# Fine-Grained Cryptography Revisited

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Abstract. Fine-grained cryptographic primitives are secure against adversaries with bounded resources and can be computed by honest users with less resources than the adversaries. In this paper, we revisit the results by Degwekar, Vaikuntanathan, and Vasudevan in Crypto 2016 on fine-grained cryptography and show constructions of three key fundamental fine-grained cryptographic primitives: one-way permutations, hash proof systems (which in turn implies a public-key encryption scheme against chosen chiphertext attacks), and trapdoor one-way functions. All of our constructions are computable in NC<sup>1</sup> and secure against (non-uniform) NC<sup>1</sup> circuits under the widely believed worst-case assumption NC<sup>1</sup>  $\subsetneq \oplus L$ /poly.

**Keywords:** Fine-grained cryptography  $\cdot NC^1$  circuit  $\cdot$  One-way permutation  $\cdot$  Hash proof system  $\cdot$  Trapdoor one-way function

# 1 Introduction

## 1.1 Background

To prove the security of a cryptographic scheme, we typically reduce the security to some computational hardness assumption with a precise security definition. Due to the fact that most assumptions are unproven, it is desirable to make the underlying assumptions as weak as possible. However, it turns out to be very hard to construct a public-key cryptographic scheme without assuming the existence of one-way functions (OWF). Moreover, for a vast majority of primitives (including public-key encryption (PKE)), we further need to assume the hardness of specific problems such as factoring, discrete-logarithm, learning with errors, etc.. It still remains open whether it is possible to construct even basic cryptographic primitives under no assumptions, or at least mild complexity-theoretic assumptions. For instance, the complexity-theoretic assumption NP  $\not\subseteq$  BPP, which is strictly weaker than the assumption of OWFs, has been proven to be insufficient for constructing even OWFs as shown by Akavia et al. [4].

Due to the difficulty of directly constructing cryptographic primitives against any polynomial probabilistic time adversaries based on mild complexity-theoretic assumptions such as NP  $\not\subseteq$  BPP, a line of beautiful works focused on fine-grained cryptographic primitives [17], where (1) the resource of an adversary is a-prior

bounded, (2) an honest party can run the algorithms with less resource than an adversary, and (3) the underlying assumption is extremely mild.

Merkle [37] initialized the study in this field by constructing a non-interactive key exchange scheme, which can be run in time O(n) and adversaries running in time  $o(n^2)$  cannot break the security. The construction only requires random functions (i.e., the random oracle). Subsequent to his work, Biham et al. [10] showed the existence of strong OWFs based on the same assumption.

While Merkle restricted adversaries in the term of running time, Maurer considered a model where adversaries have infinite computing power but only restricted storage [36]. Afterwards, he proposed a key exchange protocol in this model [38]. Following these works, Cachin and Maurer [13] constructed a symmetric-key encryption scheme and a key exchange protocol which can be run with storage O(s) and are unconditionally secure against adversaries with storage  $o(s^2)$ . Besides, there have been many other works focusing on primitives in this model [9, 44, 19, 8, 20, 21].

In the constant depth circuit model, Ajtai and Wigderson [3] constructed an unconditional secure pseudo-random generator. Then, Boppana and Lagarias [12] exploited the results by Ajitai [2] and Furst et al. [23], which shows that parity cannot be computed in size-bounded circuits, to achieve OWFs. The proposed OWF can be computed in  $AC^0$  (constant-depth polynomial-sized) circuits consisting of AND, OR, and NOT gates of unbounded fan-in, while the inverse cannot. Afterwards, several works treating the same model have been proposed [30, 6, 45, 46].

Recently, Degwekar et al. [17] proposed fine-grained cryptographic primitives against adversaries captured by two (non-uniform) classes of adversaries, which are  $AC^0$  and  $NC^1$  (logarithmic-depth polynomial-sized) circuits consisting of AND, OR, and NOT gates of fan-in 2. They first constructed an unconditionally secure pseudorandom generator with arbitrary polynomial stretch, a weak pseudorandom function, and a secret-key encryption scheme, all of which are computable in  $AC^0$  and secure against adversaries that are  $AC^0$  circuits. Then, under the widely believed separation assumption  $NC^1 \subsetneq \oplus L/poly$ , they constructed a OWF, a pseudorandom generator, a collision-resistant hash function, and a semantically secure PKE scheme that are computable in  $NC^1$  and secure against  $NC^1$  circuits.

Following the above work, Campanelli and Gennaro [14] constructed a somewhat homomorphic encryption and a verifiable computation against NC<sup>1</sup> circuits. As in [17], the underlying assumption is NC<sup>1</sup>  $\subsetneq \oplus L/poly$ .

While the above sequence of works have achieved amazing success, it still remains open whether it is possible to construct other fine-grained primitives, such as one-way permutation (OWP), PKE against chosen ciphertext attacks (CCA), and even trapdoor one-way function (TDF).

### 1.2 Our Results and Techniques

In this paper, we propose several fine-grained cryptographic primitives under the assumption  $NC^1 \subsetneq \oplus L/poly$ . Specifically, we propose a OWP, a hash proof

system (HPS) (which in turn derives a CCA-secure PKE scheme), and a TDF. All of them are computable in  $NC^1$  and secure against adversaries captured by the class of  $NC^1$  circuits. Since a lot of results have been devoted to constructing advanced primitives from these fundamental ones, our results greatly alleviate the efforts to achieve more fine-grained primitives from scratch.

Our constructions rely on the fact shown in the papers by Appelebaum, Ishai, and Kushilevitz [31,5], that if  $NC^1 \subsetneq \oplus L/poly$ , there exist a distribution  $D_0^n$  over  $n \times n$  matrices of rank (n-1) and a distribution  $D_1^n$  over  $n \times n$  matrices of rank n, which are indistinguishable for  $NC^1$  circuits.

**One-way permutation.** As one of the most fundamental cryptographic primitives, OWP has been shown to be sufficient for constructing many primitives (e.g. pseudorandom generators [11] and universal one-way hash functions [39]). Compared with primitives built from OWFs which are not bijective (e.g., [42, 28]), ones built from OWPs are usually more efficient [7, 35].

In the previous work, Degwekar et al. [17] showed a construction of finegrained OWFs in NC<sup>1</sup>. Their construction relies on a randomized encoding of a boolean function f, which is a randomized function outputting the distribution related only to f(x). Specifically, let  $\hat{f} : \{0,1\}^n \times \{0,1\}^m \to \{0,1\}^{m+1} \in \mathbb{NC}^1$ be the randomized encoding of  $f \in \bigoplus L/poly$ , where the existence of  $\hat{f}$  is shown in [5]. Then, their construction of a OWF is  $g(x) = \hat{f}(0^n, x)$ .<sup>3</sup> However, the domain and range of g are  $\{0,1\}^m$  and  $\{0,1\}^{m+1}$  respectively, i.e., the domain and range of g are inconsistent. Thus their construction is not a permutation. Moreover, since they define OWFs using randomized encoding directly, it is difficult to make their construction a permutation, i.e., it is not clear how to further achieve OWPs under the same worst-case assumption.

In this work, we propose a collection of OWPs and extend it to a OWP, both of which are computable in  $NC^1$  and secure against  $NC^1$  circuits under the assumption  $NC^1 \subsetneq \oplus L/poly$ .

To achieve the goal, we exploit the two distributions  $D_0^n$  and  $D_1^n$  described above. Essentially, our idea is to construct a "lossy function family"  $\{f_{\mathbf{M}}(\mathbf{x}) = \mathbf{M}\mathbf{x}\}_{\mathbf{M}\in D_1^n}$ . We let  $\mathbf{M} \leftarrow D_1^n$  and  $\mathbf{M} \leftarrow D_0^n$  in the injective and lossy model respectively, and the indistinguishability between the two models can be reduced to the indistinguishability between  $D_1^n$  and  $D_0^n$ . Then we follow the Peikert-Waters [40] approach to prove that  $f_{\mathbf{M}}$  in the injective model satisfies onewayness. Furthermore, since a matrix  $\mathbf{M} \leftarrow D_1^n$  is of full rank, it holds that  $f_{\mathbf{M}}$ in the injective model is a permutation. Therefore,  $\{f_{\mathbf{M}}(\mathbf{x}) = \mathbf{M}\mathbf{x}\}_{\mathbf{M}\in D_1^n}$  is a collection of OWPs. Next, we extend it to a OWP, i.e., we give a construction of OWP based on a collection of OWPs which satisfies the distribution of index sample algorithm is identical to the uniform distribution over index set as follows. For a collection of OWPs  $\{f_i : D_i \to D_i\}_{i\in I}$  where I is an index set, define a function g with the domain  $D := \bigcup_{i\in I} (\{i\} \times D_i\}$  and  $g((i, x) \in D) = (i, f_i(x))$ .

<sup>&</sup>lt;sup>3</sup> The one-wayness of g is based on the indistinguishability of the output distributions of  $\hat{f}$  conditioned on f(x) = 0 and f(x) = 1, which can be reduced to  $NC^1 \subsetneq \oplus L/poly$ .

Since  $f_i$  is a permutation and one-way, g is a permutation and one-way as well, i.e., g is a OWP.

Hash proof system and CCA secure PKE scheme. The notion of HPS, which can be treated as designated verifier non-interactive zero-knowledge proof system for a language, was first introduced by Cramer and Shoup [16] for the purpose of constructing a CCA secure PKE scheme. An HPS allows one to generate a valid proof  $\pi$  proving that a statement x is in a language L by using a witness w and a public key pk. Also, one can generate a valid proof for x (not necessarily in L) by using only a secret key sk. For  $x \in L$ , proofs generated in these two ways should be the same. Typically, L is required to be a hard subset membership one, i.e., statements sampled from inside and outside the language should be indistinguishable. Furthermore, an HPS usually satisfies universality and smoothness. Universality means that for fixed x outside L and pk, the entropy of  $\pi$  is high enough (due to the entropy of sk). Smoothness means that for x outside L, the distribution of  $\pi$  honestly generated with sk is close to the uniform distribution in the proof space. HPSs are very versatile. Besides the application of PKE schemes, they play important roles in constructing various primitives, such as password authenticated key exchange [26, 34], oblivious transfer [33, 1], and zero-knowledge arguments [32].

In previous works, there has been no known way to construct HPSs that is computable in NC<sup>1</sup> and secure against adversaries bounded in NC<sup>1</sup> yet. Note that HPS is a quite different primitive from the ones in [17, 14], and its instantiation cannot be achieved via some simple extension. The main bottleneck is that it is not clear how to construct an HPS, where we can reduce the hardness of the subset membership problem to the indistinguishability between  $D_0^n$  and  $D_1^n$ . To overcome this problem, we define two sets L and L' that are identical to the supported language in a somewhat sophisticated way. The interesting part is that we can reduce the indistinguishability between L' and X/L to that between  $D_0^n$  and  $D_1^n$ . Also we did very careful analysis on the entropy of secret keys with respect to fixed public keys the to prove smoothness and universality. More details are given as follows.

In this work, we propose the first HPS that is computable in  $NC^1$  and secure against  $NC^1$  adversaries based on the worst-case assumption  $NC^1 \subsetneq \oplus L/poly$ .

Our idea is to let a proof in the HPS be of the form  $\mathbf{sk}^{\top}\mathbf{M}^{\top}\mathbf{w}$ , where  $\mathbf{M} \leftarrow D_0^n$ ,  $\mathbf{x} = \mathbf{M}^{\top}\mathbf{w}$  is the statement with witness  $\mathbf{w}$ ,  $\mathbf{sk}$  is the secret key, and  $\mathbf{M}\mathbf{sk} = \mathbf{pk}$  is the public key. A proof can be generated as either  $\mathbf{pk}^{\top}\mathbf{w}$  or  $\mathbf{sk}^{\top}\mathbf{x}$ . The language that our HPS supports is  $\mathrm{Im}(\mathbf{M}^{\top})$ . To achieve the hardness of our subset membership problem, we exploit the fact that  $\mathrm{Im}(\mathbf{M}^{\top})$  is identical to

$$L = \{ \mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{n-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w} \} = \{ \mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{n-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w} \},\$$

which is indistinguishable from

$$L' = \{ \mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w} \}$$
$$= \{ \mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w} \},$$

where we sample  $\mathbf{M}$  as  $\mathbf{M} \leftarrow D_1^n$  instead of  $\mathbf{M} \leftarrow D_0^n$  for L'. Then, we can reduce the indistinguishability between the uniform distributions over L' and  $X \setminus L$  to that between  $D_0^n$  and  $D_1^n$ . To prove the universality and smoothness, we show that for one  $\mathbf{pk}$ , there exist different valid secret keys, which lead to different outputs for any statement not in the language. Hence, the entropy of the proof is high due to the entropy of the secret key for a fixed  $\mathbf{pk}$  and statement. We refer the reader to Section 4 for further details.

