the real world is the same as in the ideal world but there are some difference as follows. A randomness
$r_{i}$ is held by each party $P_{i}$. After finishes corrupting, in addition to the knowledge $A$ learns in ideal world, cheating parties' randomness $\vec{r}\langle I\rangle$ is also learn by $A$, where $\vec{r} \stackrel{\text { def }}{=}\left(r_{1}, r_{2}\right)$.

Computing $f$. In the real world, computing $f$ is finished by all entities' interaction. Each honest party strictly follows the prescribed protocol (i.e. the concrete protocol, usually denoted $\pi$, for $O T_{h}^{n}$ ). The cheating parties have to follow $A$ 's instructions and may arbitrarily deviate from prescribed protocol.

Outputs. Each honest party $P_{i}$ always outputs what the prescribed protocol instructs. Each cheating party $P_{i}$ outputs nothing. The adversary outputs something generated by executing arbitrary function of the information he gathers by far. Without loss of generality, this can be assumed to be string consisting of $1^{k}, I, r_{A}, \vec{r}\langle I\rangle, z_{k}, \vec{x}\langle I\rangle$ and messages addressed to the cheating parties.

The output, denoted $\operatorname{Real}_{\pi, I, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, S\right)$, of $O T_{h}^{n}$ in the real world is a three-dimensional vector orderly consisting of the outputs of $A, P_{1}, P_{2}$. Obviously, $\operatorname{Real}_{f, I, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, S\right)$ is a random variable whose randomnesses are $r_{A}$ and $\vec{r}$.

### 2.2.5 Security definition

The security of a protocol for $O T_{h}^{n}$ is formally captured by the following definition.

Definition 1 (The security of a protocol for $O T_{h}^{n}$ ). Let $f$ denotes the functionality of $O T_{h}^{n}$ and let $\pi$ be a concrete protocol for $O T_{h}^{n}$. We say $\pi$ securely computes $f$, if and only if for any non-uniform probabilistic polynomial-time adversary $A$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ in the real world, there exists a non-uniform probabilistic expected polynomial-time adversary $A^{\prime}$ with the same sequence in the ideal world such that, for any $I \subseteq[2]$, it holds that

$$
\begin{align*}
& \left\{\operatorname{Real}_{\pi, I, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}} \stackrel{c}{=} \\
& \left\{\operatorname{Ideal}_{f, I, A^{\prime}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{\substack{\in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}\right.}} \tag{1}
\end{align*}
$$

where the parameters input to the two probability ensembles are the same and each $\vec{m}\langle i\rangle$ is of the same length.

The concept, non-uniform probabilistic expected polynomial-time, mentioned in Definition 1 is formulated in distinct way in distinct literature such
as $[6,15]$. We prefer to the following definition [23], because it is clearer in formulation and more closely related to our issue.

Definition 2 ( $M_{1}$ runs in expected polynomial-time with respect to $M_{2}$ ). Let $M_{1}, M_{2}$ be two interactive Turing machines running a protocol. By $<$ $M_{1}\left(x_{1}, r_{1}, z_{1}\right), M_{2}\left(x_{2}, r_{2}, z_{2}\right)>\left(1^{k}\right)$, we denote a running which starts with $M_{i}$ holding a private input $x_{i}$, a randomness $r_{i}$, an auxiliary input $z_{i}$, the public security parameter $k$. By $I D N_{M_{1}}\left(<M_{1}\left(x_{1}, r_{1}, z_{1}\right), M_{2}\left(x_{2}, r_{2}, z_{2}\right)>\left(1^{k}\right)\right)$, we denote the number of total direct deduction steps $M_{1}$ takes in the whole running. We say $M_{1}$ runs in expected polynomial-time with respect to $M_{2}$, if and only if there exists a polynomial poly(.) such that for every $k \in \mathbb{N}$, it holds that

$$
\begin{aligned}
& \max \left(\left\{E _ { R _ { 1 } , R _ { 2 } } \left(I D N _ { M _ { 1 } } \left(<M_{1}\left(x_{1}, R_{1}, z_{1}\right)\right.\right.\right.\right. \\
& \\
& \left.\left.M_{2}\left(x_{2}, R_{2}, z_{2}\right)>\left(1^{k}\right)\right)\right) \mid \\
& \left.\left.\quad\left|x_{1}\right|=\left|x_{2}\right|=k, z_{1}, z_{2} \in\{0,1\}^{*}\right\}\right) \leq \operatorname{poly}(k)
\end{aligned}
$$

where $R_{1}, R_{2}$ are random variables with uniform distribution over $\{0,1\}^{*}$.
For Definition 1, it in fact requires that adversary $A$ 's simulator $A^{\prime}$ should run in expected polynomial-time with respect to TTP who computes $O T_{h}^{n}$,s functionality $f$.

We point out that the security definition presented in $[5,15,16]$ requires the simulator $A^{\prime}$ to run in strictly polynomial-time, but the one presented in [25, 26] allows $A^{\prime}$ to run in expected polynomial-time. Definition 1 follows the latter. We argue that this is justified, since [2] shows that there is no (nontrivial) constant-round zero-knowledge proof or argument having a strictly polynomial-time black-box simulator, which means allowing simulator to run in expected polynomial-time is essential for achieving constant-round protocols. See [23] for further discussion.

### 2.3 Commitment Scheme

In this section, we briefly introduce commitment scheme [15, 17] which will be used in our framework. Loosely speaking, commitment scheme is a twoparty protocol involving two phases. In the first phase, a sender $U_{1}$ sends a commitment, which hides his private input (i.e. the value he wants to commit to), to a receiver $U_{2}$. In the second phase, $U_{1}$ reveals its commitment to $U_{2}$, and $U_{2}$ knows the value $U_{1}$ commits to.

Definition 3 (Commitment Scheme). A commitment scheme is defined as follows.

- Initial Inputs. At the beginning, all parties know the public security parameter $k$. The sender $U_{1}$ holds a randomness $r_{1} \in\{0,1\}^{*}$, a value $m \in\{0,1\}^{\text {poly }(k)}$ to be committed to, where the polynomial poly(.) is public. The receiver $U_{2}$ holds a randomness $r_{2} \in\{0,1\}^{*}$.
- Commit Phase. $U_{1}$ computes a commitment, denoted $\alpha$, based on his knowledge, ie $\alpha \leftarrow U_{1}\left(1^{k}, m, r_{1}\right)$, then $U_{1}$ send $\alpha$ to $U_{2}$.

The security for $U_{1}$ is implied by the commitment scheme's hiding, which guarantees that for any PPT malicious $\tilde{U}_{2}$, the probability that he deduces the knowledge of $m$ from information he have gathered by far is negligible. More formally, for any PPT $\tilde{U}_{2}$, for any string $m^{\prime} \in$ $\{0,1\}^{\text {poly }(k)}$, it holds that,

$$
\begin{aligned}
\left\{V i e w C P _ { \tilde { U } _ { 2 } } \left(<U_{1}(m), \tilde{U}_{2}>\right.\right. & \left.\left.\left(1^{k}\right)\right)\right\} \\
& \stackrel{c}{=}\left\{\operatorname{ViewCP} P_{\tilde{U}_{2}}\left(<U_{1}\left(m^{\prime}\right), \tilde{U}_{2}>\left(1^{k}\right)\right)\right\}
\end{aligned}
$$

where ViewCP $P_{\tilde{U}_{2}}($.$) denotes \tilde{U}_{2}$ 's view at the end of commit phase.

- Reveal Phase. $U_{1}$ computes and sends a de-commitment, which typically consists of $m, r_{1}$, to $U_{2}$ to let $U_{2}$ know $m$. Receiving de-commitment, $U_{2}$ checks its validity. Typically $U_{2}$ checks that $\alpha=U_{1}\left(1^{k}, m, r_{1}\right)$ holds. If de-commitment pass the check, $U_{2}$ knows and accepts $m$.

The security for $U_{2}$ is implied by the commitment scheme's binding, which guarantees that for any PPT malicious $\tilde{U}_{1}$, the probability that $\tilde{U}_{1}$ cheats to interpret $\alpha$ as a commitment to a value which is different from $m$ without being caught is negligible. More formally, for any PPT $\tilde{U}_{1}$, any $m, \tilde{U}_{1}$ do the following experiment, experiment: $\alpha \leftarrow \tilde{U}_{1}\left(1^{k}, m\right), r_{1} \leftarrow \tilde{U}_{1}\left(1^{k}, m\right),\left(m^{\prime}, r_{1}^{\prime}\right) \leftarrow \tilde{U}_{1}\left(1^{k}, m\right)$. it holds that

$$
\begin{aligned}
& \operatorname{Pr}\left(\operatorname{ViewC} P_{U_{2}}\left(<\tilde{U}_{1}(m), U_{2}>\left(1^{k}\right)\right)=\right. \\
& \quad \operatorname{ViewC} P_{U_{2}}\left(<\tilde{U}_{1}\left(m^{\prime}\right), U_{2}>\left(1^{k}\right)\right) \wedge \\
& \alpha=U_{1}\left(1^{k}, m, r_{1}\right) \wedge \\
& \left.\quad \alpha=U_{1}\left(1^{k}, m^{\prime}, r_{1}^{\prime}\right)\right)=\mu(k)
\end{aligned}
$$

We are to use two stronger versions of commitment scheme to construct the framework. One, called perfectly hiding commitment scheme (PHC),
provides security for a sender against computationally unbound malicious receivers. The other, called perfectly binding commitment scheme (PBC), provides security for a receiver against computationally unbound malicious sender. For notational simplicity, we let $P H C_{r}(m)\left(P B C_{r}(m)\right)$ denote a commitment to $m$ which generated by using PHC (PBC) scheme and randomness $r$.

## 3 A New Smooth Projective Hash - $S P W H_{h, t}$

### 3.1 The Definition Of $S P W H_{h, t}$

In this section, we define a new smooth projective hash - $h$-smooth $t$ projective hash family with witnesses and hard subset membership, denoted $S P W H_{h, t}$, which will be used to construct our framework. In section 6, we instantiate $S P W H_{h, t}$ respectively under four distinct intractable problems.

Let us recall some related works before defining $S P W H_{h, t}$. [7. 41] present the classic notation of "universal hashing". Based on "universal hashing", [8] first introduces the concept of universal projective hashing, smooth projective hashing and hard subset membership problem in terms of languages and sets. In order to construct a framework for password-based authenticated key exchange, [14] modifies such definition to some extent. That is, smoothness is defined over every instance of a language rather than a randomly chosen instance. [22] refines the modified version in terms of the procedures used to implement it. What is more, a new requirement called verifiable smoothness is added to the hashing so as to construct a framework for $O T_{1}^{2}$. The resulting hashing is called verifiablely-smooth projective hash family that has hard subset membership property. A corresponding universal version also is presented by [22]. Note that, the framework presented by [22] is not fully-simulatable. Our $S P W H_{h, t}$ is an ameliorated version of 8, 14, 22]. The difference between $S P W H_{h, t}$ and the works mentioned above will be under a detailed discussion after we define $S P W H_{h, t}$. Usually constructing a hashing with property universality is the first step to gain a hashing with property smoothness. The definition of the property universality will be given in section 6.1.1 where we deal with how to gain smoothness for a hash system.

For clarity in presentation, we assume $n=h+t$ always holds and introduce additional notations. Let $R=\left\{(x, w) \mid x, w \in\{0,1\}^{*}\right\}$ be a relation, then $L_{R} \stackrel{\text { def }}{=}\left\{x \mid x \in\{0,1\}^{*}, \exists w((x, w) \in R)\right\}, R(x) \stackrel{\text { def }}{=}\{w \mid(x, w) \in R\}$. $\Pi \stackrel{\text { def }}{=}\{\pi \mid \pi:[n] \rightarrow[n], \pi$ is a permutation $\}$. Let $\pi \in \Pi$ (to comply with other literature, we also use $\pi$ somewhere to denote a protocol without bring-
ing any confusion). Let $\vec{x}$ be an arbitrary vector. By $\pi(\vec{x})$, we denote a vector resulted from applying $\pi$ to $\vec{x}$. That is, $\vec{y}=\pi(\vec{x})$, if and only if $\forall i(i \in[d] \rightarrow \vec{x}\langle i\rangle=\vec{y}\langle\pi(i)\rangle) \wedge \forall i(i \notin[d] \rightarrow \vec{x}\langle i\rangle=\vec{y}\langle i\rangle)$ holds, where $d \stackrel{\text { def }}{=} \min (\# \vec{x}, n)$.
Definition 4 ( $h$-Smooth $t$-Projective Hash Family With Witnesses And Hard Subset Membership). $\mathcal{H}=(P G, I S, V F, K G$, Hash,pHash) is an $h-$ smooth t-projective hash family with witnesses and hard subset membership ( $S P W H_{h, t}$ ), if and only if $\mathcal{H}$ is specified as follows

- The parameter-generator PG is a PPT algorithm that takes a security parameter $k$ as input and outputs a family parameter $\Lambda$, ie $\Lambda \leftarrow$ $P G\left(1^{k}\right) . \Lambda$ will be used as a parameter to define three relations $R_{\Lambda}, \dot{R}_{\Lambda}$ and $\ddot{R}_{\Lambda}$, where $R_{\Lambda}=\dot{R}_{\Lambda} \cup \ddot{R}_{\Lambda}$ and $\dot{R}_{\Lambda} \cap \ddot{R}_{\Lambda}=\emptyset$ are supposed to hold.
- The instance-sampler IS is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$ as input and outputs a vector $\vec{a} \in$ $\left(\dot{R}_{\Lambda}\right)^{h} *\left(\ddot{R}_{\Lambda}\right)^{t}$, ie $\vec{a} \leftarrow I S\left(1^{k}, \Lambda\right)$.
Let $\vec{a} \stackrel{\text { def }}{=}\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots\right.$,
$\left.\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)^{T}$ be a vector generated by IS. We call each $\dot{x}_{i}$ or $\ddot{x}_{i}$ an instance of $L_{R_{\Lambda}}$. For each pair $\left(\dot{x}_{i}, \dot{w}_{i}\right)\left(\left(\ddot{x}_{i}, \ddot{w}_{i}\right)\right)$, $\dot{w}_{i}\left(\ddot{w}_{i}\right)$ is called a witness of $\dot{x}_{i} \in L_{\dot{R}_{\Lambda}}\left(\ddot{x}_{i} \in L_{\ddot{R}_{\Lambda}}\right)$.
For simplicity in formulation later, we introduce some additional notations here. $\vec{x}^{\vec{a}} \stackrel{\text { def }}{=}\left(\dot{x}_{1}, \ldots, \dot{x}_{h}, \ddot{x}_{h+1}, \ldots, \ddot{x}_{n}\right)^{T}$, where for each $i, \vec{x}^{\vec{a}}\langle i\rangle=$ $\vec{a}\langle i\rangle\langle 1\rangle . \vec{w}^{\vec{a}} \stackrel{\text { def }}{=}\left(\dot{w}_{1}, \ldots, \dot{w}_{h}, \ddot{w}_{h+1}, \ldots, \ddot{w}_{n}\right)^{T}$, where for each $i$, $\vec{w}^{\vec{a}}\langle i\rangle=$ $\vec{a}\langle i\rangle\langle 2\rangle$. What is more, we abuse notation $\in$ to some extent. We write $\vec{x} \in \operatorname{range}\left(I S\left(1^{k}, \Lambda\right)\right)$ if and only if there exists a vector $\vec{x}^{a}$ such that $\vec{x}^{\vec{a}}=\vec{x}$ and $\vec{a} \in \operatorname{Range}\left(\operatorname{IS}\left(1^{k}, \Lambda\right)\right)$. We write $x \in \operatorname{Range}\left(\operatorname{IS}\left(1^{k}, \Lambda\right)\right)$ if and only if there exists a vector $\vec{x}$ such that $\vec{x} \in \operatorname{Range}\left(\operatorname{IS}\left(1^{k}, \Lambda\right)\right)$ and $x$ is an entry of $\vec{x}$.
- The verifier VF is a PPT algorithm that computes the following function

$$
\begin{aligned}
& \zeta: \mathbb{N} \times\left(\{0,1\}^{*}\right)^{3} \rightarrow\{0,1\} \\
& \zeta\left(1^{k}, \Lambda, x, w\right)= \begin{cases}0 & \text { if }(x, w) \in \dot{R}_{\Lambda}, \\
1 & \text { if }(x, w) \in \ddot{R}_{\Lambda}, \\
\text { undefined } & \text { otherwise } .\end{cases}
\end{aligned}
$$

$V F$ takes a security parameter $k$, a family parameter $\Lambda$ and a pair strings $(x, w)$ as input and outputs an indicator bit $b$, ie $b \leftarrow V F\left(1^{k}, \Lambda, x, w\right)$.

