# Practical Frameworks For $h$-Out-Of- $n$ Oblivious Transfer With Security Against Covert and Malicious Adversaries 

Zeng Bing, Tang Xueming, Xu Peng, and Jing Jiandu


#### Abstract

We present two practical frameworks for $h$-out-of- $n$ oblivious transfer $\left(O T_{h}^{n}\right)$. The first one is secure against covert adversaries who are not always willing to cheat at any price. The security is proven under the ideal/real simulation paradigm (call such security fully simulatable security). The second one is secure against malicious adversaries who are always willing to cheat. It provides fully simulatable security and privacy respectively for the sender and the receiver (call such security one-sided simulatable security). The two frameworks can be implemented from the decisional Diffie-Hellman (DDH) assumption, the decisional $N$-th residuosity assumption, the decisional quadratic residuosity assumption and so on.

The DDH-based instantiation of our first framework costs the minimum communication rounds and the minimum computational overhead, compared with existing practical protocols for oblivious transfer with fully simulatable security against covert adversaries or malicious adversaries.

Though our second framework is not efficient, compared with existing practical protocols with one-sided simulatable security against malicious adversaries. However, it first provides a way to deal with general $O T_{h}^{n}$ on this security level. What is more, its DDH-based instantiation is more efficient than the existing practical protocols for oblivious transfer with fully simulatable security against malicious adversaries.


Index Terms—oblivious transfer (OT) protocols, secure two-party computation.

## 1 IntRODUCTION

### 1.1 Oblivious transfer

OBLIVIOUS transfer (OT), first introduced by [39] and later defined in another way with equivalent effect [14] by [15], is a fundamental primitive in cryptography and a concrete problem in the filed of secure multi-party computation. Considerable cryptographic protocols can

[^0]be built from it. Most remarkable, [23], [27], [29], [41] 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 $O T_{h}^{n}$.

It is known that, by using Goldreich's compiler [21], [23], we can gain a protocol for $O T_{h}^{n}$ with security against malicious adversaries (a malicious adversary act in any arbitrary malicious way to learn as much extra information as possible) from a protocol for $O T_{h}^{n}$ with security against semi-honest adversaries (a semihonest 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 him), which can be built from any collection of enhanced trapdoor permutations [21]. The security of the resulting protocol can be proven under the real/ideal simulation paradigm. The paradigm requires that for any adversary in the real world, there exists a corresponding adversary in the ideal world who can simulate him, where the ideal world holds desirable security level. That is, for any malicious adversary, he can not do more harm in the real world than in the ideal world, which implies that the proven protocol is secure. However, the mentioned resulting protocol for $O T_{h}^{n}$ is prohibitive expensive for practical use, because it is embedded with so many invocations of zero-knowledge proof for NP and invocations of commitment scheme. Thus, directly constructing the protocol based on specific intractable assumptions seems more feasible.

### 1.2 Our Contribution

In this paper, we aim to construct practical and efficient frameworks for $O T_{h}^{n}$ which can be instantiated from the decisional Diffie-Hellman (DDH) assumption, the decisional $N$-th residuosity (DNR) assumption, the decisional quadratic residuosity ( DQR ) assumption and so on. To this end, compared with the security mentioned above (for simplicity, we call this security standard security), the security we guarantee is relaxed to some extent. Specifically, we construct two frameworks respectively with the following two security level.

- Security against covert adversaries. This security, presented by [3], guarantees that in case a adversary commits an effective cheating, then the adversary can be caught with probability at least $\epsilon$, where $\epsilon$ is called the deterrence factor. By effective cheating, we refer to a cheating in which a adversary learning new knowledge is possible. The security definition follows the real/ideal simulation paradigm. Therefore, compared with the standard security, the only relaxed point in this security is that it does not require to rule out any possibility that the adversaries learns new knowledge in an execution of a protocol. In any setting that the malicious adversaries are not always willing to cheat at any price, this security makes sense.
In our first framework, in case a adversary is caught in committing an effective cheating, he will learn no new knowledge. What is more, the deterrence factor can be set to be very close to 1 , e.g., $\epsilon=1-7.25 \times$ $10^{-12}$.
- Security against malicious adversaries based on onesided simulation. This security, called the security based on half-simulation by [25] and also called the security based on half-simulation by [7], guarantees that, in case one party is corrupted, it provides privacy to the honest party; in case another party is corrupted, it provides standard security to the honest party. Specifically, in case one party is corrupted, the security only requires to prove that the adversary controlling the corrupted party can't distinguish his views obtained in the different interactions with the honest party who uses different privates inputs; in case another party is corrupted, the security proof requires as same as the standard security does. Therefore, compared with the standard security, the properties correctness and the independence of private inputs are not guaranteed any more in this security. However, this security indeed is useful in the computation in which only one party receives a output. For example, [16], [25], |28], [35] take this security definition to design efficient protocols. In these works, the party receiving a output is provided with the privacy, and the other party is provided with the full security. In our second framework, this is also the case.

To our best knowledge, our first framework is the first known framework for $O T_{h}^{n}$ with security against covert adversaries. Though a framework for $O T_{1}^{2}$ with security against malicious against malicious adversaries based on one-sided simulation is by [28], to our best knowledge, our second framework is the first known framework for general $O T_{h}^{n}$ with this security level.

Our two frameworks have the following features.
First, practical. On one side, the securities of our frameworks are obtained without restoring to a random oracle. Though the schemes presented by [26|, [37] are efficient, their security is proven under the Random Oracle Model. However, the Random Oracle Model seems risky. [10] shows that a scheme is secure in the Random Oracle Model does not necessarily imply that a particular implementation of it (in the real world) is secure, or even that this scheme does not have any "structural flaws". [10] also shows efficient implementing the random oracle is impossible. Later, [30] finds that the random oracle instantiations proposed by Bellare and Rogaway from 1993 and 1996, and the ones implicit in IEEE P1363 and PKCS standards are weaker than a random oracle. What is worse, [30] shows that how the defects of the random oracle instantiations deadly damage the securities of the cryptographic schemes presented in [5], [6]. Therefore, considering practical use, our frameworks are better.

On one side, the securities of our frameworks are obtained without restoring to a trusted common reference string (CRS). Though the framework for $O T_{1}^{2}$ presented by [38] and its adaptive instantiations presented by [18] hold higher level that is secure against malicious adversaries even under universal composition, they don't work without a trusted CRS. How to provide a trusted CRS before the protocol run still is a unsolved problem. The existing possible solutions, such as natural process suggested by [38], are only conjectures without formal proofs. What is worse, [9], [11| show that even given a authenticated communication channel, implementing a universal composable protocol providing useful trusted CRS in the presence of malicious adversaries is impossible. Therefore, considering practical use, our frameworks are better.

Second, efficient. Our first framework costs four communication rounds. Fixing the deterrence factor to be $\frac{1}{2}$, this framework costs $n$ public key encryption operations, $h$ public key decryption operations. Compared with the existing practical protocols for $O T$ with standard security or with security against covert adversaries, i.e. the protocols presented by [3], [7], [24], [31], [42], the DDH-based instantiation of our first framework is the most efficient one in the sense that it costs the minimum number of communication rounds and costs the minimum computational overhead. Please see Section 3.4 and Section 3.5 for the detailed comparisons.

Our second framework costs four communication rounds. To be securely used in practice, our framework, in the worst case, costs the sender $40 \cdot n$ public key encryption operations and costs the receiver $40 \cdot h$ public
key decryption operations. Compared with the existing practical protocols for $O T$ with the security based on one-sided simulation, i.e. the protocols presented by [2], [28], [33]-[36], our second framework is not efficient. However, it still makes sense since it does not require $k$ is far less than $n$ and is the first framework/protocol dealing with general $O T_{k}^{n}$ on this security level. Compared with the existing practical protocols for $O T$ with standard security, i.e. the protocols presented by [7], [24], [31], [42], the DDH-based instantiation of our second framework is the most efficient one in the sense mentioned above. Please see Section 4.3 and Section 4.4 for the details.

Third, abstract and modular. The framework is described using only one high-level cryptographic tool, i.e., a variant of smooth projective hash presented by [42] (for simplicity, refer to it as $S P H D H C_{t, h}$. This allows a simple and intuitive understanding of its security.

Fourth, generally realizable. [42] shows that $S P H D H C_{t, h}$ are realizable from a variety of known specific assumption, such as DDH, DNR, DQR. We remark that $S P H D H C_{t, h}$ also is possible to be realized from future assumptions. This makes our framework generally realizable. Considering the future progress in breaking a specific intractable problem, generally realizability is vital to make framework live long. If this considered case happen, replacing the instantiation based on the broken problem with that based on a unbroken problem suffices.

### 1.3 Construction Overview

Our first framework is described with high-level as follows.

1) Let $K, g$ be two predetermined positive integers. The receiver generates hash parameters and $K$ instance vectors, then sends them to the sender after disordering each vector.
2) The receiver chooses $g$ instance vectors at random to open.
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 $n$ message he holds) using the hash system, and sends the ciphertext together with some information 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 property hard subset membership of $S P H D H C_{t, h}$. This property guarantees that the receiver can securely encode his private input by reordering each unchosen instance vector. The sender's security is
implied by the cut-and-choose technique, which guarantees that the probability that the adversaries controlling a corrupted receiver learns new knowledge is small enough.

Our second framework is obtained by modifying the first framework by letting $K$ be a polynomial in security parameter and granting the sender to uniformly choose arbitrary many instance vectors to open.

In some sense, our frameworks are based on the framework presented in [28] for $O T_{1}^{2}$ with security based on one-sided simulation. However, the technique and tool we use are very different. In our frameworks, we use cut-and-choose to guarantee the sender's security rather than use the property verifiability of a tool. What is more, the tool we uses is different from the smooth projective hash [28] uses. These differences are keys to make our first framework fully simulatable. With the help of cut-and-choose, the property feasible cheating and distinguishability of $S P H D H C_{t, h}$ respectively enable the simulator to extract the adversary's real input in case the sender is corrupted and in case the receiver is corrupted. The differences also are keys to make our second framework able to deal with $O T_{h}^{n}$. Since $S P H D H C_{t, h}$ can directly manipulate $n$ instances and provide a novel way to guarantee the sender's security.

### 1.4 Related Work

The first step to construct practical and efficient protocol for $O T$ is independently made by [35] and [2]. The protocols are for $O T_{1}^{2}$, based on the DDH assumption, with security based on one-sided simulation. Later, using the tool smooth projective hash, [28] generalizes the ideas of the two previous works and presents a framework for $O T_{1}^{2}$. |28] shows the framework can be instantiated from not only the DDH assumption but also the DNR assumption and the DQR assumption.

There are some work aiming to build protocols for $O T_{h}^{n}$ from known protocols for $O T_{1}^{2}$ with security based on one-sided simulation. [33] shows how to implementation $O T_{h}^{n}$ using $\log n$ invocation of $O T_{1}^{2}$ with security based on one-sided simulation. A similar implementation for adaptive $O T_{h}^{n}$ can be seen in [34]. These mentioned works later are described in [36]. With the help of a random oracle, [26] shows how to extend $k$ oblivious transfers (for some security parameter $k$ ) into many more, without much additional effort. [37] proposes an efficient adaptive $h$-out- $n$ oblivious transfer (denoted by $O T_{h \times 1}^{n}$ in related literature) schemes under Random Oracle Model.

The protocols for oblivious transfer with standard security is first presented by [7]. This protocol is for $O T_{k \times 1}^{n}$ and based on $q$-Power Decisional Diffie-Hellman and $q$ Strong Diffie-Hellman assumptions which are not standard assumptions. Later, using a blind identity-based encryption, [24] first presents a protocol for $O T_{h}^{n}$ based on a standard assumption with standard security. [31] presents a DDH-based protocol for $O T_{1}^{2}$ with standard
security and with efficiency over the two mentioned works. Recently, [42] presents a framework for $O T_{h}^{n}$ with standard security, which also can be instantiated from all the assumptions used by [28], [31]. The DDH-based instantiation of this framework is the most efficient one of the protocols that works in the setting that there are no set-up assumption such as a trusted CRS available.

Under CRS model, [38] presents a framework for $O T_{1}^{2}$ with higher security level that the security is preserved even other arbitrary malicious protocols concurrently run with it. [38] shows that the framework be instantiated from the DDH, DQR and worst-case lattice assumption. We remark that if a trusted CRS is available, then the DDH-based instantiation of the framework is the most efficient protocol for $O T_{1}^{2}$. Later [18] strengthen the security of the instantiations based on $\overline{\mathrm{DDH}}, \mathrm{DQR}$ to against adaptive malicious adversaries.

Using homomorphic encryption, [3] first presents a protocol for $O T_{1}^{2}$ with security against covert adversaries.

### 1.5 Organization

In Section 2, we describe the notations used in this paper, the security definitions of $O T_{h}^{n}$ and a variant of smooth projective hash which is the basic tool we use. In Section 3, we construct the first framework for $O T_{h}^{n}$, prove its security and analyze its performance. In Section 4. we construct the second framework for $O T_{h}^{n}$, prove its security and analyze its performance. In Section 5 , we discuss the technique cut-and-choose in the frameworks.

## 2 Preliminaries

### 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 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 $\operatorname{poly}($.$) and any k$ 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 polynomialtime (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 $Y \quad \stackrel{\text { def }}{=}$ $\left\{Y\left(1^{k}, a\right)\right\}_{k \in \mathbb{N}, a \in\{0,1\}^{*}}$ 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

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

They are 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 $M_{r}($.$) we denote M^{\prime}$ s output generated at the end of an execution using randomness $r$.

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

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 Against Covert Adversaries

Security against covert adversaries is presented by [3]. The key observation of this security is that in many real-world setting, such as business, political settings and playing remote games, the assumption that the adversaries are always ready to do cheating at any price and under any circumstance, implicitly used in traditional model of secure multiparty computation [8], [21], is overly pessimistic and unnecessary. Therefore, using great amount of resource to preclude such covert adversaries from commenting cheating seems unnecessary. Instead, this security aims at catching the cheating of the covert adversaries with some probability, called deterrence factor and denoted $\varepsilon$, rather than aims at eliminating any successful cheating of covert adversaries. [3] presents three versions of covert security. In this paper, we deals with the strongest one, which implies the other two versions of covert security. For clarity and simplicity, we tailor it to the need of dealing with $O T_{h}^{n}$.
[3] shows that if $1-\varepsilon$ is negligible, then this security turns to be the standard security against malicious adversaries; if $\varepsilon \geq 1 / \operatorname{poly}(k)$, then this security implies the security against semi-honest adversaries. It is easy to deduce that, if $\varepsilon \in(0,1)$ is a constant value, then this security also implies the security against semi-honest adversaries.