The proof size of the above scheme is only one single bit, while we can extend it to an HPS with multi-bit proofs by running many HPSs in parallel and show that the extension is still computable in  $NC^1$  and secure against  $NC^1$  circuits.

We now can instantiate the generic constructions [16] of a CCA-secure PKE scheme with our HPSs. The resulting scheme is secure against  $NC^1$  circuits allowed to make constant rounds of adaptive decryption queries, while in each round, it can make arbitrary polynomial number of queries. This restriction is natural and defined in the same way as the adversaries for the  $NC^1$ -verifiable computation scheme in [14].

As far as we know, this is the first PKE that is CCA secure against  $NC^1$  circuits under a mild complexity-theoretic assumptions, and there is no known way to make the PKE in [17] and the somewhat homomorphic encryption scheme in [14], which are malleable, CCA secure.

**Trapdoor one-way function.** TDF is a fundamental primitive introduced by Diffie and Hellman [18]. Unlike PKE schemes, where the decryption algorithm only recovers the plaintext (not including the internal randomness used in the encryption procedure), the inversion algorithm of a TDF recovers the entire preimage. The property of TDF mentioned above is useful in many applications, where proofs of well-formedness are required [24]. However, in the same time, it makes constructing TDFs very challenging.

In the previous works [17, 14], the PKEs use randomness in the encrypting procedures and it is difficult to recover the randomness in the decrypting procedures since the constructions recover the plaintexts by canceling the randomness using the property of the kernel of  $\mathbf{M} \leftarrow D_0^n$ . Namely, it is not easy to extend their construction to achieve a TDF. In fact, it has been shown that a TDF cannot be built from a PKE scheme in a black-box way [27]<sup>4</sup>. On the other hand, it seems that there is a naive approach to construct a TDF f by defining it in the same way as our OWP, i.e.,  $f(\mathbf{x}) = \mathbf{M}_1 \mathbf{x}$  where  $\mathbf{M}_1$  is a sampled from  $D_1^n$ , and sample the inverse  $\mathbf{M}_1^{-1}$  or some other elements that can be used to solve linear equations efficiently as the trapdoor. However, there is no known way to perform such a sampling procedure in  $\mathbf{NC}^1$  circuits. Therefore, some more sophisticated approach has to be taken.

In this work, we propose a TDF computable in  $NC^1$  and secure against  $NC^1$  circuits based on  $NC^1 \subseteq \bigoplus L/poly$ . The intuition is as follows.

<sup>&</sup>lt;sup>4</sup> There is no rigorous proof showing that the separation holds for NC<sup>1</sup>, while it is an evidence that TDF is not easy to achieve.

We first change the domain to  $\{0,1\}^t \times (L \times X \setminus L)^t$  where  $\mathbf{M} \leftarrow D_0^n$ ,  $L = \operatorname{Im}(\mathbf{M})$ , and  $X = \{0,1\}^n$ . On input  $(x, (\mathbf{c}_1, \mathbf{c}'_1), \cdots, (\mathbf{c}_t, \mathbf{c}'_t)) \in \{0,1\}^t \times (L \times X \setminus L)^t$ , our TDF computes  $y = f(\mathbf{x})$ , and additionally outputs  $(\mathbf{c}_i, \mathbf{c}'_i)$  if  $x_i = 0$  and  $(\mathbf{c}'_i, \mathbf{c}_i)$  otherwise for all *i*. Here, *f* is a OWF that is computable in  $\mathbf{NC}^1$  and secure against  $\mathbf{NC}^1$  and  $x_i$  denotes the *i*th bit of *x*. Then, if we have a non-zero vector  $\mathbf{k}$  in the kernel of  $\mathbf{M}$ , which is samplable in  $\mathbf{NC}^1$  [17], we can determine whether  $\mathbf{x} \in \{0,1\}^n$  is in  $\operatorname{Im}(\mathbf{M}^{\top})$  or  $\{0,1\}^n \setminus \operatorname{Im}(\mathbf{M}^{\top})$  and recover  $x_i$  by checking whether  $\mathbf{c}_i$  and  $\mathbf{c}'_i$  are swapped. This provides us an efficiently samplable trapdoor. Due to the subset membership problem for  $L = \operatorname{Im}(\mathbf{M})$  we described before, the uniform distributions over  $\operatorname{Im}(\mathbf{M}^{\top})$  and  $\{0,1\}^n \setminus \operatorname{Im}(\mathbf{M}^{\top})$ are indistinguishable when  $\mathbf{M}$  is a matrix sampled from  $D_0^n$ . Therefore, the adversary in the one-wayness game can only obtain information on f(x) (which is one-way) and the additional pairs do little help to it.

The above technique of sampling additional pairs is called *bits planting* which was used by Garg et al. [25] to construct a TDF based on the computational Diffie-Hellman problem. Although both our construction and the one in [25] aim at constructing trapdoor TDFs, we use the *bits planting* in a different way. In [25], this technique is exploited to recover the randomness used in the computation procedure of the TDF (see [25] for details), while in our work, we use it to avoid sampling the inverse of  $\mathbf{M}$  so that every operation can be performed in  $\mathsf{NC}^1$ .

# 1.3 Possibility on the Extension from Our Proposed NC<sup>1</sup> Fine-Grained Primitives

As described above, the fundamental cryptographic primitives we considered play key roles in a great deal of applications. Hence, our results directly imply the existence of more advanced  $NC^1$ -fine-grained primitives. As a simple instance, besides CCA secure PKE schemes, our HPS immediately implies the existence of a non-interactive key exchange scheme according to the recent construction by [29]. However, some  $NC^1$  primitives can not be directly derived from existing ones by adopting previous generic conversions in the polynomial-time world since the resulting primitive may not be in  $NC^1$  any more. For example, although it is well known that pseudorandom functions can be constructed from OWF/OWPs, ones in  $NC^1$  are neither implied by our  $NC^1$ -OWP nor the OWF in [17]. It remains open how to construct such fine-grained primitives, and we believe that our works will serve a good starting point.

#### 1.4 Comparison with the Proceedings Version

This is the full version of the paper that appeared at ASIACRYPT 2019 [22]. In this full version, we give the transformation from collections of OWPs to OWPs. We also give the extension to a multi-bit perfectly smooth and strong universal<sub>1</sub>  $NC^{1}$ -HPS against  $NC^{1}$  and a further extension to a strong universal<sub>2</sub>  $NC^{1}$ -HPS against  $NC^{1}$ , where the language supports  $\{0, 1\}^{n}$ . Based on the resulting HPS, we give an  $NC^{1}$ -CCA-secure PKE scheme against  $NC^{1}$ . These results are respectively given in Sections A, B, C, and D in Supplementary Material.

We also note that in the previous version, there is a subtle flaw in the proof for the subset membership problem of our HPS. Roughly, the matrix **M** included in the public parameter is sampled from  $OneSamp(\lambda)$  in the final game, while it should be sampled from  $ZeroSamp(\lambda)$  in the original definition. In this version, we fix the proof by exploiting the underlying assumption twice when switching the distribution of **M**. Compared with the previous version, the proof is more technical, while the construction remains the same. We refer the reader to Section 4 for details.

# 2 Preliminaries

#### 2.1 Notation

For a distribution D, we denote sampling x according to D by  $x \leftarrow D$ . For a set S, we denote sampling x from S uniformly at random by  $x \leftarrow S$ . We denote the set  $\{1, \dots, n\}$  by [n] and the *i*th element of a vector  $\mathbf{x}$  by  $x_i$ . For a vector  $\mathbf{x} \in \{0, 1\}^*$ ,  $\mathbf{x}$  will be regarded by default as a column vector. For a matrix  $\mathbf{M}$ , we denote the sets  $\{\mathbf{y} \mid \exists \mathbf{x} \ s.t. \ \mathbf{y} = \mathbf{M}\mathbf{x}\}$  and  $\{\mathbf{x} \mid \mathbf{M}\mathbf{x} = 0\}$  by Im( $\mathbf{M}$ ) and Ker( $\mathbf{M}$ ) respectively. Let X and Y be random variables over a finite set S. The statistical distance between X and Y is defined to be

$$Dist(X, Y) = \frac{1}{2} \sum_{s \in S} |Pr[X = s] - Pr[Y = s]|.$$

We say that X and Y are  $\epsilon$ -close if  $\text{Dist}(X, Y) \leq \epsilon$ 

We note that all arithmetic computations are over GF(2) in this work. Namely, all arithmetic computations are performed with a modulus of 2. By negl we denote an unspecified negligible function.

#### 2.2 Definitions

In this section, we recall the definitions of a function family,  $NC^1$  circuits, and  $\oplus L/poly$ .

**Definition 1 (Function Family)** A function family is a family of (possibly randomized) functions  $\mathcal{F} = \{f_{\lambda}\}_{\lambda \in \mathbb{N}}$ , where for each  $\lambda$ ,  $f_{\lambda}$  has a domain  $D_{\lambda}^{f}$  and a range  $R_{\lambda}^{f}$ .

**Definition 2 (NC<sup>1</sup>)** The class of (non-uniform) NC<sup>1</sup> function families is the set of all function families  $\mathcal{F} = \{f_{\lambda}\}$  for which there is a polynomial p and constant c such that for each  $\lambda$ ,  $f_{\lambda}$  can be computed by a (randomized) circuit of size  $p(\lambda)$ , depth  $c \log(\lambda)$  and fan-in 2 using AND, OR, and NOT gates.

**Definition 3** ( $\oplus$ L/poly)  $\oplus$ L/poly is the set of all boolean function families  $\mathcal{F} = \{f_{\lambda}\}$  for which there is a constant c such that for each  $\lambda$ , there is a non-deterministic Turing machine  $M_{\lambda}$  such that for each input x with length  $\lambda$ ,  $M_{\lambda}(x)$  uses at most  $c \log(\lambda)$  space, and  $f_{\lambda}(x)$  is equal to the parity of the number of accepting paths of  $M_{\lambda}(x)$ .

We now give the lemma about the number of solutions for the linear equations defined by a matrix. It is straightforwardly follows from the fact that the rank of  $\mathcal{A}$  is n-1.

**Lemma 1** For any  $n \times n$  matrix  $\mathbf{A}$ , if the rank of  $\mathbf{A}$  is n-1 and all arithmetic computations are over GF(2), then for any  $\mathbf{y} \in \text{Im}(\mathbf{A})$ , there exist and only exist two different vectors  $\mathbf{x}$  and  $\mathbf{x}'$  such that  $\mathbf{Ax} = \mathbf{Ax}' = \mathbf{y}$ .

# 2.3 Definitions in Fine-Grained Cryptography

In this section, we define several cryptographic primitives which are secure against restricted complexity classes of adversaries and easy to run for honest parties. In the following definitions, we denote the class of honest parties by  $C_1$  i.e., function families that compose the primitive are in the class  $C_1$  and the class of adversaries by  $C_2$ , and the condition  $C_1 \subseteq C_2$  is implicit in each definition and hence left unmentioned.