- The key generator KG is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$ and an instance $x$ as input and outputs a hash key and a projection key, ie $(h k, p k) \leftarrow K G\left(1^{k}, \Lambda, x\right)$.
- The hash Hash is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$, an instance $x$ and a hash key hk as input and outputs a value $y$, ie $y \leftarrow \operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right)$.
- The projection pHash is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$, an instance $x$ and a projection key $p k$ as input and outputs a value $y$, ie $y \leftarrow \operatorname{Hash}\left(1^{k}, \Lambda, x, p k\right)$.
and $\mathcal{H}$ has the following properties

1. Projection. $\mathcal{H}$ is projective on every pair $(\Lambda, \dot{x})$, where $\dot{x} \in L_{\dot{R}_{\Lambda}}$. That is, for any sufficiently large $k$, any $\Lambda \in \operatorname{Range}\left(P G\left(1^{k}\right)\right)$, any $(\dot{x}, \dot{w})$ generated by $I S\left(1^{k}, \Lambda\right)$, any $(h k, p k) \in \operatorname{Range}\left(K G\left(1^{k}, \Lambda, \dot{x}\right)\right)$, it holds that

$$
\operatorname{Hash}\left(1^{k}, \Lambda, \dot{x}\right)=p H a s h\left(1^{k}, \Lambda, \dot{x}, \dot{w}\right)
$$

2. Smoothness. Loosely speaking, it requires that for any pair $(\Lambda, \overrightarrow{\vec{x}})$, where $\overrightarrow{\ddot{x}} \in L_{\vec{R}_{\Lambda}}^{t}$, the hash values of $\overrightarrow{\ddot{x}}$ are random and unobtainable unless their hash keys are known. That is, for any $\pi \in \Pi$, the two probability ensembles $S m_{1} \stackrel{\text { def }}{=}\left\{S m_{1}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ and $S m_{2} \stackrel{\text { def }}{=}\left\{\operatorname{Sm}_{2}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ defined as follows, are computationally indistinguishable, ie $S m_{1} \stackrel{c}{=} S m_{2}$.
$\operatorname{SmGen}_{1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right), \vec{a} \leftarrow I S\left(1^{k}, \Lambda\right), \vec{x} \leftarrow \vec{x}^{\vec{a}}$, for each $j \in[n]$ operates as follows: $\left(h k_{j}, p k_{j}\right) \leftarrow K G\left(1^{k}, \Lambda, \vec{x}\langle j\rangle\right), y_{j} \leftarrow \operatorname{Hash}\left(1^{k}, \Lambda, \vec{x}\langle j\rangle, h k_{j}\right)$, $\overrightarrow{x p k y}\langle j\rangle \leftarrow\left(\vec{x}\langle j\rangle, p k_{j}, y_{j}\right)$. Finally outputs $(\Lambda, \overrightarrow{x p k y})$.
$\operatorname{SmGen}_{2}\left(1^{k}\right)$ : compared with $\operatorname{SmGen}_{1}\left(1^{k}\right)$, the only difference is that $y_{j} \in_{U} \operatorname{Range}\left(\operatorname{Hash}\left(1^{k}, \Lambda, \vec{x}\langle j\rangle,.\right)\right)$ for each $j \in[n]-[h]$.
$\operatorname{Sm}_{i}\left(1^{k}\right):(\Lambda, \overrightarrow{x p k y}) \leftarrow \operatorname{SmGen}_{i}\left(1^{k}\right), \overrightarrow{\overrightarrow{x p k y}} \leftarrow \pi(\overrightarrow{x p k y})$, finally outputs
$(\Lambda, \overrightarrow{x p k y})$.
3. Hard Subset Membership. Loosely speaking, it requires that the instances of $L_{\dot{R}_{\Lambda}}$ and that of $L_{\ddot{R}_{\Lambda}}$ are computationally indistinguishable. To be more precise, it requires $\mathcal{H}$ to meet the following conditions.
(a) For any $\pi \in \Pi$, the two probability ensembles $H S M_{1} \stackrel{\text { def }}{=}\left\{H S M_{1}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ and $H S M_{2} \stackrel{\text { def }}{=}\left\{H S M_{2}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ are computationally indistinguishable, i.e. $H S M_{1} \stackrel{c}{=} H S M_{2}$.
$\operatorname{HSM}_{1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right), \vec{a} \leftarrow I S\left(1^{k}, \Lambda\right)$, finally outputs $\left(\Lambda, \vec{x}^{\vec{a}}\right)$. $H S M_{2}\left(1^{k}\right)$ : Operates in the same way as $\operatorname{HSM}_{1}\left(1^{k}\right)$, but finally outputs $\left(\Lambda, \pi\left(\vec{x}^{\vec{a}}\right)\right)$.
(b) For any $\pi \in \Pi$, for any $\pi^{\prime} \in \Pi$, the two probability ensembles $H S M_{2}$ and $H S M_{3} \stackrel{\text { def }}{=}\left\{\operatorname{HSM}_{3}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ are computationally indistinguishable, i.e. $H S M_{2} \stackrel{c}{=} H S M_{3}$, where $H S M_{2}$ is defined above and $\mathrm{HSM}_{3}$ is defined as follows.
Cheat $\left(1^{k}\right)$ : Generates $n$ instances of $L_{\dot{R}_{\Lambda}}$ in the following way. $\Lambda \leftarrow P G\left(1^{k}\right), e \leftarrow\llcorner n / h\lrcorner, r \leftarrow n \bmod h, \vec{a}_{i} \leftarrow I S\left(1^{k}, \Lambda\right)$ for each $i \in[e+1], \vec{b}\langle(i-1) h+j\rangle \leftarrow \vec{a}_{i}\langle j\rangle$ for each $i \in[e]$ and $j \in[h], \vec{b}\langle$ eh $+j\rangle \leftarrow \vec{a}_{e+1}\langle j\rangle$ for each $j \in[r]$, finally outputs $(\Lambda, \vec{b})$. $H S S M_{3}\left(1^{k}\right):(\Lambda, \vec{b}) \leftarrow C h e a t\left(1^{k}\right)$, finally outputs $\left(\Lambda, \pi^{\prime}(\vec{b})\right)$.

Remark 5 (The Witnesses Of The Instances). The main use of the witnesses of an instance $\dot{x} \in L_{\dot{R}_{\Lambda}}$ is to project and gain the hash value of $x$ rather than to service as a proof of $\dot{x} \in L_{\dot{R}_{A}}$. However, with respect to an instance $\ddot{x} \in L_{\ddot{R}_{\Lambda}}$, it is on the contrary. For $O T_{h}^{n}$, this means that a receiver uses the witnesses of $\ddot{x}$ to persuade a sender to believe that the receiver is unable to gain the hash value of $\ddot{x}$.
Remark 6 (Hard Subset Membership). The property $3 a$ guarantees that for any $\vec{x} \in \operatorname{Range}\left(\operatorname{IS}\left(1^{k}, \Lambda\right)\right)$, any $\pi \in \Pi$, any PPT adversary $A$, the advantage of $A$ identifying an entry of $\pi(\vec{x})$ falling into $L_{\dot{R}_{\Lambda}}\left(L_{\vec{R}_{\Lambda}}\right)$ with probability over $h / n(t / n)$ is negligible. That is, seen from $A$, every entry of $\pi(\vec{x})$ seems the same. With respect to $O T_{h}^{n}$, this means that the receiver can encode his private input into a permutation of a vector $\vec{x} \in L_{R_{\Lambda}}^{n}$ without leaking any information. For example, if the receiver expects to gain $\vec{m}\langle H\rangle$, then he may generates $a \vec{x}$ and randomly chooses a permutation $\pi \in \Pi$ such that $\pi(\vec{x})\langle i\rangle \in L_{\dot{R}_{\Lambda}}$ for each $i \in H$. Any PPT adversary knows nothing about $H$ if only given $\pi(\vec{x})$.

The property $3 b$ guarantees that there is a probability that one can cheat to generate a $\vec{x}$ which is supposed to fall into $L_{\dot{R}_{\Lambda}}^{h} * L_{\ddot{R}_{\Lambda}}^{t}$ but actually $L_{\dot{R}_{\Lambda}}^{n}$. Note that, for $O T_{h}^{n}$, this property is a key for the simulator to extract the real input of the corrupted sender and conductive to construct a fully-simulatable $O T_{h}^{n}$.

### 3.2 The Difference Between $S P W H_{h, t}$ And Related Hash Systems

Now we discuss the difference between our $S P W H_{h, t}$ and related hash systems previous works present or use. For simplicity, we only compare our
$S P W H_{h, t}$ with the hash system (denoted by $V S P H$ for simplicity) which is presented by [22]. We argue that this is justified, on the one hand, the version of $[22]$ is the most complete version among previous works. On the other hand, the aim of $[22$ is the closest to ours. They aim to construct a framework for $O T_{1}^{2}$ which actually is half-simulatable, while we aim to establish a framework for fully-simulatable $O T_{h}^{n}$.

Loosely speaking, our $S P W H_{h, t}$ can be viewed as a generalized version of $V S P H$. Indeed, $V S P H$ resembles $S P W H_{1,1}$ very much and can be converted into $S P W H_{1,1}$ through some straightforward modifications. The essential differences are listed as follows.

1. To deal with $O T_{h}^{n}, S P W H_{h, t}$ extends the IS algorithm to generate $h$ $\dot{x}$ s and $t \ddot{x}$ s in a invocation. What is more, besides each $\dot{x}$ should hold a witness $\dot{w}, S P W H_{h, t}$ also requires each $\ddot{x}$ to hold a witness $\ddot{w}$.
2. $S P W H_{h, t}$ discards $V S P H$ 's the instance test IT algorithm and provide a new verification (VF) algorithm which is more useful for applying cut-and-choose technique.

We observe that the VSPH indeed is easy to be extended to deal with $O T_{1}^{n}$, but seems difficult to be extended to deal with the more general $O T_{h}^{n}$. The reason is that, on one hand, the $\ddot{x}$ lacks a direct witness, which result in $\dot{x}$ and $\ddot{x}$ being generated in a dependent way. This makes designing IT for $O T_{h}^{n}$ difficult without leaking information which is conductive to distinguish such $\dot{x}$ s and $\ddot{x}$ s. Thus, even constructing a framework for $O T_{h}^{n}$ which is half-simulatable as [22] seems difficult. On the other hand, to use the technique cut-and-choose, a direct witness for $\ddot{x}$ indeed is needed. Because the simulator have to use such witness to extract the receiver's real input which is encoded as a permutation of $\dot{x}$ s and $\ddot{x}$ s. The difficulties mentioned above can be overcome by requiring each $\ddot{x}$ to hold a direct witness. What is more, the implementation of VF is easier than that of its predecessor IT. Because the operated object essentially is a pair of the form $(x, w)$ which is simpler than $\left(x_{1}, \ldots, x_{h+t}, w_{1}, \ldots, w_{h}\right)$ which is the general form of operated objects of IT.
3. $S P W H_{h, t}$ extends KG algorithm such that the information of the instance is available to it. This makes constructing hash system easier. In indeed, this makes lattice-based hash system come true which is thought difficult by [25].

## 4 Constructing A Framework For Fully-simulatable $O T_{h}^{n}$

In this section, we construct a framework for $O T_{h}^{n}$. In the framework, we will use a PPT algorithm, denoted $\Gamma$, that receiving $B_{1}, B_{2} \in \Psi$, outputs a uniformly chosen permutation $\pi \epsilon_{U} \Pi$ such that $\pi\left(B_{1}\right)=B_{2}$, i.e. $\pi \leftarrow$ $\Gamma\left(B_{1}, B_{2}\right)$. We give an example implementation of $\Gamma$ as follows.
$\Gamma\left(B_{1}, B_{2}\right)$ : First, $E \leftarrow \emptyset$. Second, for each $j \in B_{2}$, then $i \in_{U} B_{1}$, $B_{1} \leftarrow B_{1}-\{i\}, E \leftarrow E \cup\{j \rightleftharpoons i\}$. Third, $C \leftarrow[n]-B_{1}, D \leftarrow[n]-B_{2}$, for each $j \in D$, then $i \in_{U} C, C \leftarrow C-\{i\}, E \leftarrow E \cup\{j \rightleftharpoons i\}$. Fourth, define $\pi$ as $\pi(i)=j$ if and only if $j \rightleftharpoons i \in E$. Finally, outputs $\pi$.

### 4.1 The Framework For $O T_{h}^{n}$

- Common inputs: All entities know the public security parameter $k$, an positive polynomial polys(.), a $S P W H_{h, t}$ (where $n=h+t$ ) hash system $\mathcal{H}$, a perfectly hiding commitment scheme, a perfectly binding commitment scheme.
- Private Inputs: Party $P_{1}$ (i.e. the sender) holds a private input $\vec{m} \in$ $\left(\{0,1\}^{*}\right)^{n}$ and a randomness $r_{1} \in\{0,1\}^{*}$. Party $P_{2}$ (i.e. the receiver) holds a private input $H \in \Psi$ and a randomness $r_{2} \in\{0,1\}^{*}$. The adversary $A$ holds a name list $I \subseteq[2]$ and a randomness $r_{A} \in\{0,1\}^{*}$.
- Auxiliary Inputs: The adversary $A$ holds an infinite auxiliary input sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}, z_{k} \in\{0,1\}^{*}$.

The protocol works as follow. For clarity, we omit some trivial errorhandlings such as $P_{1}$ refusing to send $P_{2}$ something which is supposed to be sent. Handling such errors is easy. $P_{2}$ halting and outputting abort $t_{1}$ suffices.

- Receiver's step (R1): $P_{2}$ generates hash parameters and samples instances.

1. $P_{2}$ samples $\operatorname{poly}_{s}(k)$ instance vectors. $K \stackrel{\text { def }}{=}$ poly $(k)$. For each $i \in$ $[K], P_{2}$ does: $\Lambda_{i} \leftarrow P G\left(1^{k}\right), \vec{a}_{i} \leftarrow I S\left(1^{k}, \Lambda_{i}\right)$. Without loss of generality, we assume $\vec{a}_{i}=\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots\right.$, $\left.\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)^{T}$.
2. $P_{2}$ disorders each instance vector.

For each $i \in[K], P_{2}$ uniformly chooses a permutation $\pi_{i}^{1} \in_{U} \Pi$, then $\tilde{\vec{a}}_{i} \leftarrow \pi_{i}^{1}\left(\vec{a}_{i}\right)$.
3. $P_{2}$ sends the instances and the corresponding hash parameters, i.e. $\left(\left(\Lambda_{1}, \tilde{\vec{x}}_{1}\right), \ldots,\left(\Lambda_{K}, \tilde{\vec{x}}_{K}\right)\right)$, to $P_{1}$, where $\tilde{\vec{x}}_{i} \stackrel{\text { def }}{=} \vec{x}^{\tilde{a}_{i}}$ (correspondingly, $\left.\tilde{\vec{w}}_{i} \stackrel{\text { def }}{=} \vec{w}^{\tilde{a}_{i}}\right)$.

- Receiver's step (R2-R3)/Sender's step (S1-S2): $P_{1}$ and $P_{2}$ cooperate to toss coin to choose instance vectors to open.

1. $P_{1}: s \in_{U}\{0,1\}^{K}$, sends $P H C(s)$ to $P_{2}$.
2. $P_{2}: s^{\prime} \in_{U}\{0,1\}^{K}$, sends $P B C\left(s^{\prime}\right)$ to $P_{1}$.
3. $P_{1}$ and $P_{2}$ respectively sends each other the de-commitments to $P H C(s)$ and $P B C\left(s^{\prime}\right)$, and respectively checks the received decommitments are valid. If the check fails, $P_{1}$ ( $P_{2}$ respectively) halts and outputs abort $t_{2}$ (abort $t_{1}$ respectively). If no check fails, then they proceed to next step.
4. $P_{1}$ and $P_{2}$ share a common randomness $r=s \oplus s^{\prime}$. The instance vectors whose index fall into $C S \stackrel{\text { def }}{=}\{i \mid r\langle i\rangle=1, i \in[K]\}$ (correspondingly, $\overline{C S} \stackrel{\text { def }}{=}[K]-C S)$ are chosen to be opened.

- Receiver's step (R4): $P_{2}$ opens the chosen instances to $P_{1}$, encodes and sends his private input to $P_{1}$.

1. $P_{2}$ opens the chosen instances to prove that the instances he generates are legal.
$P_{2}$ sends $\left(\left(i, j, \tilde{\vec{w}}_{i}\langle j\rangle\right)\right)_{i \in C S, j \in J_{i}}$ to $P_{1}$, where $J_{i} \stackrel{\text { def }}{=}\left\{j \mid \tilde{\vec{x}}_{i}\langle j\rangle \in L_{\tilde{R}_{\Lambda_{i}}}, j \in\right.$ $[n]\}$.
2. $P_{2}$ encodes his private input and sends the resulting code to $P_{1}$. Let $G_{i} \stackrel{\text { def }}{=}\left\{j \mid \tilde{\vec{x}}_{i}\langle j\rangle \in L_{\dot{R}_{\Lambda}}, i \in \overline{C S}\right\}$. For each $i \in \overline{C S}, P_{2}$ does $\pi_{i}^{2} \leftarrow \Gamma\left(G_{i}, H\right)$, sends $\left(\pi_{i}^{2}\right)_{i \in \overline{C S}}$ to $P_{1}$. That is, $P_{2}$ encode his private input into sequences such as $\pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$ where $i \in \overline{C S}$.

Note that $P_{2}$ can send $\left(\left(i, j, \tilde{\vec{w}}_{i}\langle j\rangle\right)\right)_{i \in C S, j \in J_{i}}$ and $\left(\pi_{i}^{2}\right)_{i \in \overline{C S}}$ in one step.

- Sender's step (S3): $P_{1}$ checks the chosen instances, encrypts and sends his private input to $P_{2}$.

1. $P_{1}$ verifies that each chosen instance vectors is legal, i.e. the number of the entries belonging to $L_{\dot{R}_{\Lambda_{i}}}$ is not more than $h$.
$P_{1}$ checks that, for each $i \in C S, \# J_{i} \geq n-h$, and for each $j \in J_{i}, V F\left(1^{k}, \Lambda_{i}, \tilde{\vec{x}}_{i}\langle j\rangle, \tilde{\vec{w}}_{i}\langle j\rangle\right)$ is 1 . If the check fails, $P_{1}$ halts and outputs abort ${ }_{2}$, otherwise $P_{1}$ proceeds to next step.
2. $P_{1}$ reorders the entries of each unchosen instance vector in the way told by $P_{2}$.
For each $i \in \overline{C S}, P_{1}$ does $\tilde{\tilde{\vec{x}}_{i}} \leftarrow \pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$.
3. $P_{1}$ encrypts and sends his private input to $P_{2}$ together with some auxiliary messages.
For each $i \in \overline{C S}, j \in[n], P_{1}$ does: $\left(h k_{i j}, p k_{i j}\right) \leftarrow K G\left(1^{k}, \Lambda_{i}, \tilde{\vec{x}}_{i}\langle j\rangle\right)$,
 $\left(\oplus_{i \in \overline{C S}} \vec{\beta}_{i}\right), \overrightarrow{p k}_{i} \stackrel{\text { def }}{=}\left(p k_{i 1}, p k_{i 2}, \ldots, p k_{i n}\right)^{T}$, sends $\vec{c}$ and $\left(\overrightarrow{p k}_{i}\right)_{i \in \overline{C S}}$ to $P_{2}$.

- Receiver's step (R5): $P_{2}$ decrypts the ciphertext $\vec{c}$ and gains the message he want.
For each $i \in \overline{C S}, j \in H, P_{2}$ operates: $\beta_{i j}^{\prime} \leftarrow p \operatorname{Hash}\left(1^{k}, \Lambda_{i}, \tilde{\vec{x}}_{i}\langle j\rangle, \tilde{\vec{w}}_{i}\langle j\rangle, \overrightarrow{p k}_{i}\langle j\rangle\right)$, $m_{j}^{\prime} \leftarrow \vec{c}\langle j\rangle \oplus\left(\oplus_{i \in \overline{C S}} \beta_{i j}^{\prime}\right)$. Finally, $P_{2}$ gains the messages $\left(m_{j}^{\prime}\right)_{j \in H}$.