### 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

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

and also be denoted by

$$
(\vec{m}, H) \mapsto(\lambda, \vec{m}\langle H\rangle)
$$

for simplicity, 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 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 [21] 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, $P_{2}{ }^{\prime}$ 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, the ideal world, real world were introduced. Since the security target here is to be secure against non-adaptive covert adversaries, we only referred to concepts related to this case in the following.

### 2.2.2 Non-Adaptive Covert Adversary

Before running $O T_{h}^{n}$, the adversary $\mathcal{A}$ has to corrupt all parties listed in $I \subseteq[2]$. In case $P_{i} \in\left\{P_{1}, P_{2}\right\}$ is not corrupted, $P_{i}$ will strictly follow the prescribed protocol as an honest party. In case party $P_{i}$ is corrupted, $P_{i}$ will be fully controlled by $\mathcal{A}$ as a corrupted party. In this case, $P_{i}$ will have to pass all his knowledge to $\mathcal{A}$ before the protocol runs and follows $\mathcal{A}^{\prime}$ s instructions from then on - so there is a probability that $P_{i}$ arbitrarily deviates from prescribed protocol. In fact, after $\mathcal{A}$ finishes corrupting, $\mathcal{A}$ and all cheating parties have formed a coalition led by $\mathcal{A}$ 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. $\mathcal{A}$ receives messages addressed to the members of the coalition and sends messages on behalf of the members.

### 2.2.3 $O T_{h}^{n}$ 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 $P_{i}$. An execution of $O T_{h}^{n}$ with deterrence factor $\epsilon \in(0,1]$ proceeds as follows.

1) Initial Inputs. All entities know the public security parameter $k$. Party $P_{1}$ holds a private input $\vec{m} \in$ $\left(\{0,1\}^{*}\right)^{n}$. Party $P_{2}$ holds a private input $H \in \Psi$. Adversary $\mathcal{A}$ holds a name list $I \subseteq[2]$, a randomness $r_{\mathcal{A}} \in\{0,1\}^{*}$ and an infinite auxiliary input sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$, where $z_{k} \in\{0,1\}^{*}$. Before proceeding to the next stage, $\mathcal{A}$ corrupts parties listed in $I$ and learns $\vec{x}\langle I\rangle$, where $\vec{x} \stackrel{\text { def }}{=}(\vec{m}, H)$.
2) The parties submitting inputs to TTP. Each honest party $P_{i}$ submits $\vec{x}\langle i\rangle$ to TTP. $\mathcal{A}$ submits arbitrary string based on his knowledge to TTP for cheating parties. The inputs TTP receives, denoted by $\vec{y}$, can be described as follows.

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

where

$$
\begin{aligned}
& \alpha_{i} \leftarrow \mathcal{A}\left(1^{k}, I, r_{\mathcal{A}}, z_{k}, \vec{x}\langle I\rangle, i\right) \\
& \alpha_{i} \in\{\vec{x}\langle i\rangle\} \cup\{0,1\}^{|\vec{x}\langle i\rangle|} \cup\left\{\text { abort }_{i}\right\} \cup\left\{\text { corrupted }_{i}\right\} \\
& \cup\left\{\text { cheat }_{i}\right\}
\end{aligned}
$$

Obviously, there is a probability that $\vec{x} \neq \vec{y}$.
3) TTP computing $f$. TTP checks $\vec{y}$ and takes the following actions in order.
a) In case $\exists i\left(\vec{y}\langle i\rangle=a b o r t_{i}\right)$. For $i \in I, \vec{w}\langle i\rangle \leftarrow \lambda$; for $i \in[n]-I, \vec{w}\langle i\rangle \leftarrow a$ abrrt $_{i}$. If there are multiple abort $_{i}$ s, TTP chooses the one whose $i$ is the smallest. TTP deals with multiple corrupted $_{i} \mathrm{~s}$ and cheat $_{i} \mathrm{~s}$, which will meet later, in a similar way.
b) In case $\exists i\left(\vec{y}\langle i\rangle=\right.$ corrupted $\left._{i}\right)$. For $i \in I, \vec{w}\langle i\rangle \leftarrow$ $\lambda$; for $i \in[n]-I, \vec{w}\langle i\rangle \leftarrow$ corrupted $_{i}$.
c) In case $\exists i\left(\vec{y}\langle i\rangle=\right.$ cheat $\left._{i}\right)$. TTP does nothing.
d) In other cases, $\vec{w} \leftarrow f\left(1^{k}, \vec{y}\right)$.
4) TTP delivering results to the parties.
a) In case $\exists i\left(\vec{y}\langle i\rangle=a b o r t_{i}\right)$ or in case $\exists i(\vec{y}\langle i\rangle=$ corrupted $\left._{i}\right)$. TTP sends each $\vec{w}\langle i\rangle$ to each honest party $P_{i}(i \in[n]-I)$. Finally, TTP halts.
b) In case $\exists i\left(\vec{y}\langle i\rangle=\right.$ cheat $\left._{i}\right)$.

- With probability $\epsilon$, TTP sends corrupted ${ }_{i}$ to $\mathcal{A}$ and honest parties. That is, TTP sets $\vec{w}\langle i\rangle \leftarrow\left(\right.$ corrupted $_{i}$, corrupted $\left._{i}\right)$. TTP sends each $\vec{w}\langle i\rangle$ to each honest party $P_{i}(i \in[n]-I)$ , sends $\vec{w}\langle I\rangle$ to $\mathcal{A}$. Finally, TTP halts.
- With probability $1-\epsilon$, TTP sets $\vec{w}\langle I\rangle \leftarrow$ ( $\vec{x}\langle[n]-I\rangle$, undetected) and sends $\vec{w}\langle I\rangle$ to $\mathcal{A}$. On receiving $\vec{w}\langle I\rangle, \mathcal{A}$ determines the results the honest parties will receive as follows.

$$
c_{j} \leftarrow A\left(1^{k}, I, r_{\mathcal{A}}, z_{k}, \vec{x}\langle I\rangle, \vec{w}\langle I\rangle, j\right)
$$

$\mathcal{A}$ sends them to TTP.
For each $j \in[n]-I$, TTP sets $\vec{w}\langle j\rangle \leftarrow c_{j}$ and sends $\vec{w}\langle j\rangle$ to $P_{j}$. Finally, TTP halts.
We remark that the inputs of honest parties are leaked to $\mathcal{A}$ only if $\mathcal{A}$ commits the cheating without being caught.
c) In other cases, TTP first sends back the corrupted parties' results, i.e., sends $\vec{w}\langle I\rangle$ to $\mathcal{A}$. $\mathcal{A}$ computes

$$
\begin{aligned}
& \beta \leftarrow A\left(1^{k}, I, r_{\mathcal{A}}, z_{k}, \vec{x}\langle I\rangle, \vec{w}\langle I\rangle\right) \\
& \beta \in\left\{\text { abor }_{i} \mid i \in I\right\} \cup\{\text { continue }\}
\end{aligned}
$$

and sends $\beta$ to TTP.
TTP checks $\beta$. If $\beta \in\left\{a b o r_{i} \mid i \in I\right\}$, for each $i \in[n]-I$, TTP sets $\vec{w}\langle i\rangle \leftarrow \beta$. Then, TTP sends each $\vec{w}\langle i\rangle(i \in[n]-I)$ to each honest party $P_{i}$. Finally, TTP halts.
5) Outputs. Each honest party $P_{i}$ outputs $\vec{w}\langle i\rangle$. Each corrupted party outputs nothing (i.e., $\lambda$ ). The adversary outputs something generated by executing an arbitrary function of the information he gathers during the execution. Without loss of generality, this can be assumed to be $\left(1^{k}, I, r_{\mathcal{A}}, z_{k}, \vec{x}\langle I\rangle, \vec{w}\langle I\rangle\right)$.
The whole execution is denoted by Ideal $l_{f, \mathcal{A}(z), I}^{\epsilon}\left(1^{k}, \vec{x}, r_{\mathcal{A}}\right)$. The output of the execution is defined by the outputs of all parties and the adversary as follows.

$$
\begin{aligned}
& \text { Ideal }{ }_{f, \mathcal{A}(z), I}^{\epsilon}\left(1^{k}, \vec{x}, r_{\mathcal{A}}\right)\langle i\rangle \\
& \stackrel{\text { def }}{=} \begin{cases}\mathcal{A}^{\prime} \text { 's output, i.e., }\left(1^{k}, I, r_{\mathcal{A}},\right. & i=0 ; \\
\left.z_{k}, \vec{x}\langle I\rangle, \vec{w}\langle I\rangle\right), & i \in I ; \\
P_{i}^{\prime} \text { 's output, i.e., } \lambda, & i \in[n]-I . \\
P_{i} \text { 's output, i.e., } \vec{w}\langle i\rangle, & \end{cases}
\end{aligned}
$$

Obviously, $\operatorname{Ideal}_{f, \mathcal{A}(z), I}^{\epsilon}\left(1^{k}, \vec{x}\right)$ is a random variable whose randomness is $r_{\mathcal{A}}$.

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

In real world, there is no TTP. Let $\pi$ be a protocol for $O T_{h}^{n}$. A execution of $O T_{h}^{n}$ proceeds as follows.

1) Initial Inputs. Initial input each entity holds in real world is the same as in the ideal world with following exceptions. A randomness $r_{i}$ is held by each party $P_{i}$. After finishing the corrupting, in addition to the knowledge $\mathcal{A}$ learns in the ideal world, the corrupted parties' randomness $\vec{r}\langle I\rangle$ is also learn by $\mathcal{A}$, where $\vec{r} \stackrel{\text { def }}{=}\left(r_{1}, r_{2}\right)$.
2) Computing $f$. In the real world, computing $f$ is finished by all entities' interaction. Each honest party strictly follows the prescribed protocol $\pi$. The corrupted parties have to follow $\mathcal{A}$ 's instructions and may arbitrarily deviate from $\pi$.
3) Outputs. Each honest party $P_{i}$ always outputs what $\pi$ instructs. Each corrupted party $P_{i}$ outputs nothing. The adversary outputs something generated by executing an arbitrary function of
the information he gathers during the execution. Without loss of generality, this can be assumed to be $\left(1^{k}, I, r_{\mathcal{A}}, \vec{r}\langle I\rangle, z_{k}, \vec{x}\langle I\rangle\right)$ and the messages addressed to the corrupted parties (We denote these messages by $m s g_{I}$ ).
The whole execution is denoted by $\operatorname{Real}_{\pi, I, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H, r_{\mathcal{A}}, \vec{r}\right)$. The output of the execution is defined by the outputs of all parties and the adversary as follows.

$$
\begin{aligned}
& \operatorname{Real}_{\pi, I, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H, r_{\mathcal{A}}, \vec{r}\right)\langle i\rangle \\
& \stackrel{\text { def }}{=}\left\{\begin{array}{cl}
\mathcal{A}^{\prime} s \text { output, i.e., }\left(1^{k}, I, r_{\mathcal{A}},\right. & i=0 \\
\left.\vec{r}\langle I\rangle, z_{k}, \vec{x}\langle I\rangle, m s g_{I}\right), & i \in I \\
P_{i}^{\prime} s \text { output, i.e., } \lambda, & i \in[n]-I . \\
P_{i}^{\prime} s \text { output, i.e., what } \\
\text { instructed by } \pi,
\end{array}\right.
\end{aligned}
$$

Obviously, $\operatorname{Real}_{\pi, I, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)$ is a random variable whose randomnesses are $r_{\mathcal{A}}$ and $\vec{r}$.

### 2.2.5 Security definition

Loosely speaking, we say that protocol $\pi$ securely computes $O T_{h}^{n}$ in the presence of covert adversaries, if and only if, for any covert adversary $\mathcal{A}$, the knowledge $\mathcal{A}$ 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 covert adversary $\mathcal{A}$, what harm $\mathcal{A}$ can do in the real world is not more than what harm he can do in the ideal world.
Definition 1 (security for $O T_{h}^{n}$ against covert adversaries). Let $f$ denote the functionality of $O T_{h}^{n}$. Let $\pi$ be a concrete protocol for $O T_{h}^{n}$. Let $\epsilon \in(0,1]$. We say $\pi$ securely computes $f$ in the presence of covert adversaries with deterrence factor of $\epsilon$, if and only if for any non-uniform probabilistic polynomial-time adversary $\mathcal{A}$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ in the real world, there exists a nonuniform probabilistic polynomial-time adversary $\mathcal{S}$ with the same sequence in the ideal world such that, for any $I \subseteq[2]$, the following equation holds.

$$
\begin{array}{r}
\left\{\text { Real }_{\pi, I, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n}} \stackrel{c}{=} \\
\left\{\in \Psi, z_{k} \in\{0,1\}^{*}\right.  \tag{1}\\
\left\{\text { Ideal }_{f, I, \mathcal{S}\left(z_{k}\right)}^{\epsilon}\left(1^{k}, \vec{m}, H\right)\right\}_{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n}}^{H \in \Psi, z_{k} \in\{0,1\}^{*}}
\end{array}
$$

where the parameters input to the two probability ensembles are same, $|\vec{m}|=|H|$, and each $\vec{m}\langle i\rangle$ is of the same length. The adversary $\mathcal{S}$ in the ideal world is called a simulator of the adversary $\mathcal{A}$ in the real world.

### 2.3 Security Against Malicious Adversaries Based On One-Sided Simulation

The standard security against malicious adversaries in the field of multi-party computation is first presented by [8] and also can be seen in [21]. Compared with the security against covert adversaries, the essential difference is that this security rules outs any successful cheating of the adversaries. In other worlds, there are no choices of
cheat $_{i}$ and corrupted ${ }_{i}$ available for the adversary in the ideal world. Therefore, removing such choice and related processes, we gain the execution of $O T_{h}^{n}$ in the ideal world with the security against malicious adversaries. Similarly modifying Definition 1, we gain the definition of this standard security.