**Definition 4 (One-Way Function [17])** Let l be a polynomial in  $\lambda$ . Let  $\mathcal{F} = \{f_{\lambda} : \{0,1\}^{\lambda} \to \{0,1\}^{l(\lambda)}\}$  be a function family.  $\mathcal{F}$  is a  $\mathcal{C}_1$ -one-way function (OWF) against  $\mathcal{C}_2$  if:

- Computability: For each  $\lambda$ ,  $f_{\lambda}$  is deterministic.
- **One-wayness:** For any  $\mathcal{G} = \{g_{\lambda} : \{0,1\}^{l(\lambda)} \to \{0,1\}^{\lambda}\}$  and any  $\lambda \in \mathbb{N}$ :

$$\Pr\left[f_{\lambda}(g_{\lambda}(y)) = y \middle| \begin{array}{c} x \leftarrow \{0,1\}^{\lambda} \\ y = f_{\lambda}(x) \end{array} \right] \le \operatorname{\mathsf{negl}}(\lambda)$$

**Definition 5 (One-Way Permutation)** Let  $\mathcal{F} = \{f_{\lambda} : D_{\lambda} \to D_{\lambda}\}$  be a function family.  $\mathcal{F}$  is a  $\mathcal{C}_1$ -one-way permutation (OWP) against  $\mathcal{C}_2$  if:

- **Permutation:** For each  $\lambda$ ,  $f_{\lambda}$  is a permutation.
- **One-wayness:** For any  $\mathcal{G} = \{g_{\lambda} : D_{\lambda} \to D_{\lambda}\}$  and any  $\lambda \in \mathbb{N}$ :

$$\Pr\left[g_{\lambda}(y) = x \middle| \begin{array}{c} x \leftarrow D_{\lambda} \\ y = f_{\lambda}(x) \end{array}\right] \le \operatorname{\mathsf{negl}}(\lambda).$$

**Definition 6 (Collection of OWPs)** Let  $KeyGen = \{KeyGen_{\lambda} : \phi \rightarrow K_{\lambda}\}$  and  $\mathcal{E}val = \{Eval_{\lambda} : K_{\lambda} \times \{0,1\}^{\lambda} \rightarrow \{0,1\}^{\lambda}\}$  be function families. ( $KeyGen, \mathcal{E}val$ ) is a collection of  $C_1$ -OWPs against  $C_2$  if:

- **Permutation:** For each  $\lambda$  and  $k \leftarrow \text{KeyGen}_{\lambda}$ ,  $\text{Eval}_{\lambda}(k, \cdot) : D_{\lambda,k} \to D_{\lambda,k}$  is a permutation where  $D_{\lambda,k} \subseteq \{0,1\}^{\lambda}$ .
- **One-wayness:** For any  $\mathcal{G} = \{g_{\lambda} : K_{\lambda} \times \{0,1\}^{\lambda} \to \{0,1\}^{\lambda}\}$  and any  $\lambda \in \mathbb{N}$ :

$$\Pr\left[g_{\lambda}(k,y) = x \middle| \begin{array}{c} k \leftarrow \mathsf{KeyGen}_{\lambda} \\ x \leftarrow D_{\lambda,k} \subseteq \{0,1\}^{\lambda} \\ y = \mathsf{Eval}_{\lambda}(k,x) \end{array} \right] \leq \mathsf{negl}(\lambda).$$

**Definition 7 (Hash Proof System)** Let  $PP_{\lambda} = (X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, \Pi_{\lambda}, H_{\lambda}, \alpha_{\lambda}, \mathsf{aux}_{\lambda})$  where  $X_{\lambda}$  is a finite non-empty set,  $L_{\lambda}$  is a subset of X such that  $x \in L_{\lambda}$  iff there exists a witness  $w \in W_{\lambda}$  with  $(x, w) \in R_{\lambda} \subset X_{\lambda} \times W_{\lambda}$ ,  $SK_{\lambda}$  is a secret key space,  $PK_{\lambda}$  is a public key space,  $\Pi_{\lambda}$  is a proof space,  $H_{\lambda} : SK_{\lambda} \times X_{\lambda} \to \Pi_{\lambda}$  is a hash function,  $\alpha_{\lambda} : SK_{\lambda} \to PK_{\lambda}$  is a projective map, and  $\mathsf{aux}_{\lambda}$  is an auxiliary information. Define the following function families.

- ·  $Setup = {Setup_{\lambda} : \phi \to PP_{\lambda}}$  where  $Setup_{\lambda}$  outputs a public parameter  $pp \in PP_{\lambda}$ .
- ·  $Samp\mathcal{Y}es = \{Samp\mathsf{Y}es_{\lambda} : PP_{\lambda} \to R_{\lambda}\}\$  where  $Samp\mathsf{Y}es_{\lambda}$  on input  $pp \in PP_{\lambda}$ outputs a random element  $x \in L_{\lambda}$  with a witness  $w \in W_{\lambda}$ , i.e., a random element  $(x, w) \in R_{\lambda}$ .
- ·  $SampNo = {SampNo_{\lambda} : PP_{\lambda} \to X_{\lambda} \setminus L_{\lambda}} where SampNo_{\lambda} on input pp \in PP_{\lambda} outputs a random element <math>x \in X_{\lambda} \setminus L_{\lambda}$ .
- ·  $\mathcal{K}ey\mathcal{G}en = \{\mathsf{KeyGen}_{\lambda} : PP_{\lambda} \to PK_{\lambda} \times SK_{\lambda}\}$  where  $\mathsf{KeyGen}_{\lambda}$  on input  $\mathsf{pp} \in PP_{\lambda}$  outputs a public key pk and secret key sk such that  $pk = \alpha_{\lambda}(sk)$ .
- ·  $\mathcal{P}riv = \{\mathsf{Priv}_{\lambda} : PP_{\lambda} \times SK_{\lambda} \times X_{\lambda} \to \Pi_{\lambda}\}$  where  $\mathsf{Priv}_{\lambda}$  on input  $\mathsf{pp} \in PP_{\lambda}$ ,  $sk \in SK_{\lambda}$ , and an instance  $x \in X_{\lambda}$  outputs its proof  $\pi = H_{\lambda}(sk, x)$ .
- $\mathcal{P}ub = \{ \mathsf{Pub}_{\lambda} : PP_{\lambda} \times PK_{\lambda} \times R_{\lambda} \to \Pi_{\lambda} \}$  where  $\mathsf{Pub}_{\lambda}$  on input  $\mathsf{pp} \in PP_{\lambda}$ ,  $pk \in PK_{\lambda}$ , and an instance with a witness  $(x, w) \in R_{\lambda}$  outputs its proof  $\pi \in \Pi_{\lambda}$ .

 $(Setup, Samp \mathcal{Y}es, Samp \mathcal{N}o, \mathcal{K}ey \mathcal{G}en, \mathcal{P}riv, \mathcal{P}ub)$  is a  $C_1$ -hash proof system (HPS) against  $C_2$  if for any  $\lambda \in \mathbb{N}$ , it holds that:

- Correctness: For any  $(x, w) \in R_{\lambda}$ , we have

$$\mathsf{Priv}_{\lambda}(\mathsf{pp}, sk, x) = H_{\lambda}(sk, x) = \mathsf{Pub}_{\lambda}(\mathsf{pp}, pk, x, w)$$

where  $pp \leftarrow \mathsf{Setup}_{\lambda}$  and  $(pk, sk) \leftarrow \mathsf{KeyGen}_{\lambda}(pp)$ .

- Subset membership problem:
  - The distributions of x and x' are identical where pp ← Setup<sub>λ</sub>, (x, w) ← SampYes<sub>λ</sub>(pp), and x' ← L<sub>λ</sub>.
  - The distributions of x and x' are identical where  $pp \leftarrow Setup_{\lambda}$ ,  $x \leftarrow SampNo_{\lambda}(pp)$ , and  $x' \leftarrow X_{\lambda} \setminus L_{\lambda}$ .
  - For any  $\mathcal{G} = \{g_{\lambda}\} \in \mathcal{C}_2$ ,

$$|\Pr[g_{\lambda}(\mathsf{pp}, x_0) = 1] - \Pr[g_{\lambda}(\mathsf{pp}, x_1) = 1]| \le \mathsf{negl}(\lambda)$$

where  $pp \leftarrow Setup_{\lambda}$ ,  $(x_0, w) \leftarrow SampYes_{\lambda}(pp)$ , and  $x_1 \leftarrow SampNo_{\lambda}(pp)$ .

 $(Setup, Samp\mathcal{Y}es, Samp\mathcal{N}o, \mathcal{K}ey\mathcal{G}en, \mathcal{P}riv, \mathcal{P}ub)$  is perfectly smooth  $C_1$ -HPS against  $C_2$  if it satisfies the following property.

- **Perfect smoothness**: For any  $pp \leftarrow Setup_{\lambda}$ , the following random variables are identical, *i.e.*, 0-close.

$$(x, pk, \pi), (x, pk, \pi')$$

where  $\mathbf{x} \leftarrow \mathsf{SampNo}_{\lambda}(\mathsf{pp}), (pk, sk) \leftarrow \mathsf{KeyGen}_{\lambda}(\mathsf{pp}), \pi = \mathsf{Priv}_{\lambda}(\mathsf{pp}, sk, x), and \pi' \leftarrow \Pi.$ 

 $(Setup, Samp\mathcal{Y}es, Samp\mathcal{N}o, \mathcal{K}ey\mathcal{G}en, \mathcal{P}riv, \mathcal{P}ub)$  is  $\epsilon$ -universal<sub>1</sub>  $\mathcal{C}_1$ -HPS against  $\mathcal{C}_2$  if it satisfies the following property.

-  $\epsilon$ -universality<sub>1</sub>: For any pp  $\leftarrow$  Setup<sub> $\lambda$ </sub>,  $pk \in PK_{\lambda}$ ,  $x \in X_{\lambda} \setminus L_{\lambda}$  and  $\pi \in \Pi_{\lambda}$ , *it holds that* 

 $\Pr[\mathsf{Priv}_{\lambda}(\mathsf{pp}, sk, x) = \pi \mid \alpha_{\lambda}(sk) = pk] \le \epsilon.$ 

If  $\epsilon$  is a negligible function, then (Setup, SampYes, SampNo, KeyGen, Priv, Pub) is a strong universal<sub>1</sub>  $C_1$ -HPS against  $C_2$ .

 $(Setup, Samp\mathcal{Y}es, Samp\mathcal{N}o, \mathcal{K}ey\mathcal{G}en, \mathcal{P}riv, \mathcal{P}ub)$  is  $\epsilon$ -universal<sub>2</sub>  $\mathcal{C}_1$ -HPS against  $\mathcal{C}_2$  if it satisfies the following property.

-  $\epsilon$ -universality<sub>2</sub>: For any pp  $\leftarrow$  Setup<sub> $\lambda$ </sub>,  $pk \in PK_{\lambda}$ ,  $x, x^* \in X_{\lambda}$  and  $\pi, \pi^* \in \Pi_{\lambda}$  with  $x \notin L_{\lambda} \cup \{x^*\}$ , it holds that

 $\Pr[\mathsf{Priv}_{\lambda}(\mathsf{pp}, sk, x) = \pi \mid \mathsf{Priv}_{\lambda}(\mathsf{pp}, sk, x^*) = \pi^* \land \alpha_{\lambda}(sk) = pk] \leq \epsilon.$ 

If  $\epsilon$  is a negligible function, then (Setup, SampYes, SampNo, KeyGen, *Priv*, *Pub*) is a strong universal<sub>2</sub>  $C_1$ -HPS against  $C_2$ .

**Definition 8 (Trapdoor One-Way Function)** Let  $KeyGen = \{KeyGen_{\lambda} : \phi \to EK_{\lambda} \times TK_{\lambda}\}, Eval = \{Eval_{\lambda} : EK_{\lambda} \times \mathcal{D}_{\lambda} \to \mathcal{R}_{\lambda}\} and Inverse = \{Inverse_{\lambda} : TK_{\lambda} \times \mathcal{D}_{\lambda} \to \mathcal{R}_{\lambda}\}$  be function families where  $\mathcal{D}_{\lambda}$  and  $\mathcal{R}_{\lambda}$  are determined by the key pair (ek,tk) generated by KeyGen\_{\lambda}. (KeyGen, Eval, Inverse) is a  $C_1$ -trapdoor one-way function (TDF) against  $C_2$  if:

- Correctness: For any  $\lambda \in \mathbb{N}$ , any  $(ek, tk) \leftarrow \mathsf{KeyGen}_{\lambda}$ , and any  $X \in \mathcal{D}_{\lambda}$ :

Inverse<sub>$$\lambda$$</sub>(tk, Eval <sub>$\lambda$</sub> (ek, X)) = X.

- **One-wayness:** For any  $\mathcal{G} = \{g_{\lambda}\} \in \mathcal{C}_2$ , and any  $\lambda \in \mathbb{N}$ :

$$\Pr\left[ \mathsf{Eval}_{\lambda}(ek,g_{\lambda}(ek,Y)) = Y \, \left| \begin{array}{c} (ek,tk) \leftarrow \mathsf{KeyGen}_{\lambda} \\ X \leftarrow \mathcal{D}_{\lambda} \\ Y = \mathsf{Eval}_{\lambda}(ek,X) \end{array} \right] \leq \mathsf{negl}(\lambda).$$

# 2.4 Sampling Procedure

In this section, we recall the sampling procedure in [17], and then show several lemmas on the sampling procedure that will be used later in the security proofs.

**Construction 1 (Sampling Procedure)** Let  $\mathbf{M}_0^n$  and  $\mathbf{M}_1^n$  be the following  $n \times n$  matrices:

$$\mathbf{M}_{0}^{n} = \begin{pmatrix} 0 & \cdots & 0 & 0 \\ 1 & 0 & & 0 \\ 0 & 1 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix}, \ \mathbf{M}_{1}^{n} = \begin{pmatrix} 0 & \cdots & 0 & 1 \\ 1 & 0 & & 0 \\ 0 & 1 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix}.$$

- LSamp(n):

1. Output the following  $n \times n$  upper triangular matrix:

$$\begin{pmatrix} 1 & r_{1,2} & \cdots & r_{1,n-1} & r_{1,n} \\ 0 & 1 & r_{2,3} & \cdots & r_{2,n} \\ 0 & 0 & \ddots & & \vdots \\ \vdots & \vdots & \ddots & 1 & r_{n-1,n} \\ 0 & \cdots & 0 & 0 & 1 \end{pmatrix}$$

where  $r_{i,j} \leftarrow \{0,1\}$ . - RSamp(n):

1. Output the following  $n \times n$  matrix:

/1			$0 r_1 $
0	1		$r_2$
0	0	·	÷
	÷	۰.	$1 r_{n-1}$
$\setminus 0$	• • •	0	01/

where  $r_i \leftarrow \{0, 1\}$ .