### 4.2 The Correctness Of The Framework

Now let us check the correctness of the framework, i.e. the framework works in case $P_{1}$ and $P_{2}$ are honest. For each $i \in \overline{C S}, j \in H$, we know

$$
\begin{gathered}
\vec{c}\langle j\rangle=\vec{m}\langle j\rangle \oplus\left(\oplus_{i \in \overline{C S}} \vec{\beta}_{i}\langle j\rangle\right) \\
m_{j}^{\prime}=\vec{c}\langle j\rangle \oplus\left(\oplus_{i \in \overline{C S}} \beta_{i j}^{\prime}\right)
\end{gathered}
$$

Because of the projection of $\mathcal{H}$, we know

$$
\vec{\beta}_{i}\langle j\rangle=\beta_{i j}^{\prime}
$$

So we have

$$
\vec{m}\langle j\rangle=m_{j}^{\prime}
$$

This means what $P_{2}$ gets is $\vec{m}\langle H\rangle$ that indeed is $P_{2}$ wants.

### 4.3 The Security Of The Framework

With respect to the security of the framework, we have the following theorem.
Theorem 7 (The protocol is secure against the malicious adversaries). Assume that $\mathcal{H}$ is an $h$-smooth $t$-projective hash family with witnesses and hard subset membership, PHC is a perfectly hiding commitment, PBC is a perfectly binding commitment. Then, the protocol securely computes the oblivious transfer functionality in the presence of non-adaptive malicious adversaries.

We defer the strick proof of Theorem 7 to section 5 and first give an intuitive analysis here as a warm-up. For the security of $P_{1}$, the framework should prevent $P_{2}$ from gaining more than $h$ messages. Using cut and choose technique, $P_{1}$ makes sure with some probability that each instance vector contains no more than $h$ projective instance, which leads to $P_{2}$ learning extra messages is difficult. The following theorem guarantees that this probability is overwhelming.

Theorem 8. In case $P_{1}$ is honest and $P_{2}$ is corrupted, the probability that $P_{2}$ cheats to obtain more than $h$ messages is at most $1 / 2^{\text {polys }_{s}(k)}$.

Proof. According to the framework, there are two necessary conditions for $P_{2}$ 's success in the cheating.

1. $P_{2}$ has to generate at least one illegal $\vec{x}_{i}$ which contains more than $h$ entries belonging to $L_{\dot{R}_{\Lambda_{i}}}$. If not, $P_{2}$ cann't correctly decrypt more than $h$ entries of $\vec{c}$, because of the smoothness of $\mathcal{H}$. Without loss of generality, we assume the illegal instance vectors are $\vec{x}_{l_{1}}, \vec{x}_{l_{2}}, \ldots, \vec{x}_{l_{d}}$.
2. The illegal instance vectors are lucky not to be chosen, i.e. $\overline{C S}=$ $\left\{l_{1}, l_{2}, \ldots, l_{d}\right\}$. We prove this claim in two case.
(a) In case $\overline{C S} \neq\left\{l_{1}, l_{2}, \ldots, l_{d}\right\}$ and $\overline{C S}-\left\{l_{1}, l_{2}, \ldots, l_{d}\right\}=\emptyset$, there exists $j\left(j \in[d] \wedge l_{j} \in C S\right)$. So $P_{1}$ can detect $P_{2}$ 's cheating and $P_{2}$ will gain nothing.
(b) In case $\overline{C S} \neq\left\{l_{1}, l_{2}, \ldots, l_{d}\right\}$ and $\overline{C S}-\left\{l_{1}, l_{2}, \ldots, l_{d}\right\} \neq \emptyset$, there exists $j\left(j \in \overline{C S} \wedge \vec{x}_{j}\right.$ is legal). Because of the smoothness of $\mathcal{H}$, $P_{2}$ cannot correctly decrypt more than $h$ entries of $\vec{c}$.

Now, let us estimate the probability that the second necessary condition is met. Note that, $\operatorname{PHC}(s)$ is a perfectly hiding commitment and $P_{2}$ is honest, so the shared randomness is uniformly distributed. We have

$$
\begin{aligned}
\operatorname{Pr}\left(\overline{C S}=\left\{l_{1}, l_{2}, \ldots, l_{d}\right\}\right) & =(1 / 2)^{d}(1 / 2)^{\text {poly }}(k)-d \\
& =1 / 2^{\text {poly }}(k)
\end{aligned}
$$

This means that the probability that $P_{2}$ cheats to obtain more than $h$ messages is at most $1 / 2^{\text {polys }(k)}$.

For the security of $P_{2}$, the framework first should prevent $P_{1}$ from learning $P_{2}$ 's private input. There is a potential risk in Step R4 where $P_{2}$ encodes his private input. From Remark 6, we know that hard subset membership
guarantees that for any PPT malicious $P_{1}$, without being given $\pi_{i}^{1}$, the probability that $P_{1}$ learns any new knowledge is negligible. Thus $P_{2}$ 's encoding is safe. Besides cheating $P_{2}$ of private input, it seems there is another obvious attack that malicious $P_{1}$ sends invalid messages, e.g. $p k_{i j}$ which $\left(h k_{i j}, p k_{i j}\right) \notin \operatorname{Range}\left(K G\left(1^{k}, \Lambda_{i}, x_{i j}\right)\right)$, to $P_{2}$. This attack in fact doesn't matter. Its effect is equal to that of $P_{1}$ 's altering his real input, which is allowed in the ideal world too.

### 4.4 The Communication Complexity Of The Framework

Step R1 and Step R2 can be taken in one round. Step R5 is taken without communication. Each of other steps is taken in one round. Therefore, the total number of the communication rounds is six.

Compared with existing protocols for $O T_{h}^{n}$ which are fully-simulatable without restoring to a random oracle, our protocol is round-efficient. The round number of [4]'s $O T_{h \times 1}^{n}$ is $4+4 h$. The round number of [19]'s $O T_{h}^{n}$ is $a+h \cdot b$, where $a, b \geq 2$ respectively is the round number of two zero-knowledge proof of knowledge protocol used in the protocol.

### 4.5 The Computational Overhead Of The Framework

We measure the computational overhead of the framework in terms of the number of public key operations (i.e. operations based on trapdoor functions, or similar operations), because the overhead of public key operations, which depends on the length of their inputs, is greater than that of symmetric key operations (i.e. operations based on one-way functions) by orders of magnitude. Please see [27] to know which cryptographic operation is public key operation or private key operation.

As to the framework, the public key operations are $\operatorname{Hash}($.$) and p H a s h($.$) ,$ and the symmetric key operations are $P H C($.$) and P B C($.$) . In Step S3, P_{1}$ takes $n \cdot \# \overline{C S}$ invocations of $\operatorname{Hash}($.$) s to encrypt his private input. In Step$ R5, $P_{2}$ takes $h \cdot \# \overline{C S}$ invocations of $p H a s h($.$) to decrypt the messages he$ want. The value of $\# \overline{C S}$ is poly $_{s}(k)$, polys $(k) / 2$, respectively, in the worst case and in the average case. Thus, fixing the problem we tackle (i.e. fixing the values of $n$ and $h$ ), the efficiency only depends on the value of poly $(k)$. In Section 5 where we strictly prove the security of the framework, we'll see the probability that the simulator fails is at most $1 / 2^{\text {polys }_{s}(k)-1}$ in case $P_{2}$ is corrupted. Thus, conditioning on the cryptographic primitives without being broken, the real world and the ideal world can be distinguished at
$\operatorname{most} 1 / 2^{\text {poly }}(k)-2$. Setting $\operatorname{poly}_{s}(k)$ to be 40 , we obtains such a probability $3.6 \times 10^{-12}$, which is secure enough to be used in practice. By the way, our simulator also may fail (with negligible probability) in case $P_{1}$ is corrupted, but the probability of this event arising depends on the computational hiding of PBC and on the computational binding of PHC rather than the value of $\operatorname{poly}_{s}(k)$. So we don't need to take this case into consideration here.

Note that the operations, based on the non-standard assumptions and used by [4], ie $q$-Power Decisional Diffie-Hellman and $q$-Strong Diffie-Hellman assumptions, are very expensive. What is more, since bilinear curves are considerably more expensive than regular Elliptic curves 12 and DDH is obtainable from Elliptic curves, the operations, based on Decisional Bilinear Diffie-Hellman and used by [19], are also considerably more expensive than that based on DDH. Thus the DDH-based instantiation, presented in Section 6.3, of our framework is the most efficient protocol for $O T_{h}^{n}$.

We have to admit that, in the context of a trusted CRS is available and only $O T_{1}^{2}$ is needed, [35]'s DDH-based instantiation, which is two-round efficient and of two public key encryption operations and one public key decryption operation, is most efficient no matter seen from the number of communication rounds or the computational overhead.

## 5 A Security Proof Of The Framework

We prove Theorem 7 holds in this section. For notational clarity, we denote the entities, the parties and the adversary in the real world by $P_{1}, P_{2}, A$, and denote the corresponding entities in the ideal world by $P_{1}^{\prime}, P_{2}^{\prime}, A^{\prime}$. In the light of the parties being corrupted, there are four cases to be considered and we prove Theorem 7 holds in each case.

We don't know how to construct a strictly polynomial-time simulator for the adversary in the real world, in case only $P_{1}$ or $P_{2}$ is corrupted. Instead, expected polynomial-time simulators are constructed, which results in a failure of standard black-box reduction technique. Fortunately, the problem and its derived problems can be solved using the technique given by (17].

### 5.1 In Case $P_{1}$ Is Corrupted

In case $P_{1}$ is corrupted, $A$ takes the full control of $P_{1}$ in the real world. Correspondingly, $A$ 's simulator, $A^{\prime}$, takes the full control of $P_{1}^{\prime}$ in the ideal world, where $A^{\prime}$ is constructed as follow.

- Initial input: $A^{\prime}$ holds the same $k, I \stackrel{\text { def }}{=}\{1\}, z=\left(z_{k}\right)_{k \in \mathbb{N}}$, as $A$. What is more, $A^{\prime}$ holds a uniform distributed randomness $r_{A^{\prime}} \in\{0,1\}^{*}$. The
parties $P_{1}^{\prime}$ and $P_{1}$, whom $A^{\prime}$ and $A$ respectively is to corrupt, hold the same $\vec{m}$.
- $A^{\prime}$ works as follows.
- Step Sm1: $A^{\prime}$ corrupts $P_{1}^{\prime}$ and learns $P_{1}^{\prime}$ 's private input $\vec{m}$. Let $\bar{A}$ be a copy of $A$, i.e. $\bar{A}=A$. $A^{\prime}$ use $\bar{A}$ as a subroutine. $A^{\prime}$ fixes the initial inputs of $\bar{A}$ to be identical to his except that fixes the randomness of $\bar{A}$ to be a uniformly distributed value. $A^{\prime}$ activates $\bar{A}$, and supplies $\bar{A}$ with $\vec{m}$ before $\bar{A}$ engages in the protocol for $O T_{h}^{n}$.
In the following steps, $A^{\prime}$ builds an environment for $\bar{A}$ which simulates the real world. That is, $A^{\prime}$ disguises himself as $P_{1}$ and $P_{2}$ at the same time to interact with $\bar{A}$.
- Step Sm2: $A^{\prime}$ uniformly chooses a randomness $r \in_{U}\{0,1\}^{K}$ ( $\left.K \stackrel{\text { def }}{=} \operatorname{poly}_{s}(k)\right)$ as the shared randomness. Let $C S$ and $\overline{C S}$ be the sets decided by $r$. For each $i \in C S, A^{\prime}$ honestly generates the hash parameters and instance vectors. For each $i \in \overline{C S}, A^{\prime}$ calls Cheat $\left(1^{k}\right)$ which is defined in Definition 4 to generate these thing. $A^{\prime}$ sends these hash parameters and instance vectors to $\bar{A}$.

Remark 9. From the remark 6, we know that each entry of the instance vector generated by Cheat $\left(1^{k}\right)$ is projective. If such instance vectors are not chosen to be open, then the probability of $\bar{A}$ detecting this fact is negligible, and $A^{\prime}$ can extract the real input of $\bar{A}$, which is we want.

- Step Sm3: $A^{\prime}$ plays the role of $P_{2}$ and executes Step R2-R3 of the framework to cooperate with $\bar{A}$ to toss coin. When tossing coin is completed successfully, $A^{\prime}$ learns and records the value $s \bar{A}$ commits to.
Remark 10. The aim of doing this tossing coin is to know the randomness s $\bar{A}$ choses. What $A^{\prime}$ will do next is to take $P B C(r \oplus$ s) as his commitment to redo tossing coin.
- Step Sm4: $A^{\prime}$ repeats the following procedure, denoted $\Upsilon$, until $\bar{A}$ correctly reveals the recorded value $s$.
$\Upsilon: A^{\prime}$ rewinds $\bar{A}$ to the end of Step S1 of the framework. Then, taking $P B C_{\gamma}(r \oplus s)$ as his commitment, $A^{\prime}$ executes Step R2 and R 3 of the framework, where $\gamma$ is a fresh randomness uniformly chosen.
- Step Sm5: Now $A^{\prime}$ and $\bar{A}$ shares the common randomness $r$. $A^{\prime}$ executes Step R4 of the framework as the honest $P_{2}$ do. On receiving $\vec{c}$ and $\left(\overrightarrow{p k}_{i}\right)_{i \in \overline{C S}}, A^{\prime}$ correctly decrypts all entries of $\vec{c}$ and gains $\bar{A}$ 's full real private input $\vec{m}$. Then $A^{\prime}$ sends $\vec{m}$ to the TTP.
- Step Sim6: When $\bar{A}$ halts, $A^{\prime}$ halts with outputing what $\bar{A}$ outputs.

Without considering Step Sim4, $A^{\prime}$ is polynomial-time. However, taking Step Sim4 into consideration, this is not true any more. Let $q(\alpha), p(\alpha)$ respectively denotes the probability that $\bar{A}$ correctly reveals his commitment in Step Sim3 and in Procedure $\Upsilon$, where $\alpha \stackrel{\text { def }}{=}\left(1^{k}, z_{k}, I, \vec{m}, r_{\bar{A}}\right)$. Then, the expected times of repeating $\Upsilon$ in Step $\operatorname{Sim} 4$ is $q(\alpha) / p(\alpha)$. Since the view $\bar{A}$ holds before revealing his commitment in Step Sim3 is different from that in procedure $\Upsilon, q(\alpha), p(\alpha)$ are distinct. What the computational secrecy of $P B C$ guarantees and only guarantees is $|q(\alpha)-p(\alpha)|=\mu($.$) . However,$ there is a risk that $q(\alpha) / p(\alpha)$ is not bound by a polynomial. For example, $q(\alpha)=1 / 2^{k}, p(\alpha)=1 / 2^{2 k}$, which result in $q(\alpha) / p(\alpha)=2^{k}$. This is a big problem and gives rise to many other difficulties we will encounter later.

Fortunately, [17] encounters and solves the same problem and its derived problem as ours. In a little more details, [17] presents a protocol, in which $P_{1}, P_{2}$ respectively sends a perfectly hiding commitment, a perfectly binding commitment, and the corresponding de-commitments to each other as the situation of tossing coin of our framework. To prove the security in case $P_{1}$ is corrupted, [17] constructs a simulator in the same way as ours and encounters the same problem as ours.

Using the idea of [17], we can overcome such problem too. Specifically, an expected polynomial-time simulator can be obtained by replacing Step Sim4 with Step Sim4.1, Sim4.2 given as follow.

- Step Sim4.1: $A^{\prime}$ estimates the value of $q(\alpha) . A^{\prime}$ repeats the following procedure, denoted $\Phi$, until the number of the time of $\bar{A}$ correctly revealing his commitment is up to poly $(k)$, where poly(.) is a big enough polynomial.
$\Phi: A^{\prime}$ rewinds $\bar{A}$ to the end of Step S1 of the framework and $A^{\prime}$ honestly executes Step R2 and R3 of the framework to interact with it.
Denote the number of times that $\Phi$ is repeated by $d$, then $q(\alpha)$ is estimated as $\tilde{q}(\alpha) \stackrel{\text { def }}{=} \operatorname{poly}(k) / d$.
- Step Sim4.2: $A^{\prime}$ repeats the procedure $\Upsilon$. In case $\bar{A}$ correctly reveals the recorded value $s, A^{\prime}$ proceeds to the next step. In case $\bar{A}$ correctly
reveals a value which is different from $s, A^{\prime}$ outputs ambiguity $_{1}$ and halts. In case the number of the time of repeating $\Upsilon$ exceeds the value of $\operatorname{poly}(k) / \tilde{q}(\alpha), A^{\prime}$ outputs timeout and halts.

Proposition 11. The simulator $A^{\prime}$ is expected polynomial-time.
Proof. Conditioning on Step Sim4.1 is executed, the expected value of $d$ is $\operatorname{poly}(k) / q(\alpha)$. Choosing a big enough poly(.), $\tilde{q}(\alpha)$ is within a constant factor of $q(\alpha)$ with probability $1-2^{\text {poly }(k)}$. Therefore, the expected running time of $A^{\prime}$,

$$
\begin{aligned}
\text { ExpTime }_{A^{\prime}} & \leq \text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2}+\text { Time }_{\text {Sim } 3} \\
& +q(\alpha) \cdot\left(\text { Time }_{\Phi} \cdot \operatorname{poly}(k) / q(\alpha)+\right. \\
& \text { Time } \left._{\Upsilon} \cdot \operatorname{poly}(k) / \tilde{q}(\alpha)\right) \\
& + \text { Time }_{\text {Sim } 5}+\text { Time }_{\text {Sim } 6}
\end{aligned}
$$

, is bounded by a polynomial.
What is more, we have

1. The probability that $A^{\prime}$ outputs timeout is negligible.
2. The probability that $A^{\prime}$ outputs ambiguity $y_{1}$ is negligible.
3. The output of $A^{\prime}$ in the ideal world and the output of $A$ in the real world are computationally indistinguishable, ie

$$
\begin{aligned}
& \left\{\text { deale }_{f,\{1\}, A^{\prime}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle\right\}_{\substack{\in \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,\}^{*}}} \stackrel{c}{=} \\
& \left\{\operatorname{Real}_{\pi,\{1\}, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}}
\end{aligned}
$$

Since the propositions above can be proven in the same way as [17], we don't iterate such details here.

Proposition 12. In case $P_{1}$ was corrupted, i.e. $I=\{1\}$, the equation (1) holds.