One-sided simulation security, named by [25], is a slightly weaker security against malicious adversaries than the standard security. Specifically, if $P_{2}$ is corrupted, the security is proven in the same way as that under the standard security; if $P_{1}$ is corrupted, the security is proven in a way that only guarantees that the adversaries know nothing about $P_{2}$ 's input. With respect to $O T_{h}^{n}$, the definition is specified as follows.
Definition 2 (security for $O T_{h}^{n}$ against malicious adversaries under one-sided simulation). Let $f$ denote the functionality of $O T_{h}^{n}$. Let $\pi$ be a concrete protocol for $O T_{h}^{n}$. We say $\pi$ securely computes $f$ in the presence of malicious adversaries under one-sided simulation, if and only if the following holds.

1) In case $P_{2}$ (i.e., the receiver) is corrupted, for any nonuniform probabilistic polynomial-time adversary $\mathcal{A}$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ in the real world, there exists a expected non-uniform probabilistic polynomialtime adversary $\mathcal{S}$ with the same sequence in the ideal world such that, the following equation holds.

$$
\begin{gather*}
\left\{\operatorname{Real}_{\pi,\{2\}, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}} \stackrel{c}{=} \\
\left\{\text { deal }_{f,\{2\}, \mathcal{S}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}} \tag{2}
\end{gather*}
$$

where the parameters input to the two probability ensembles are same, $|\vec{m}|=|H|$, and each $\vec{m}\langle i\rangle$ is of the same length.
2) In case $P_{1}$ (i.e., the sender) is corrupted, for any nonuniform probabilistic polynomial-time adversary $\mathcal{A}$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ in the real world, the following equation holds.

$$
\begin{gather*}
\left\{\operatorname{View}_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}^{\mathcal{A}}\left(1^{k}, \vec{m}, H\right)\right\}_{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n}} \stackrel{c}{H \in \Psi, z_{k} \in\{0,1\}^{*}} \\
\left.\left\{\text { View }_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}^{\mathcal{A}}\left(1^{k}, \vec{m}, \tilde{H}\right)\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n}}}^{\tilde{H} \in \Psi}\right\} \tag{3}
\end{gather*}
$$

One-sided simulation security is first considered by [35] and later considered by [25], [28]. Both [35] and [28] provide $O T_{1}^{2}$ with such security. [25] provides this security to the problem set intersection. Compared with standard security, some security properties, such as independence of inputs and correctness, are not guaranteed. However, the standard security against semi-honest adversaries is indeed guaranteed. What is more, onesided simulation security enable us to construct efficient protocol against malicious adversaries.

We point out that here we allow the simulator to run in expected polynomial-time, rather than require the simulator to run in strictly polynomial-time as the one-sided simulation security considered in [25], [28],
[35]. We argue that this is justified, first, [4] shows that constant-round proof system for sets outside $\mathcal{B P} \mathcal{P}$ do not have strict polynomial-time black-box simulators; second, in many cases (also when strict polynomialtime simulators exist), the expected running time of the simulator provides a better bound than the worst-case running time of the simulator [22].

### 2.4 A Variant Of Smooth Projective Hash

Basic smooth projective hash is first presented by [13]. Though it is originally used to design chosen-ciphertext secure encryption schemes, now it and its variants are used as tools to solve other problems, such as passwordbased authenticated key exchange [19], oblivious transfer [28], [42], extractable commitment |1]. In this paper, we use the variant of smooth projective hash presented by [42] to construct frameworks for $h$-out-of- $n$ oblivious transfer. In this section, we introduce this tool.

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 bringing 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 3 ( $t$-smooth $h$-projective hash family that holds properties distinguishability, hard subset membership, feasible cheating, $S P H D H C_{t, h}$, [42]). $\mathcal{H}=$ (PG,IS, DI, KG, Hash, pHash, Cheat) is an t-smooth $h$ projective hash family with witnesses and hard subset membership (SPHDHC $C_{t, h}$ ), if and only if $\mathcal{H}$ is specified as follows

- The parameter-generator $P G$ is a PPT algorithm that takes a security parameter $k$ as input and outputs a family parameter $\Lambda$, i.e., $\Lambda \leftarrow P G\left(1^{k}\right)$. $\Lambda$ will be used as a parameter to define three relations $R_{\Lambda}, \dot{R}_{\Lambda}$ and $\ddot{R}_{\Lambda}$, where $R_{\Lambda}=\dot{R}_{\Lambda} \cup \ddot{R}_{\Lambda}$. Moreover, $\dot{R}_{\Lambda} \cap \ddot{R}_{\Lambda}=\emptyset$ 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 $\vec{a}$, i.e., $\vec{a} \leftarrow I S\left(1^{k}, \Lambda\right)$.
Let $\vec{a}=\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots,\left(\dot{x}_{h}, \dot{w}_{h}\right),\left(\ddot{x}_{h+1}, \ddot{w}_{h+1}\right), \ldots\right.$,
$\left.\left(\ddot{x}_{n}, \ddot{w}_{n}\right)\right)^{T}$ 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 $\left(\dot{x}_{i}, \dot{w}_{i}\right)$ (resp., $\left.\left(\ddot{x}_{i}, \ddot{w}_{i}\right)\right), \dot{w}_{i}$ (resp., $\ddot{w}_{i}$ ) is called a witness of $\dot{x}_{i} \in L_{\dot{R}_{\Lambda}}$ (resp., $\ddot{x}_{i} \in L_{\ddot{R}_{\Lambda}}$ ). Note that, by this way we indeed have defined the relationship $R_{\Lambda}, \dot{R}_{\Lambda}$ and $\ddot{R}_{\Lambda}$ here. The properties smoothness and projection we will mention later makes sure $\dot{R}_{\Lambda} \cap \ddot{R}_{\Lambda}=\emptyset$ holds.
For simplicity in formulation later, we introduce some additional notations here. For $\vec{a}$ mentioned above, $\quad \vec{x}^{\vec{a}} \stackrel{\text { def }}{=} \quad\left(\dot{x}_{1}, \ldots, \dot{x}_{h}, \ddot{x}_{h+1}, \ldots, \ddot{x}_{n}\right)^{T}$,
$\overrightarrow{w^{a}} \stackrel{\text { def }}{=}\left(\dot{w}_{1}, \ldots, \dot{w}_{h}, \ddot{w}_{h+1}, \ldots, \ddot{w}_{n}\right)^{T}$. What is more, we abuse notation $\in$ to some extent. We write $\vec{x} \in \operatorname{Range}\left(\operatorname{IS}\left(1^{k}, \Lambda\right)\right)$ if and only if there exists a vector $\vec{x}^{\vec{a}}$ 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 distinguisher DI is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$ and a pair strings $(x, w)$ as input and outputs an indicator bit $b$, i.e., $b \leftarrow D I\left(1^{k}, \Lambda, x, w\right)$.
- The key generator $K G$ 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, i.e., $(h k, p k) \leftarrow K G\left(1^{k}, \Lambda, x\right)$.
- 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 $y$, i.e., $y \leftarrow$ $\operatorname{Hash}\left(1^{k}, \Lambda, x, h k\right)$.
- The projection pHash is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$, an instance $x$, a witness $w$ of $x$ and a projection key $p k$ as input and outputs a value $y$, i.e., $y \leftarrow p \operatorname{Hash}\left(1^{k}, \Lambda, x, w, p k\right)$.
- The cheat Cheat is a PPT algorithm that takes a security parameter $k$, a family parameter $\Lambda$ as input and outputs $n$ elements of $\dot{R}_{\Lambda}$, i.e., $\left(\left(\dot{x}_{1}, \dot{w}_{1}\right), \ldots\left(\dot{x}_{n}, \dot{w}_{n}\right)\right) \leftarrow$ Cheat $\left(1^{k}, \Lambda\right)$.
and $\mathcal{H}$ has the following properties

1) Projection. Intuitively speaking, it requires that for any instance $\dot{x} \in L_{\dot{R}_{\Lambda}}$, the hash value of $\dot{x}$ is obtainable with the help of its witness $\dot{w}$. 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

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

2) Smoothness. Intuitively speaking, it requires that for any instance vector $\overrightarrow{\ddot{x}} \in L^{t} \ddot{R}_{\Lambda}$, 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, i.e., $S m_{1} \stackrel{c}{=} S m_{2}$.
$\operatorname{SmGen}_{1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right), \vec{a} \leftarrow I S\left(1^{k}, \Lambda\right), \vec{x} \leftarrow \vec{x}^{\vec{a}}$, for each $j \in[n]$ operates as follows: $\left(h k_{j}, p k_{j}\right) \leftarrow$ $\left.\xrightarrow{K G\left(1^{k}\right.}, \Lambda, \vec{x}\langle j\rangle\right), \quad y_{j} \quad \leftarrow \quad \operatorname{Hash}\left(1^{k}, \Lambda, \vec{x}\langle j\rangle, h k_{j}\right)$, $\overrightarrow{x p k y}\langle j\rangle \leftarrow\left(\vec{x}\langle j\rangle, p k_{j}, y_{j}\right)$. 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 $j \in[n]-[h]$.
$\operatorname{Sm}_{i}\left(1^{k}\right):(\Lambda, \overrightarrow{x p k y}) \leftarrow \underset{\longrightarrow}{\operatorname{SmGen}_{i}\left(1^{k}\right), \overrightarrow{\overrightarrow{x p k y}} \leftarrow} \leftarrow$ $\pi(\overrightarrow{x p k y})$, finally outputs $(\Lambda, \overrightarrow{x p k y})$.
3) Distinguishability. Intuitively speaking, it requires that the DI can distinguish the projective instances and smooth instances with help of their witnesses. That is, it requires that the DI correctly computes the following
function.

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

4) Hard Subset Membership. Intuitively speaking, it requires that for any $\vec{x} \in \operatorname{Range}\left(\operatorname{IS}\left(1^{k}, \Lambda\right)\right), \vec{x}$ can be disordered without being detected. That is, 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 $H S M_{2} \stackrel{\text { def }}{=}\left\{\operatorname{HSM}_{2}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$, specified as follows, are computationally indistinguishable, i.e., $H S M_{1} \stackrel{c}{=} H S M_{2}$.
$H S M_{1}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right), \vec{a} \leftarrow I S\left(1^{k}, \Lambda\right)$, finally outputs $\left(\Lambda, \vec{x}^{\vec{a}}\right)$.
$H S M_{2}\left(1^{k}\right)$ : Operates as same as $H S M_{1}\left(1^{k}\right)$ with an exception that finally outputs $\left(\Lambda, \pi\left(\vec{x}^{\vec{a}}\right)\right)$.
5) Feasible Cheating. Intuitively speaking, it requires that there is a way to cheat to generate a $\vec{x}$ which is supposed to fall into $L_{\dot{R}_{\Lambda}}^{h} \times L_{\tilde{R}_{A}}^{t} \quad$ but actually falls into $L_{\dot{R}_{\Lambda}}^{n}$ without being caught. That is, for any $\pi \in \Pi$, for any $\pi^{\prime} \in \Pi$, the two probability ensembles $\mathrm{HSM}_{2}$ and $\mathrm{HSM}_{3} \stackrel{\text { def }}{=}\left\{H S M_{3}\left(1^{k}\right)\right\}_{k \in \mathbb{N}}$ 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.
$H S M_{3}\left(1^{k}\right): \Lambda \leftarrow P G\left(1^{k}\right), \vec{a} \leftarrow$ Cheat $\left(1^{k}\right)$, finally outputs $\left(\Lambda, \pi^{\prime}\left(\vec{x}^{\vec{a}}\right)\right)$.
[42] shows that $S P H D H C_{t, h}$ can be instantiated under various hardness assumptions, such as the decisional Diffie-Hellman assumption, the decisional $N$-th residuosity assumption, the decisional quadratic residuosity assumption. Please see [42] for such instantiations.

## 3 Constructing A Framework For $O T_{h}^{n}$ Against Covert Adversaries

In this section, we construct a framework for $O T_{h}^{n}$ against covert adversaries. In the framework, we will use a PPT algorithm, denoted $\Gamma$, that receiving $B_{1}, B_{2} \in \Psi$, outputs a uniformly chosen permutation $\pi \in_{U} \Pi$ 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, $E \leftarrow \emptyset, C \leftarrow[n]-B_{1}$. 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, $D \leftarrow[n]-B_{2}$, 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 $\pi$.

### 3.1 The Detailed Framework For $O T_{h}^{n}$ Against Covert Adversaries

- Common inputs: the public security parameter $k$, a $S P H D H C_{t, h}$ (where $n=h+t$ ) hash system $\mathcal{H}$, the number of instance vector $P_{2}$ (i.e., the receiver) should generate $K$, the number of instance vector $P_{2}$ should open $g$, where $K, g$ are positive integer, $g<K$, and $K \leq \operatorname{poly}(k)$.
- 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 $P_{2}$ holds a private input $H \in \Psi$ and a randomness $r_{2} \in\{0,1\}^{*}$. The adversary $\mathcal{A}$ holds a name list $I \subseteq[2]$ and a randomness $r_{\mathcal{A}} \in\{0,1\}^{*}$.
- Auxiliary Inputs: The adversary $\mathcal{A}$ holds an infinite auxiliary input sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}, z_{k} \in\{0,1\}^{*}$.
The protocol are described as follow.
- Convention: For clarity in description, we make conventions here about some trivial error-handlings such as $P_{1}$ refusing to send $P_{2}$ some message which is supposed to be sent, $P_{1}$ sending a invalid message that $P_{2}$ can not process. Handling such errors is easy. $P_{2}$ halting and outputting abort $_{1}$ suffices. $P_{1}$ can handle such errors in a similar way. In the following description, we will not explicitly iterate such details any more.
- Receiver's step (R1): $P_{2}$ generates hash parameters and samples instances.

1) $P_{2}$ samples $K$ instance vectors.
$P_{2}$ does: $\Lambda \leftarrow P G\left(1^{k}\right)$; for each $i \in[K], \quad \vec{a}_{i} \leftarrow \quad I S\left(1^{k}, \Lambda\right)$. 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 instance vectors and the corresponding hash parameters to $P_{1}$.
$P_{2}$ sends $\left(\Lambda, \tilde{\vec{x}}_{1}, \tilde{\vec{x}}_{2}, \ldots, \tilde{\vec{x}}_{K}\right)$, where $\tilde{\vec{x}}_{i} \stackrel{\text { def }}{=} \vec{x}^{\tilde{\vec{a}}_{i}}$ (correspondingly, $\tilde{\vec{w}}_{i} \stackrel{\text { def }}{=} \vec{w}^{\tilde{\vec{a}}_{i}}$ ), to $P_{1}$.