- ZeroSamp(n):
  - 1. Sample  $\mathbf{R}_1 \leftarrow \mathsf{LSamp}(n)$  and  $\mathbf{R}_2 \leftarrow \mathsf{RSamp}(n)$ .
  - 2. Output  $\mathbf{R}_1 \mathbf{M}_0^n \mathbf{R}_2$ .
- OneSamp(n):
  - 1. Sample  $\mathbf{R}_1 \leftarrow \mathsf{LSamp}(n)$  and  $\mathbf{R}_2 \leftarrow \mathsf{RSamp}(n)$ .

2. Output  $\mathbf{R}_1 \mathbf{M}_1^n \mathbf{R}_2$ .

Here, the output of  $\mathsf{ZeroSamp}(n)$  is always a matrix of rank n-1 and the output of OneSamp(n) is always a matrix of full rank.

**Lemma 2** ( [31,5]) If NC<sup>1</sup>  $\subsetneq \oplus L/poly$ , then there is a polynomial n such that for any family  $\mathcal{F} = \{f_{\lambda}\}$  in NC<sup>1</sup> and any  $\lambda \in \mathbb{N}$ , we have

$$|\Pr[f_{\lambda}(\mathbf{M}) = 1 | \mathbf{M} \leftarrow \mathsf{ZeroSamp}(n(\lambda))] - \Pr[f_{\lambda}(\mathbf{M}') = 1 | \mathbf{M}' \leftarrow \mathsf{OneSamp}(n(\lambda))]| \le \mathsf{negl}(\lambda).$$

**Lemma 3** For any  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(n)$ , it holds that  $\operatorname{Ker}(\mathbf{M}) = \{\mathbf{0}, \mathbf{k}\}$  where  $\mathbf{k}$ is a vector such that  $\mathbf{k} \in \{0,1\}^{n-1} \times 1$ .

*Proof.* **M** is a matrix sampled from  $\mathsf{ZeroSamp}(n)$ , i.e.,

$$\mathbf{M} = \mathbf{R}_{1} \mathbf{M}_{0}^{n} \mathbf{R}_{2}$$

$$= \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 0 \\ 1 & 0 & & 0 \\ 0 & 1 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & \cdots & 0 & r_{1} \\ 0 & 1 & & r_{2} \\ 0 & 0 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 1 & r_{n-1} \\ 0 & \cdots & 0 & 0 & 1 \end{pmatrix}$$

where  $\mathbf{R}_1 \leftarrow \mathsf{LSamp}(n)$  and  $\mathbf{R}_2 \leftarrow \mathsf{RSamp}(n)$ . Then, we have  $\mathbf{k} = (r_1 r_2 \cdots 1)^\top \in \mathrm{Ker}(\mathbf{M})$  since

$$\mathbf{M} = \mathbf{R}_{1} \mathbf{M}_{0}^{n} \mathbf{R}_{2} \mathbf{k}$$

$$= \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 0 \\ 1 & 0 & & 0 \\ 0 & 1 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & \cdots & 0 & r_{1} \\ 0 & 1 & r_{2} \\ 0 & 0 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 1 & r_{n-1} \\ 0 & \cdots & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} r_{1} \\ r_{2} \\ \vdots \\ \vdots \\ \vdots \\ \vdots \\ r_{n-1} \\ 1 \end{pmatrix}$$

$$= \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 0 \\ 1 & 0 & & 0 \\ 0 & 1 & \ddots & \vdots \\ \vdots & \vdots & \ddots & 0 \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix} = \mathbf{R}_{1} \mathbf{0} = \mathbf{0}.$$

Moreover, according to Lemma 1, there are only two vectors v such that  $\mathbf{M}v = \mathbf{0}$ . Therefore, we have  $\text{Ker}(\mathbf{M}) = \{\mathbf{0}, \mathbf{k}\}$ , completing the proof of Lemma 3.

**Lemma 4** For any  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(n)$ , it holds that  $\operatorname{Ker}(\mathbf{M}^{\top}) = \{\mathbf{0}, \mathbf{k}\}$  where  $\mathbf{k}$  is a vector such that  $\mathbf{k} \in 1 \times \{0, 1\}^{n-1}$ .

*Proof.* **M** is a matrix sampled from  $\mathsf{ZeroSamp}(n)$  i.e.,  $\mathbf{M} = \mathbf{R}_1 \mathbf{M}_0^n \mathbf{R}_2$ , where  $\mathbf{R}_1 \leftarrow \mathsf{LSamp}(n)$ ,  $\mathbf{R}_2 \leftarrow \mathsf{RSamp}(n)$ . Since  $\mathbf{R}_1^\top$  has full rank, the equation  $\mathbf{R}_1^\top \mathbf{x} = (1 \ 0 \ \cdots \ 0)^\top$  has a unique solution  $\mathbf{x}^*$ .  $\mathbf{x}^*$  is in the kernel of  $\mathbf{M}^\top$  since  $\mathbf{R}_2^\top \mathbf{M}_0^\top \mathbf{R}_1^\top \mathbf{x}^* = \mathbf{R}_2^\top \mathbf{M}_0^{n\top} (1 \ 0 \ \cdots \ 0)^\top = \mathbf{R}_2^\top \mathbf{0} = \mathbf{0}$ . According to the following equation

$$\mathbf{R}_{1}^{\top}\mathbf{x}^{*} = \begin{pmatrix} 1 & 0 & \cdots & 0 & 0 \\ r_{2,1} & 1 & 0 & \cdots & 0 \\ r_{3,1} & r_{3,2} & \ddots & & \vdots \\ \vdots & \vdots & \ddots & 1 & 0 \\ r_{n,1} & \cdots & r_{n,n-1} & 1 \end{pmatrix} \begin{pmatrix} x_{1}^{*} \\ x_{2}^{*} \\ x_{3}^{*} \\ \vdots \\ x_{n}^{*} \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

we have  $x_1^* = 1$ , i.e.,  $\mathbf{x} \in 1 \times \{0, 1\}^{n-1}$ .

Moreover, according to Lemma 1 and the fact that the rank of  $\mathbf{M}^{\top}$  is n-1, there are only two vectors  $\boldsymbol{v}$  such that  $\mathbf{M}^{\top}\boldsymbol{v} = \mathbf{0}$ . Therefore, we have  $\operatorname{Ker}(\mathbf{M}^{\top}) = \{\mathbf{0}, \mathbf{x}^*\}$ , completing the proof of Lemma 4.

**Lemma 5** For any  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ , it holds that

$$\operatorname{Im}(\mathbf{M}^{\top}) = \{\mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{\lambda - 1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\} = \{\mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{\lambda - 1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\}$$

*Proof.* Let  $U = {\mathbf{x} | \mathbf{w} \in 0 \times {0,1}^{\lambda-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}}$  and  $V = {\mathbf{x} | \mathbf{w} \in 1 \times {0,1}^{\lambda-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}}$ . Let  $\mathbf{k}$  be a non-zero vector such that  $\mathbf{k} \in \text{Ker}(\mathbf{M}^{\top})$ .

According to Lemma 4, we have  $\mathbf{k} \in 1 \times \{0, 1\}^{\lambda-1}$ . Therefore, for any  $\mathbf{x} \in U$  such that  $\mathbf{x} = \mathbf{M}^{\top} \mathbf{w}$  where  $\mathbf{w} \in 0 \times \{0, 1\}^{\lambda-1}$ , we have  $\mathbf{x} = \mathbf{M}^{\top} \mathbf{w} = \mathbf{M}^{\top} (\mathbf{w} + \mathbf{k}) \in V$  since  $(\mathbf{w} + \mathbf{k}) \in 1 \times \{0, 1\}^{\lambda-1}$ . Moreover, for any  $\mathbf{x} \in V$  such that  $\mathbf{x} = \mathbf{M}^{\top} \mathbf{w}$  where  $\mathbf{w} \in 1 \times \{0, 1\}^{\lambda-1}$ , we have  $\mathbf{x} = \mathbf{M}^{\top} \mathbf{w} = \mathbf{M}^{\top} (\mathbf{w} + \mathbf{k}) \in U$  since  $(\mathbf{w} + \mathbf{k}) \in 0 \times \{0, 1\}^{\lambda-1}$ , we have  $\mathbf{x} = \mathbf{M}^{\top} \mathbf{w} = \mathbf{M}^{\top} (\mathbf{w} + \mathbf{k}) \in U$  since  $(\mathbf{w} + \mathbf{k}) \in 0 \times \{0, 1\}^{\lambda-1}$ . Therefore, we have U = V and it follows that  $\operatorname{Im}(\mathbf{M}^{\top}) = U \cup V = U \cup U = U = \{\mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{\lambda-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\}$ . In the same way, we have  $\operatorname{Im}(\mathbf{M}^{\top}) = U \cup V = V \cup V = V = \{\mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{\lambda-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\}$ . As a result, we have

$$\operatorname{Im}(\mathbf{M}^{\top}) = \{\mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{\lambda - 1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\} = \{\mathbf{x}' | \mathbf{w}' \in 1 \times \{0, 1\}^{\lambda - 1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\}$$

completing the proof of Lemma 5.

**Lemma 6** For any  $\mathbf{M} \in \mathsf{ZeroSamp}(\lambda)$ , the distributions of  $\mathbf{x}$  and  $\mathbf{x}'$  are identical where  $\mathbf{w} \leftarrow 0 \times \{0,1\}^{\lambda-1}$ ,  $\mathbf{w}' \leftarrow 0 \times \{0,1\}^{\lambda-1}$ ,  $\mathbf{x} = \mathbf{M}^{\top}\mathbf{w}$ , and  $\mathbf{x}' = \mathbf{M}^{\top}\mathbf{w}'$ .

*Proof.* Let  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and let  $U = \{\mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{\lambda-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\}$ and  $V = \{\mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{\lambda-1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}\}$ . According to Lemma 5, we have  $U = V = \mathrm{Im}(\mathbf{M}^{\top})$ . Moreover, since the rank of  $\mathbf{M}$  is  $\lambda - 1$ , we have  $|U| = |V| = |\mathbf{M}| \mathbf{M}^{\top}| = 2^{\lambda-1}$ . It follows that for any  $\mathbf{y} \in \mathrm{Im}(\mathbf{M}^{\top})$ , there exists and only exist  $\mathbf{w} \in 0 \times \{0, 1\}^{\lambda-1}$  (resp.  $\mathbf{w}' \in 1 \times \{0, 1\}^{\lambda-1}$ ) such that  $\mathbf{y} = \mathbf{M}^{\top} \mathbf{w}$  (resp.  $\mathbf{y} = \mathbf{M}^{\top} \mathbf{w}'$ ). Then, Lemma 6 immediately follows.

**Lemma 7** The distributions of  $\mathbf{M} + \mathbf{N}$  and  $\mathbf{M}'$  are identical, where  $\mathbf{M} \leftarrow \text{ZeroSamp}(\lambda)$ ,  $\mathbf{M}' \leftarrow \text{OneSamp}(\lambda)$ , and  $\mathbf{N}$  is the following matrix.

$$\mathbf{N} = \begin{pmatrix} 0 & \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & 0 \end{pmatrix}.$$

*Proof.* For  $\mathbf{R}_1 \leftarrow \mathsf{LSamp}(\lambda)$  and  $\mathbf{R}_2 \leftarrow \mathsf{RSamp}(\lambda)$ , we have

$$\begin{split} \mathbf{M}' &= \mathbf{R}_{1} \mathbf{M}_{1}^{\lambda} \mathbf{R}_{2} = \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 1 \\ 1 & 0 & \vdots & 0 \\ \vdots & \vdots & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \mathbf{R}_{2} \\ &= \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 0 \\ 1 & 0 & \vdots & 0 \\ \vdots & \vdots & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \mathbf{R}_{2} + \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 1 \\ 0 & 0 & \vdots & 0 \\ \vdots & \vdots & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 0 & 0 \end{pmatrix} \mathbf{R}_{2} \\ &= \mathbf{R}_{1} \begin{pmatrix} 0 & \cdots & 0 & 0 \\ 1 & 0 & \vdots & 0 \\ \vdots & \vdots & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 1 & 0 \end{pmatrix} \mathbf{R}_{2} + \begin{pmatrix} 0 & \cdots & 0 & 1 \\ 0 & 0 & \vdots & 0 \\ \vdots & \vdots & \ddots & 0 & \vdots \\ 0 & \cdots & 0 & 0 & 0 \end{pmatrix} \\ &= \mathbf{R}_{1} \mathbf{M}_{0}^{\lambda} \mathbf{R}_{2} + \mathbf{N} = \mathbf{M} + \mathbf{N}. \end{split}$$

Hence, the distributions of  $\mathbf{M} + \mathbf{N}$  and  $\mathbf{M}'$  are identical for  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ and  $\mathbf{M}' \leftarrow \mathsf{OneSamp}(\lambda)$ , completing the proof of Lemma 7.  $\Box$ 

# 3 Construction of NC<sup>1</sup>-OWP against NC<sup>1</sup>

In this section, we first give our construction of a collection of  $NC^1$ -OWPs against  $NC^1$  under the assumption  $NC^1 \subsetneq \oplus L/poly$ . Next, we extend it to a  $NC^1$ -OWP against  $NC^1$  based on the same assumption.