Proof. First let us focus on the real world. A's real input can be formulated as $\gamma \leftarrow A\left(1^{k}, \vec{m}, z_{k}, r_{A}, r_{1}\right)$. Note that in this case, $P_{2}$ 's output is a determinate function of $A$ 's real input. Since $A$ 's real input is in its view, without loss of generality, we assume $A$ 's output, denoted $\alpha$, constains its real input.

Therefore, $P_{2}$ 's output is a determinate function of $A$ 's output, where the function is

$$
g(\alpha)= \begin{cases}a^{a b o r t_{1}} & \text { if } \gamma=\text { abort }_{1} \\ \gamma\langle H\rangle & \text { otherwise }\end{cases}
$$

Let $h(\alpha) \stackrel{\text { def }}{=}(\alpha, \lambda, g(\alpha))$. Then we have

$$
\operatorname{Real}_{\pi,\{1\}, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right) \equiv \quad h\left(\operatorname{Real}_{\pi,\{1\}, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle\right)
$$

Similarly, in the ideal world, we have

$$
\text { Ideal }_{f,\{1\}, A^{\prime}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right) \stackrel{c}{=} \quad h\left(\operatorname{Ideal}_{f,\{1\}, A^{\prime}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle\right)
$$

We use $\stackrel{c}{=}$ not $\equiv$ here because there is a negligible probability that $A^{\prime}$ outputs timeout or ambiguity $_{1}$, which makes $h($.$) undefined.$

Let $X\left(1^{k}, \vec{m}, H, z_{k},\{1\}\right) \stackrel{\text { def }}{=} \operatorname{Real}_{\pi,\{1\}, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle, Y\left(1^{k}, \vec{m}, H, z_{k},\{1\}\right) \stackrel{\text { def }}{=}$ deal $_{f,\{1\}, A^{\prime}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle, F \stackrel{\text { def }}{=}(h)_{k \in \mathbb{N}}$. What is more, assume that $A^{\prime}$ runs in a strictly polynomial-time. According to Proposition 19 we will present in Section 6, the proposition holds.

In fact, $A^{\prime}$ doesn't run in strictly polynomial-time, which results in a failure of above standard reduction. Fortunately, this difficulty can be overcome by truncating the rare executions of $A^{\prime}$ which are too long, then applying standard reduction techniques. Since the details is the same as (17, we don't give them here and please see [17] for them.

### 5.2 In Case $P_{2}$ Is Corrupted

In case $P_{2}$ is corrupted, $A$ takes the full control of $P_{2}$ in the real world. Correspondingly, $A^{\prime}$ takes the full control of $P_{2}^{\prime}$ in the ideal world. We construct $A$ as follows.

- Initial input: $A^{\prime}$ holds the same $k, I \stackrel{\text { def }}{=}\{2\}, z=\left(z_{k}\right)_{k \in \mathbb{N}}$ as $A$, and holds a uniformly distributed randomness $r_{A^{\prime}} \in\{0,1\}^{*}$. The parties $P_{2}^{\prime}$ and $P_{2}$ hold the same private input $H$.
- $A^{\prime}$ works as follows.
- Step Sim1: $A^{\prime}$ corrupts $P_{2}^{\prime}$ and learns $P_{2}^{\prime \prime}$ s private input $H$. $A^{\prime}$ takes $A$ 's copy $\bar{A}$ as a subroutine, fixes $\bar{A}$ 's initial input, activates $\bar{A}$, supplies $\bar{A}$ with $H$, builds an environment for $\bar{A}$ in the same way as $A^{\prime}$ does in case $P_{1}$ is corrupted.
- Step Sim2: Playing the role of $P_{1}, A^{\prime}$ honestly executes the sender's steps until reaches Step S3.3. If Step S3.3 is reached, $A^{\prime}$ records the shared randomness $r$ and the messages, denoted $m s g$, which he sends to $\bar{A}$. Then $A^{\prime}$ proceeds to next step. Otherwise, $A^{\prime}$ sends abort ${ }_{2}$ to TTP, outputs what $\bar{A}$ outputs and halts.
- Step Sim3: $A^{\prime}$ repeats the following procedure, denoted $\Xi$, until the hash parameters and the instance vectors $\bar{A}$ sends in Step R1 passes the check. $A^{\prime}$ records the shared randomness $\tilde{r}$, the messages $\bar{A}$ sends to open the chosen instance vectors.
$\Xi: A^{\prime}$ rewinds $\bar{A}$ to the beginning of Step R2, and honestly follows sender's steps until reaches Step S3.3 to interact with $\bar{A}$. Note that, the value $A^{\prime}$ commits to and the randomness used to generate the commitment in Step S1 are fresh and uniformly chosen.
- Step Sim4:

1. In case $r=\tilde{r}, A^{\prime}$ outputs failure and halts;
2. In case $r \neq \tilde{r} \wedge \forall i(r\langle i\rangle \neq \tilde{r}\langle i\rangle \rightarrow r\langle i\rangle=1 \wedge \tilde{r}\langle i\rangle=0), A^{\prime}$ runs from scratch;
3. Otherwise, i.e. in case $r \neq \tilde{r} \wedge \exists i(r\langle i\rangle=0 \wedge \tilde{r}\langle i\rangle=1), A^{\prime}$ records the first one, denoted $e$, of these is and proceeds to next step.
Remark 13. The aim of Step Sim3 and Sim4 is to prepare to extract the real input of $\bar{A}$. If the third case happens, then $A^{\prime}$ knows each entry of $\tilde{\vec{x}}_{e}$ he sees in Step Sim2 belong to $L_{\dot{R}_{\Lambda_{e}}}$ or $L_{\ddot{R}_{\Lambda_{e}}}$. What is more, $\tilde{\vec{x}}_{e}$ is indeed a legal instance vector. This is because $\tilde{\vec{x}}_{\text {e }}$ passes the check executed by $A^{\prime}$ in Step Sim3. Combing $\pi_{e}^{2}$ received in Step Sim2, $A^{\prime}$ knows the real input of $\bar{A}$.
Note that, $\bar{A}$ 's initial input is fixed by $A^{\prime}$ in Step Sim1. So receiving the same messages, $\bar{A}$ responds in the same way. Therefore, rewinding $\bar{A}$ to the beginning of Step R2, sending the message sent in Step Sim2, $A^{\prime}$ can reproduce the same scenario as he meets in Step Sim2.

- Step Sim5: $A^{\prime}$ rewinds $\bar{A}$ to the beginning of Step R2 of the framework, and sends $m s g$ previously recorded to $\bar{A}$ in order. According
to the analysis of Remark $13, A^{\prime}$ can extract $\bar{A}$ 's real input $H^{\prime} . A^{\prime}$ does so and sends $H^{\prime}$ to TTP and receives message $\vec{m}\left\langle H^{\prime}\right\rangle$.
- Step Sim6: $A^{\prime}$ constructs $\vec{m}^{\prime}$ as follows. For each $i \in H^{\prime}, \vec{m}^{\prime}\langle i\rangle \leftarrow$ $\vec{m}\langle i\rangle$. For each $i \notin H^{\prime}, \vec{m}^{\prime}\langle i\rangle \in_{U}\{0,1\}^{*}$. Playing the role of $P_{1}$ and taking $\vec{m}^{\prime}$ as his real input, $A^{\prime}$ follows Step S3.3 to complete the interaction with $\bar{A}$.
- Step Sim6: When $\bar{A}$ halts, $A^{\prime}$ halts with outputing what $\bar{A}$ outputs..

Proposition 14. The simulator $A^{\prime}$ is expected polynomial-time.
Proof. First, let us focus on Step Sim3. In each repetition of $\Xi$, because of the perfectly hiding of $P H C($.$) , and the uniform distribution of the value A^{\prime}$ commits to, the chosen instance vectors are uniformly distributed. This lead to the probability that $\bar{A}$ passes the check in each repetition is the same. Denote this probability by $p$. The expected time of Step Sim3 is

$$
\text { ExpTime }_{\text {Sim } 3}=(1 / p) \cdot \text { Time }_{\Xi}
$$

Under the same analysis, the probability that $\bar{A}$ passes the check in Step Sim2 is $p$ too. Then, the expected time that $A^{\prime}$ runs once from Step Sim1 to the beginning of Step Sim4 is

$$
\begin{aligned}
\text { OncExpTime }_{\text {Sim } 1 \rightarrow \text { Sim } 4} & \leq \text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2} \\
& +p \cdot \text { Exp Time }_{\text {Sim } 3} \\
& =\text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2} \\
& + \text { Time }_{\Xi}
\end{aligned}
$$

Second, let us focus on Step Sim4, especially the case that $A^{\prime}$ needs to run from scratch. Note that the initial inputs $A^{\prime}$ holds is the same in each trial. Thus the probability that $A^{\prime}$ runs from scratch in each trial is the same. We denote this probability by $1-q$. Then the expected time that $A^{\prime}$ runs from Step Sim1 to the beginning of Step Sim5 is

$$
\begin{aligned}
\text { Exp Time }_{\text {Sim } 1 \rightarrow \text { Sim } 5} & \leq(1+1 / q) \\
& \cdot\left(\text { OncExpTime }_{\text {Sim } 1 \rightarrow \text { Sim } 4}\right. \\
& \left.+ \text { Time }_{\text {Sim } 4}\right) \\
& =(1+1 / q) \cdot\left(\text { Time }_{\text {Sim } 1}+\right. \\
& \text { Time } \left._{\text {Sim } 2}+\text { Time }_{\Xi}+\text { Time }_{\text {Sim } 4}\right)
\end{aligned}
$$

The reason there is 1 here is that $A^{\prime}$ has to run from scratch at least one time in any case.

The expected running time of $A^{\prime}$ in a whole execution is

$$
\begin{align*}
\text { Exp Time }_{A^{\prime}} & \leq \text { ExpTime }_{\text {Sim } 1 \rightarrow \text { Sim } 5}+\text { Time }_{\text {Sim } 5} \\
& + \text { Time }_{\text {Sim } 6} \\
& =(1+1 / q) \cdot\left(\text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2}\right.  \tag{2}\\
& \left.+ \text { Time }_{\Xi}+\text { Time }_{\text {Sim } 4}\right) \\
& + \text { Time }_{\text {Sim } 5}+\text { Time }_{\text {Sim } 6}
\end{align*}
$$

Third, let us estimate the value of $q$, which is the probability that $A^{\prime}$ does not run from scratch in a trial. We denote this event by $C$. It's easy to see that event $C$ happens, if and only if one of the following events happens.

1. Event $B$ happens, where $B$ denotes the even that $A^{\prime}$ halts before reaching Step Sim3.
2. Event $\bar{B}$ happens and $R=\tilde{R}$, where $R$ and $\tilde{R}$ respectively denotes the random variable which is defined as the shared randomness $A^{\prime}$ gets in Step Sim2 and Step Sim3.
3. Event $\bar{B}$ happens and there exists $i$ such that $R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1$.

So

$$
\begin{align*}
q & =\operatorname{Pr}(C) \\
& =\operatorname{Pr}(B)+\operatorname{Pr}(\bar{B} \cap R=\tilde{R}) \\
& +\operatorname{Pr}(\bar{B} \cap \exists i(R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1))  \tag{3}\\
& =\operatorname{Pr}(B)+\operatorname{Pr}(\bar{B}) \cdot(\operatorname{Pr}(R=\tilde{R} \mid \bar{B}) \\
& +\operatorname{Pr}(\exists i(R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1) \mid \bar{B}))
\end{align*}
$$

Let $S_{1} \stackrel{\text { def }}{=}\left\{(r, \tilde{r}) \mid(r, \tilde{r}) \in\left(\{0,1\}^{K}\right)^{2}, r=\tilde{r}\right\}, S_{2} \stackrel{\text { def }}{=}\left\{(r, \tilde{r}) \mid(r, \tilde{r}) \in\left(\{0,1\}^{K}\right)^{2}, r \neq\right.$ $\tilde{r}, \forall i(r\langle i\rangle \neq \tilde{r}\langle i\rangle \rightarrow r\langle i\rangle=1 \wedge \tilde{r}\langle i\rangle=0)\}, S_{3} \stackrel{\text { def }}{=}\left\{(r, \tilde{r}) \mid(r, \tilde{r}) \in\left(\{0,1\}^{K}\right)^{2}, r \neq\right.$ $\tilde{r}, \exists i(i \in[K] \wedge r\langle i\rangle=0 \wedge \tilde{r}\langle i\rangle=1)\}$. It is easy to see that $S_{1}, S_{2}, S_{3}$ constitute a complete partition of $\left(\{0,1\}^{K}\right)^{2}$ and $\# S_{1}=2^{K}, \# S_{2}=\# S_{3}=$ $\left(2^{K} \cdot 2^{K}-2^{K}\right) / 2$.

Because of the perfectly hiding of $P H C($.$) , and the uniform distribution$ of the value $A^{\prime}$ commits to, $R$ and $\tilde{R}$ are all uniformly distributed. We have

$$
\begin{equation*}
\operatorname{Pr}(R=\tilde{R} \mid \bar{B})=\# S_{1} / \#\left(\{0,1\}^{K}\right)^{2}=1 / 2^{K} \tag{4}
\end{equation*}
$$

and

$$
\begin{align*}
\operatorname{Pr}(\exists i(R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1) \mid \bar{B}) & =\# S_{3} / \#\left(\{0,1\}^{K}\right)^{2} \\
& =1 / 2-1 / 2^{K+1} \tag{5}
\end{align*}
$$

Combining equation (3), (4) and (5), we have

$$
\begin{align*}
q & =\operatorname{Pr}(B)+\operatorname{Pr}(\bar{B})\left(1 / 2+1 / 2^{K+1}\right) \\
& =1 / 2+1 / 2^{K+1}+(1 / 2) \operatorname{Pr}(B)+\left(1 / 2^{K+1}\right) \operatorname{Pr}(\bar{B})  \tag{6}\\
& >1 / 2
\end{align*}
$$

Combining equation (2) and (6), we have

$$
\begin{aligned}
\text { ExpTime }_{A^{\prime}} & <3\left(\text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2}\right. \\
& \left.+ \text { Time }_{\Xi}+\text { Time }_{\text {Sim } 4}\right) \\
& + \text { Time }_{\text {Sim } 5}+\text { Time }_{\text {Sim } 6}
\end{aligned}
$$

which means the expected running time of $A^{\prime}$ is bound by a polynomial.
Proposition 15. The probability that $A^{\prime}$ outputs failure is less than $1 / 2^{K-1}$.
Proof. From the proof of Proposition 14, we know two fact. First, $\operatorname{Pr}(X=$ $i) \leq 1 / 2^{i-1}$, where $X$ is a random variable defined as the number of the trials in a whole execution. Second, in each trial the event $A^{\prime}$ outputs failure is the combined event of $\bar{B}$ and $R=\tilde{R}$, and this event happens with probability

$$
\operatorname{Pr}(\bar{B} \cap R=\tilde{R})=\operatorname{Pr}(\bar{B}) \operatorname{Pr}(R=\tilde{R} \mid \bar{B}) \leq \operatorname{Pr}(R=\tilde{R} \mid \bar{B})
$$

Combining equation (4), this probability is less than $1 / 2^{K}$. Therefore, the probability that $A^{\prime}$ outputs failure in a whole execution is

$$
\begin{aligned}
\sum_{i=1}^{\infty} \operatorname{Pr}(X=i) \operatorname{Pr}(\bar{B} \cap R=\tilde{R}) & <\left(1 / 2^{K}\right) \cdot \sum_{i=1}^{\infty} 1 / 2^{i-1} \\
& =1 / 2^{K-1}
\end{aligned}
$$

Proposition 16. The output of the adversary $A$ in the real world and that of the simulator $A^{\prime}$ in the ideal world are computationally indistinguishable, ie

$$
\begin{gathered}
\left\{\operatorname{Real}_{\pi,\{2\}, A\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 1\rangle\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}}=\substack{c \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}
\end{gathered}
$$

Proof. First, we claim that the outputs of $A^{\prime}$ and $\bar{A}$ are computationally indistinguishable. The only point that the output of $A^{\prime}$ is different from that of $\bar{A}$ is $A^{\prime}$ may outputs failure. Since the probability that this point arises is negligible, our claim holds.

Second, we claim that the outputs of $A$ and $\bar{A}$ are computationally indistinguishable. The only point that the view of $\bar{A}$ is different from that of $A$ is that the ciphertext $\bar{A}$ receives is generated by encrypting $\vec{m}^{\prime}$ not $\vec{m}$. Fortunately, $S P W H_{h, t}$ 's property smoothness guarantees that the ciphertext generated in the two way are computationally indistinguishable. Therefore, our claim holds.

Combining the two claims, the proposition holds.
Proposition 17. In case $P_{2}$ was corrupted, i.e. $I=\{2\}$, the equation (1) holds.

Proof. Note that the honest parties $P_{1}$ and $P_{1}^{\prime}$ end up with outputing nothing. Thus, the fact that the outputs of $A^{\prime}$ and $A$ are computationally indistinguishable, which is supported by Proposition 16, directly prove this proposition holds.

### 5.3 Other Cases

In case both $P_{1}$ and $P_{2}$ are corrupted, $A$ takes the full control of the two corrupted parties. In the ideal world, a similar situation also holds with respect to $A^{\prime}, P_{1}^{\prime}$ and $P_{2}^{\prime}$. Liking in previous cases, $A^{\prime}$ uses $A^{\prime}$ 's copy, $\bar{A}$, as a subroutine and builds a simulated environment for $\bar{A} . A^{\prime}$ provids $\bar{A}$ with $P_{1}^{\prime}$ and $P_{2}^{\prime}$ 's initial inputs before $\bar{A}$ engages in the protocol. When $\bar{A}$ halts, $A^{\prime}$ halts with outputing what $\bar{A}$ outputs. Obviously, $A^{\prime}$ runs in strictly polynomial-time and the equation (1) holds in this case.

In case none of $P_{1}$ and $P_{2}$ is corrupted. The simulator $A^{\prime}$ is constructed as follows. $A^{\prime}$ uses $\bar{A}, \bar{P}_{1}, \bar{P}_{2}$ as subroutines, where $\bar{A}, \bar{P}_{1}, \bar{P}_{2}$, respectively, is the copy of $A, P_{1}$ and $P_{2}$. $A^{\prime}$ fixes $\bar{A}$ 's initial inputs in the same way as in previous cases. $A^{\prime}$ chooses an arbitrary $\overline{\vec{m}} \in\left(\{0,1\}^{*}\right)^{n}$ and a uniformly distributed randomness $\bar{r}_{1}$ as $\bar{P}_{1}$ 's initial inputs. $A^{\prime}$ chooses an arbitrary $\bar{H} \in \Psi$ and a uniformly distributed randomness $\bar{r}_{2}$ as $\bar{P}_{2}$ 's initial inputs. $A^{\prime}$ actives these subroutines and make the communication between $\bar{P}_{1}$ and $\bar{P}_{2}$ be available to $\bar{A}$. When $\bar{A}$ halts, $A^{\prime}$ halts with outputing what $\bar{A}$ outputs. Obviously, $A^{\prime}$ runs in strictly polynomial-time and the equation (1) holds in this case.