- Sender's step (S1): $P_{1}$ choose $g$ instance vectors at random to open.
$P_{1}$ does: $r \in_{U}\{0,1\}^{K}$ such that $\#\{i \mid r\langle i\rangle=1, i \in$ $[K]\}=g$, and sends $r$ to $P_{2}$.
$r$ indicates that the instance vectors whose indexes 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 by $P_{1}$ to be open. We call such $r$ choose indicator from now on.
- Receiver's step (R2): $P_{2}$ opens the chosen instances to $P_{1}$, encodes and sends his private input to $P_{1}$.

1) $P_{2}$ verifies that the choice indicator is legal, i.e. the number of 1-bits contained in $r$ just is $g$. If $r$ is not legal, $P_{2}$ halts and outputs abort ${ }_{1}$; otherwise, $P_{2}$ proceeds to the next step.
2) $P_{2}$ sends the witnesses of the chosen instances to prove that the instances generated by him 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 $J_{i} \stackrel{\text { def }}{=}\left\{j \mid \tilde{\vec{x}}_{i}\langle j\rangle \in L_{\ddot{R}_{\Lambda}}, j \in[n]\right\}$.
3) $P_{2}$ 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, $P_{2}$ encodes his private input into
sequences such as $\pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$ where $i \in \overline{C S}$.
We comment that, to prove instances' legality, there is no need for $P_{2}$ to send the witnesses of the projective instances of the chosen instance vectors to $P_{1}$, since $P_{1}$ only care whether each chosen instance vector contains enough smooth instances, which we will see later. We also comment 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 (S2): $P_{1}$ checks the chosen instances, sends his private input after encrypting them to $P_{2}$.

1) $P_{1}$ verifies that each chosen instance vector is legal, i.e., for each chosen instance vector the number of the entries belonging to $L_{\ddot{R}_{\Lambda}}$ is at least $n-h$.
$P_{1}$ checks that, for each $i \in C S, \# J_{i} \underset{\sim}{\geq} n-h$, and for each $j \in J_{i}, D I\left(1^{k}, \Lambda, \tilde{\vec{x}}_{i}\langle j\rangle, \tilde{\vec{w}}_{i}\langle j\rangle\right)$ is 1. If $P_{2}$ does not send the opening or the check fails, $P_{1}$ halts and outputs corrupted ${ }_{2}$, 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 $\tilde{\vec{x}}_{i} \leftarrow \pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$.
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: $\left(h k_{i j}, p k_{i j}\right) \leftarrow$ $K G\left(1^{k}, \Lambda, \tilde{\tilde{\vec{x}}}_{i}\langle j\rangle\right), \beta_{i j} \leftarrow \operatorname{Hash}\left(1^{k}, \Lambda, h k_{i j}, \tilde{\tilde{\vec{x}}}_{i}\langle j\rangle\right)$, $\vec{\beta}_{i} \stackrel{\text { def }}{=}\left(\beta_{i 1}, \beta_{i 2}, \ldots, \beta_{i n}\right)^{T}, \vec{c} \leftarrow \vec{m} \oplus\left(\oplus_{i \in \overline{C S}} \vec{\beta}_{i}\right)$, $\overrightarrow{p k}_{i} \stackrel{\text { def }}{=}\left(p k_{i 1}, p k_{i 2}, \ldots, p k_{i n}\right)^{T}$, sends $\vec{c}$ and $\left(\overrightarrow{p k}_{i}\right)_{i \in \overline{C S}}$ to $P_{2}$.

- Receiver's step (R3): $P_{2}$ decrypts the ciphertext $\vec{c}$ and gains the message he want.
For each $i \in \underset{\sim}{C S}, j \in H, P_{2}$ operates: $\beta_{i j}^{\prime} \leftarrow$ $p H a s h\left(1^{k}, \Lambda, \tilde{\vec{x}}_{i}\langle j\rangle, \tilde{\tilde{\vec{w}}}_{i}\langle j\rangle, \overrightarrow{p k}_{i}\langle j\rangle\right), \quad m_{j}^{\prime} \leftarrow \vec{c}\langle j\rangle \oplus$ $\left(\oplus_{i \in \overline{C S}} \beta_{i j}^{\prime}\right)$. Finally, $P_{2}$ gains the messages $\left(m_{j}^{\prime}\right)_{j \in H}$.


### 3.2 The Correctness Of The Framework

In our framework, 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 $\dot{x}$. In contrast, with respect to an instance $\ddot{x} \in L_{\ddot{R}_{\Lambda}}$, the witness services as a proof of $\ddot{x} \in L_{\ddot{R}_{\Lambda}}$. This means that a receiver can use the witnesses of $\ddot{x}$ to persuade a sender to believe that the receiver is unable to gain the hash value of $\ddot{x}$.

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.

### 3.3 The Security Of The Framework

With respect to the security of the framework, we have the following theorem.

Theorem 4 (The framework is secure against covert adversaries). Assume that $\mathcal{H}$ is $t$-smooth $h$-projective hash family that holds properties distinguishability, hard subset membership, feasible cheating ( $S P H D H C_{t, h}$ ). Then, the framework securely computes $h$-out-of-n oblivious transfer functionality in the presence of non-adaptive covert adversaries with deterrence factor $1-1 / C_{K}^{K-g}$.

Before prove Theorem 4, we first give an intuitive analysis 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, $P_{1}$ can detect with probability $1-\frac{g}{K}$ the cheating that $P_{2}$ generates a instance vector containing more than $h$ projective instance. Unexpectedly, we find that the probability that $P_{2}$ cheats to obtain more than $h$ messages may be far less than $\frac{g}{K}$.

Theorem 5. 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 $1 / C_{K}^{K-g}$.

Proof: According to the framework, there are following 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 $L_{\dot{R}_{\Lambda}}$. 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) All illegal instance vectors are lucky not to be chosen and all the instance vectors unchosen just are the illegal instance vectors, i.e., $\overline{C S}=$ $\left\{l_{1}, l_{2}, \ldots, l_{d}\right\}$. We prove this claim in two cases.
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 $j\left(j \in[d] \wedge l_{j} \in\right.$ $C S)$. 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 $j(j \in \overline{C S} \wedge$ $\vec{x}_{j}$ is legal). Because of the smoothness of $\mathcal{H}$, $P_{2}$ cannot correctly decrypt more than $h$ entries of $\vec{c}$.
3) The number of illegal instance vectors just are $K-g$, i.e., $d=K-g$. On one side, in the framework, $P_{1}$ choose $g$ instance vectors to open, therefore the number of illegal instance vectors has to be no more than $K-g$ to avoid being caught. On another side, following the analysis of the secondary condition, to learn extra messages for $P_{2}$, the number of illegal instance vectors has to be no less than $K-g$ to make possible that all $d$ unchosen instance vectors are illegal.
Let us estimate the probability that both the second and
the third necessary conditions are met. We have

$$
\begin{aligned}
\operatorname{Pr}\left(\overline{C S}=\left\{l_{1}, l_{2}, \ldots, l_{d}\right\} \wedge d=K-g\right) & =1 / C_{K}^{d} \\
& =1 / C_{K}^{K-g}
\end{aligned}
$$

This means that the probability that $P_{2}$ cheats to obtain more than $h$ messages is at most $1 / C_{K}^{K-g}$.
From the proof of Theorem 5, we know there is a possibility that $P_{2}$ commits a not effective cheating in the sense that even if such cheating is not detected by $P_{1}, P_{2}$ still learns no new knowledge. What is more, such not effective cheating even diminish the amount of knowledge $P_{2}$ deserves. Therefore, the deterrence factor is at least $1-1 / C_{K}^{K-g}$. Following the properties of binomial factor, the maximum lower-bound of the deterrence factor can be achieved by setting $g$ to be $K-\left\llcorner\frac{K+1}{2}\right\lrcorner$. This is in contrast to our intuition that the more instance vectors chosen to be open, the higher the deterrence factor is. The essential reason is that our intuition is apt to neglect the confusion induced by $S P H D H C_{t, h}$ 's property smoothness. Since $K, g$ are predetermined constant value, the distance between the deterrence factor and 1 can not be negligibly small. However, a very small distance is achievable. For example, setting $K, g$ to be 40,20 respectively, we gain a deterrence factor $\epsilon=1-7.25 \times 10^{-12}$.
For the security of $P_{2}$, the framework first should prevent $P_{1}$ from learning $P_{2}$ 's private input. There is a possible leakage of $P_{2}$ 's private input in Step R2 where $P_{2}$ encodes his private input. However, $S P H D H C_{t, h}$ 's property hard subset membership guarantees that for any $\vec{x} \in \operatorname{Range}\left(I S\left(1^{k}, \Lambda\right)\right)$, any $\pi \in \Pi$, any PPT adversary $\mathcal{A}$, without being given $\pi$, the advantage of $\mathcal{A}$ identifying an entry of $\pi(\vec{x})$ falling into $L_{\dot{R}_{\Lambda}}$ (resp., $L_{\ddot{R}_{\Lambda}}$ ) with probability over prior knowledge $h / n$ (resp., $t / n$ ) is negligible. That is, seen from $\mathcal{A}$, every entry of $\pi(\vec{x})$ seems the same. This implies that the receiver encodes his private inputs without leaking any information in our framework.
Besides cheating $P_{2}$ of private input, it seems there is another obvious attack that malicious $P_{1}$ sends invalid messages, e.g. $p k_{i j}$ that $\left(h k_{i j}, p k_{i j}\right) \notin$ $\operatorname{Range}\left(K G\left(1^{k}, \Lambda, x_{i j}\right)\right)$, 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. Seen from intuition, it seems that there is no way for $P_{1}$ to cheat out the inputs of $P_{2}$. Indeed, our security proof later shows this intuition is correct.

We now proceeds to prove Theorem 4 holds. For notational clarity, we denote the parties and the adversary in the real world by $P_{1}, P_{2}, \mathcal{A}$, and denote the corresponding entities in the ideal world by $P_{1}^{\prime}, P_{2}^{\prime}, \mathcal{S}$. In the light of the parties corrupted by adversaries, there are four cases to be considered and we separately prove Theorem 4 holds in each case.

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

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

- Initial input: $\mathcal{S}$ holds the same $k, I \stackrel{\text { def }}{=}\{1\}, z=$ $\left(z_{k}\right)_{k \in \mathbb{N}}$, as $\mathcal{A}$. What is more, $\mathcal{S}$ holds a uniform distributed randomness $r_{\mathcal{S}} \in\{0,1\}^{*}$. The parties $P_{1}^{\prime}$ and $P_{1}$, whom $\mathcal{S}$ and $\mathcal{A}$ respectively are to corrupt, hold the same $\vec{m}$.
- $\mathcal{S}$ works as follows.
- Step Sim1: $\mathcal{S}$ performs initialization operations in this step. Specifically, first, $\mathcal{S}$ corrupts $P_{1}^{\prime}$ and learns $P_{1}^{\prime \prime}$ s private input $\vec{m}$. Second, let $\overline{\mathcal{A}}$ be a copy of $\mathcal{A}$, i.e., $\overline{\mathcal{A}}=\mathcal{A}$. $\mathcal{S}$ use $\overline{\mathcal{A}}$ as a subroutine. $\mathcal{S}$ fixes the initial inputs of $\overline{\mathcal{A}}$ to be his own initial input, i.e., $k, I, z$, with exception that fixes the randomness of $\overline{\mathcal{A}}$ to be a uniformly distributed value. Third, $\mathcal{S}$ activates $\overline{\mathcal{A}}$.
We comments that, during what follows until $\overline{\mathcal{A}}$ halts, $\mathcal{S}$ builds an environment for $\overline{\mathcal{A}}$ which simulates the real world. That is, $\mathcal{S}$ simultaneously disguises himself as $P_{1}$ and $P_{2}$ to interact with $\overline{\mathcal{A}}$.
Fourth, before engaging in the framework for $O T_{h}^{n}, \overline{\mathcal{A}}$ sends a message as in the real world indicating to corrupt $P_{1} . \mathcal{S}$ plays the role of $P_{1}$ to supplies $\overline{\mathcal{A}}$ with $\vec{m}$. Now, $\overline{\mathcal{A}}$ is engaging the framework.
- Convention: In the following interactions between $\overline{\mathcal{A}}$ and $\mathcal{S}$, in case $\overline{\mathcal{A}}$ refuses to send some message which is supposed to be sent or $\overline{\mathcal{A}}$ sends a invalid message that $\mathcal{S}$ can not process, $\mathcal{S}$ sends $a^{2}$ ort $t_{1}$ to the TTP, and halts with outputting whatever $\overline{\mathcal{A}}$ outputs. We make this convention here, and will not explicitly iterate such details any more.
- Step Sim2: $\mathcal{S}$ repeats the following procedure, denoted $\Upsilon$, until the output of $\Upsilon$ is $\lambda$. In each repeating, all randomness used by $\mathcal{S}$ is fresh. On finishing the repeating, $\mathcal{S}$ proceeds to the next step.
* Procedure $\Upsilon$ :

1) $\mathcal{S}$ rewinds $\overline{\mathcal{A}}$ to the beginning of Step S 1 of the framework.
2) $\mathcal{S}$ uniformly chooses a choice indicator as a honest party $P_{1}$ does in Step S 1 of the framework. $\mathcal{S}$ executes receiver's Step R1 of the framework with following exception, for each $\bar{r}\langle i\rangle=1, \mathcal{S}$ honestly generates instance vector $\vec{x}_{i}$; for each $\bar{r}\langle i\rangle=0$, $\mathcal{S}$ call Cheat $\left(1^{k}, \Lambda\right)$ to generate instance vector $\vec{x}_{i}$.
3) $\mathcal{S}$ receives a choice indicator $r$ from $\overline{\mathcal{A}}$.
4) If $r=\bar{r}, \alpha \leftarrow \lambda$; otherwise, $\alpha \leftarrow \perp$. Finally outputs $\alpha$.
Remark 6. Step Sim2 is a key step to extract $\overline{\mathcal{A}}$ 's real input. Note that, in case event $r=\bar{r}$ happens, then each entry of each unchosen instance vector is projective. SPHDHC $C_{t, h}$ 's property feasible
cheating guarantees that $\overline{\mathcal{A}}$ can't detect this cheating. Recalling the proof of Theorem 5. we know $\mathcal{S}$ can succeed in extracting $\overline{\mathcal{A}}$ 's real input.