**Construction 2 (Collection of** NC<sup>1</sup>-OWPs) Let  $\lambda$  be a security parameter. We define the families  $\mathcal{K}ey\mathcal{G}en = \{\mathsf{KeyGen}_{\lambda}\}$  with key spaces  $\{K_{\lambda} = \{\mathbf{M} \mid \mathbf{M} \in \mathsf{OneSamp}(\lambda)\}\}$  and  $\mathcal{E}val = \{\mathsf{Eval}_{\lambda}\}$  as follows.

- KeyGen<sub> $\lambda$ </sub>:
  - 1. Sample  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$ .
  - 2. Output **M** (which defines the domain as  $D_{\lambda,\mathbf{M}} := \{0,1\}^{\lambda}$ ).
- $\mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x})$ :
  - 1. Compute  $\mathbf{y} := \mathbf{M}\mathbf{x}$  and output  $\mathbf{y}$ .

**Theorem 1** ( $\mathcal{K}ey\mathcal{G}en, \mathcal{E}val$ ) defined as Construction 2 is a collection of NC<sup>1</sup>-OWPs against NC<sup>1</sup> under the assumption NC<sup>1</sup>  $\subseteq \oplus L/poly$ .

**Proof sketch.** As described in Introduction, our construction is essentially a "lossy function". More specifically, it is straightforward that our scheme is a permutation, since  $\mathbf{M}$  is of full rank when  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$ . Moreover, when we generate  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  instead of  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  in  $\mathsf{KeyGen}_{\lambda}$ , we can prove that an adversary  $\mathcal{A}$  breaking the one-wayness of our construction with probability  $\epsilon$  can also be used to find a second pre-image  $\mathbf{x}'$  for  $\mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x})$  such that  $\mathbf{x} \neq \mathbf{x}'$  with probability  $\frac{1}{2}\epsilon$ . This is due to the fact that  $\mathbf{M}$  is not of full rank in this case and  $\mathcal{A}$  has no information on whether the pre-image is  $\mathbf{x}$  or  $\mathbf{x}'$ . However, it is unlikely that  $\mathcal{A}$  can find such a second pre-image, since this construction is indistinguishable with the original one, where  $\mathbf{M}$  is generated as  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  and there exists no second pre-image for each  $\mathbf{M}$ . Therefore, we can conclude that this scheme is one-way, which immediately gives us the one-wayness of the original scheme (due to the indistinguishability between  $\mathsf{OneSamp}(\lambda)$  and  $\mathsf{ZeroSamp}(\lambda)$ ).

The formal proof is as follows.

*Proof.* First note that both  $\mathcal{K}ey\mathcal{G}en$  and  $\mathcal{E}val$  are computable in NC<sup>1</sup>, since they only involve operations including multiplications of a constant number of matrices, inner products, and sampling random bits. We now show that  $(\mathcal{K}ey\mathcal{G}en, \mathcal{E}val)$  satisfies computability and one-wayness.

**Permutation.** Since for  $\mathbf{M} \leftarrow \mathsf{OneSamp}_{\lambda}$ ,  $\mathbf{M}$  is a full rank matrix, we have that  $\mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x}) = \mathbf{M}\mathbf{x} \in D_{\lambda,\mathbf{M}} = \{0,1\}^{\lambda}$  is a permutation.

**One-wayness.** Let  $\mathcal{A} = \{a_{\lambda}\}$  be any adversary in NC<sup>1</sup>. We give hybrid games to show that the advantage of  $\mathcal{A}$  in breaking the one-wayness of Construction 2 is negligible.

**Game 0:** This is the original one-wayness game for  $\mathcal{A} = \{a_{\lambda}\}$ .  $\mathcal{CH}$  runs  $\mathbf{M} \leftarrow \mathsf{KeyGen}_{\lambda}$  and samples  $\mathbf{x} \leftarrow \{0,1\}^{\lambda}$ . Then, it runs  $\mathbf{y} = \mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x})$  and sends  $\mathbf{y}$  to  $a_{\lambda}$ .  $a_{\lambda}$  succeeds if it outputs  $\tilde{\mathbf{x}}$  such that  $\mathbf{x} = \tilde{\mathbf{x}}$ . Otherwise, it fails.

**Game 1:** This game is the same as **Game 0** except that CH runs  $ZeroSamp(\lambda)$  instead of  $OneSamp(\lambda)$  in the key generation procedure.

**Lemma 8** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_0$  (resp.,  $\epsilon_1$ ) in **Game 0** (resp., **Game 1**), then  $|\epsilon_0 - \epsilon_1| = \mathsf{negl}(\lambda)$ .

*Proof.* We now construct  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$  that distinguishes  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ and  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  with advantage  $|\epsilon_0 - \epsilon_1|$ , which contradicts to Lemma 2.

 $b_{\lambda}$  takes as input **M**, which is generated as  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  or  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  from its challenger. Then, it samples  $\mathbf{x} \leftarrow \{0,1\}^{\lambda}$ . Next,  $b_{\lambda}$  runs  $\mathbf{y} = \mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x})$  and sends  $\mathbf{y}$  to  $a_{\lambda}$ . When  $a_{\lambda}$  outputs  $\tilde{\mathbf{x}}$ , if  $\mathbf{x} = \tilde{\mathbf{x}}$ ,  $b_{\lambda}$  outputs 1. Otherwise, it outputs 0.

Since all operations in  $b_{\lambda}$  are performed in  $\mathsf{NC}^1$ , we have  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$ .

One can see that when  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  (resp.,  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$ ), the view of  $a_{\lambda}$  is identical to its view in **Game 0** (resp., **Game 1**), i.e.,  $b_{\lambda}$ outputs 1 with probability  $\epsilon_0$  (resp.,  $\epsilon_1$ ). Therefore,  $\mathcal{B} = \{b_{\lambda}\}$  distinguishes  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  with advantage  $|\epsilon_0 - \epsilon_1|$ , which should be negligible according to Lemma 2, completing the proof of Lemma 8.

**Game 2:** This game is the same as **Game 1** except that  $a_{\lambda}$  succeeds if  $\mathbf{x} \neq \tilde{\mathbf{x}} \wedge \mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x}) = \mathsf{Eval}(\mathbf{M}, \tilde{\mathbf{x}})$ .

**Lemma 9** If  $A = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_1$  (resp.,  $\epsilon_2$ ) in **Game 1** (resp., **Game 2**), then  $\epsilon_1 = \epsilon_2$ .

*Proof.* According to Lemma 1 and due to the fact that the rank of  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  is  $\lambda - 1$ , for any  $\mathbf{y} \in \mathrm{Im}(\mathbf{M})$ , there are two vectors  $\mathbf{x}, \mathbf{x}'$  such

that  $\mathbf{M}\mathbf{x} = \mathbf{M}\mathbf{x}' = \mathbf{y} \land \mathbf{x} \neq \mathbf{x}'$ , and we have

$$\begin{aligned} \epsilon_{1} &= \Pr\left[\tilde{\mathbf{x}} = \mathbf{x}^{*} \begin{vmatrix} \mathbf{x}^{*} \leftarrow \{\mathbf{x}, \mathbf{x}'\} \\ \mathbf{y} = \mathbf{M}\mathbf{x}^{*} \\ \tilde{\mathbf{x}} \leftarrow a_{\lambda}(\mathbf{y}) \end{vmatrix} \right] \\ &= \frac{1}{2}\Pr\left[\tilde{\mathbf{x}} = \mathbf{x} \begin{vmatrix} \mathbf{x}^{*} = \mathbf{x} \\ \mathbf{y} = \mathbf{M}\mathbf{x}^{*} \\ \tilde{\mathbf{x}} \leftarrow a_{\lambda}(\mathbf{y}) \end{vmatrix} + \frac{1}{2}\Pr\left[\tilde{\mathbf{x}} = \mathbf{x}' \begin{vmatrix} \mathbf{x}^{*} = \mathbf{x}' \\ \mathbf{y} = \mathbf{M}\mathbf{x}^{*} \\ \tilde{\mathbf{x}} \leftarrow a_{\lambda}(\mathbf{y}) \end{vmatrix} \right] \\ &= \frac{1}{2}\Pr\left[\tilde{\mathbf{x}} = \mathbf{x} \begin{vmatrix} \mathbf{x}^{*} = \mathbf{x}' \\ \mathbf{y} = \mathbf{M}\mathbf{x}^{*} \\ \tilde{\mathbf{x}} \leftarrow a_{\lambda}(\mathbf{y}) \end{vmatrix} + \frac{1}{2}\Pr\left[\tilde{\mathbf{x}} = \mathbf{x}' \begin{vmatrix} \mathbf{x}^{*} = \mathbf{x}' \\ \mathbf{y} = \mathbf{M}\mathbf{x}^{*} \\ \tilde{\mathbf{x}} \leftarrow a_{\lambda}(\mathbf{y}) \end{vmatrix} \right] \\ &= \Pr\left[\frac{\tilde{\mathbf{x}} \neq \mathbf{x}^{*} \land}{\mathbf{E}\mathsf{val}_{\lambda}(\mathbf{M}, \tilde{\mathbf{x}}) = \mathsf{E}\mathsf{val}_{\lambda}(\mathbf{M}, \mathbf{x}^{*})} \begin{vmatrix} \mathbf{x}^{*} \leftarrow \{\mathbf{x}, \mathbf{x}'\} \\ \mathbf{y} = \mathbf{M}\mathbf{x}^{*} \\ \tilde{\mathbf{x}} \leftarrow a_{\lambda}(\mathbf{y}) \end{vmatrix} \right] = \epsilon_{2}, \end{aligned}$$

completing the proof of Lemma 9.

**Lemma 10** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_2$  in **Game 2**, then  $\epsilon_2 = \operatorname{negl}(\lambda)$ .

*Proof.* We now construct  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$  that distinguishes  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ and  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  with advantage  $\epsilon_2$ , which contradicts to Lemma 2.

 $b_{\lambda}$  takes as input  $\mathbf{M}$ , which is generated as  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  or  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  from its challenger. Then, it samples  $\mathbf{x} \leftarrow \{0,1\}^{\lambda}$ . Next,  $b_{\lambda}$  runs  $\mathbf{y} = \mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x})$  and send  $\mathbf{y}$  to  $a_{\lambda}$ . When  $a_{\lambda}$  outputs  $\tilde{\mathbf{x}}$ , if  $\mathbf{x} \neq \tilde{\mathbf{x}} \wedge \mathbf{y} = \mathsf{Eval}_{\lambda}(\mathbf{M}, \mathbf{x})$ ,  $b_{\lambda}$  outputs 1. Otherwise, it outputs 0.

Since all operations in  $b_{\lambda}$  are performed in NC<sup>1</sup>, we have  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$ .

One can see that when  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ , the view of  $a_{\lambda}$  is identical to its view in **Game 2**, i.e.,  $b_{\lambda}$  outputs 1 with probability  $\epsilon_2$ .

When  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$ , since  $\mathsf{Eval}(\mathbf{M}, \tilde{\mathbf{x}})$  is permutation, there is no vector  $\tilde{\mathbf{x}}$  such that  $\mathbf{x} \neq \tilde{\mathbf{x}} \land \mathbf{y} = \mathsf{Eval}_{\lambda}(\mathbf{k}, \mathbf{x})$ , i.e.  $b_{\lambda}$  outputs 1 with probability 0.

Therefore,  $\mathcal{B} = \{b_{\lambda}\}$  distinguishes  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  with advantage  $\epsilon_2$ , which should be negligible according to Lemma 2, completing the proof of Lemma 10.