## 6 Constructing $S P W H_{h, t}$

### 6.1 How To Constructe $S P W H_{h, t}$ Easily

It is not always easy to construct $S P W H_{h, t}$ from scratch. In this section, we reduce the task of constructing $S P W H_{h, t}$ to that of constructing a hash system which seems easier.

### 6.1.1 Smoothness

In this section, we deal with how to obtain smoothness for a hash family. First, we introduce a lemma from [16.

Lemma $18(16)$. Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \stackrel{\text { def }}{=}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two polynomial-time constructible probability ensembles, and $X \stackrel{c}{=} Y$, then

$$
\vec{X} \stackrel{c}{=} \vec{Y}
$$

where $\vec{X} \stackrel{\text { def }}{=}\left\{\vec{X}\left(1^{k}, a\right)\right\} \underset{\substack{k \in \mathbb{N}, 1\}^{*}}}{ }, \vec{X}\left(1^{k}, a\right) \stackrel{\text { def }}{=}\left(X_{i}\left(1^{k}, a\right)\right)_{i \in[\text { poly }(k)]}$, each $X_{i}\left(1^{k}, a\right)=$ $X\left(1^{k}, a\right), \vec{Y} \stackrel{\text { def }}{=}\left\{\vec{Y}\left(1^{k}, a\right)\right\} \underset{\substack{k \in\{0,1\}^{*}}}{ }, \vec{Y}\left(1^{k}, a\right) \stackrel{\text { def }}{=}\left(Y_{i}\left(1^{k}, a\right)\right)_{i \in[p o l y(k)]}$, each $Y_{i}\left(1^{k}, a\right)=$ $Y\left(1^{k}, a\right)$, and all $X_{i}\left(1^{k}, a\right), Y_{i}\left(1^{k}, a\right)$ are independent.

Proposition 19. Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \xlongequal{\text { def }}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two polynomial-time constructible probability ensembles, $X \stackrel{c}{=} Y, F \stackrel{\text { def }}{=}$ $\left(f_{k}\right)_{k \in \mathbb{N}}, f_{k}:\{0,1\}^{*} \rightarrow\{0,1\}^{*}$ is polynomial-time computable, then

$$
F(X) \stackrel{c}{=} F(Y)
$$

where $F(X) \stackrel{\text { def }}{=}\left\{f_{k}\left(X\left(1^{k}, a\right)\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}, F(Y) \stackrel{\text { def }}{=}\left\{f_{k}\left(Y\left(1^{k}, a\right)\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$.
Proof. Assume the proposition is false, then there exists a non-uniform PPT distinguisher $D$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$, a polynomial poly(.), an infinite positive integer set $G \subseteq \mathbb{N}$ such that, for each $k \in G$, it holds that

$$
\begin{aligned}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, f_{k}\left(X\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, f_{k}\left(Y\left(1^{k}, a\right)\right)\right)=1\right) \mid \geq 1 / \operatorname{poly}(k)
\end{aligned}
$$

We construct a distinguisher $D^{\prime}$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ for the ensembles $X$ and $Y$ as follows.

$$
D^{\prime}\left(1^{k}, z_{k}, a, \gamma\right): \delta \leftarrow f_{k}(\gamma) \text {, finally outputs } D\left(1^{k}, z_{k}, a, \delta\right)
$$

Obviously, $D^{\prime}\left(1^{k}, z_{k}, a, X\left(1^{k}, a\right)\right)=D\left(1^{k}, z_{k}, a, f_{k}\left(X\left(1^{k}, a\right)\right), D^{\prime}\left(1^{k}, z_{k}, a, Y\left(1^{k}, a\right)\right)=\right.$ $D\left(1^{k}, z_{k}, a, f_{k}\left(Y\left(1^{k}, a\right)\right)\right.$. So we have

$$
\begin{aligned}
& \mid \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, X\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, Y\left(1^{k}, a\right)\right)=1\right) \mid \geq 1 / \operatorname{poly}(k)
\end{aligned}
$$

This contradicts the fact $X \stackrel{c}{=} Y$.
Theorem 20. Let $\mathcal{H}=(P G, I S, V F, H G$, Hash, pHash $)$ be a Hash Family. $n \stackrel{\text { def }}{=} h+t$. For each $i \in[2]$ and $j \in[n], S m_{i}^{j} \stackrel{\text { def }}{=}\left\{S m_{i}^{j}\left(1^{k}\right)\right\}_{k \in \mathbb{N}} \stackrel{\text { def }}{=}$ $\left\{\left(\operatorname{SmGen}_{i}\left(1^{k}\right)\langle 1\rangle, \operatorname{SmGen}_{i}\left(1^{k}\right)\langle 2\rangle\langle j\rangle\right)\right\}_{k \in \mathbb{N}}$, where $\operatorname{SmGen}_{i}\left(1^{k}\right)$ is defined in Definition 4. If $\mathcal{H}$ meets the following three conditions

1. All random variables $\operatorname{SmGen}_{i}\left(1^{k}\right)\langle 2\rangle\langle j\rangle$ are independent, where $i \in$ $[2], j \in[n]-[h]$.
2. $S m_{1}^{h+1}=\ldots=S m_{1}^{n}$, and $S m_{2}^{h+1}=\ldots=S m_{2}^{n}$.
3. $S m_{1}^{h+1} \stackrel{c}{=} S m_{2}^{h+1}$.
then $\mathcal{H}$ has property smoothness.
Proof. Following Lemma 18 ,

$$
\left\{\left(S m_{1}^{h+1}\left(1^{k}\right), \ldots, S m_{1}^{n}\left(1^{k}\right)\right)\right\}_{k \in \mathbb{N}} \quad \stackrel{c}{=}\left\{\left(S m_{2}^{h+1}\left(1^{k}\right), \ldots, S m_{2}^{n}\left(1^{k}\right)\right)\right\}_{k \in \mathbb{N}}
$$

holds. Let $\vec{X} \stackrel{\text { def }}{=}\left\{\left(S m_{1}^{1}\left(1^{k}\right), \ldots, S m_{1}^{n}\left(1^{k}\right)\right)\right\}_{k \in \mathbb{N}}$, and $\vec{Y} \stackrel{\text { def }}{=}\left\{\left(S m_{2}^{1}\left(1^{k}\right), \ldots, S m_{2}^{n}\left(1^{k}\right)\right)\right\}_{k \in \mathbb{N}}$. From the definition of $\operatorname{SmGen}_{i}\left(1^{k}\right)$, we notice that, for each $j \in[h] S m_{1}^{j}\left(1^{k}\right)=$ $S m_{2}^{j}\left(1^{k}\right)$. So it holds that

$$
\vec{X} \stackrel{c}{=} \vec{Y}
$$

Since each $S m_{i}^{j}\left(1^{k}\right)$ is polynomial-time constructible, thus both $\vec{X}$ and $\vec{Y}$ are polynomial-time constructible. Let $F \stackrel{\text { def }}{=}(\pi)_{k \in \mathbb{N}}$, where $\pi \in \Pi$. Following Proposition 19, we have $F(\vec{X}) \stackrel{c}{=} F(\vec{Y})$, i.e.

$$
\left\{\pi\left(S m_{1}^{1}\left(1^{k}\right), \ldots, S m_{1}^{n}\left(1^{k}\right)\right)\right\}_{k \in \mathbb{N}} \quad \stackrel{c}{=}\left\{\pi\left(\operatorname{Sm}_{2}^{1}\left(1^{k}\right), \ldots, S m_{2}^{n}\left(1^{k}\right)\right)\right\}_{k \in \mathbb{N}}
$$

Notice that $\operatorname{SmGen}_{1}\left(1^{k}\right)\langle 1\rangle=\operatorname{SmGen}_{2}\left(1^{k}\right)\langle 1\rangle$, we have

$$
\begin{aligned}
& \left\{\left(\operatorname{SmGen}_{1}\left(1^{k}\right)\langle 1\rangle, \pi\left(\operatorname{SmGen}_{1}\left(1^{k}\right)\langle 2\rangle\right)\right)\right\}_{k \in \mathbb{N}} \\
& \quad \stackrel{c}{=}\left\{\left(\operatorname{SmGen}_{2}\left(1^{k}\right)\langle 1\rangle, \pi\left(\operatorname{SmGen}_{2}\left(1^{k}\right)\langle 2\rangle\right)\right)\right\}_{k \in \mathbb{N}}
\end{aligned}
$$

That is

$$
S m_{1} \stackrel{c}{=} S m_{2}
$$

Loosely speaking, following Theorem 20, given a hash family $\mathcal{H}$, if each $\ddot{x}$ was sampled in an independent way and its projective key is useless to obtain the value of $\operatorname{Hash}\left(1^{k}, \Lambda, \ddot{x},.\right)$, then $\mathcal{H}$ is smooth.

Sometimes it is not easy to gain smoothness for hash family $\mathcal{H}$ in such a way. In this case we have to constructe a hash family, defined as follows, as the first step to our goal.

Definition 21 ( $\epsilon$ - $h$-Universal $t$-Projective Hash Family With Witnesses And Hard Subset Membership, $\epsilon-U P W H_{h, t}$. The definition of $\epsilon-U P W H_{h, t}$ is obtained by relaxing the definition of $S P W H_{h, t}$ (i.e. Definition (4) in the way of replacing the smoothness with a new property, called $\epsilon$-universality. A hash family is $\epsilon$-universal, if for any sufficiently large $k$, any $\Lambda \in \operatorname{Range}\left(P G\left(1^{k}\right)\right)$, any $\ddot{x} \in \operatorname{Range}\left(I S\left(1^{k}, \Lambda\right)\right)$, any $p k \in \operatorname{Range}\left(K G\left(1^{k}, \Lambda, \ddot{x}\right)\langle 2\rangle\right)$, any $y \in$ Range $\left(\operatorname{Hash}\left(1^{k}, \Lambda, \ddot{x},.\right)\right)$, it holds that

$$
\operatorname{Pr}\left(\operatorname{Hash}\left(1^{k}, \Lambda, \ddot{x}, H K\right)=y \mid P K=p k\right) \leq \epsilon
$$

where $(H K, P K)$ is uniformly chosen from Range $\left(K G\left(1^{k}, \Lambda, \ddot{x}\right)\right)$, i.e. $(H K, P K) \leftarrow$ $K G_{R}\left(1^{k}, \Lambda, \ddot{x}\right)$, where $R$ is random variable defined over $\{0,1\}^{*}$ with a uniform distribution.

Compared with $S P W H_{h, t}, \epsilon-U P W H_{h, t}$ relaxes the requirement of the randomness of the hash value of $\ddot{x}$ to some extent and requires that, given $\ddot{x}, p k$, the probability of guessing the value of $\operatorname{Hash}\left(1^{k}, \ddot{x}, h k\right)$ is at most $\epsilon$. Assume $\epsilon<1$, as 8, 22], we can efficiently gain a $S P W H_{h, t}$ from a $\epsilon-U P W H_{h, t}$.

Theorem 22. There exists an efficient algorithm that receiving a $\epsilon-U P W H_{h, t}$ $\mathcal{H}$, where $\epsilon<1$, it outputs a $S P W H_{h, t} \mathcal{H}^{\prime}$.

The way to prove this theorem is to construct a required algorithm, which can be gained by a simply application of the Leftover Hash Lemma (please see [28] for this lemma). The detailed construction essentially is the same as $[8]$. Considering the space, we don't iterate it here.

Theorem 22 implies that to construct a $S P W H_{h, t}$, what we need to do is only to construct a $\epsilon-U P W H_{h, t}$, where $\epsilon<1$.

### 6.1.2 Hard Subset Membership

In this section, we deal with how to obtain hard subset membership for a hash family.
Proposition 23. Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \stackrel{\text { def }}{=}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two polynomial-time constructible probability ensembles, and $X \stackrel{c}{=} Y$. Then

$$
\overrightarrow{X Y} \stackrel{c}{=} \Phi(\overrightarrow{\widetilde{X Y}})
$$

where $\overrightarrow{X Y}$ and $\Phi(\overrightarrow{\overrightarrow{X Y}})$ are two probability ensembles defined as follows.

- $\overrightarrow{X Y} \stackrel{\text { def }}{=}\left\{\overrightarrow{X Y}\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}, \overrightarrow{X Y}\left(1^{k}, a\right) \stackrel{\text { def }}{=}\left(X_{1}\left(1^{k}, a\right), \ldots, X_{\text {poly }}^{1}(k)\left(1^{k}, a\right), Y_{\text {poly }}(k)+1\left(1^{k}\right.\right.$, $\left.Y_{\text {poly(k) }}\left(1^{k}, a\right)\right)$, each $X_{i}\left(1^{k}, a\right)=X\left(1^{k}, a\right)$, each $Y_{i}\left(1^{k}, a\right)=Y\left(1^{k}, a\right)$, poly $y_{1}($.$) \leq poly ($.$) , all X_{i}\left(1^{k}, a\right)$ and $Y_{i}\left(1^{k}, a\right)$ are independent;
- $\Phi(\overrightarrow{\widetilde{X Y}}) \stackrel{\text { def }}{=}\left\{\Phi_{k}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}, \overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)=\overrightarrow{X Y}\left(1^{k}, a\right), \Phi \stackrel{\text { def }}{=}$ $\left(\Phi_{k}\right)_{k \in \mathbb{N}}$, each $\Phi_{k}$ is a permutation over $[$ poly $(k)]$.

Proof. In case $\Phi_{k}\left(\left[\operatorname{poly}_{1}(k)\right]\right) \subseteq\left[\operatorname{poly}_{1}(k)\right]$, it obviously holds. We proceed to prove it also holds in case $\Phi_{k}\left(\left[\operatorname{poly}_{1}(k)\right]\right) \nsubseteq\left[\operatorname{poly}_{1}(k)\right]$. Assume it does not hold in this case, then there exists a non-uniform PPT distinguisher $D$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$, a polynomial poly $2($.$) , a infinite positive$ integer set $G \subseteq \mathbb{N}$ such that, for each $k \in G$,

$$
\begin{align*}
\mid \operatorname{Pr}\left(D \left(1^{k}, z_{k}, a,\right.\right. & \left.\left.\overrightarrow{X Y}\left(1^{k}, a\right)\right)=1\right) \\
& \quad-\operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)=1\right) \mid\right. \\
& \geq 1 / \operatorname{poly}_{2}(k) \tag{7}
\end{align*}
$$

$V \stackrel{\text { def }}{=}\left\{i \mid i \in\left[\operatorname{pol}_{1}(k)\right], \Phi_{k}(i) \in[\operatorname{poly}(k)]-\left[\operatorname{pol}_{1}(k)\right]\right\}$. We order the elements of $V$ as $i_{1}<\ldots<i_{j} \ldots<i_{\# V}$. Let $V_{j} \stackrel{\text { def }}{=}\left\{i_{1}, \ldots, i_{j}\right\}$. We define the following permutations over $[\operatorname{poly}(k)]$.

$$
\begin{gathered}
\Phi_{k}^{0^{\prime}}(i)=i \quad i \in[\operatorname{poly}(k)] \\
\Phi_{k}^{0}(i)= \begin{cases}i & i \in V \cup \Phi_{k}(V) \\
\Phi_{k}(i) & i \in[\operatorname{poly}(k)]-V-\Phi_{k}(V)\end{cases}
\end{gathered}
$$

For $j \in[\# V]$,

$$
\Phi_{k}^{j}(i)= \begin{cases}i & i \in\left(V-V_{j}\right) \cup \Phi_{k}\left(V-V_{j}\right), \\ \Phi_{k}(i) & i \in[\operatorname{poly}(k)]-\left(V-V_{j}\right)-\Phi_{k}\left(V-V_{j}\right) .\end{cases}
$$

It is easy to see that $\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)=\Phi_{k}^{0^{\prime}}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right) \equiv \Phi_{k}^{0}\left(\overrightarrow{\overrightarrow{X Y}}\left(\underline{\left.\underline{1^{k}}, a\right)}\right)\right.$, and $\Phi_{k}=\Phi_{k}^{\# V}$. Since $\overrightarrow{X Y}\left(1^{k}, a\right)=\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)$, then $\overrightarrow{X Y}\left(1^{k}, a\right) \stackrel{c}{=} \Phi_{k}^{0}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)$. So we have

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{X Y}\left(1^{k}, a\right)\right)=1\right)- \\
& \qquad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \\
& =\mid \\
& P r\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{0}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)\right)=1\right)-  \tag{8}\\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{\# V}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)\right)=1\right) \mid
\end{align*}
$$

Following triangle inequality, we have

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{0}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{\# V}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \leq \\
& \sum_{j=1}^{\# V} \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \tag{9}
\end{align*}
$$

Combining equation (7) (8) (9), we have

$$
\begin{aligned}
& \sum_{j=1}^{\# V} \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \geq 1 / \operatorname{poly}_{2}(k)
\end{aligned}
$$

So there exists $j \in[\# V]$ such that

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \\
& \geq 1 /\left(\# V \cdot \operatorname{poly}_{2}(k)\right) \tag{10}
\end{align*}
$$

According to the definition of $\Phi_{k}^{j-1}, \Phi_{k}^{j}$, the differences between them are the values of points $i_{j}, \Phi_{k}\left(i_{j}\right)$. Similarly, the only differences between $\Phi_{k}^{j-1}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)$ and $\left.\Phi_{k}^{j} \stackrel{\widetilde{X Y}}{ }\left(1^{k}, a\right)\right)$ are the $i_{j}$-th and $\Phi_{k}\left(i_{j}\right)$-th entries, i.e. $\Phi_{k}^{j-1}\left(\stackrel{\rightharpoonup}{X Y}\left(1^{k}, a\right)\right)\left\langle i_{j}\right\rangle=$
$X\left(1^{k}, a\right), \Phi_{k}^{j-1}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)\left\langle\Phi_{k}\left(i_{j}\right)\right\rangle=Y\left(1^{k}, a\right), \Phi_{k}^{j}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)\left\langle i_{j}\right\rangle=Y\left(1^{k}, a\right)$, $\Phi_{k}^{j}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)\left\langle\Phi_{k}\left(i_{j}\right)\right\rangle=X\left(1^{k}, a\right)$.
Let $\overrightarrow{M X Y} \stackrel{\text { def }}{=}\left\{\overrightarrow{M X Y}\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$, where $\overrightarrow{M X Y}\left(1^{k}, a\right)$ is defined as follows. For each $d \in[\operatorname{poly}(k)]$,

$$
\overrightarrow{M X Y}\left(1^{k}, a\right)\langle d\rangle= \begin{cases}\Phi_{k}^{j-1}(\overrightarrow{X Y Y} \\ \left.\left.1^{k}, a\right)\right)\langle d\rangle & d \neq \Phi_{k}\left(i_{j}\right) \\ X\left(1^{k}, a\right) & d=\Phi_{k}\left(i_{j}\right)\end{cases}
$$