- Step Sim3: Playing the role of $P_{2}$ who takes an arbitrary element of $\Psi$ as private input, $\mathcal{S}$ honestly execute receiver's Step R 2 to interact with $\overline{\mathcal{A}}$. On receiving $\vec{c},\left(\overrightarrow{p k}_{i}\right)_{i \in \overline{C S}}$ sent by $\overline{\mathcal{A}}, \mathcal{S}$ decrypts every entry of $\vec{c}$, and gains $\overrightarrow{\mathcal{A}^{\prime} \text { s real }}$ input $\vec{m}^{\prime}$.
- Step Sim4: $\mathcal{S}$ sends $\vec{m}^{\prime}$ to TTP and receives back a message $\alpha \in\left\{\right.$ abort $_{1}$, corrupt $\left._{1}, \lambda\right\}$.
- Step Sim5: When $\overline{\mathcal{A}}$ halts, $\mathcal{S}$ halts with outputting what $\overline{\mathcal{A}}$ outputs.


## Lemma 7. The simulator $\mathcal{S}$ is expected polynomial-time.

Proof: First, let us focus on Step Sim2. Note that the number of legal choice indicators is $C_{K}^{g}$. So we have $\operatorname{Pr}(r=\bar{r})=1 / C_{K}^{g}$. Therefore, the expected number of repeating $\Upsilon$ is $C_{K}^{g}$ which is a predetermined constant value. Then, Step Sim 2 runs expected polynomial-time.

Second, let us focus on other steps of $\mathcal{S}$. Obviously, they all run strict polynomial-time. Following the above analysis, $\mathcal{S}$ is expected polynomial-time.

However, the security definition requires a strictly polynomial-time simulator. Observer that, the reason why the above $\mathcal{S}$ doesn't run strictly polynomial-time is that the probability that $\Upsilon$ outputs $\perp$ is relatively too high. The following lemma guarantees that we can get a new version of $\Upsilon$ that run strictly polynomial-time and outputs $\perp$ negligible probability.

Lemma 8. Given a PPT machine $M$ that outputs $\perp$ with probability $p$, where $p \in(0,1)$ is a constant value, we can have a PPT machine $\widetilde{M}$ such that

$$
\begin{gathered}
\forall x \rightarrow \operatorname{Pr}(\widetilde{M}(x)=\perp)=\mu(k) \\
\forall x \forall \alpha \rightarrow \operatorname{Pr}(\widetilde{M}(x)=\alpha \mid \widetilde{M}(x) \neq \perp) \\
\quad=\operatorname{Pr}(M(x)=\alpha \mid M(x) \neq \perp)
\end{gathered}
$$

holds.
Proof: Let $v$ be some polynomial that $v(k)>1 / p$. We construct $\widetilde{M}$ as follows.
$\widetilde{M}$ : on receiving input $x$, repeats $M(x)$ at most $v$ times. If $M(x)$ outputs something different from $\perp$ in a repetition, then sets $\alpha$ to be this output, and stops repeating; otherwise, set $\alpha$ to be $\perp$. Finally outputs $\alpha$.

Let $M_{i}(x)$ be the output of $M(x)$ in $i$-th repetition. Let $X_{i}$ be $0 / 1$ random variable defined as follows.

$$
X_{i} \stackrel{\text { def }}{=} \begin{cases}1, & M_{i}(x)=\perp \\ 0, & M_{i}(x) \neq \perp\end{cases}
$$

Then, we know $E\left(X_{i}\right)=p, D\left(X_{i}\right)=p(1-p)$.

$$
\begin{aligned}
\operatorname{Pr}(\widetilde{M}(x)=\perp) & =\operatorname{Pr}\left(\sum_{i}^{v} X_{i}<1\right) \\
& =\operatorname{Pr}\left(p-\frac{\sum_{i}^{v} X_{i}}{v}>p-\frac{1}{v}\right) \\
& \leq \operatorname{Pr}\left(\left|p-\frac{\sum_{i}^{v} X_{i}}{v}\right| \geq p-\frac{1}{v}\right)
\end{aligned}
$$

Following Chernoff Bound (can be seen in [20|), we have

$$
\operatorname{Pr}(\widetilde{M}(x)=\perp) \leq \frac{2}{e^{\frac{v}{2 \sigma^{2}}\left(p-\frac{1}{v}\right)^{2}}}
$$

Therefore, $\operatorname{Pr}(\widetilde{M}(x)=\perp)$ is negligible.
Modifying Step Sim2 as follows

- Step Sim2: $\mathcal{S}$ executes the new version of $\Upsilon$ guaranteed by Lemma 8. If $\Upsilon$ outputs $\perp$, then $\mathcal{S}$ outputs $\perp$ and halts; otherwise, $\mathcal{S}$ proceeds to the next step.
, we gain a new version of simulator. Combining the proof of Lemma 7, we have
Proposition 9. The new simulator $\mathcal{S}$ is polynomial-time.
Note that $\mathcal{S}$ never sends cheat ${ }_{1}$ to TTP. Thus, we actually construct a standard simulator for $\mathcal{A}$, and so provide a standard security against malicious adversaries to $P_{2}$. This means we need to prove for any nonuniform probabilistic polynomial-time adversary $\mathcal{A}$ with an infinite sequence $z=\left(z_{k}\right)_{k \in \mathbb{N}}$ in the real world, the following equation holds.

$$
\begin{align*}
& \left\{\text { deal }_{f,\{1\}, \mathcal{S}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}} \stackrel{c}{=} \\
& \left\{\operatorname{Real}_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\right\}_{\substack{k \in \mathbb{N}, \vec{m} \in\left(\{0,1\}^{*}\right)^{n} \\
H \in \Psi, z_{k} \in\{0,1\}^{*}}} \tag{4}
\end{align*}
$$

Proposition 10. In case $P_{1}$ was corrupted, i.e., $I=\{1\}$, the equation (4) holds.

Before proving Proposition 10. we first prove the some lemmas.

Lemma 11. The output of $\mathcal{S}$ in the ideal world and the output of $\mathcal{A}$ in the real world are computationally indistinguishable, i.e. the following equation holds.

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

Proof: First, we claim that the outputs of $\mathcal{S}$ and $\overline{\mathcal{A}}$ are computationally indistinguishable. From the construction of $\mathcal{S}$, we know that the event that $\mathcal{S}$ doesn't take $\overline{\mathcal{A}}$ 's output as its output arises only if the event that $\Upsilon$ outputs $\perp$ arises. Since the latter arises with negligible probability, this claim holds.

Second, we claim that the views of $\overline{\mathcal{A}}$ and $\mathcal{A}$ are computationally indistinguishable. The only point $\overline{\mathcal{A}}$ 's view is different from $\mathcal{A}$ 's is that the unchosen instance vectors $\overline{\mathcal{A}}$ sees are generated by calling $\operatorname{Cheat}\left(1^{k}, \Lambda\right)$. $S P H D H C_{t, h}$ 's property feasible cheating guarantees
such instance vectors are indistinguishable from the that generated honestly. therefore, this claim holds.

Combining the above two claims, this lemma holds.

Lemma 12. 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, $X \stackrel{c}{=} Y, F \stackrel{\text { def }}{=}\left(f_{k}\right)_{k \in \mathbb{N}}, f_{k}:\{0,1\}^{*} \rightarrow$ $\{0,1\}^{*}$ 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

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

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

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

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

This contradicts the fact $X \stackrel{c}{=} Y$.
Now we proceed to prove Proposition 10.
Proof: First let us focus on the real world. $\mathcal{A}^{\prime}$ s real input can be formulated as $\gamma \leftarrow A\left(1^{k}, \vec{m}, z_{k}, r_{\mathcal{A}}, r_{1}\right)$. Note that in this case, $P_{2}$ 's output is a determinate function of $\mathcal{A}^{\prime}$ s real input. Since $\mathcal{A}$ 's real input is in its view, without loss of generality, we assume $\mathcal{A}$ 's output, denoted $\alpha$, constains its real input. Therefore, $P_{2}$ 's output is a determinate function of $\mathcal{A}^{\prime}$ s output, where the function is

$$
g(\alpha)= \begin{cases}\text { abort }_{1} & \text { if } \gamma=\text { abort }_{1} \\ \text { corrupted }_{1} & \text { if } \gamma=\text { corrupted }_{1} \\ \gamma\langle H\rangle & \text { otherwise }\end{cases}
$$

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

$$
\begin{aligned}
& \operatorname{Real}_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right) \equiv \\
& \quad h\left(\operatorname{Real}_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 0\rangle\right)
\end{aligned}
$$

Similarly, in the ideal world, we have

$$
\begin{aligned}
\operatorname{Ideal}_{f,\{1\}, \mathcal{S}\left(z_{k}\right)}\left(1^{k}, \vec{m},\right. & H) \stackrel{c}{=} \\
& h\left(\text { Ideal }_{f,\{1\}, \mathcal{S}\left(z_{k}\right)}\left(1^{k}, \vec{m}, H\right)\langle 0\rangle\right)
\end{aligned}
$$

We use $\stackrel{c}{=}$ not $\equiv$ here because there is a negligible probability that $\mathcal{S}$ outputs $\perp$, which makes $h($.$) undefined.$

| Let | $X\left(1^{k}, \vec{m}, H, z_{k},\{1\}\right)$ |
| :---: | :---: |
| Real | $\left(1^{k}, \vec{m}, H\right)\langle 0\rangle, \quad Y\left(1^{k}, \vec{m}, H, z_{k},\{1\}\right)$ |
| Ideal $_{f}$ | $\left(1^{k}, \vec{m}, H\right)\langle 0\rangle$. Following Lemma |
| $X \stackrel{c}{=}$ | $\stackrel{\text { def }}{=}(h)_{k \in \mathbb{N}}$. According to Lemma 12 |

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

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

- Initial input: $\mathcal{S}$ holds the same $k, I \stackrel{\text { def }}{=}\{2\}, z=$ $\left(z_{k}\right)_{k \in \mathbb{N}}$ as $\mathcal{A}$, and holds a uniformly distributed randomness $r_{\mathcal{S}} \in\{0,1\}^{*}$. The parties $P_{2}^{\prime}$ and $P_{2}$ hold the same private input $H$.
- $\mathcal{S}$ works as follows.
- Step Sim1: In this step, $\mathcal{S}$ performs initialization operations in a similar way that the simulator does in case $P_{1}$ is corrupted. That is, $\mathcal{S}$ corrupts $P_{2}^{\prime}$ and learns $P_{2}^{\prime \prime}$ s private input $H . \mathcal{S}$ takes $\mathcal{A}^{\prime}$ s copy $\overline{\mathcal{A}}$ as a subroutine, fixes $\overline{\mathcal{A}}$ 's initial input, activates $\overline{\mathcal{A}}$, supplies $\overline{\mathcal{A}}$ with $H$. Now, $\overline{\mathcal{A}}$ is engaging the framework.
- Convention: In the following interactions between $\overline{\mathcal{A}}$ and $\mathcal{S}$, in case $\overline{\mathcal{A}}$ refuses to send some message which is supposed to be sent or $\overline{\mathcal{A}}$ sends a invalid message that $\mathcal{S}$ can not process, $\mathcal{S}$ sends abort $_{2}$ to the TTP, and halts with outputting whatever $\overline{\mathcal{A}}$ outputs. We make this convention here, and will not explicitly iterate such details any more.
- Step Sim2: On receiving $\left(\Lambda, \tilde{\vec{x}}_{1}, \tilde{\vec{x}}_{2}, \ldots, \tilde{\vec{x}}_{K}\right)$ sent from $\overline{\mathcal{A}}, \mathcal{S}$ rewinds $\overline{\mathcal{A}} C_{K}^{g}$ times to knows $\overline{\mathcal{A}}$ 's distinct responses to distinct choice indicators. Specifically, at the $i$-th time, $\mathcal{S}$ works as follows.

1) $\mathcal{S}$ choose a choice indicator $r$ never used. Note that there are just $C_{K}^{g}$ choice indicators.
2) Playing the role of $P_{1}, \mathcal{S}$ sends $r$ to $\overline{\mathcal{A}}$, records $\overline{\mathcal{A}}$ 's response (i.e., a rejection or a corresponding opening $\left.\left(\left(i, j, \vec{w}_{i}\langle j\rangle\right)\right)_{i \in C S, j \in J_{i}}\right)$.
3) $\mathcal{S}$ rewinds $\overline{\mathcal{A}}$ to the beginning of Step R2 of the framework.