Since  $|\epsilon_0 - \epsilon_1| = \mathsf{negl}(\lambda)$ ,  $\epsilon_1 = \epsilon_2$ , and  $\epsilon_2 = \mathsf{negl}(\lambda)$ , we have

$$\epsilon_0 \le |\epsilon_0 - \epsilon_1| + \epsilon_1 = \mathsf{negl}(\lambda) + \epsilon_2 = \mathsf{negl}(\lambda),$$

i.e., Construction 2 satisfies one-wayness. This completes the proof of Theorem 1.  $\hfill \Box$ 

**Extension to NC<sup>1</sup>-OWPs against NC<sup>1</sup>.** We now show a transformation from collections of NC<sup>1</sup>-OWPs against NC<sup>1</sup>, where the output distributions of the key generation algorithms are uniformly random over key space, to NC<sup>1</sup>-OWPs against NC<sup>1</sup>. Specifically, given a collection of OWPs  $\{f_k : D_k \to D_k\}_{k \in K}$  where K is the key space, we construct a OWP  $g : D \to D$  where  $D := \bigcup_{k \in K} (\{k\} \times D_k\}_{k \in K})$ 

 $D_k$  and  $g((k,x) \in D) = (k, f_k(x))$ . This transformation can be applied in  $NC^1$ , and the properties of permutation and one-wayness of g hold due to those properties of f. Note that in [5], it is shown that  $OneSamp(\lambda)$  samples  $\mathbf{M} \leftarrow$  $\{\mathbf{M} \in \mathsf{OneSamp}(\lambda)\}\$  uniformly. Thus,  $\mathsf{KeyGen}_{\lambda}$  of our construction samples  $k \leftarrow$  $K_{\lambda} = \{\mathbf{M} \mid \mathbf{M} \in \mathsf{OneSamp}(\lambda)\}$  uniformly, and we can apply this transformation to our collection of  $NC^1$ -OWPs against  $NC^1$ . We refer the reader to Section A in the supplementary material for the details.

**Computability in AC<sup>0</sup>**[2]. Perhaps interestingly, our one-way permutation can be run by an even smaller class of circuits  $AC^{0}[2]$ , which satisfies  $AC^{0}[2] \subseteq$ NC<sup>1</sup> [41, 43] and consists of constant-depth circuits with MOD2 gates. The reason is that it only involves multiplications of a constant number of matrices, inner products, and sampling random bits. Due to the same reason, our constructions of single-bit HPS introduced later in Section 4 is also computable in  $AC^{0}[2]$ .

#### Construction of NC<sup>1</sup>-HPS against NC<sup>1</sup> 4

In this section, we start by giving a construction of perfectly smooth and  $\frac{1}{2}$ universal<sub>1</sub>  $NC^1$ -HPS against  $NC^1$  such that the proof space is one-bit. Next, we turn this construction into a perfectly smooth and strong universal<sub>1</sub>  $NC^{1}$ -HPS against NC<sup>1</sup> such that the proof space is multi-bit. Finally, we construct a strong universal<sub>2</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup> such that the language L supports  $\{0,1\}^n$ .

#### 4.1Perfectly Smooth and Universal<sub>1</sub> for One-Bit

In this section, we give our construction of perfectly smooth and  $\frac{1}{2}$ -universal<sub>1</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup> circuits under the assumption NC<sup>1</sup>  $\subseteq \oplus L/poly$ .

**Construction 3** (NC<sup>1</sup>-HPS) Let  $\lambda$  be a security parameter. We define the families  $Setup = {Setup_{\lambda}}, Samp\mathcal{Y}es = {SampYes_{\lambda}}, Samp\mathcal{N}o = {SampNo_{\lambda}},$  $\mathcal{K}ey\mathcal{G}en = \{\mathsf{KeyGen}_{\lambda}\}, \ \mathcal{P}riv = \{\mathsf{Priv}_{\lambda}\}, \ and \ \mathcal{P}ub = \{\mathsf{Pub}_{\lambda}\} \ as \ follows.$ 

- Setup<sub>3</sub>:
  - 1. Sample  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ .
  - 2. Output  $pp = (X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, \Pi_{\lambda}, H_{\lambda}, \alpha_{\lambda}, aux_{\lambda})$  where
    - $X_{\lambda} := \{0, 1\}^{\lambda}$ .
    - $L_{\lambda} := \{ \mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{\lambda 1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w} \} = \operatorname{Im}(\mathbf{M}^{\top}) \ (\because Lemma \ 5).$   $W_{\lambda} := 1 \times \{0, 1\}^{\lambda 1}.$

    - $R_{\lambda} := \{(\mathbf{x}, \mathbf{w}) | \mathbf{w} \in 1 \times \{0, 1\}^{\lambda 1}, \mathbf{x} = \mathbf{M}\mathbf{w}\}.$
    - $SK_{\lambda} := \{0, 1\}^{\lambda}$ .
    - $PK_{\lambda} := \operatorname{Im}(\mathbf{M}).$
    - $\Pi_{\lambda} := \{0, 1\}.$
    - $H_{\lambda}(\mathbf{sk}, \mathbf{x}) := \mathbf{sk}^{\top} \mathbf{x}.$
    - $\alpha_{\lambda}(\mathbf{sk}) := \mathbf{M} \mathbf{sk}.$
    - $\operatorname{aux}_{\lambda} := \mathbf{M}.$
- SampYes<sub> $\lambda$ </sub>(pp):

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  - 1. Parse pp =  $(X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, \Pi_{\lambda}, H_{\lambda}, \alpha_{\lambda}, \mathsf{aux}_{\lambda})$  and let  $\mathsf{aux}_{\lambda} = \mathbf{M}$ .
  - 2. Sample  $\mathbf{w} \leftarrow 1 \times \{0, 1\}^{\lambda 1}$ .
  - 3. Compute  $\mathbf{x} := \mathbf{M}^{\top} \mathbf{w}$  and output  $\mathbf{x}$ .
  - SampNo<sub> $\lambda$ </sub>(pp):
    - 1. Parse pp =  $(X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, \Pi_{\lambda}, H_{\lambda}, \alpha_{\lambda}, \mathsf{aux}_{\lambda})$  and let  $\mathsf{aux}_{\lambda} = \mathbf{M}$ .
    - 2. Sample  $\mathbf{w} \leftarrow 1 \times \{0, 1\}^{\lambda 1}$ .
    - 3. Compute  $\mathbf{M}'$  as

$$\mathbf{M}' = \mathbf{M} + \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix}.$$

4. Compute  $\mathbf{x} := \mathbf{M}^{\prime \top} \mathbf{w}$  and output  $\mathbf{x}$ .

- KeyGen<sub>λ</sub>(pp):
  - 1. Parse  $pp = (X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, H_{\lambda}, \Pi_{\lambda}, \alpha_{\lambda}, \mathsf{aux}_{\lambda}).$
  - 2. Sample  $\mathbf{sk} \leftarrow SK_{\lambda}$ .
  - 3. Compute  $\mathbf{pk} := \alpha_{\lambda}(\mathbf{sk})$  and output  $(\mathbf{pk}, \mathbf{sk})$ .
- $\mathsf{Priv}_{\lambda}(\mathsf{pp}, \mathbf{sk}, \mathbf{x})$ :
  - 1. Parse  $pp = (X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, \Pi_{\lambda}, H_{\lambda}, \alpha_{\lambda}, \mathsf{aux}_{\lambda}).$
  - 2. Compute  $\pi := H_{\lambda}(\mathbf{sk}, \mathbf{x})$  and output  $\pi$ .
- $\mathsf{Pub}_{\lambda}(\mathsf{pp},\mathbf{pk},\mathbf{x},\mathbf{w}):$ 
  - 1. Compute  $\pi := \mathbf{p} \mathbf{k}^\top \mathbf{w}$  and output  $\pi$ .

**Theorem 2** If  $NC^1 \subsetneq \oplus L/poly$ , then (Setup, SampYes, SampNo, KeyGen, Priv, Pub) defined as Construction 3 is a perfectly smooth and  $\frac{1}{2}$ -universal<sub>1</sub>  $NC^1$ -HPS against  $NC^1$  circuits.

# **Proof sketch.** It is straightforward that this HPS is correct.

To show the subset membership problem of our construction, we first give two observations: (1) for any **M** sampled from  $ZeroSamp(\lambda)$ , the distribution of  $\mathbf{M} + \mathbf{N}$  is identical to  $OneSamp(\lambda)$ , where

$$\mathbf{N} = \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix},$$

and (2) perhaps interestingly, for any  $\mathbf{w} \in 0 \times \{0,1\}^{n-1}$  (respectively,  $\mathbf{w} \in 1 \times \{0,1\}^{n-1}$ ), there is a vector  $\mathbf{k}$  in the kernel of  $\mathbf{M}$  such that  $\mathbf{M}\mathbf{w}^{\top} = \mathbf{M}(\mathbf{w} + \mathbf{k})^{\top}$ and  $(\mathbf{w} + \mathbf{k}) \in 1 \times \{0,1\}^{n-1}$  (respectively,  $(\mathbf{w} + \mathbf{k}) \in 0 \times \{0,1\}^{n-1}$ ), which implies  $\mathrm{Im}(\mathbf{M}^{\top}) = \{\mathbf{x} | \mathbf{w} \in 0 \times \{0,1\}^{n-1}, \mathbf{x} = \mathbf{M}^{\top}\mathbf{w}\} = \{\mathbf{x} | \mathbf{w} \in 1 \times \{0,1\}^{n-1}, \mathbf{x} = \mathbf{M}^{\top}\mathbf{w}\}$ . Since for any vector  $\mathbf{w} \in 0 \times \{0,1\}^{n-1}$ , it holds that  $(\mathbf{M} + \mathbf{N})^{\top}\mathbf{w} = \mathbf{M}^{\top}\mathbf{w} + \mathbf{N}^{\top}\mathbf{w} = \mathbf{M}^{\top}\mathbf{w} + \mathbf{O} = \mathbf{M}^{\top}\mathbf{w}$ , we have

$$L = \operatorname{Im}(\mathbf{M}^{\top}) = {\mathbf{x} | \mathbf{w} \in 0 \times {\{0,1\}}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w}}$$

due to observation (2). We now switch the distribution of **M** from  $\text{ZeroSamp}(\lambda)$  to  $\text{OneSamp}(\lambda)$  by exploiting the indistinguishability between  $\text{ZeroSamp}(\lambda)$  and  $\text{OneSamp}(\lambda)$ . Then one can see that the language is now

$$L' = \{ \mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w} \}$$
$$= \{ \mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w} \},\$$

since  $(\mathbf{M} + \mathbf{N})^{\top} \in \mathsf{ZeroSamp}(\lambda)$  now. Next we return the distribution of  $\mathbf{M}$  to  $\mathsf{ZeroSamp}(\lambda)$  by exploiting the indistinguishability between  $\mathsf{ZeroSamp}(\lambda)$  and  $\mathsf{OneSamp}(\lambda)$  again, and the language becomes exactly  $X \setminus L$ . The reason is that when  $(\mathbf{M} + \mathbf{N})^{\top} \in \mathsf{ZeroSamp}(\lambda)$ , we have

$$X = \{0, 1\}^n = \{\mathbf{x} | \mathbf{w} \in \{0, 1\}^n, \mathbf{x} = (\mathbf{M} + \mathbf{N})^\top \mathbf{w}\}$$

since  $\mathbf{M} + \mathbf{N}$  is of full rank due to observation (1), i.e., we have

$$X \setminus L = \{\mathbf{x} | \mathbf{w} \in \{0, 1\}^n, \mathbf{x} = (\mathbf{M} + \mathbf{N})^\top \mathbf{w}\} \setminus \{\mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^\top \mathbf{w}\}$$
$$= \{\mathbf{x} | \mathbf{w} \in 1 \times \{0, 1\}^{n-1}, \mathbf{x} = (\mathbf{M} + \mathbf{N})^\top \mathbf{w}\}.$$

As a result, L and  $X \setminus L$  are indistinguishable in NC<sup>1</sup>.