The difference between $\overrightarrow{M X Y}\left(1^{k}, a\right)$ and $\Phi_{k}^{j-1}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)$ is that $\overrightarrow{M X Y}\left(1^{k}, a\right)\left\langle\Phi_{k}\left(i_{j}\right)\right\rangle=$ $X\left(1^{k}, a\right), \Phi_{k}^{j-1}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)\left\langle\Phi_{k}\left(i_{j}\right)\right\rangle=Y\left(1^{k}, a\right)$. The difference between $\overrightarrow{M X Y}\left(1^{k}, a\right)$ and $\Phi_{k}^{j}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)$ is that $\left.\overrightarrow{M X Y}\left(1^{k}, a\right)\left\langle i_{j}\right\rangle=X\left(1^{k}, a\right), \Phi_{k}^{j} \overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\left\langle i_{j}\right\rangle=$ $Y\left(1^{k}, a\right)$. Following triangle inequality, we have

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{M X Y}\left(1^{k}, a\right)\right)=1\right) \mid+ \\
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{M X Y}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \\
& \geq \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \tag{11}
\end{align*}
$$

Combining (10) (11), we know that

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\overrightarrow{X Y}}\left(1^{k}, a\right)\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{M X Y}\left(1^{k}, a\right)\right)=1\right) \mid \\
& \quad \geq 1 /\left(2 \# V \cdot \operatorname{poly}_{2}(k)\right) \tag{12}
\end{align*}
$$

or

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{M X Y}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j}\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)\right)=1\right) \mid \\
& \geq 1 /\left(2 \# V \cdot \operatorname{poly}_{2}(k)\right) \tag{13}
\end{align*}
$$

holds. Without loss of generality, we assume equation (12) holds (in case equation (13) holds, the proof can be done in similar way). We can construct
a distinguisher $D^{\prime}$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ for the probability ensembles $X$ and $Y$ as follows.

$$
D^{\prime}\left(1^{k}, z_{k}, a, \gamma\right): \overrightarrow{x y}\left\langle\Phi_{k}^{j-1}(i)\right\rangle \leftarrow S_{X}\left(1^{k}, a\right) \forall i \in\left[\operatorname{poly}_{1}(k)\right], \overrightarrow{x y}\left\langle\Phi_{k}^{j-1}(i)\right\rangle \leftarrow
$$ $S_{Y}\left(1^{k}, a\right) \forall i \in[\operatorname{poly}(k)]-\left[\operatorname{poly}_{1}(k)\right]-\left\{\Phi_{k}\left(i_{j}\right)\right\}, \overrightarrow{x y}\left\langle\Phi_{k}\left(i_{j}\right)\right\rangle \leftarrow \gamma$, finally outputs $D\left(1^{k}, z_{k}, a, \overrightarrow{x y}\right)$.

Obviously, if $\gamma$ is sampled from $Y\left(1^{k}, a\right)$, then $\overrightarrow{x y}$ is an instance of $\Phi_{k}^{j-1}\left(\overrightarrow{\widetilde{X Y}}\left(1^{k}, a\right)\right)$; if $\gamma$ is sampled from $X\left(1^{k}, a\right)$, then $\overrightarrow{x y}$ is an instance of $\overrightarrow{M X Y}\left(1^{k}, a\right)$. So we have

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, X\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, Y\left(1^{k}, a\right)\right)=1\right) \mid= \\
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{M X Y}\left(1^{k}, a\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \Phi_{k}^{j-1}\left(\overrightarrow{\widehat{X Y}}\left(1^{k}, a\right)\right)\right)=1\right) \mid \tag{14}
\end{align*}
$$

Combining (12) (14), we have

$$
\begin{aligned}
& \mid \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, X\left(1^{k}, a\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, Y\left(1^{k}, a\right)\right)=1\right) \mid \geq 1 /\left(2 \# V \cdot \operatorname{poly}_{2}(k)\right)
\end{aligned}
$$

This contradicts the fact $X \stackrel{c}{=} Y$.
Proposition 24. Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \stackrel{\text { def }}{=}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two polynomial-time constructible probability ensembles, and $X \stackrel{c}{=} Y$. Then

$$
\vec{X} \stackrel{c}{=} \overrightarrow{X Y}
$$

where $\vec{X}$ is defined in Lemma 18 and $\overrightarrow{X Y}$ is defined in Proposition 23. All random variables $\vec{X}\left(1^{k}, a\right)\langle i\rangle$ and $\overrightarrow{X Y}\left(1^{k}, a\right)\langle i\rangle$ are independent.

Proof. Assume the proposition is false, then there exists a non-uniform PPT distinguisher $D$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$, a polynomial poly $y_{2}($.$) ,$ an infinite positive integer set $G \subseteq \mathbb{N}$ such that, for each $k \in G$,

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \vec{X}\left(1^{k}, a\right)\right)=1\right)- \\
& \quad \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{X Y}\left(1^{k}, a\right)\right)=1\right) \mid \geq 1 / \operatorname{poly}_{2}(k) \tag{15}
\end{align*}
$$

Let $\operatorname{Hybird}_{j} \stackrel{\text { def }}{=}\left\{\operatorname{Hybird}_{j}\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}, \operatorname{Hybird}_{j}\left(1^{k}, a\right) \stackrel{\text { def }}{=}\left(X_{1}\left(1^{k}, a\right), \ldots, X_{\text {poly }_{1}(k)+j}\left(1^{k}, a\right)\right.$, $\left.Y_{\text {poly }_{1}(k)+j+1}\left(1^{k}, a\right), \ldots, Y_{\text {poly }(k)}\left(1^{k}, a\right)\right)$. Let $d \stackrel{\text { def }}{=} \operatorname{poly}(k)-\operatorname{poly}_{1}(k)$. Obvi-
ously, $\operatorname{Hybird}_{0}\left(1^{k}, a\right)=\overrightarrow{X Y}\left(1^{k}, a\right), \operatorname{Hybird}_{d}\left(1^{k}, a\right)=\vec{X}\left(1^{k}, a\right)$, so we have

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \vec{Y}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \overrightarrow{X Y}\left(1^{k}, a\right)\right)=1\right) \mid= \\
& \mid \operatorname{Pr}\left(D \left(1^{k}, z_{k}, a,\right.\right.\left.\left.\operatorname{Hybird}_{0}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{d}\left(1^{k}, a\right)\right)=1\right) \mid \tag{16}
\end{align*}
$$

Following triangle inequality, we have

$$
\begin{align*}
& \sum_{j=1}^{d} \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{j-1}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{j}\left(1^{k}, a\right)\right)=1\right) \mid \geq \\
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{0}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{d}\left(1^{k}, a\right)\right)=1\right) \mid \tag{17}
\end{align*}
$$

Combining (15) (16) (17), we know that there exists a constant $j \in[d]$ such that

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{j-1}\left(1^{k}, a\right)\right)=1\right)- \\
& \qquad \begin{aligned}
\operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{j}\left(1^{k}, a\right)\right)\right. & =1) \mid \\
& \geq 1 /\left(d \cdot \operatorname{poly}_{2}(k)\right)
\end{aligned}
\end{align*}
$$

The difference between $\operatorname{Hybird}_{j-1}\left(1^{k}, a\right)$ and $\operatorname{Hybird}_{j-1}\left(1^{k}, a\right)$ is the poly $y_{1}(k)+$ $j$-th entry, i.e. $\operatorname{Hybird}_{j-1}\left(1^{k}, a\right)\left\langle\operatorname{poly}_{1}(k)+j\right\rangle=Y\left(1^{k}, a\right), \operatorname{Hybird}_{j}\left(1^{k}, a\right)\left\langle\right.$ poly $_{1}(k)+$ $j\rangle=X\left(1^{k}, a\right)$. We can construct a distinguisher $D^{\prime}$ with an infinite sequence $z^{\prime}=\left(z_{k}^{\prime}\right)_{k \in \mathbb{N}}$ for the probability ensembles $X$ and $Y$ as follows.
$D^{\prime}\left(1^{k}, z_{k}^{\prime}, a, \gamma\right): \overrightarrow{x y}\langle i\rangle \leftarrow S_{X}\left(1^{k}, a\right) \forall i \in\left[\operatorname{pol}_{1}(k)+j-1\right], \overrightarrow{x y}\langle i\rangle \leftarrow \gamma i=$ $\operatorname{poly}_{1}(k)+j, \overrightarrow{x y}\langle i\rangle \leftarrow S_{Y}\left(1^{k}, a\right) \quad \forall i \in[\operatorname{poly}(k)]-\left[\operatorname{pol}_{1}(k)+j\right]$.

Obviously,

$$
\begin{align*}
& \mid \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, Y\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, X\left(1^{k}, a\right)\right) \mid=\right. \\
& \mid \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird} d_{j-1}\left(1^{k}, a\right)\right)=1\right)- \\
& \operatorname{Pr}\left(D\left(1^{k}, z_{k}, a, \operatorname{Hybird}_{j}\left(1^{k}, a\right)\right)=1\right) \mid \tag{19}
\end{align*}
$$

Combining (18) (19), we have

$$
\begin{aligned}
\mid \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, Y\left(1^{k}, a\right)\right)=1\right)- & \\
& \operatorname{Pr}\left(D^{\prime}\left(1^{k}, z_{k}, a, X\left(1^{k}, a\right)\right) \mid \geq 1 /\left(d \cdot \operatorname{poly}_{2}(k)\right.\right.
\end{aligned}
$$

This contradicts the fact $X \stackrel{c}{=} Y$.

Proposition 25. Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \stackrel{\text { def }}{=}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two polynomial-time constructible probability ensembles, $X \stackrel{c}{=} Y, F=$ $\left(f_{k}\right)_{k \in \mathbb{N}}, f_{k}:\{0,1\}^{*} \rightarrow\{0,1\}^{*}$ is a polynomial-time computable function, then

$$
F(\vec{X}) \stackrel{c}{=} F(\vec{Y})
$$

where $F(\vec{X}) \stackrel{\text { def }}{=}\left\{f_{k}\left(\vec{X}\left(1^{k}, a\right)\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}, F(\vec{Y}) \stackrel{\text { def }}{=}\left\{f_{k}\left(\vec{Y}\left(1^{k}, a\right)\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$, $\vec{X}\left(1^{k}, a\right)$ and $\vec{Y}\left(1^{k}, a\right)$ are defined in Lemma 18 .

Proof. Following lemma $18, \vec{X} \stackrel{c}{=} \vec{Y}$ holds. Since the probability ensemble $X$ is polynomial-time constructible, so we can gain a PPT sampling algorithm for the probability ensemble $\vec{X}$ by invocating $S_{X}($.$) poly (k)$ times, thus $\vec{X}$ also is polynomial-time constructible. We can prove $\vec{Y}$ is polynomial-time constructible in the same way. Following Proportion 19, we have $F(\vec{X}) \stackrel{c}{=}$ $F(\vec{Y})$.

Proposition 26. Let $X \stackrel{\text { def }}{=}\left\{X\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ and $Y \stackrel{\text { def }}{=}\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ be two polynomial-time constructible probability ensembles, $X \stackrel{c}{=} Y, F=$ $\left(f_{k}\right)_{k \in \mathbb{N}}, f_{k}:\{0,1\}^{*} \rightarrow\{0,1\}^{*}$ is a polynomial-time computable function, then

$$
F(\vec{X}) \stackrel{c}{=} F(\overrightarrow{X Y})
$$

where the probability ensemble $\vec{X}, \overrightarrow{X Y}$ respectively is defined in Lemma 18 and Proposition 23. $F(\vec{X})$ and $F(\overrightarrow{X Y})$ are defined in similar way as Proposition 25. All variables $F\left(\vec{X}\left(1^{k}, a\right)\right)\langle i\rangle$ and $F\left(\overrightarrow{X Y}\left(1^{k}, a\right)\right)\langle i\rangle$ are independent.
Proof. Following Proposition 24, we know $\vec{X} \stackrel{c}{=} \overrightarrow{X Y}$. As in the proof of Proposition 25, we can prove that $\vec{X}$ and $\overrightarrow{X Y}$ are polynomial-time constructible. Following proposition 25, we have $F(\vec{X}) \stackrel{c}{=} F(\overrightarrow{X Y})$.

Theorem 27. Let $\mathcal{H}=(P G, I S, V F, H G, H a s h, p H a s h)$ be a hash family. Let $n \stackrel{\text { def }}{=} h+t$. For each $i \in[n], H S M^{i} \stackrel{\text { def }}{=}\left\{H S M^{i}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}, H S M^{i}\left(1^{k}\right) \stackrel{\text { def }}{=}$ $\left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, H S M_{1}\left(1^{k}\right)\langle i+1\rangle\right)$, where $H S M_{1}\left(1^{k}\right)$ is defined in Definition 4. If $\mathcal{H}$ meets the following three conditions,

1. All variables $H S M_{1}\left(1^{k}\right)\langle i+1\rangle$ are independent, where $i \in[n]$.
2. $H S M^{1}=\ldots=H S M^{h}, H S M^{h+1}=\ldots=H S M^{n}$.
3. $H S M^{1} \stackrel{c}{=} H S M^{h+1}$.
then $\mathcal{H}$ has property hard subset membership.

Proof. First, we prove $\mathcal{H}$ has property hard subset membership 3 a. Let $\pi \in \Pi, X \stackrel{\text { def }}{=} H S M^{1}, Y \stackrel{\text { def }}{=} H S M^{h+1}, \Phi=(\pi)_{k \in \mathbb{N}}$, poly $(.) \stackrel{\text { def }}{=} h$, $\operatorname{poly}(.) \stackrel{\text { def }}{=} n$. Following Proposition 23, we know

$$
\overrightarrow{X Y} \stackrel{c}{=} \Phi(\overrightarrow{\widetilde{X Y}})
$$

That is

$$
\begin{aligned}
& \left(\left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, H S M_{1}\left(1^{k}\right)\langle 2\rangle\right), \ldots\right. \\
& \left.\left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, H S M_{1}\left(1^{k}\right)\langle n+1\rangle\right)\right) \stackrel{c}{=} \\
& \left(H S M_{2}\left(1^{k}\right)\langle 1\rangle, H S M_{2}\left(1^{k}\right)\langle 2\rangle\right), \ldots \\
& \left.\quad\left(H S M_{2}\left(1^{k}\right)\langle 1\rangle, H S M_{2}\left(1^{k}\right)\langle n+1\rangle\right)\right)
\end{aligned}
$$

where $\operatorname{HSM}_{1}\left(1^{k}\right), \operatorname{HSM}_{2}\left(1^{k}\right)$ are taken from Definition 4 . Note that $H S M_{1}\left(1^{k}\right)\langle 1\rangle=$ $H S M_{2}\left(1^{k}\right)\langle 1\rangle$, so

$$
\begin{aligned}
& \left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, H S M_{1}\left(1^{k}\right)\langle 2\rangle, \ldots, H S M_{1}\left(1^{k}\right)\langle n+1\rangle\right) \stackrel{c}{=} \\
& \quad\left(H S M_{2}\left(1^{k}\right)\langle 1\rangle, H S M_{2}\left(1^{k}\right)\langle 2\rangle, \ldots, H S M_{2}\left(1^{k}\right)\langle n+1\rangle\right)
\end{aligned}
$$

i.e.

$$
H S M_{1} \stackrel{c}{=} H S M_{2}
$$

Second, we prove $\mathcal{H}$ has property hard subset membership 3 b , Let $\pi^{\prime} \in \Pi$, $F=\left(\pi^{\prime}\right)_{k \in \mathbb{N}}$. Following Proposition 26, we have

$$
F(\vec{X}) \stackrel{c}{=} F(\overrightarrow{X Y})
$$

That is

$$
\begin{aligned}
& \left(\left(\operatorname{HSM}_{3}\left(1^{k}\right)\langle 1\rangle,\right.\right. \\
& \left.\quad H S M_{3}\left(1^{k}\right)\langle 2\rangle\right), \ldots \\
& \left.\qquad\left(H S M_{3}\left(1^{k}\right)\langle 1\rangle, \operatorname{HSM}_{3}\left(1^{k}\right)\langle n+1\rangle\right)\right) \stackrel{c}{=} \\
& \pi^{\prime}\left(\left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, H S M_{1}\left(1^{k}\right)\langle 2\rangle\right), \ldots\right. \\
& \left.\quad\left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, \operatorname{HSM}_{1}\left(1^{k}\right)\langle n+1\rangle\right)\right)
\end{aligned}
$$

where $\operatorname{HSM}_{3}\left(1^{k}\right)$ is taken from Definition 4. Since $\operatorname{HSM}_{3}\left(1^{k}\right)\langle 1\rangle=\operatorname{HSM}_{1}\left(1^{k}\right)\langle 1\rangle$, so

$$
\begin{aligned}
& \left(H S M_{3}\left(1^{k}\right)\langle 1\rangle, H S M_{3}\left(1^{k}\right)\langle 2\rangle, \ldots,\right. \\
& \left.H S M_{3}\left(1^{k}\right)\langle n+1\rangle\right) \stackrel{c}{=} \\
& \left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, \pi^{\prime}\left(H S M_{1}\left(1^{k}\right)\langle 2\rangle, \ldots,\right.\right. \\
& \left.\left.\quad H S M_{1}\left(1^{k}\right)\langle n+1\rangle\right)\right)
\end{aligned}
$$

We further have

$$
\begin{align*}
& \left(H S M_{3}\left(1^{k}\right)\langle 1\rangle, \pi^{\prime-1}\left(H S M_{3}\left(1^{k}\right)\langle 2\rangle, \ldots,\right.\right. \\
& \left.\left.\quad H S M_{3}\left(1^{k}\right)\langle n+1\rangle\right)\right) \\
& \stackrel{c}{=}\left(H S M_{1}\left(1^{k}\right)\langle 1\rangle, \operatorname{HSM}_{1}\left(1^{k}\right)\langle 2\rangle, \ldots,\right. \\
& \left.\quad H S M_{1}\left(1^{k}\right)\langle n+1\rangle\right) \tag{20}
\end{align*}
$$

Because $H S M_{3}\left(1^{k}\right)\langle 2\rangle=\ldots=\operatorname{HSM}_{3}\left(1^{k}\right)\langle 1+n\rangle$, we have

$$
\begin{align*}
& \left(H S M_{3}\left(1^{k}\right)\langle 1\rangle, \pi^{\prime-1}\left(H S M_{3}\left(1^{k}\right)\langle 2\rangle, \ldots,\right.\right. \\
& \left.\left.\quad H S M_{3}\left(1^{k}\right)\langle n+1\rangle\right)\right) \\
& \equiv\left(\operatorname{HSM}_{3}\left(1^{k}\right)\langle 1\rangle, \operatorname{HSM}_{3}\left(1^{k}\right)\langle 2\rangle, \ldots,\right. \\
& \left.\quad H S M_{3}\left(1^{k}\right)\langle n+1\rangle\right) \tag{21}
\end{align*}
$$

Combining equation (20) (21), we have

$$
H S M_{1} \stackrel{c}{=} H S M_{3}
$$

Remember that we have proven $H S M_{1} \stackrel{c}{=} H S M_{2}$, so we have

$$
H S M_{2} \stackrel{c}{=} H S M_{3}
$$

Loosely speaking, Theorem 27 shows that, given a hash family $\mathcal{H}$, if random variables $I S\left(1^{k}, \Lambda\right)\langle 1\rangle, \ldots, I S\left(1^{k}, \Lambda\right)\langle n\rangle$ are independent, $I S\left(1^{k}, \Lambda\right)\langle 1\rangle, \ldots, I S\left(1^{k}, \Lambda\right)\langle h\rangle$ sample $\dot{x}$ from $L_{\dot{R}_{\Lambda}}$ in the same way $, I S\left(1^{k}, \Lambda\right)\langle h+1\rangle, \ldots, I S\left(1^{k}, \Lambda\right)\langle n\rangle$ sample $\ddot{x}$ from $L_{\ddot{R}_{\Lambda}}$ in the same way, $L_{\dot{R}_{\Lambda}}$ and $L_{\ddot{R}_{\Lambda}}$ are computationally indistinguishable, then $\mathcal{H}$ has hard subset membership.