- Step Sim3: Now $\mathcal{S}$ knows $\overline{\mathcal{A}}$ 's all responses. $\mathcal{S}$ proceeds as follows basing on such knowledge.
* Case 1, no $\overline{\mathcal{A}}$ 's response will cause honest party $P_{1}$ outputting corrupted ${ }_{2}$. $\mathcal{S}$ proceeds to Step Sim4.
* Case 2, $\overline{\mathcal{A}}$ 's all responses will cause honest party $P_{1}$ outputting corrupted $_{2}$. $\mathcal{S}$ sends corrupted $_{2}$ to TTP and receives back corrupted $_{2}$ from TTP. Then, playing the role of $P_{1}, \mathcal{S}$ honestly follows the sender's steps of the framework which begins from Step S 1 to interact with $\overline{\mathcal{A}}$. When $\overline{\mathcal{A}}$ halts, $\mathcal{S}$ halts with outputting what $\overline{\mathcal{A}}$ outputs.
* Case 3, the number of $\overline{\mathcal{A}}$ 's responses that will cause honest party $P_{1}$ outputting corrupted $_{2}$ is in the domain $\left[1, C_{K}^{g}-1\right] . \mathcal{S}$ send cheat $_{2}$ to TTP. Since $\mathcal{S}$ knows $\overline{\mathcal{A}}$ 's all responses, $\mathcal{S}$ knows the the probability that $\overline{\mathcal{A}}$ is being caught in cheating by honest party $P_{1}$ in the real world. We denote this probability By $\epsilon$. It is easy to see that $0<\epsilon<1$.
- Case 3.1, with probability $\epsilon$, TTP replies $\mathcal{S}$ with corrupted 2 . $\mathcal{S}$ uniformly chooses one of the choice indicator that will cause $\overline{\mathcal{A}}$ 's cheating to be caught and sends it to $\overline{\mathcal{A}}$. Then, playing the role of $P_{1}, \mathcal{S}$ honestly follows the sender's steps which begins from Step S2 to interact with $\overline{\mathcal{A}}$.
- Case 3.2 , with probability $1-\epsilon$, TTP replies $\mathcal{S}$ with undetected and honest party $P_{1}$ 's private input $\vec{m}$. Playing the role of $P_{1}$ with private input $\vec{m}, \mathcal{S}$ honestly follows the sender's steps which begins from Step S1 to interact with $\overline{\mathcal{A}}$.
When $\overline{\mathcal{A}}$ halts, $\mathcal{S}$ halts with outputting what $\overline{\mathcal{A}}$ outputs.
- Step Sim4: Playing the role of $P_{1}, \mathcal{S}$ honestly follows the sender's steps Step S1, Step S2.1, Step S2.2 to interact with $\overline{\mathcal{A}}$. Basing on the instances' witnesses $\mathcal{S}$ records in Step Sim2 and the codes $\left(\pi_{i}^{2}\right)_{i \in \overline{C S}}$ sent by $\overline{\mathcal{A}}, \mathcal{S}$ extracts $\overline{\mathcal{A}}$ 's real input $H^{\prime}$. Specifically, $\mathcal{S}$ operates as follows: for each $i \in \overline{C S}, H_{i}^{\prime} \leftarrow \emptyset$; for each $i \in \overline{C S}$ and $j \in[n]$, if $\tilde{\tilde{x}}_{i}\langle j\rangle$ is projective, then $H_{i}^{\prime} \leftarrow H_{i}^{\prime} \cup\{j\}$; finally, $H^{\prime} \leftarrow \cap_{i \in \overline{C S}} H_{i}^{\prime}$.
- Step $\operatorname{Sim} 5: \mathcal{S}$ sends $H^{\prime}$ to TTP and receives back $\vec{m}\left\langle H^{\prime}\right\rangle$. To proceed to the interaction with $\overline{\mathcal{A}}$, $\mathcal{S}$ fabricates $\vec{m}^{\prime}$ as follows. For each $i \in H^{\prime}$, $\vec{m}^{\prime}\langle i\rangle \leftarrow \vec{m}\langle i\rangle$. For each $i \notin H^{\prime}$, set $\vec{m}^{\prime}\langle i\rangle$ to be an arbitrary message of appropriate length.
- Step Sim6: Playing the role of $P_{1}$ with private input $\vec{m}^{\prime}, \mathcal{S}$ follows Step S 2.3 to complete the interaction with $\overline{\mathcal{A}}$. When $\overline{\mathcal{A}}$ halts, $\mathcal{S}$ halts with outputting what $\overline{\mathcal{A}}$ outputs.
Proposition 13. The simulator $\mathcal{S}$ is polynomial-time.
Looking at the construction of $\mathcal{S}$, each step is polynomial-time. Obviously, Proposition 13 holds.

Lemma 14. If Case 1 of Step Sim3 happens, then the output of $\mathcal{A}$ in the real world and that of $\overline{\mathcal{A}}$ in the ideal world are indistinguishable.

Proof: Note that the only point makes $\overline{\mathcal{A}}$ 's view different from $\mathcal{A}^{\prime}$ s view is that the ciphertext $\vec{c} \overline{\mathcal{A}}$ receives is generated by encrypting fabricated $\vec{m}^{\prime}$ rather than party $P_{1}$ 's private input $\vec{m}$.

In case $\# H^{\prime}=h, S P W H_{h, t}$ 's property smoothness directly guarantees that $\overline{\mathcal{A}}$ 's view and $\mathcal{A}^{\prime}$ s view are indistinguishable.

In case $\# H^{\prime}<h, \overline{\mathcal{A}}$ 's view and $\mathcal{A}^{\prime}$ s are indistinguishable too. This is because $S P W H_{h, t}$ 's property smoothness guarantees that for each $j \in[n]-\left[H^{\prime}\right]$, what $\overline{\mathcal{A}}, \mathcal{A}$ can respectively get from $\vec{c}\langle j\rangle, \vec{c}\langle j\rangle$ are values uniformly distributed over appropriate domains.

There is no case $\# H^{\prime}>h$, since all instance vectors are legal.

Combining above analysis, we know that $\mathcal{A}$ 's output and $\overline{\mathcal{A}}$ 's output are indistinguishable.

We remark that the proof of Lemma 14 also shows that both $\mathcal{A}^{\prime}$ s effective private input and $\overline{\mathcal{A}^{\prime}}$ s are $H^{\prime}$.
Lemma 15. If Case 2 or Case 3 of Step Sim3 happens, then the outputs of $\mathcal{A}$ in the real world and $\overline{\mathcal{A}}$ in the ideal world are identical.

Proof: Looking at the construction, we know that, in case $\mathcal{S}$ receives back corruptted ${ }_{2}$ from TTP, during the following interaction with $\overline{\mathcal{A}}, \mathcal{S}$ plays the role of honest party $P_{1}$ without the need to know $\vec{m}$. In case $\mathcal{S}$ receives back undetected from TTP, $\mathcal{S}$ knows $\vec{m}$ and perfectly plays the role of honest party $P_{1}$. Therefore, $\mathcal{A}^{\prime}$ s view and $\overline{\mathcal{A}}$ 's view are identical and their outputs are identical.
Lemma 16. The output of the adversary $\mathcal{A}$ in the real world and that of the simulator $\mathcal{S}$ in the ideal world are computationally indistinguishable, i.e.,

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

Proof: First, we claim that the outputs of $\mathcal{S}$ and $\overline{\mathcal{A}}$ are identical. This follows from the fact that $\mathcal{S}$ always takes $\overline{\mathcal{A}}$ 's output as his own output.

Second, we claim that the outputs of $\mathcal{A}$ and $\overline{\mathcal{A}}$ are computationally indistinguishable. Looking at Step Sim3 and Step Sim4, the probability that each case happens is identical to that in the real world. Following Lemma 14 and Lemma 15, this claim holds.

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

Proof: Note that the honest parties $P_{1}$ and $P_{1}^{\prime}$ end up with outputting nothing. Thus, the fact that the outputs of $\mathcal{S}$ and $\mathcal{A}$ are computationally indistinguishable, which is supported by Proposition 24, directly prove this proposition holds.

### 3.3.3 In Other Cases

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

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

### 3.4 The Communication Rounds

Step R3 is performed without communication. Each of the rest steps is performed in one round. Thus, the number of the total communication rounds is four.

Seen from practical use, there is no sense to compare our work to the works whose security are not guaranteed in the implementation or the works that can not be implemented at present. Therefore, we do not consider any protocols presented by [18], [26], [37], [38] in the rest of this paper.

Compared with existing fully-simulatable protocols for $O T_{h}^{n}$, our framework is the most efficient one in communications rounds. On counting the total communication rounds of a protocol, we count that of the modified version. In the modified version, the consecutive communications of the same direction are combined into one round. The protocol for $O T_{h \times 1}^{n}$ of [7] costs one, two zero-knowledge proofs of knowledge respectively in initialization and in transfer a message, where each zeroknowledge proofs of knowledge is performed in four rounds. The whole protocol costs at least ten rounds. The protocol for $O T_{h}^{n}$ of [24] costs one zero-knowledge proof of knowledge in initialization which is performed in three rounds at least, one protocol to extract a secret key corresponding to the identity of a message which is performed in four rounds, one zero-knowledge proof of knowledge in transfer a message which is performed in three rounds at least. We point out that the interactive proof of knowledge of a discrete logarithm modulo a prime, presented by [40] and taken as a zero-knowledge proof of knowledge protocol in [24], to our best knowledge, is not known to be zero-knowledge. However, turning to the techniques of $\Sigma$-protocol, [12] make it zero-knowledge at cost of increment of three rounds in communication, which in turn induces the increment in communication rounds of the protocol of [24]. Taking all into consideration, this protocol costs at least ten rounds. The framework for $O T_{h}^{n}$ of [42] costs six rounds.

Compared with existing fully-simulatable protocols for $O T_{1}^{2}$ that works without a trusted common reference
string (CRS), our framework is also the most efficient one in communications rounds. The protocol presented by [31] can be viewed as a special DDH-based instantiation of the framework in [42], though modifying the protocol to some extent is needed. Therefore, there is no need to considered this protocol separately. In the rest of this paper, we will not repeat this point again. The protocol presented by $[3]$ is secure against covert adversaries with deterrence factor $\frac{1}{2}$ and costs four rounds.

We point that the protocols presented in [7], [24], [31], [42] are secure against malicious adversaries, which leads to the security level of the protocols are higher than that of ours. However, as pointed out by [3], the adversaries are not always malicious and ready to cheat at any price. Therefore, in such setting, the mentioned protocol seems too pessimistic and our framework indeed makes sense.

### 3.5 The Computational Overhead

We measure the computational overhead of a protocol mainly 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. However, in comparison of two protocols, the number of private key operations is also taken into consideration, if their overhead of public key operations are same. Please see [32] to know which cryptographic operation is public key operation or private key operation.

As to our framework, the public key operations are $H a s h($.$) and p H a s h($.$) . In Step S2, P_{1}$ takes $n \cdot(K-g)$ invocations of $\operatorname{Hash}($.$) to encrypt his private input. In$ Step R3, $P_{2}$ takes $h \cdot(K-g)$ invocations of $p H a s h($. to decrypt the messages he want. To compare with existing fully-simulatable protocols in computational overhead, we instantiate our framework with the DDH-based $S P H D H C_{t, h}$ presented by [42] and set $K, g$ to be 2,1 respectively. In the resulting DDH-based instantiation, $P_{1}$ costs $n$ public key encryption operations, $P_{2}$ costs $h$ public key decryption operations, and the deterrence factor is $\frac{1}{2}$.
Compared with existing fully-simulatable protocols for $O T_{h}^{n}$, our DDH-based instantiation is the most efficient one in computational overhead. The operations of the protocol in [7] are based on the non-standard assumptions, i.e., $q$-Power Decisional Diffie-Hellman and $q$-Strong Diffie-Hellman assumptions. Such operations are more expensive than DDH-based operations. The operations of the protocol in [24] are based on Decisional Bilinear Diffie-Hellman (DBDH) assumption. Since bilinear curves are considerably more expensive than regular Elliptic curves [17] and DDH is obtainable from Elliptic curves, the DBDH-based operations are also considerably more expensive than that based on DDH. The most
efficient instantiation of the framework presented by [42] are DDH-based too. However, used in practice, the instantiation costs $P_{1} 40 n$ public key encryption operations and costs $P_{2} 40 h$ public key decryption operations in the worst case. The computational overhead in the average case is half of that in the worst case. Therefore, our DDHbased instantiation are the most efficient protocol for $O T_{h}^{n}$.

Compared with existing fully-simulatable protocols for $O T_{1}^{2}$ that works without a trusted CRS, our DDHbased instantiation for $O T_{1}^{2}$ is also the most efficient one in computational overhead. The operations of the protocol for $O T_{1}^{2}$ in [3] are based on homomorphic encryption. Fixing the same functionality $O T_{1}^{2}$, the number of public operations of the protocol and our instantiation are equivalent. However, to our best knowledge, there no homomorphic encryption whose operations are more efficient than or are as efficient as that based on DDH.

### 3.6 Extensions

In our original framework, we require $K, g$ to be predetermined constant values. However, $K, g$ can be relaxed to be any functions in $k$ with a limitation that $C_{K}^{g}$ is polynomial in $k$. The reason why we place such limitation on $K, g$ is to make sure that the simulators runs in strictly polynomial time.

## 4 Constructing A Framework For $O T_{h}^{n}$ Against Malicious Adversaries

### 4.1 The Detailed Framework For $O T_{h}^{n}$ Against Malicious Adversaries

In this section, we present a framework for $O T_{h}^{n}$ against malicious adversaries. This framework is obtained by modifying the framework against covert adversaries presented in Section 3.1 as follows.

- Common inputs: Remove the limitation that $K$ is a predetermined constant. Instead, $K$ is relaxed to be a function $K=\operatorname{poly}(k)$. Remove the parameter $g$ from the common inputs.
- Sender's step (S1): Remove the limitation placed on way of $P_{1}$ choosing the choice indicator. Instead, $P_{1}$ is granted to uniformly choose choice indicator over $\{0,1\}^{K}$.
- Receiver's step (R2): Remove the step that $P_{2}$ checks the legality of the choice indicator.
It is easy to verify the correctness of the framework. So, we omit such details.


### 4.2 The Security Of The Framework

Theorem 18 (The framework is secure against malicious adversaries). Assume that $\mathcal{H}$ is a $t$-smooth $h$-projective hash family that holds properties distinguishability, hard subset membership, feasible cheating ( $S P H D H C_{t, h}$ ). Then, the protocol securely computes h-out-of-n oblivious transfer functionality in the presence of malicious adversaries under one-sided simulation.

As previous, before proving Theorem 18 , we first give an intuitive analysis as a warm-up. With respect to the security of $P_{1}$, we have Theorem 19

Theorem 19. 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 $1 / 2^{K}$.

Proof: Compared with previous framework, the necessary conditions for malicious $P_{2}$ 's succeeding in the cheating here are the same ones mentioned in the proof of Theorem 5 except the third necessary condition. Let us estimate the probability that the second necessary condition is met. We have

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

This means that the probability that $P_{2}$ cheats to obtain more than $h$ messages is at most $1 / 2^{K}$.

With respect to the security of $P_{2}$, the intuitive analysis given to the framework against covert adversary also holds in this case. Therefore, seen from intuition, $P_{2}{ }^{\prime}$ s security is guaranteed by this new framework. Now, we proceeds to prove Theorem 18 .

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

In case $P_{1}$ is corrupted, we don't need to provide a simulator $\mathcal{S}$ for the real adversary $\mathcal{A}$ which is required by the ideal/real simulation paradigm. Instead, we prove that honest party $P_{2}$ 's privacy is guaranteed.
Proposition 20. In case $P_{1}$ was corrupted, i.e., $I=\{1\}$, the equation (3) required by Definition 2 holds.