We now explain the intuition of the proof of universal<sub>1</sub>. Since the rank of **M** is n - 1, when we fix the public key **pk**, there are two different secret keys **sk** and **sk'** such that  $\mathbf{pk} = \mathbf{M} \mathbf{sk} = \mathbf{M} \mathbf{sk'}$ . As explained before, for any  $\mathbf{x} \in X \setminus L$ , there exists  $\mathbf{w} \in 1 \times \{0,1\}^{n-1}$  such that  $\mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top}\mathbf{w}$ , and  $(\mathbf{M} + \mathbf{N})$  is a full rank matrix. Therefore, we have  $(\mathbf{M} + \mathbf{N})\mathbf{sk} \neq (\mathbf{M} + \mathbf{N})\mathbf{sk'}$  which implies  $\mathbf{N} \mathbf{sk} \neq \mathbf{N} \mathbf{sk'}$ , i.e., either  $\mathbf{N} \mathbf{sk}$  or  $\mathbf{N} \mathbf{sk'}$  is zero-vector and the other is  $(1 \ 0 \cdots 0)^{\top}$ . Therefore, when we let  $\mathbf{N} \mathbf{sk} = (0 \cdots 0)^{\top}$  and  $\mathbf{N} \mathbf{sk'} = (1 \ 0 \cdots 0)^{\top}$ , it holds that  $H(\mathbf{sk}, \mathbf{x}) = \mathbf{sk}^{\top}(\mathbf{M} + \mathbf{N})^{\top}\mathbf{w} = \mathbf{sk}^{\top}\mathbf{M}^{\top}\mathbf{w} + (0 \cdots 0)\mathbf{w} = \mathbf{sk}^{\top}\mathbf{M}^{\top}\mathbf{w}$  and  $H(\mathbf{sk'}, \mathbf{x}) = \mathbf{sk'}^{\top}(\mathbf{M} + \mathbf{N})^{\top}\mathbf{w} = \mathbf{sk'}^{\top}\mathbf{M}^{\top}\mathbf{w} + (1 \ 0 \cdots 0)\mathbf{w} = \mathbf{sk'}^{\top}\mathbf{M}^{\top}\mathbf{w} + 1$ , which implies  $H(\mathbf{sk}, \mathbf{x}) \neq H(\mathbf{sk'}, \mathbf{x})$ . As a result, for fixed **pk**, one can guess the proof for an instance  $\mathbf{x} \in X \setminus L$  with probability at most  $\frac{1}{2}$  since there is no information on whether the secret key is **sk** or **sk'**.

The formal proof is as follows.

*Proof.* First note that all of the algorithms Setup,  $Samp\mathcal{Y}es$ ,  $Samp\mathcal{N}o$ ,  $Key\mathcal{G}en$ ,  $\mathcal{P}riv$ , and  $\mathcal{P}ub$  are in  $\mathsf{NC}^1$ , since they only involve operations including multiplications of a constant number of matrices, inner products, and sampling random bits.

Next we prove that Construction 3 satisfies correctness, subset membership problem, perfect smoothness, and  $\frac{1}{2}$ -universality<sub>1</sub>.

**Correctness.** Since  $\mathsf{Priv}_{\lambda}(\mathsf{pp}, \mathbf{sk}, \mathbf{x}) = H_{\lambda}(\mathbf{sk}, \mathbf{x}) = \mathbf{sk}^{\top}\mathbf{x} = \mathbf{sk}^{\top}\mathbf{M}^{\top}\mathbf{w} = (\mathbf{Msk})^{\top}\mathbf{w} = \mathbf{pk}^{\top}\mathbf{w} = \mathsf{Pub}_{\lambda}(\mathsf{pp}, \mathbf{pk}, \mathbf{x}, \mathbf{w})$ , Construction 3 satisfies correctness.

**Subset membership problem.** We now propose and prove three propositions corresponding to the three properties in the definition of subset membership problem (see Definition 7) respectively.

**Proposition 1** The distributions of  $\mathbf{x}$  and  $\mathbf{x}'$  are identical where  $pp \leftarrow \mathsf{Setup}_{\lambda}$ ,  $\mathbf{x} \leftarrow \mathsf{SampYes}_{\lambda}(pp)$ , and  $\mathbf{x}' \leftarrow L_{\lambda}$ .

*Proof.* Let  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  be a matrix generated in the procedure of  $\mathsf{Setup}_{\lambda}$ . Let f be a map  $f: 1 \times \{0, 1\}^{\lambda-1} \to \operatorname{Im}(\mathbf{M}^{\top})$  such that  $f(\mathbf{w}) = \mathbf{M}^{\top}\mathbf{w}$ . One can see that for any  $\mathsf{pp} \leftarrow \mathsf{Setup}_{\lambda}$ , the distributions of  $\mathbf{x}$  and  $\mathbf{x}'$  are identical where  $\mathbf{x} \leftarrow \mathsf{SampYes}_{\lambda}(\mathsf{pp}), \mathbf{w}' \leftarrow 1 \times \{0, 1\}^{\lambda-1}$ , and  $\mathbf{x}' = f(\mathbf{w}')$ . Moreover, if f is bijective, the distributions of  $\mathbf{x}'$  and  $\mathbf{x}''$  are identical for  $\mathbf{w}' \leftarrow 1 \times \{0, 1\}^{\lambda-1}$ ,  $\mathbf{x}' = f(\mathbf{w}')$ , and  $\mathbf{x}'' \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$ . Therefore, if f is bijective, the distributions of  $\mathbf{x}$  and  $\mathbf{x}''$  are identical. Namely, to show Proposition 1, we only have to show that f is bijective.

**Injectivity.** We now show that for any  $\mathbf{w}, \mathbf{w}' \leftarrow 1 \times \{0, 1\}^{\lambda-1}$  such that  $\mathbf{w} \neq \mathbf{w}'$ , we have  $f(\mathbf{w}) \neq f(\mathbf{w}')$ . We prove by contradiction, i.e. we show that if there are  $\mathbf{w}, \mathbf{w}' \leftarrow 1 \times \{0, 1\}^{\lambda-1}$  such that  $\mathbf{w} \neq \mathbf{w}'$  and  $f(\mathbf{w}) = f(\mathbf{w}')$ , then it contradicts on Lemma 4.

Since  $\mathbf{M}^{\top}\mathbf{w} = \mathbf{M}^{\top}\mathbf{w}'$ , we have  $\mathbf{M}^{\top}(\mathbf{w} - \mathbf{w}') = \mathbf{0}$ . Moreover, since  $\mathbf{w} \neq \mathbf{w}'$ and  $\mathbf{w}, \mathbf{w}' \in 1 \times \{0, 1\}^{\lambda-1}, \mathbf{w} - \mathbf{w}'$  is the non-zero vector in the kernel of  $\mathbf{M}^{\top}$  and  $\mathbf{w} - \mathbf{w}' \in 0 \times \{0, 1\}^{\lambda-1}$ . However, according to Lemma 4, we have  $\operatorname{Ker}(\mathbf{M}^{\top}) = \{\mathbf{0}, \mathbf{k}\}$  where  $\mathbf{k} \in 1 \times \{0, 1\}^{\lambda}$ , which gives us the conflict.

Surjectivity. We now show that for any  $\mathbf{x} \in \text{Im}(\mathbf{M}^{\top})$ , there exists a vector  $\mathbf{w} \in 1 \times \{0,1\}^{\lambda-1}$  such that  $\mathbf{x} = f(\mathbf{w})$ , i.e.,  $\mathbf{x} = \mathbf{M}^{\top}\mathbf{w}$ . According to Lemma 5, we have  $\text{Im}(\mathbf{M}^{\top}) = \{\mathbf{x} | \mathbf{w} \in 1 \times \{0,1\}^{\lambda}, \mathbf{x} = \mathbf{M}^{\top}\mathbf{w}\}$ . Therefore, it holds that for any  $\mathbf{x} \in \text{Im}(\mathbf{M}^{\top})$ , there exists  $\mathbf{w} \in 1 \times \{0,1\}^{\lambda}$  such that  $\mathbf{x} = \mathbf{M}^{\top}\mathbf{w}$ , i.e.,  $\mathbf{x} = f(\mathbf{w})$ , completing the proof of surjectivity.

Putting all the above together, Proposition 1 immediately follows.  $\Box$ 

**Proposition 2** The distributions of  $\mathbf{x}$  and  $\mathbf{x}'$  are identical for  $pp \leftarrow \mathsf{Setup}_{\lambda}$ ,  $\mathbf{x} \leftarrow \mathsf{SampNo}_{\lambda}(pp)$ , and  $\mathbf{x}' \leftarrow X_{\lambda} \setminus L_{\lambda}$ .

*Proof.* Let  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and  $\mathbf{M}'$  be

$$\mathbf{M}' = \mathbf{M} + \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix}.$$

We first show that for any  $\mathbf{w} \in 1 \times \{0, 1\}^{\lambda-1}$ , we have  $\mathbf{M}^{\prime \top} \mathbf{w} \in \{0, 1\}^{\lambda} \setminus \mathrm{Im}(\mathbf{M}^{\top})$ .

 $(\bigstar)$  For any  $\mathbf{w} \in 0 \times \{0,1\}^{\lambda-1}$ , we have

$$\mathbf{M}'\mathbf{w} = \left(\mathbf{M} + \begin{pmatrix} 0 \cdots & 0 & 1\\ 0 & 0 & \cdots & 0\\ \vdots & \ddots & \vdots\\ 0 \cdots & 0 & 0 \end{pmatrix}\right)^{\top} \mathbf{w}$$
$$= \mathbf{M}^{\top}\mathbf{w} + \begin{pmatrix} 0 \cdots & 0\\ \vdots & 0 & \cdots & 0\\ 0 & \ddots & \vdots\\ 1 & 0 & \cdots & 0 \end{pmatrix} \begin{pmatrix} 0\\ w_{2}\\ \vdots\\ w_{\lambda} \end{pmatrix} = \mathbf{M}^{\top}\mathbf{w} + \mathbf{0} = \mathbf{M}^{\top}\mathbf{w}$$

Moreover, according to Lemma 5, we have  $\operatorname{Im}(\mathbf{M}^{\top}) = {\mathbf{x} | \mathbf{w} \in 0 \times \{0, 1\}^{\lambda - 1}, \mathbf{x} = \mathbf{M}^{\top} \mathbf{w}}.$  Hence, we have

$$\{ \mathbf{x} \mid \mathbf{w} \in 0 \times \{0,1\}^{\lambda-1}, \ \mathbf{x} = \mathbf{M}^{\prime \top} \mathbf{w} \} = \{ \mathbf{x} \mid \mathbf{w} \in 0 \times \{0,1\}^{\lambda-1}, \ \mathbf{x} = \mathbf{M}^{\top} \mathbf{w} \} \ (\bigstar)$$
$$= \operatorname{Im}(\mathbf{M}^{\top}).$$

As a result, for all  $\mathbf{x} \in \operatorname{Im}(\mathbf{M}^{\top})$ , there exists  $\mathbf{w}' \in 0 \times \{0,1\}^{\lambda-1}$  such that  $\mathbf{x} = \mathbf{M}'^{\top}\mathbf{w}'$ . Moreover, according to Lemma 7,  $\mathbf{M}'$  is a full rank matrix, which means that for any  $\mathbf{w} \in 1 \times \{0,1\}^{\lambda-1}$  and any  $\mathbf{x} \in \operatorname{Im}(\mathbf{M}^{\top})$ , we have  $\mathbf{M}'^{\top}\mathbf{w} \neq \mathbf{x}$ . Namely, for any  $\mathbf{w} \in 1 \times \{0,1\}^{\lambda-1}$ , we have  $\mathbf{M}'^{\top}\mathbf{w} \in \{0,1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$ .

It is straightforward that for any  $\mathbf{pp} \leftarrow \mathsf{Setup}_{\lambda}$ , the distributions of  $\mathbf{x} \leftarrow \mathsf{SampNo}_{\lambda}(\mathbf{pp})$  and  $\mathbf{x}' = \mathbf{M}'^{\top}\mathbf{w}'$  are identical where  $\mathbf{w}' \leftarrow 1 \times \{0, 1\}^{\lambda-1}$ . Moreover, since  $\mathbf{M}'^{\top}$  is of full rank, the map  $f : 1 \times \{0, 1\}^{n-1} \rightarrow \{0, 1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$  such that  $f(\mathbf{w}) = \mathbf{M}'^{\top}\mathbf{w}$  is bijective, i.e., the distributions of  $\mathbf{x}' = f(\mathbf{w}')$  and  $\mathbf{x}'' \leftarrow \{0, 1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$  are identical for  $\mathbf{w}' \leftarrow 1 \times \{0, 1\}^{\lambda-1}$ , completing the proof of Proposition 2.

**Proposition 3** For any  $\mathcal{A} = \{a_{\lambda}\} \in \mathsf{NC}^1$ ,

$$|\Pr[a_{\lambda}(\mathsf{pp}, \mathbf{x}_0) = 1] - \Pr[a_{\lambda}(\mathsf{pp}, \mathbf{x}_1) = 1]| \le \mathsf{negl}(\lambda)$$

where  $pp \leftarrow Setup_{\lambda}$ ,  $(\mathbf{x}_0, \mathbf{w}) \leftarrow SampYes_{\lambda}(pp)$ , and  $\mathbf{x}_1 \leftarrow SampNo_{\lambda}(pp)$ .

*Proof.* Let  $\mathcal{A} = \{a_{\lambda}\}$  be any adversary in NC<sup>1</sup> and N be the following  $\lambda \times \lambda$  matrix.

$$\mathbf{N} = \begin{pmatrix} 0 & \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & 0 \end{pmatrix}.$$

We give hybrid games to show that the advantage of  $\mathcal{A}$  in breaking the hardness of subset membership problem is negligible.