### 6.2 A Construction Under Lattice

### 6.2.1 Background

Learning with errors (LWE) is an average-case problem. [37] shows that its hardness is implied by the worst-case hardness of standard lattice problem for quantum algorithms.

In lattice, the modulo operation is defined as $x \bmod y \stackrel{\text { def }}{=} x-\llcorner x / y\lrcorner y$. Then we know $x \bmod 1 \stackrel{\text { def }}{=} x-\llcorner x\lrcorner$. Let $\beta$ be an arbitrary positive real number. Let $\Psi_{\beta}$ be a probability density function whose distribution is over $[0,1)$
and obtained by sampling from a normal variable with mean 0 and standard deviation $\beta / \sqrt{2 \pi}$ and reducing the result modulo 1 , more specifically

$$
\begin{aligned}
\Psi_{\beta} & :[0,1) \rightarrow R^{+} \\
\Psi_{\beta}(r) & \stackrel{\text { def }}{=} \sum_{k=-\infty}^{\infty} \frac{1}{\beta} \exp \left(-\pi\left(\frac{r-k}{\beta}\right)^{2}\right)
\end{aligned}
$$

Given an arbitrary integer $q \geq 2$, an arbitrary probability destiny function $\phi:[0,1) \rightarrow R^{+}$, the discretization of $\phi$ over $Z_{q}$ is defined as

$$
\begin{gathered}
\bar{\phi}: Z_{q} \rightarrow R^{+} \\
\bar{\phi}(i) \stackrel{\text { def }}{=} \int_{(i-1 / 2) / q}^{(i+1 / 2) / q} \phi(x) d x
\end{gathered}
$$

$L W E$ can be formulated as follows.
Definition 28 (Learning With Errors). Learning with errors problem (LWE $E_{q, \chi}$ ) is how to construct an efficient algorithm that receiving $q, g, m, \chi,\left(\vec{a}_{i}, b_{i}\right)_{i \in[m]}$, outputs $\vec{s}$ with nonnegligible probability. The input and the output is specified in the following way.
$q \leftarrow q\left(1^{k}\right), g \leftarrow g\left(1^{k}\right), m \leftarrow \operatorname{poly}\left(1^{k}\right), \chi \leftarrow \chi\left(1^{k}\right), \vec{s} \in_{U}\left(Z_{q}\right)^{k}$. For each $i \in[m], \vec{a}_{i} \in_{U}\left(Z_{q}\right)^{k}, e_{i} \in_{\chi} Z_{q}, b_{i} \leftarrow \vec{s}^{T} \cdot \vec{a}_{i}+e_{i} \bmod q$.
where $q, g$ are positive integers, $\chi: Z_{q} \rightarrow R^{+}$is a probability density function.

With respect to the hardness of $L W E, 37$ proves that setting appropriate parameters, we can reduce two worst-case standard lattice problems to $L W E$, which means $L W E$ is a very hard problem.

Lemma 29 ( 37 ). Setting security parameter $k$ to be a value such that $q$ is a prime, $\beta \leftarrow \beta\left(1^{k}\right)$, and $\beta \cdot q>2 \sqrt{k}$. Then the lattice problems SIVP and GapSVP can be reduced to $L W E_{q, \bar{\Psi}_{\beta}}$. More specifically, if there exists an efficient (possibly quantum) algorithm that solves $L W E_{q, \bar{\Psi}_{\beta}}$, then there exists an efficient quantum algorithm solving the following worst-case lattice problems in the $l_{2}$ norm.

- SIVP: In any lattice $\Lambda$ of dimension $k$, find a set of $k$ linearly independent lattice vectors of length within at most $\tilde{O}(k / \beta)$ of optimal.
- GapSVP: In any lattice $\Lambda$ of dimension m, approximate the length of a shortest nonzero lattice vector to within a $\tilde{O}(k / \beta)$ factor.

We emphasize the fact that the reduction of Lemma 29 is quantum, which implies that any algorithm breaking any cryptographic schemes which only based on $L W E$ is an algorithm solving at least one of the problems SIVP and GapSVP.

How to precisely set the parameters as values to gain a concrete $L W E$, which is as hard as required in Lemma 29 is beyond the scope of this paper. To see such examples and more details, we recommend [37] and [35].

The instantiation of $S P W H_{h, t}$, which we will present soon, needs to use a public key cryptosystem based on $L W E$. [37] and [35] respectively presents such an cryptosystem. Considering the cost, we choose the one presented by the latter and slightly tailor it to our need. The LWE-based cryptosystem with message space $Z_{p}$ is defined as follow, where $p \geq 2$ is polynomial in $k$.

- $\operatorname{Setup}\left(1^{k}, p\right)$ : Generates the public parameters as follows. $q \in_{U}\{q \mid q \in$ $\mathbb{P}, q$ is polynomial in $k, q>p\}, m \leftarrow \operatorname{pol} y\left(1^{k}\right), \chi \leftarrow \chi\left(1^{k}\right)$ and $\chi$ is a probability density function over $Z_{q}$, para $\leftarrow(q, p, m, \chi)$, finally outputs para.
- KeyGen(1 ${ }^{k}$,para): $A \in_{U}\left(Z_{q}\right)^{m \times k}, \vec{s} \in_{U}\left(Z_{q}\right)^{k}, \vec{e} \in_{\chi}\left(Z_{q}\right)^{m}, \vec{p} \leftarrow A \vec{s}+\vec{e}$ $\bmod q, p u b k \leftarrow(A, \vec{p}), s k \leftarrow \vec{s}$, finally outputs a public-secret key pair ( $p u b k, s k$ ).
- Enc(.), $\operatorname{Dec}($.$) : Since \operatorname{Enc}(),. \operatorname{Dec}($.$) are immaterial to understand this$ paper, we omit their detailed procedure here.


### 6.2.2 Detailed Construction

We now present our construction of a $S P W H_{h, t}$ based on $L W E$.

- $P G\left(1^{k}\right):$ para $\leftarrow \operatorname{Setup}\left(1^{k}, p\right),(q, p, m, \chi) \leftarrow \operatorname{para}, A \in_{U}\left(Z_{q}\right)^{m \times k}$, $\Lambda \leftarrow(p, q, m, A, \chi)$, finally outputs $\Lambda$.
- IS $\left(1^{k}, \Lambda\right):(p, q, m, A, \chi) \leftarrow \Lambda, \forall i \in[n] \vec{s}_{i} \in_{U}\left(Z_{q}\right)^{k}, \forall i \in[n] \vec{e}_{i} \in_{\chi}$ $\left(Z_{q}\right)^{m}, \forall i \in[h] \dot{x}_{i} \leftarrow A \vec{s}_{i}+\vec{e}_{i} \bmod q, \forall i \in[h] \dot{w}_{i} \leftarrow \vec{s}_{i}, \forall i \in[n]-[h]$ $\ddot{x}_{i} \leftarrow A \vec{s}_{i}+\vec{e}_{i}+(1,1, \ldots, 1)^{T} \bmod q, \forall i \in[n]-[h] \ddot{w}_{i} \leftarrow\left(\vec{s}_{i}, \vec{e}_{i}\right)$, finally outputs $\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots\right.$, $\left.\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)$.
- VF $\left(1^{k}, \Lambda, x, w\right):(p, q, m, A, \chi) \leftarrow \Lambda,\left(\vec{s}_{i}, \vec{e}_{i}\right) \leftarrow w$, if $x=A \vec{s}_{i}+\vec{e}_{i}+$ $(1,1, \ldots, 1)^{T} \bmod q$ holds, then outputs 1 ; otherwise outputs 0 .
- $K G\left(1^{k}, \Lambda, x\right):(p, q, m, A, \chi) \leftarrow \Lambda, a \in_{U} Z_{p}, \vec{s} \in_{U}\left(Z_{q}\right)^{k}, \vec{p} \leftarrow A \vec{s}+x$ $\bmod q, \alpha \leftarrow E n c_{A, \vec{p}}(a), h k \leftarrow a, p k \leftarrow(\vec{s}, \alpha)$, finally outputs $(h k, p k)$.
- $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right):(p, q, m, A, \chi) \leftarrow \Lambda, a \leftarrow h k$, finally outputs $a$.
- $p H a s h\left(1^{k}, \Lambda, x, p k, w\right):(k, m, p, q, \chi, A) \leftarrow \Lambda,(\vec{s}, \alpha) \leftarrow p k, \vec{u} \leftarrow \vec{s}+w$, $a \leftarrow D e c_{\vec{u}}(\alpha)$, finally outputs $a$.

We remark that the choice of $(1,1, \ldots, 1)^{T}$ is arbitrary. It is used to separate $\dot{R}$ from $\ddot{R}$. From the proof of the following proposition, we can see that any constant vector $\vec{c} \in\left(Z_{p}\right)^{m}-\{(0,0, \ldots, 0)\}$ is good too.
Proposition 30. Assume LWE is a hard problem, then VF computes the function $\zeta$ defined in Definition 4.
Proof. It is easy to see that in case $(x, w) \in \ddot{R}_{\Lambda}, V F$ correctly computes $\zeta$. It remains to prove that in case $(x, w) \in \dot{R}_{\Lambda}, V F$ correctly computes $\zeta$. Assume that $V F$ outputs 1 in the latter case. Then there exists an efficient adversary such that on receiving $\left(1^{k},(k, m, p, q, \chi, A), A \vec{s}_{i}+\vec{e}_{i}, \vec{s}\right)$ outputs $\left(\vec{s}_{i}, \vec{e}_{i}\right)$, where $A \vec{s}_{i}+\vec{e}_{i} \bmod q=A \vec{s}_{i}^{\prime}+\vec{e}_{i}+(1,1, \ldots, 1)^{T} \bmod q$. That is,

$$
A \vec{s}_{i}+\vec{e}_{i}-(1,1, \ldots, 1)^{T} \quad \bmod q=A \vec{s}_{i}^{\prime}+\vec{e}_{i}^{\prime} \quad \bmod q
$$

which implies that the adversary is an efficient algorithm breaking $L W E$.
Proposition 31. Assume LWE is a hard problem, the hash system holds the property projection.

Proof. Let $\dot{x}_{i} \in \operatorname{Range}\left(I S\left(1^{k}, \Lambda\right)\right)$. Looking at $I S\left(1^{k}, \Lambda\right), \dot{x}_{i}$ in fact is a public key whose corresponding secret key is $\vec{s}_{i}$. The ciphertext $\alpha$ in $K G\left(1^{k}, \Lambda, x\right)$ is encrypted using the public key whose corresponding secret key is $\overrightarrow{s_{i}}+\vec{s}$. The value of $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right)$ is the plaintext of $\alpha$. Using $\vec{s}_{i}+\vec{s}$ as a secret key, $p H a s h\left(1^{k}, \Lambda, x, p k, w\right)$ correctly outputs $\alpha$ 's plaintext. This means that for any ( $\dot{x}, \dot{w}, \Lambda$ ) generated by the hash system, it holds that

$$
\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right)=p \operatorname{Hash}\left(1^{k}, \Lambda, x, p k, w\right)
$$

Proposition 32. The hash system holds the property smoothness.
Proof. We are to use Theorem 20 to prove this. It is easy to see that this $S P W H_{h, t}$ meets the first two requirements, so it remains to prove that $S P W H_{h, t}$ also meets the last requirement, i.e. $S m_{1}^{h+1} \stackrel{c}{=} S m_{2}^{h+1}$. For this case, $S m_{1}^{h+1}, S m_{2}^{h+1}$ are

- $S m_{1}^{h+1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right),(p, q, m, k, A, \chi) \leftarrow \Lambda, \vec{s}_{h+1} \in_{U}\left(Z_{q}\right)^{k}, \vec{e}_{h+1} \in_{\chi}$ $\left(Z_{q}\right)^{m}, \ddot{x}_{h+1} \leftarrow A \vec{s}_{h+1}+\vec{e}_{h+1}+(1,1, \ldots, 1)^{T} \bmod q, a \in_{U} Z_{p}, \vec{s} \in_{U}$ $\left(Z_{q}\right)^{k}, \vec{p} \leftarrow A \vec{s}+\ddot{x}_{h+1} \bmod q, \alpha \leftarrow E n c_{A, \vec{p}}(a), h k \leftarrow a, p k \leftarrow(\vec{s}, \alpha)$. Finally outputs $\left(\Lambda, \ddot{x}_{h+1}, p k, a\right)$.
- $S m_{2}^{h+1}\left(1^{k}\right)$ : Outputs $\left(\Lambda, \ddot{x}_{h+1}, p k, y\right)$, where $\left(\Lambda, \ddot{x}_{h+1}, p k\right)$ is generated in the same way as $S m_{1}^{h+1}\left(1^{k}\right)$ and $y \in_{U} Z_{p}$.
Obviously, $S m_{1}^{h+1}\left(1^{k}\right)$ and $S m_{2}^{h+1}\left(1^{k}\right)$ are identically distributed, which implies that $S m_{1}^{h+1} \stackrel{c}{=} S m_{2}^{h+1}$.

Proposition 33. Assume LWE is a hard problem, then hash system holds the property hard subset membership.

Proof. We are to use Theorem 27 to prove this proposition holds. It is easy to see that this $S P W H_{h, t}$ meets the first two requirements, so it remains to prove that $S P W H_{h, t}$ also holds the last requirement, i.e. $H S M^{1} \stackrel{c}{=} H S M^{h+1}$. For this case, $H S M^{1}, H S M^{h+1}$ are where

- $H S M^{1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right),(p, q, m, k, A, \chi) \leftarrow \Lambda, \vec{s}_{1} \in_{U}\left(Z_{q}\right)^{k}, \vec{e}_{1} \in_{\chi}$ $\left(Z_{q}\right)^{m}, \dot{x}_{1} \leftarrow A \vec{s}_{1}+\vec{e}_{1} \bmod q$. Finally outputs $\left(\Lambda, \dot{x}_{1}\right)$.
- $H S M^{h+1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right),(p, q, m, k, A, \chi) \leftarrow \Lambda,(p, q, m, k, A, \chi) \leftarrow$ $\Lambda, \vec{s}_{h+1} \in_{U}\left(Z_{q}\right)^{k}, \vec{e}_{h+1} \in_{\chi}\left(Z_{q}\right)^{m}, \ddot{x}_{h+1} \leftarrow A \vec{s}_{h+1}+\vec{e}_{h+1}+(1,1, \ldots, 1)^{T}$ $\bmod q$. Finally outputs $\left(\Lambda, \ddot{x}_{h+1}\right)$.

Obviously, $H S M^{1}$ and $H S M^{h+1}$ are identically distributed, which implies that $H S M^{1} \stackrel{c}{=} H S M^{h+1}$.

Combining propositions above and Lemma 29, we have the following theorem.

Theorem 34. If SIVP or GapSVP is a hard problem, then the hash system is a $S P W H_{h, t}$.

### 6.2.3 A Concrete Protocol For $O T_{h}^{n}$ With Security Against Quantum Algorithms

The security proof of the framework guarantees that, any algorithm breaking the framework is an algorithm breaking at least one of cryptographic tools used in the framework. Therefore, to gain an instantiation of our framework with security against quantum algorithms, it suffices to adopt instantiations of commitment schemes and $S P W H_{h, t}$ in our framework which are secure against quantum algorithms.
[37] shows that the problems SIVP and GapSVP are hard for quantum algorithms at present. Combining Theorem 34, our LWE-based $S P W H_{h, t}$ is secure against quantum algorithms. It remains to find a $P H C$ and a $P B C$ with such security level. [3] presents a commitment scheme, which is provably unbreakable by both parties with unlimited computation power and algorithmic sophistication. So we have,

Theorem 35. Assuming that one of the problems SIVP and GapSVP is hard for quantum algorithms, instantiating the $O T_{h}^{n}$ framework with our LWE-based SPW $H_{h, t}$ and the commitment scheme presented by [3], the resulting concrete protocol for $O T_{h}^{n}$ is secure against quantum algorithms.
[37] points out that the problem of LWE and the problem of decoding random linear code (DRLC) are essentially the same. This implies that instantiating the commitment scheme with the PHC and $P B C$ based on DRLC, Theorem 35 also holds. What is more, [15] shows that, first, assuming that DRLC is hard, there exists a one-way function; second, assuming the existence of a one-way function, then there exists perfectly binding scheme and perfectly hiding scheme. Therefore, we have

Theorem 36. Assuming that one of the problems SIVP and GapSVP is hard for quantum algorithms, then there exists a protocol for $O T_{h}^{n}$ with security against quantum algorithms.