Proof: Looking at the framework, we know the only difference between $\operatorname{View}_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}^{\mathcal{A}}\left(1^{k}, \vec{m}, H\right)$ and View $_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}^{\mathcal{A}}\left(1^{k}, \vec{m}, \tilde{H}\right)$ are the codes of honest party $P_{2}$ 's private input, i.e., $\pi_{i}^{2}$ s and $\pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$ s, where $i \in \overline{C S}$.

First, we claim that each $\pi_{i}^{2}$ is uniformly distributed over $\Pi$. Observe that $\pi_{i}^{1}$ is uniformly distributed. This leads to that $G_{i}$ is uniformly distributed. Therefore, for any image $j \in[n]$, its pre-image under $\pi_{i}^{2}$ is is uniformly distributed. That is, $\pi_{i}^{2}$ is uniformly distributed.

Second, we claim that $\vec{x}_{i} \stackrel{c}{=} \pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$. Following SPHDHC ${\underset{\tilde{x}}{t, h}}$ 's property hard membership, we have $\vec{x}_{i} \stackrel{c}{c}$ $\pi_{i}^{1}\left(\vec{x}_{i}\right)=\tilde{\vec{x}}_{i}$, and $\tilde{\vec{x}}_{i} \stackrel{c}{=} \pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$. Therefore, $\vec{x}_{i} \stackrel{c}{=} \pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$.

Combining the above two claims, we know the codes in $\operatorname{View}_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}^{\mathcal{A}}\left(1^{k}, \vec{m}, H\right) \quad$ and View $_{\pi,\{1\}, \mathcal{A}\left(z_{k}\right)}^{\mathcal{A})}\left(1^{k}, \vec{m}, \tilde{H}\right)$ are indistinguishable. Thus the proposition holds.

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

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

- Initial input: $\mathcal{S}$ holds the same $k, I \stackrel{\text { def }}{=}\{2\}, z=$ $\left(z_{k}\right)_{k \in \mathbb{N}}$ as $\mathcal{A}$, and holds a uniformly distributed
randomness $r_{\mathcal{S}} \in\{0,1\}^{*}$. The parties $P_{2}^{\prime}$ and $P_{2}$ hold the same private input $H$.
- $\mathcal{S}$ works as follows.
- Step Sim1: In this step, $\mathcal{S}$ performs initialization operations in a similar way that the simulator for the first framework does in case $P_{1}$ is corrupted. That is, $\mathcal{S}$ corrupts $P_{2}^{\prime}$ and learns $P_{2}^{\prime \prime}$ s private input $H . \mathcal{S}$ takes $\mathcal{A}^{\prime}$ s copy $\overline{\mathcal{A}}$ as a subroutine, fixes $\overline{\mathcal{A}}$ 's initial input, activates $\overline{\mathcal{A}}$, supplies $\overline{\mathcal{A}}$ with $H$. Now, $\overline{\mathcal{A}}$ is engaging the framework.
- Convention: In the following interactions between $\overline{\mathcal{A}}$ and $\mathcal{S}$, in case $\overline{\mathcal{A}}$ refuses to send some message which is supposed to be sent or $\overline{\mathcal{A}}$ sends a invalid message that $\mathcal{S}$ can not process, $\mathcal{S}$ sends abort $_{2}$ to the TTP, and halts with outputting whatever $\overline{\mathcal{A}}$ outputs. We make this convention here, and will not explicitly iterate such details any more.
- Step Sim2: Playing the role of $P_{1}, \mathcal{S}$ honestly executes the sender's steps until reaching the beginning of Step S2.3. If Step S2.3 is reached, $\mathcal{S}$ records the choice indicator $r$ and the messages, denoted $m s g$, which he sends to $\overline{\mathcal{A}}$. Then $\mathcal{S}$ proceeds to next step. Otherwise, $\mathcal{S}$ halts with outputting what $\overline{\mathcal{A}}$ outputs.
- Step Sim3: $\mathcal{S}$ repeats the following procedure, denoted $\Xi$, until the hash parameters and the instance vectors $\bar{A}$ sends in Step R1 of the framework passes the check. In each repeating, the randomness $\mathcal{S}$ uses is fresh. After the repeating, $\mathcal{S}$ records the choice indicator $\tilde{r}$ and the messages $\overline{\mathcal{A}}$ sent to open the chosen instance vectors in the last repeating.
$\Xi: \mathcal{S}$ rewinds $\overline{\mathcal{A}}$ to the beginning of Step R2, and honestly follows sender's steps to interact with $\overline{\mathcal{A}}$ which from Step S 1 to the beginning of Step S2.3.
- Step Sim4:
* Case $1, r=\tilde{r}, \mathcal{S}$ outputs failure and halts;
* Case 2, $r \neq \tilde{r}$ and $\forall i(r\langle i\rangle \neq \tilde{r}\langle i\rangle \rightarrow r\langle i\rangle=$ $1 \wedge \tilde{r}\langle i\rangle=0), \mathcal{S}$ runs from scratch;
* Case 3, $r \neq \tilde{r}$ and $\exists i(r\langle i\rangle=0 \wedge \tilde{r}\langle i\rangle=1), \mathcal{S}$ records arbitrary one of these $i$ s, denotes it by $e$, and proceeds to next step.
Remark 21. The aim of Step Sim3 and Sim4 is to extract one of $\overline{\mathcal{A}}$ 's possible private inputs. If Case 3 of Step Sim4 happens, then $\mathcal{S}$ knows which instances in $\tilde{\vec{x}}_{e}$ are smooth. What is more, $\tilde{\vec{x}}_{e}$ is indeed a legal instance vector. This is because $\tilde{\vec{x}}_{e}$ passes the check executed by $\mathcal{S}$ in Step Sim3. Combing $\pi_{e}^{2}$ received in Step Sim2, $\mathcal{S}$ knows the private input of $\overline{\mathcal{A}}$ which corresponds to $\pi_{e}^{2}\left(\tilde{\vec{x}}_{e}\right)$.
We stress the fact that this is one of $\overline{\mathcal{A}}$ 's possible private inputs. Because the elements in $\left\{\pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right) \mid i \in\right.$ $[K], r\langle i\rangle=0\}$ may correspond to distinct private inputs. However, as we will see in the proof of

Proportion 24 this inconsistence does not bring any trouble to the correctness of our simulation because of SPHDHC $C_{t, h}$ 's property smoothness.
Note that, $\overline{\mathcal{A}}$ 's initial input is fixed by $\mathcal{S}$ in Step Sim1. So receiving the same messages, $\overline{\mathcal{A}}$ responds in the same way. Therefore, rewinding $\overline{\mathcal{A}}$ to the beginning of Step R2, sending the message sent in Step Sim $2, \mathcal{S}$ can reproduce the same scenario as he meets in Step Sim2.

- Step Sim5: $\mathcal{S}$ rewinds $\overline{\mathcal{A}}$ to the beginning of Step R2 of the framework, and sends $m s g$ previously recorded to $\overline{\mathcal{A}}$ in order. According to the analysis of Remark $21, \mathcal{S}$ can extract one of $\overline{\mathcal{A}}$ 's possible private inputs. We denote this one by $H_{e}^{\prime} . \mathcal{S}$ does so and gets $H_{e}^{\prime}$.
- Step Sim6: $\mathcal{S}$ sends $H_{e}^{\prime}$ to TTP and receives back $\vec{m}\left\langle H_{e}\right\rangle$. To proceed to the interaction with $\overline{\mathcal{A}}$, $\mathcal{S}$ fabricates $\vec{m}^{\prime}$ as follows. For each $i \in H_{e}^{\prime}$, $\vec{m}^{\prime}\langle i\rangle \leftarrow \vec{m}\langle i\rangle$. For each $i \notin H_{e}^{\prime}$, set $\vec{m}^{\prime}\langle i\rangle$ to be an arbitrary message of appropriate length.
- Step Sim7: Playing the role of $P_{1}$ with private input $\vec{m}^{\prime}, \mathcal{S}$ follows Step S 2.3 to complete the interaction with $\overline{\mathcal{A}}$. When $\overline{\mathcal{A}}$ halts, $\mathcal{S}$ halts with outputting what $\overline{\mathcal{A}}$ outputs.
Lemma 22. The simulator $\mathcal{S}$ is expected polynomial-time.
Proof: First, let us focus on Step Sim3. In each repetition of $\Xi$, the chosen instance vectors are uniformly distributed. This leads to the probability that $\overline{\mathcal{A}}$ passes the check in each repetition is same. Denote this probability by $p$. The expected time of Step Sim3 is

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

Under the same analysis, the probability that $\overline{\mathcal{A}}$ passes the check in Step Sim2 is $p$ too. Then, the expected time that $\mathcal{S}$ 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 Case 2. Note that the initial inputs $\mathcal{S}$ holds is same in each trial. Thus the probability that $\mathcal{S}$ runs from scratch in each trial is same. We denote this probability by $1-q$. Then the expected time that $\mathcal{S}$ runs from Step Sim1 to the beginning of Step Sim5 is

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

The reason why there is 1 in $(1+1 / q)$ is that $\mathcal{S}$ has to run from scratch at least one time in any case.

The expected running time of $\mathcal{S}$ in a whole execution is

$$
\begin{align*}
\text {ExpTime }_{\mathcal{S}} & \leq \text { ExpTime }_{\text {Sim } 1 \rightarrow \text { Sim } 5}+\text { Time }_{\text {Sim } 5} \\
& + \text { Time }_{\text {Sim } 6}+\text { Time }_{\text {Sim } 7} \\
& =(1+1 / q) \cdot\left(\text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2}\right.  \tag{5}\\
& \left.+ \text { Time }_{\Xi}+\text { Time }_{\text {Sim } 4}\right) \\
& + \text { Time }_{\text {Sim } 5}+\text { Time }_{\text {Sim } 6}+\text { Time }_{\text {Sim } 7}
\end{align*}
$$

Third, let us estimate the value of $q$, which is the probability that $\mathcal{S}$ 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 mutually disjoint events happens.

1) Event $B$ happens, where $B$ denotes the even that $\mathcal{S}$ halts before reaching Step $\operatorname{Sim} 3$.
2) Event $\bar{B}$ happens and $R=\tilde{R}$, where $R$ and $\tilde{R}$ respectively denote the random variables which are defined as the choice indicators $\mathcal{S}$ records 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

$$
\begin{align*}
q= & \operatorname{Pr}(C) \\
= & \operatorname{Pr}(B)+\operatorname{Pr}(\bar{B} \cap R=\tilde{R}) \\
& +\operatorname{Pr}(\bar{B} \cap \exists i(R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1))  \tag{6}\\
= & \operatorname{Pr}(B)+\operatorname{Pr}(\bar{B}) \cdot(\operatorname{Pr}(R=\tilde{R} \mid \bar{B}) \\
& +\operatorname{Pr}(\exists i(R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1) \mid \bar{B}))
\end{align*}
$$

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

Since both $R$ and $\tilde{R}$ are uniformly distributed, we have

$$
\begin{equation*}
\operatorname{Pr}(R=\tilde{R} \mid \bar{B})=\# S_{1} / \#\left(\{0,1\}^{K}\right)^{2}=1 / 2^{K} \tag{7}
\end{equation*}
$$

and

$$
\begin{align*}
\operatorname{Pr}(\exists i(R\langle i\rangle=0 \wedge \tilde{R}\langle i\rangle=1) \mid \bar{B}) & =\# S_{3} / \#\left(\{0,1\}^{K}\right)^{2}  \tag{8}\\
& =1 / 2-1 / 2^{K+1}
\end{align*}
$$

Combining equation (6), (7) and (8), we have

$$
\begin{align*}
q & =\operatorname{Pr}(B)+\operatorname{Pr}(\bar{B})\left(1 / 2+1 / 2^{K+1}\right) \\
& =1 / 2+1 / 2^{K+1}+\operatorname{Pr}(B) / 2+\operatorname{Pr}(\bar{B}) / 2^{K+1}  \tag{9}\\
& >1 / 2
\end{align*}
$$

Combining equation (5) and (9), we have

$$
\begin{aligned}
\text { Exp Time }_{A^{\prime}} & <3\left(\text { Time }_{\text {Sim } 1}+\text { Time }_{\text {Sim } 2}\right. \\
& \left.+ \text { Time }_{\Xi}+\text { Time }_{\text {Sim } 4}\right) \\
& + \text { Time }_{\text {Sim } 5}+\text { Time }_{\text {Sim } 6}+\text { Time }_{\text {Sim } 7}
\end{aligned}
$$

which means the expected running time of $\mathcal{S}$ is bound by a polynomial.

Lemma 23. The probability that $\mathcal{S}$ outputs failure is less than $1 / 2^{K-1}$.

Proof: Let $X$ be a random variable defined as the number of the trials in a whole execution. From the proof of Proposition 22. we know two facts. First, $\operatorname{Pr}(X=i)=$ $(1-q)^{i-1} q \leq 1 / 2^{i-1}$. Second, in each trial the event $\mathcal{S}$ 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 (7), this probability is less than $1 / 2^{K}$. Therefore, the probability that $\mathcal{S}$ outputs failure in a whole execution is

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

Lemma 24. The output of the adversary $\mathcal{A}$ in the real world and that of the simulator $\mathcal{S}$ in the ideal world are computationally indistinguishable, i.e.,

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

Proof: First, we claim that the outputs of $\mathcal{S}$ and $\overline{\mathcal{A}}$ are computationally indistinguishable. The only point that the output of $\mathcal{S}$ is different from that of $\overline{\mathcal{A}}$ is $\mathcal{S}$ may outputs failure. Since Lemma 23 shows that this point arises with negligible probability, our claim holds.

Second, we claim that the outputs of $\mathcal{A}$ and $\overline{\mathcal{A}}$ are computationally indistinguishable. The only point makes $\overline{\mathcal{A}}$ 's view different from $\mathcal{A}^{\prime}$ s view is that the ciphertext $\vec{c}^{\prime} \overline{\mathcal{A}}$ receives is generated by encrypting fabricated $\vec{m}^{\prime}$ rather than party $P_{1}$ 's private input $\vec{m}$. This is the situation we meet in the proof of Lemma 14 . Therefore, this proof can be done in a similar way with little modification.