### 6.3 A Construction Under The Decisional Diffie-Hellman Assumption

### 6.3.1 Background

Let $\operatorname{Gen}\left(1^{k}\right)$ be an algorithm such that randomly chooses a cyclic group and outputs the group's description $G=<g, q, *>$, where $g, q, *$ respectively is the generator, the order, the operation of the group.

The DDH problem is how to construct an algorithm to distinguish the two probability ensembles $D D H_{1} \stackrel{\text { def }}{=}\left\{D D H_{1}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ and $D D H_{2} \stackrel{\text { def }}{=}\left\{D D H_{2}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ which are formulate as follows.

- $D D H_{1}\left(1^{k}\right):<g, q, *>\leftarrow G e n\left(1^{k}\right), a \in_{U} Z_{q}, b \in_{U} Z_{q}, c \leftarrow a b$, finally outputs $\left(<g, q, *>, g^{a}, g^{b}, g^{c}\right)$.
- $D D H_{2}\left(1^{k}\right)$ : Basically operates in the same way as $D D H_{1}\left(1^{k}\right)$ except that $c \in_{U} Z_{q}$.

At present, there is no efficient algorithm solving the problem. Therefore, it is assumed that $D D H_{1} \stackrel{c}{=} D D H_{2}$.

### 6.3.2 Detailed Construction

We describe our construction of hash system based on DDH as follows.

- $P G\left(1^{k}\right): \Lambda \leftarrow G e n\left(1^{k}\right)$, finally outputs $\Lambda$.
- $I S\left(1^{k}, \Lambda\right):(g, q, *) \leftarrow \Lambda, a_{i} \in_{U} Z_{q} \forall \in[n], b_{i} \in_{U} Z_{q} \forall i \in[n], c_{i} \leftarrow a_{i} b_{i}$ $\forall i \in[h], \dot{x}_{i} \leftarrow\left(g^{a_{i}}, g^{b_{i}}, g^{c_{i}}\right) \forall i \in[h], \dot{w}_{i} \leftarrow\left(a_{i}, b_{i}\right) \forall i \in[h], c_{i} \in_{U} Z_{q}$ $\forall i \in[n]-[h], \ddot{x}_{i} \leftarrow\left(g^{a_{i}}, g^{b_{i}}, g^{c_{i}}\right) \forall i \in[n]-[h], \ddot{w}_{i} \leftarrow\left(a_{i}, b_{i}\right) \forall i \in[n]-[h]$, finally outputs $\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots,\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)$.
- $V F\left(1^{k}, \Lambda, x, w\right):(g, q, *) \leftarrow \Lambda,(\alpha, \beta, \gamma) \leftarrow x,(a, b) \leftarrow w$, if $(\alpha, \beta, \gamma)=$ $\left(g^{a}, g^{b}, g^{a b}\right)$ holds, then outputs 0 ; if $(\alpha, \beta)=\left(g^{a}, g^{b}\right)$ and $\gamma \neq g^{a b}$ holds, then outputs 1 .
- $K G\left(1^{k}, \Lambda, x\right):(g, q, *) \leftarrow \Lambda,(\alpha, \beta, \gamma) \leftarrow x, u \in_{U} Z_{q}, v \in_{U} Z_{q}, p k \leftarrow$ $\alpha^{u} g^{v}, h k \leftarrow \gamma^{u} \beta^{v}$, finally outputs ( $h k, p k$ ).
- $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right): y \leftarrow h k$, outputs $y$.
- $p \operatorname{Hash}\left(1^{k}, \Lambda, x, p k, w\right):(a, b) \leftarrow w, y \leftarrow p k^{b}$, finally outputs $y$.

Theorem 37. Assuming $D D H$ is a hard problem, the hash system is a $S P W H_{h, t}$.

The proof of this theorem can be done in the same way as that of Theorem 34. So we don't iterate here.

### 6.3.3 A Concrete Protocol For $O T_{h}^{n}$ Based On DDH Only

To gain a concrete protocol for $O T_{h}^{n}$ based only on DDH , it remains to instantiate $P H C$ and $P B C$ with the ones builded on DDH . The commitment scheme [34] presents is an concrete PHC we need. The encryption scheme [10] presents is directly based on the problem of discrete log. Since the task of solving the problem DDH can be reduced to that of solving the problem discrete log, the encryption scheme is based on DDH essentially. What is more, this encryption scheme can be used as an concrete $P B C$. Therefore, using those two commitment schemes and our DDH-based $S P W H_{h, t}$, we gain a concrete protocol for $O T_{h}^{n}$ based only on DDH . To reach the best efficiency, we should use the DDH of the group which is on elliptic curves.

### 6.4 A Construction Under The Decisional N-th Residuosity Assumption

### 6.4.1 Background

Let $\operatorname{Gen}\left(1^{k}\right)$ be an algorithm that operates as follows.

- Gen $\left(1^{k}\right):(p, q) \in_{U}\{(p, q)|(p, q) \in(\mathbb{P}, \mathbb{P}), p, q>2,|p|=|q|=k, \operatorname{gcd}(p q,(p-$ 1) $(q-1))=1\}, N \leftarrow p q$, finally outputs $N$.

The problem decisional N-th residuosity (DNR), presented in [33], is how to construct an algorithm to distinguish two probability ensembles $D N R_{1} \stackrel{\text { def }}{=}$ $\left\{D N R_{1}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ and $D N R_{2} \stackrel{\text { def }}{=}\left\{D N R_{2}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ which are formulate as follows.

- $D N R_{1}\left(1^{k}\right): N \leftarrow G e n\left(1^{k}\right), a \in_{U} Z_{N^{2}}^{*}, b \leftarrow a^{N} \bmod N^{2}$, finally outputs $(N, b)$.
- $D N R_{2}\left(1^{k}\right): N \leftarrow \operatorname{Gen}\left(1^{k}\right), b \in_{U} Z_{N^{2}}^{*}$, finally outputs $(N, b)$.

The DNR assumption is that there is no efficient algorithm solving the problem. In other words, it is assumed that $D N R_{1} \stackrel{c}{=} D N R_{2}$.

The hash system we will construct is an instantiation of $\epsilon-U P W H_{h, t}$. We will build it on a DNR-based instantiation, presented by [22, of $\varepsilon$-VUPH. $\epsilon$ - UPW $H_{1,1}$ is different from $\varepsilon$-VUPH in a similar way that $S P W H_{1,1}$ is different from $V S P H$. Thus, the definition of $\varepsilon$-VUPH is easy to be deduced, and we omit them here. Please see $[22]$ for its detailed definition.

The instantiation of $\varepsilon$-VUPH [22] is stated as follows, where $\varepsilon<1$.

- $P G\left(1^{k}\right): N \leftarrow G e n\left(1^{k}\right), a \in_{U} Z_{N^{2}}^{*}, T \leftarrow N^{\ulcorner 2 \log N\urcorner}, g \leftarrow a^{N \cdot T} \bmod N^{2}$, $\Lambda \leftarrow(N, g)$, finally outputs $\Lambda$.
- $I S\left(1^{k}, \Lambda\right):(N, g) \leftarrow \Lambda, r, v \in_{U} Z_{N}^{*}, w \leftarrow r, \dot{x} \leftarrow g^{r} \bmod N^{2}, \ddot{x} \leftarrow$ $\dot{x}(1+v N) \bmod N^{2}$, finally outputs $(w, \dot{x}, \ddot{x})$.
- $I T\left(1^{k}, \Lambda, \dot{x}, \ddot{x}\right):(N, g) \leftarrow \Lambda$. Checks that $N>2^{2 k}, g, \dot{x} \in Z_{N^{2}}^{*}$. $d \leftarrow \ddot{x} / \dot{x} \bmod N^{2}$ and checks $N \mid(d-1) . v \leftarrow(d-1) / N$ and checks $\operatorname{gcd}(v, N)=1$. Outputs 1 if all the test pass and 0 otherwise.
- $K G\left(1^{k}, \Lambda\right):(N, g) \leftarrow \Lambda, h k \in_{U} Z_{N^{2}}, p k \leftarrow g^{h k} \bmod N^{2}$, finally outputs ( $h k, p k$ ).
- $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right):(N, g) \leftarrow \Lambda, y \leftarrow x^{h k} \bmod N^{2}$, finally outputs $y$.
- $p H a s h\left(1^{k}, \Lambda, x, p k, w\right):(N, g) \leftarrow \Lambda, y \leftarrow p k^{w} \bmod N^{2}$, finally outputs $y$.

The $I T$ holds a property called verifiability, which is described as follows.

1. For any $\Lambda \in \operatorname{Range}\left(P G\left(1^{k}\right)\right)$, any $(w, \dot{x}, \ddot{x}) \in \operatorname{Range}\left(I S\left(1^{k}, \Lambda\right)\right)$, it holds that $I T(\Lambda, \dot{x}, \ddot{x})=I T(\Lambda, \ddot{x}, \dot{x})=1$.
2. For any $\Lambda, \dot{x}, \ddot{x}$ such that $I T(\Lambda, \ddot{x}, \dot{x})=1$, it holds that the hash system is either $\varepsilon(|\Lambda|)$ - universal on $(\Lambda, \dot{x})$ or $\varepsilon(|\Lambda|)$ - universal on $(\Lambda, \ddot{x})$.

Note that in the hash system, $\dot{x}$ is projective and $\ddot{x}$ is universal. What is more, the properties $\varepsilon$-universality and projection are contradictory. Therefore, $[22]$ indirectly proves the following lemma.

Lemma 38. Let $\dot{L} \stackrel{\text { def }}{=}\left\{\dot{x} \mid(N, g) \leftarrow P G\left(1^{k}\right), r \in_{U} Z_{N}^{*}, w \leftarrow r, \dot{x} \leftarrow g^{w}\right.$ $\left.\bmod N^{2}\right\}$ and $\ddot{L} \stackrel{\text { def }}{=}\left\{\ddot{x} \mid(N, g) \leftarrow P G\left(1^{k}\right), r, v \in_{U} Z_{N}^{*}, w \leftarrow r, \ddot{x} \leftarrow g^{w}(1+v N)\right.$ $\left.\bmod N^{2}\right\}$. Then

$$
\dot{L} \cap \ddot{L}=\emptyset
$$

### 6.4.2 Detailed Construction

Recalling Theorem 22 , to gain a $S P W H_{h, t}$, what we need to do is to construct a $\epsilon-U P W H_{h, t}$ first, then transform it into a $S P W H_{h, t}$ using the algorithm guaranteed by the theorem. In this section, we construct an instantiation of $\epsilon-U P W H_{h, t}$ based on DNR.

We now present our construction of hash system under the DNR assumption as follows.

- $P G\left(1^{k}\right): N \leftarrow G e n\left(1^{k}\right), a \in_{U} Z_{N^{2}}^{*}, T \leftarrow N^{\ulcorner 2 \log N\urcorner}, g \leftarrow a^{N \cdot T} \bmod N^{2}$, $\Lambda \leftarrow(N, g)$, finally outputs $\Lambda$.
- $I S\left(1^{k}, \Lambda\right):(N, g) \leftarrow \Lambda, r_{i} \in_{U} Z_{N}^{*} \forall i \in[n], \dot{x}_{i} \leftarrow g^{r_{i}} \bmod N^{2} \forall i \in[h]$, $\dot{w}_{i} \leftarrow\left(r_{i}, 0\right) \forall i \in[n]-[h], v_{i} \in_{U} Z_{N}^{*} \forall i \in[n]-[h], \ddot{x}_{i} \leftarrow g^{r_{i}}\left(1+v_{i} N\right)$ $\bmod N^{2} \forall i \in[n]-[h], \ddot{w}_{i} \leftarrow\left(r_{i}, v_{i}\right) \forall i \in[n]-[h]$, finally outputs $\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots,\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)$.
- $V F\left(1^{k}, \Lambda, x, w\right):(N, g) \leftarrow \Lambda,(r, v) \leftarrow w$,

1. if $v=0 \bmod N$, operates as follows: checks that $N>2^{2 k}, g, x \in$ $Z_{N^{2}}^{*}, r \in Z_{N}^{*}, x=g^{r} \bmod N^{2}$. Outputs 0 if all the test pass.
2. if $v \neq 0 \bmod N$, operates as follows: checks that $N>2^{2 k}, g, x \in$ $Z_{N^{2}}^{*}, r \in Z_{N}^{*}, x=g^{r}(1+v n) \bmod N^{2}$. Outputs 1 if all the test pass.

- $K G\left(1^{k}, \Lambda, x\right):(N, g) \leftarrow \Lambda, h k \in_{U} Z_{N^{2}}, p k \leftarrow g^{h k} \bmod N^{2}$, finally outputs ( $h k, p k$ ).
- $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right):(N, g) \leftarrow \Lambda, y \leftarrow x^{h k} \bmod N^{2}$, finally outputs $y$.
- $p H a s h\left(1^{k}, \Lambda, x, p k, w\right):(N, g) \leftarrow \Lambda, y \leftarrow p k^{w} \bmod N^{2}$, finally outputs $y$.

We now prove that the scheme above is a $\epsilon-U P W H_{h, t}$, where $\epsilon<1$.

Proposition 39. Assume that $D N R$ is a hard problem, then VF computes the function $\zeta$ defined in Definition 4.

According to Lemma 38, it is easy to see this proposition holds.
Proposition 40. The hash system holds the property projection and $\varepsilon$-universality, where $\varepsilon<1$.

These properties holded by the hash system are directly inherited from the instantiation of $\epsilon-U P W H_{h, t}$.

Proposition 41. Assuming $D N R$ is a hard problem, the hash system holds the property hard subset membership.

Proof. We are to use Theorem 27 to prove this proposition holds. It is easy to see that this hash system meets the first two requirements, so it remains to prove that this hash system also holds the last requirement, i.e. $H S M^{1} \stackrel{c}{=} H S M^{h+1}$. For this case, $H S M^{1}, H S M^{h+1}$ are

- $H S M^{1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right),(N, g) \leftarrow \Lambda, r_{1} \in_{U} Z_{N}^{*}, \dot{x}_{1} \leftarrow g^{r_{1}} \bmod N^{2}$. Finally outputs $\left(\Lambda, \dot{x}_{1}\right)$.
- $H S M^{h+1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right),(N, g) \leftarrow \Lambda, r_{h+1}, v_{h+1} \in_{U} Z_{N}^{*}, \ddot{x}_{h+1} \leftarrow$ $g^{r_{h+1}}\left(1+v_{h+1} N\right) \bmod N^{2}$. Finally outputs $\left(\Lambda, \ddot{x}_{h+1}\right)$.

It is clear that $H S M^{1} \stackrel{c}{=} H S M^{h+1}$.
Combining propositions above, we have the following theorem.
Theorem 42. Assuming $D N R$ is a hard problem, the hash system is a $\epsilon$ $U P W H_{h, t}$, where $\epsilon<1$.

### 6.5 A Construction Under The Decisional Quadratic Residuosity Assumption

We reuse $\operatorname{Gen}\left(1^{k}\right)$ defined in section 6.4.1. The problem decisional quadratic residuosity ( DQR ) is how to construct an algorithm to distinguish the two probability ensembles $Q R_{1} \stackrel{\text { def }}{=}\left\{Q R_{1}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ and $Q R_{2} \stackrel{\text { def }}{=}\left\{Q R_{2}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ which are formulated as follows.

- $Q R_{1}\left(1^{k}\right): N \leftarrow G e n\left(1^{k}\right), x \in_{U} Z_{N}^{*}$, finally outputs $(N, x)$.
- $Q R_{2}\left(1^{k}\right): N \leftarrow \operatorname{Gen}\left(1^{k}\right), r \in_{U} Z_{N}^{*}, x \leftarrow r^{2} \bmod N$, finally outputs $(N, x)$.

The DQR assumption is that there is no efficient algorithm solving the problem. That is, it is assumed that $D Q R_{1} \stackrel{c}{=} D Q R_{2}$.

As in section 6.4, the hash system we aim to achieve is an instantiation of $\epsilon-U P W H_{h, t}$. We will build it on an instantiation of $\varepsilon$-VUPH presented by 22 which is constructed under DQR assumption.

Our hash system is described as follows.

- $P G\left(1^{k}\right):(p, q) \in_{U}(\mathbb{P}, \mathbb{P})$, where $|p|=|q|=k, p<q<2 p-1, p=q=3$ $\bmod 4, a \in_{U} Z_{N}^{*}, T \leftarrow 2^{\lceil\log N\urcorner}, g \leftarrow a^{2 \cdot T} \bmod N, \Lambda \leftarrow(N, g)$, finally outputs $\Lambda$.
- $I S\left(1^{k}, \Lambda\right):(N, g) \leftarrow \Lambda, r_{i} \in_{U} Z_{N} \forall i \in[n], \dot{x}_{i} \leftarrow g^{r_{i}} \bmod N \forall i \in[h]$, $\ddot{x}_{i} \leftarrow N-g^{r_{i}} \bmod N \forall i \in[n]-[h], \ddot{w}_{i} \leftarrow r_{i} \forall i \in[n]$, finally outputs $\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots,\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)$.
- $V F\left(1^{k}, \Lambda, x, w\right):(N, g) \leftarrow \Lambda, r \leftarrow w$; checks that $N>2^{2 k}, g, x \in Z_{N}^{*}$. Outputs 0, if $x=g^{r} \bmod N$ and all the test pass. Outputs 1, if $x=N-g^{r} \bmod N$ and all the test pass.
- $K G\left(1^{k}, \Lambda, x\right):(N, g) \leftarrow \Lambda, h k \in_{U} Z_{N}, p k \leftarrow g^{h k} \bmod N$, finally outputs ( $h k, p k$ ).
- $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right):(N, g) \leftarrow \Lambda, y \leftarrow x^{h k} \bmod N$, finally outputs $y$.
- $p H a s h\left(1^{k}, \Lambda, x, p k, w\right):(N, g) \leftarrow \Lambda, y \leftarrow p k^{w} \bmod N$, finally outputs $y$.

The fact that the hash system above is a $\epsilon-U P W H_{h, t}(\epsilon<1)$ can be proven in a similar way in which Theorem 42 is proven.

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