Let $H^{\prime} \stackrel{\text { def }}{=} \cap_{i \in \overline{C S}} H_{i}^{\prime}$. The proof of the Lemma 14 shows $H^{\prime}$ is the effective private input of $\overline{\mathcal{A}}$ and $\mathcal{A}$. If $H_{e}^{\prime}=$ $H^{\prime}$, then the proof of the second claim can be done in the same way as the proof of the Lemma 14 is done. If $H_{e}^{\prime} \supset H^{\prime}$, for each $j \in H_{e}^{\prime}-H^{\prime}$, there exists $\pi_{i}^{2}\left(\tilde{\vec{x}}_{i}\right)$ whose $j$-th entry is a smooth instance, which leads to both $\vec{c}\langle j\rangle \mathcal{A}$ sees and $\vec{c}\langle j\rangle \overline{\mathcal{A}}$ sees are values distributed over appropriate domains. Therefore, $\mathcal{A}^{\prime}$ s view and $\overline{\mathcal{A}}$ 's also are indistinguishable in this case.

Combining the above two claims, this proposition holds.

Proposition 25. In case $P_{2}$ was corrupted, i.e., $I=\{2\}$, the equation (2) required by Definition 2 holds.

Proof: Note that the honest parties $P_{1}$ and $P_{1}^{\prime}$ end up with outputting nothing. Thus, the fact that the outputs of $\mathcal{S}$ and $\mathcal{A}$ are computationally indistinguishable, which is supported by Proposition 24. directly prove this proposition holds.

### 4.3 The Communication Rounds

It is easy to see that this framework costs four communication rounds. Since this framework is secure against malicious adversaries under one-sided simulation, we only compare it with protocols that are secure with this level, and protocols that are secure against malicious adversaries under full simulation. Compared with the protocols for $O T_{1}^{2}$ with security based on one-sided simulation, i.e. the protocols presented by [2], [28], [35], our framework is not efficient in communications, since these protocols only cost two rounds. However, we argue that our framework still makes sense. Since the known protocols for $O T_{h}^{n}$ on this security level [33], |34], [36] are not constant-round and only work when $h$ is far less that $n$, e.g., $n=10^{6}, h \leq 6$. To our best knowledge, our framework is the first protocol for general $O T_{h}^{n}$ on this security level.

Compared with the protocols that are secure against malicious adversaries under full simulation, the conclusions for the framework against covert adversaries still holds for this framework. Because the two framework cost the same number of communication rounds.

### 4.4 The Computational Overhead

In Step S2, $P_{1}$ takes $n \cdot \# \overline{C S}$ invocations of $\operatorname{Hash}($.$) to$ encrypt his private input. In Step R3, $P_{2}$ 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 $K, K / 2$, 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 $K$. Lemma 23 shows that the simulator may fails with probability at most $1 / 2^{K-1}$ 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 most $1 / 2^{K-2}$. Setting $K$ to be 40 , we obtains such a probability $3.6 \times 10^{-12}$, which is secure enough to be used in practice. As a result, in the worst case, the computational overhead mainly consists of $40 n$ invocations of $\operatorname{Hash}()$ taken by $P_{1}$ and $40 h$ invocations of $p H a s h()$ taken by $P_{2}$; in the worst case, the computational overhead mainly consists of $20 n$ invocations of $\operatorname{Hash}()$ taken by $P_{1}$ and $20 h$ invocations of $p H a s h()$ taken by $P_{2}$.

Compared with the protocols with the same security level, i.e., the protocols in [2], [28], [33|-[36], our framework is not efficient. However, as pointed in Section 4.3. our framework still makes sense.

Compared with the protocols for $O T_{h}^{n}$ that are secure against malicious adversaries under full simulation, our DDH-based instantiation is the most efficient one. Following the analysis in Section 3.5, the operations of the protocol in [7], [24] are far more expensive than DDHbased operations. Though the DDH-based instantiation in [42] costs the same amount of pubic-key operations, it costs two additional private-key operations, i.e., two commitment operations.

## 5 Discussion On The Technique Cut-And-Choose

Looking at the application of the technique cut-andchoose in our two frameworks, it contradicts our intuition that the more the instance vectors chosen to be open, the lower the probability that $P_{2}$ succeeds in cheating. Theorem 5 and Theorem 19 show that the minimum probability can be much less than our intuitive value $1 / K$. The essential reason is that our intuition is apt to neglect the $S P H D H C_{t, h}$ 's property smoothness. This indeed inspire us that, to make the best of this technique, something seems beyond the technique self should be taken into account.

Comparing the two frameworks, the difference between them is the way of applying the technique cut-and-choose, i.e., the way $P_{1}$ chooses instance vectors to open. Though this difference seems trivial, its influence is not trivial at all. Specifically, for the framework against covert adversary, we cannot reduce the probability of $P_{2}$ 's successful cheating to be negligibly small. The reason is that, on one side, the possible minimum value of this probability always is a constant value; on one side, to the make sure the simulator runs in strictly polynomial time, this probability has to be noticeable. For the framework against malicious adversary, we indeed solve the mentioned problem. However, we can not provide a security proof under the ideal/real simulation paradigm in case $P_{1}$ is corrupted. Because a simulator extracting $\overline{\mathcal{A}}$ 's real input by cheating, which is the idea used for the framework against covert adversary, needs expected $2^{K}$ times of rewinding $\overline{\mathcal{A}}$. This problem seems unsolvable if the instance vectors to be open are only chosen by $P_{1}$. Therefore, $P_{2}$ 's participating in choosing such instance vectors seems essential. In fact, [42] uses this idea to solve this problem. However, [42] also needs two additional tools, a perfectly hiding commitment scheme and a perfectly binding commitment schemes. The resulting framework costs two more communication round and two more commitment operations.

## References

[1] M. Abdalla, C. Chevalier, and D. Pointcheval. Smooth projective hashing for conditionally extractable commitments. Advances in Cryptology - Crypto 2009, 5677:671-689 689, 2009. Blc86 Times Cited:0 Cited References Count:31 Lecture Notes in Computer Science.
[2] B. Aiello, Y. Ishai, and O. Reingold. Priced oblivious transfer: How to sell digital goods. In Advances in Cryptology-Eurocrypt'2001, pages 119-135. Springer, 2001.
[3] Y. Aumann and Y. Lindell. Security against covert adversaries: Efficient protocols for realistic adversaries. Journal of Cryptology, 23(2):281-343, 2010.
[4] B. Barak and Y. Lindell. Strict Polynomial-time in Simulation and Extraction. SIAM Journal on Computing, 33(4):783-818, 2004.
[5] D. Bernstein. Proving tight security for Rabin-Williams signatures. pages 70-87. Springer, 2008.
[6] D. Boneh, C. Gentry, and M. Hamburg. Space-efficient identity based encryptionwithout pairings. In Foundations of Computer Science, 2007. FOCS'07. 48th Annual IEEE Symposium on, pages 647-657. IEEE, 2007.
[7] J. Camenisch, G. Neven, and A. Shelat. Simulatable adaptive oblivious transfer. In Advances in Cryptology-Eurocrypt'2007, page 590. Springer-Verlag, 2007.
[8] R. Canetti. Security and composition of multiparty cryptographic protocols. Journal of Cryptology, 13(1):143-202, 2000.
[9] R. Canetti and M. Fischlin. Universally composable commitments. In Advances in CryptologyłCRYPTO 2001, pages 19-40. Springer, 2001.
[10] R Canetti, O Goldreich, and S Halevi. The random oracle methodology, revisited. Journal of the ACM (JACM), 51(4):557-594, 2004.
[11] R. Canetti, E. Kushilevitz, and Y. Lindell. On the limitations of universally composable two-party computation without set-up assumptions. Journal of Cryptology, 19(2):135-167, 2006.
[12] R. Cramer, I. Damgård, and P. MacKenzie. Efficient zeroknowledge proofs of knowledge without intractability assumptions. In Public Key Cryptography, pages 354-373. Springer, 2000.
[13] R. Cramer and V. Shoup. Universal hash proofs and a paradigm for adaptive chosen ciphertext secure public-key encryption. In L. Knudsen, editor, Advances in Cryptology - Eurocrypt'2002, pages 45-64, Amsterdam, NETHERLANDS, 2002. Springer-Verlag Berlin.
[14] C. Crépeau. Equivalence Between Two Flavours of Oblivious Transfers. In Advances in Cryptology-Crypto'87, page 354. SpringerVerlag, 1987.
[15] S. Even, O. Goldreich, and A. Lempel. A randomized protocol for signing contracts. Communications of the ACM, 28(6):647, 1985.
[16] M.J. Freedman, Y. Ishai, B. Pinkas, and O. Reingold. Keyword search and oblivious pseudorandom functions. Theory of Cryptography, pages 303-324, 2005.
[17] S.D. Galbraith, K.G. Paterson, and N.P. Smart. Pairings for cryptographers. Discrete Applied Mathematics, 156(16):3113-3121, 2008.
[18] J.A. Garay, D. Wichs, and H.S. Zhou. Somewhat non-committing encryption and efficient adaptively secure oblivious transfer. In Advances in Cryptology-Crypto'2009, page 523. Springer, 2009.
[19] R. Gennaro and Y. Lindell. A framework for password-based authenticated key exchange. ACM Transactions on Information and System Security (TISSEC), 9(2):234, 2006.
[20] O. Goldreich. Foundations of cryptography,volume 1. Cambridge university press, 2001.
[21] O. Goldreich. Foundations of cryptography, volume 2. Cambridge university press, 2004.
[22] O. Goldreich. On expected probabilistic polynomial-time adversaries: A suggestion for restricted definitions and their benefits. Journal of Cryptology, 23(1):1-36, 2010.
[23] O. Goldreich, S. Micali, and A. Wigderson. How to play any mental game. In Proceedings of the nineteenth annual ACM symposium on Theory of computing, pages 218-229. ACM, 1987.
[24] M. Green and S. Hohenberger. Blind identity-based encryption and simulatable oblivious transfer. In K. Kurosawa, editor, Advances in Cryptology-Asiacrypt'2007, pages 265-282, Kuching, MALAYSIA, 2007. Springer-Verlag Berlin.
[25] Carmit Hazay and Yehuda Lindell. Efficient protocols for set intersection and pattern matching with security against malicious and covert adversaries. Journal of Cryptology, 23(3):422-456, 2010.
[26] Y. Ishai, J. Kilian, K. Nissim, and E. Petrank. Extending oblivious transfers efficiently. In Advances in Cryptology-Crypto'03, pages 145-161. Springer.
[27] Y. Ishai, M. Prabhakaran, and A. Sahai. Founding cryptography on oblivious transfer - efficiently. In D. Wagner, editor, Advances in Cryptology-Crypto'2008, pages 572-591, Santa Barbara, CA, 2008. Springer-Verlag Berlin.
[28] Yael Tauman Kalai. Smooth projective hashing and two-message oblivious transfer. In Advances in Cryptology C EUROCRYPT 2005, volume 3494, pages 78-95. Springer, 2005.
[29] J Kilian. Founding crytpography on oblivious transfer. In Proceedings of the twentieth annual ACM symposium on Theory of computing, pages 20-31, Inc, One Astor Plaza, 1515 Broadway, New York, NY,10036-5701, USA, 1988. ACM New York, NY, USA. STOC.
[30] Gatan Leurent and Phong Nguyen. How risky is the randomoracle model? In Shai Halevi, editor, Advances in Cryptology Crypto 2009, volume 5677 of Lecture Notes in Computer Science, pages 445-464. Springer Berlin / Heidelberg, 2009.
[31] A.Y. Lindell. Efficient Fully-Simulatable Oblivious Transfer. In Topics in cryptology: CT-RSA 2008: the cryptographers' track at the RSA conference 2008, San Francisco, CA, USA, April 8-11, 2008: proceedings, page 52. Springer-Verlag New York Inc, 2008.
[32] Chi-Jen Lu. On the security loss in cryptographic reductions. In Advances in Cryptology-Eurocrypt'2009, volume 5479, pages 72-87. Springer, 2009.
[33] M. Naor and B. Pinkas. Oblivious transfer and polynomial evaluation. In Proceedings of the thirty-first annual ACM symposium on Theory of computing, pages 245-254. ACM New York, NY, USA, 1999.
[34] M. Naor and B. Pinkas. Oblivious transfer with adaptive queries. In Advances in Cryptology-Crypto'99, pages 573-590. Springer, 1999.
[35] M. Naor and B. Pinkas. Efficient oblivious transfer protocols. In Proceedings of the twelfth annual ACM-SIAM symposium on Discrete algorithms, page 457. Society for Industrial and Applied Mathematics, 2001.
[36] M. Naor and B. Pinkas. Computationally secure oblivious transfer. Journal of Cryptology, 18(1):1-35, 2005.
[37] W. Ogata and K. Kurosawa. Oblivious keyword search. Journal of Complexity, 20(2-3):356-371, 2004.
[38] C. Peikert, V. Vaikuntanathan, and B. Waters. A framework for efficient and composable oblivious transfer. In D. Wagner, editor, Advances in Cryptology-CRYPTO'2008, pages 554-571, Santa Barbara, CA, 2008. Springer-Verlag Berlin.
[39] M. Rabin. How to exchange secrets by oblivious transfer. Technical report, Technical Report TR-81, Harvard Aiken Computation Laboratory, 1981, 1981.
[40] CP Schnorr. Efficient signature generation by smart cards. Journal of Cryptology, 4(3):161-174, 1991.
[41] A.C.C. Yao. How to generate and exchange secrets. In Foundations of Computer Science, 1985., 27th Annual Symposium on, pages 162167, 1986.
[42] B. Zeng, X. Tang, and C. Hsu. A framework for fully-simulatable h-out-of-n oblivious transfer. Cryptology ePrint Archive, Report 2010/199, 2010. http://eprint.iacr.org/


[^0]:    This work is supported by China Postdoctoral Science Foundation (No. 20100480900).

    Zeng Bing is with the College of Computer Science and Technology, Huazhong University of Science and Technology, Wuhan City, Hubei 430074 China (email: zeng.bing@smail.hust.edu.cn, zeng.bing.zb@gmail.com).
    Tang Xueming is with the College of Computer Science and Technology, Huazhong University of Science and Technology, Wuhan City, Hubei 430074 China (e-mail:tang.xueming.txm@gmail.com).
    Xu Peng is with the College of Computer Science and Technology, Huazhong University of Science and Technology, Wuhan City, Hubei 430074 China (e-mail:xupeng@mail.hust.edu.cn).
    Jing Jiandu is with the College of Computer Science and Technology, Huazhong University of Science and Technology, Wuhan City, Hubei 430074 China (e-mail: jingjd@smail.hust.edu.cn).