**Game 0:** This is the original SampYes game for  $\mathcal{A}$ .  $\mathcal{CH}$  runs  $pp \leftarrow Setup_{\lambda}$ ,  $(\mathbf{x}, \mathbf{w}) \leftarrow SampYes_{\lambda}(pp)$ . Then it sends  $(pp, \mathbf{x})$  to  $a_{\lambda}$ .  $a_{\lambda}$  succeeds if  $a_{\lambda}$  outputs 1. Otherwise, it fails.

**Game 1:** This game is the same as **Game 0** except that  $\mathcal{CH}$  samples  $\mathbf{w} \leftarrow 0 \times \{0,1\}^{\lambda-1}$  in the procedure of  $\mathsf{SampYes}_{\lambda}(\mathsf{pp})$ .

**Lemma 11** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_0$  (resp.,  $\epsilon_1$ ) in **Game 0** (resp., **Game 1**), then  $\epsilon_0 = \epsilon_1$ .

*Proof.* Lemma 11 follows from the fact that the distributions of  $\mathbf{x}$  and  $\mathbf{x}'$  are identical where  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}_{\lambda}$ ,  $\mathbf{w} \leftarrow 1 \times \{0,1\}^{\lambda-1}$ ,  $\mathbf{x} = \mathbf{M}^{\top}\mathbf{w}$ ,  $\mathbf{w}' \leftarrow 0 \times \{0,1\}^{\lambda-1}$ , and  $\mathbf{x}' = \mathbf{M}^{\top}\mathbf{w}'$  (according to Lemma 6).

**Game 2:** This game is the same as **Game 1** except that CH computes  $\mathbf{x} := (\mathbf{M} + \mathbf{N})^\top \mathbf{w}$  in the procedure of SampYes<sub> $\lambda$ </sub>(pp).

**Lemma 12** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_1$  (resp.,  $\epsilon_2$ ) in **Game 1** (resp., **Game 2**), then  $\epsilon_1 = \epsilon_2$ .

*Proof.* Lemma 12 follows from the fact that for  $\mathbf{w} \in 0 \times \{0, 1\}^{\lambda-1}$ , we have

$$\mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w} = \mathbf{M}^{\top} \mathbf{w} + \mathbf{N}^{\top} \mathbf{w} = \mathbf{M}^{\top} \mathbf{w}.$$

**Game 3:** This game is the same as **Game 2** except that CH runs  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  in the procedure of  $\mathsf{Setup}_{\lambda}$ .

**Lemma 13** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_2$  (resp.,  $\epsilon_3$ ) in **Game 2** (resp., **Game 3**), then  $|\epsilon_2 - \epsilon_3| = \mathsf{negl}(\lambda)$ .

*Proof.* We now construct  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$  that distinguishes  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ and  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  with advantage  $|\epsilon_2 - \epsilon_3|$ , which contradicts to Lemma 2.

 $b_{\lambda}$  takes as input **M**, which is generated as  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  or  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  from its challenger. Then, it runs  $\mathsf{pp} \leftarrow \mathsf{Setup}_{\lambda}$  using **M**, samples  $\mathbf{w} \leftarrow 0 \times \{0,1\}^{\lambda-1}$ , and sets  $x := (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w}$ . Next,  $b_{\lambda}$  gives  $(\mathsf{pp}, \mathbf{x})$  to  $a_{\lambda}$ . When  $a_{\lambda}$  outputs b, then  $b_{\lambda}$  outputs b.

Since all operations in  $b_{\lambda}$  are performed in NC<sup>1</sup>, we have  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$ .

One can see that when  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  (resp.,  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$ ), the view of  $a_{\lambda}$  is identical to its view in **Game 2** (resp., **Game 3**), i.e.,  $b_{\lambda}$ outputs 1 with probability  $\epsilon_2$  (resp.,  $\epsilon_3$ ). Therefore,  $\mathcal{B} = \{b_{\lambda}\}$  distinguishes  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$  with advantage  $|\epsilon_2 - \epsilon_3|$ , which should be negligible according to Lemma 2, completing the proof of Lemma 13. **Game 4:** This game is the same as **Game 3** except that  $\mathcal{CH}$  samples  $\mathbf{w} \leftarrow 1 \times \{0,1\}^{\lambda-1}$  in the procedure of  $\mathsf{SampYes}_{\lambda}(\mathsf{pp})$ .

**Lemma 14** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_3$  (resp.,  $\epsilon_4$ ) in **Game 3** (resp., **Game 4**), then  $\epsilon_3 = \epsilon_4$ .

*Proof.* Let  $\mathbf{M} \leftarrow \mathsf{OneSamp}(\lambda)$ . According to Lemma 7, we have  $\mathbf{M} + \mathbf{N} \in \mathsf{ZeroSamp}(\lambda)$ . Therefore, according to Lemma 6, the distributions of  $\mathbf{x}$  and  $\mathbf{x}'$  are identical where  $\mathbf{w} \leftarrow 1 \times \{0,1\}^{\lambda-1}$ ,  $\mathbf{x} = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w}$ ,  $\mathbf{w}' \leftarrow 0 \times \{0,1\}^{\lambda-1}$ , and  $\mathbf{x}' = (\mathbf{M} + \mathbf{N})^{\top} \mathbf{w}'$ , completing the proof of Lemma 14.

**Game 5:** This game is the same as **Game 4** except that CH runs  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  in the procedure of  $\mathsf{Setup}_{\lambda}$ , which is identical to the original SampNo game.

**Lemma 15** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_4$  (resp.,  $\epsilon_5$ ) in **Game 4** (resp., **Game 5**), then  $|\epsilon_4 - \epsilon_5| = \mathsf{negl}(\lambda)$ .

*Proof.* We can prove Lemma 15 in the same way as Lemma 11.

Note that,  $\epsilon_0 = \Pr[a_\lambda(\mathsf{pp}, \mathbf{x}) = 1]$  and  $\epsilon_5 = \Pr[a_\lambda(\mathsf{pp}, \mathbf{x}') = 1]$  where  $\mathsf{pp} \leftarrow \mathsf{Setup}_\lambda$ ,  $(\mathbf{x}, \mathbf{w}) \leftarrow \mathsf{SampYes}_\lambda(\mathsf{pp})$ , and  $\mathbf{x}' \leftarrow \mathsf{SampNo}_\lambda(\mathsf{pp})$ . Moreover, since  $\epsilon_0 = \epsilon_1 = \epsilon_2$ ,  $|\epsilon_2 - \epsilon_3| = \mathsf{negl}(\lambda)$ ,  $\epsilon_3 = \epsilon_4$ , and  $|\epsilon_4 - \epsilon_5| = \mathsf{negl}(\lambda)$  we have

$$|\epsilon_0 - \epsilon_5| = |\epsilon_2 - \epsilon_5| \le |\epsilon_2 - \epsilon_3| + |\epsilon_3 - \epsilon_4| + |\epsilon_4 - \epsilon_5| = \mathsf{negl}(\lambda).$$

According to Proposition 1, 2, and 3, Construction 3 satisfies the subset membership problem, completing this part of proof.

**Perfect smoothness.** We now show that for any  $pp \leftarrow \mathsf{Setup}_{\lambda}$ , the random variables  $(\mathbf{x}, \mathbf{pk}, \pi)$  and  $(\mathbf{x}, \mathbf{pk}, \pi')$  are identical where  $\mathbf{x} \leftarrow X_{\lambda} \setminus L_{\lambda}$ ,  $(\mathbf{pk}, \mathbf{sk}) \leftarrow \mathsf{KeyGen}_{\lambda}(\mathsf{pp})$ , and  $\pi' \leftarrow \Pi_{\lambda}$ .

According to Lemma 1, for any  $\mathbf{pk}^* \in PK_{\lambda}$ , there are only two secret keys  $\mathbf{sk}$  and  $\mathbf{sk}'$  such that  $\mathbf{pk}^* = \alpha_{\lambda}(\mathbf{sk}) = \mathbf{M} \mathbf{sk} = \alpha_{\lambda}(\mathbf{sk}') = \mathbf{M} \mathbf{sk}'$ . Moreover, according to Lemma 3, we have  $\mathbf{sk} = \mathbf{sk}' + \mathbf{k}$  where  $\mathbf{k}$  is a vector such that  $\mathbf{k} \in \text{Ker}(\mathbf{M})$  and  $\mathbf{k} \in \{0, 1\}^{\lambda-1} \times 1$ , i.e., the last elements in  $\mathbf{sk}$  and  $\mathbf{sk}'$  are different (one is 1 and other is 0). Therefore, for any  $\mathbf{x}^* \in X_{\lambda} \setminus L_{\lambda}$  and  $\mathbf{pk}^* \in PK_{\lambda}$ , we

have

 $\pi$ 

$$= \operatorname{Priv}_{\lambda}(\operatorname{pp}, \operatorname{sk}, \mathbf{x}^{*}) = \operatorname{sk}^{\top} \mathbf{M}'^{\top} \mathbf{w}^{*}$$
$$= \operatorname{sk}^{\top} \left( \mathbf{M} + \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix} \right)^{\top} \mathbf{w}^{*}$$
$$= \operatorname{sk}^{\top} \mathbf{M}^{\top} \mathbf{w}^{*} + (sk_{1} \cdots sk_{\lambda}) \begin{pmatrix} 0 \cdots & 0 \\ \vdots & 0 & \cdots & 0 \\ 0 & \ddots & \vdots \\ 1 & 0 & \cdots & 0 \end{pmatrix} \begin{pmatrix} 1 \\ w_{2}^{*} \\ \vdots \\ w_{\lambda}^{*} \end{pmatrix}$$
$$= \operatorname{pk}^{*^{\top}} \mathbf{w}^{*} + sk_{\lambda},$$

and it follows that for any  $\mathbf{x}^* \in X_{\lambda} \setminus L_{\lambda}$  and  $\mathbf{pk}^* \in PK_{\lambda}$ , there are two secret key  $\mathbf{sk}, \mathbf{sk}'$  such that  $\mathbf{pk}^* = \alpha_{\lambda}(\mathbf{sk}) = \alpha_{\lambda}(\mathbf{sk}')$ ,  $\mathsf{Priv}_{\lambda}(\mathbf{sk}, \mathbf{x}^*) = 0$ , and  $\mathsf{Priv}_{\lambda}(\mathsf{pp}, \mathbf{sk}', \mathbf{x}^*) = 1$ . Namely, the number of secret keys satisfying  $\mathbf{pk}^* = \alpha_{\lambda}(\mathbf{sk}) \land \pi^* = \mathsf{Priv}_{\lambda}(\mathbf{sk}, \mathbf{x}^*)$  is 1. Therefore, we have

$$\Pr[(\mathbf{x}, \mathbf{pk}, \pi) = (\mathbf{x}^*, \mathbf{pk}^*, \pi^*)] = \Pr\left[\frac{\mathbf{pk} = \mathbf{pk}^* \land}{\pi = \pi^*} \middle| \mathbf{x} = \mathbf{x}^*\right] \Pr[\mathbf{x} = \mathbf{x}^*]$$
$$= \Pr\left[\frac{\mathbf{pk}^* = \alpha_\lambda(\mathbf{sk}) \land}{\pi^* = \mathsf{Priv}_\lambda(\mathsf{pp}, \mathbf{sk}, \mathbf{x}^*)}\right] \Pr[\mathbf{x} = \mathbf{x}^*]$$
$$= \frac{1}{|SK_\lambda|} \Pr[\mathbf{x} = \mathbf{x}^*]$$

where  $\mathbf{sk} \leftarrow SK_{\lambda}$  and  $\mathbf{x} \leftarrow X_{\lambda} \times L_{\lambda}$ . Similarly, we have

$$\begin{aligned} \Pr[(\mathbf{x}, \mathbf{pk}, \pi') &= (\mathbf{x}^*, \mathbf{pk}^*, \pi^*)] = \Pr[\pi' = \pi^*] \Pr[\mathbf{pk} = \mathbf{pk}^*] \Pr[\mathbf{x} = \mathbf{x}^*] \\ &= \frac{1}{2} \frac{2}{|SK_{\lambda}|} \Pr[\mathbf{x} = \mathbf{x}^*] \\ &= \frac{1}{|SK_{\lambda}|} \Pr[\mathbf{x} = \mathbf{x}^*]. \end{aligned}$$

Therefore, we have  $\Pr[(\mathbf{x}, \mathbf{pk}, \pi) = (\mathbf{x}^*, \mathbf{pk}^*, \pi^*)] = \Pr[(\mathbf{x}, \mathbf{pk}, \pi') = (\mathbf{x}^*, \mathbf{pk}^*, \pi^*)]$  and it follows that Construction 3 satisfies perfect smoothness.

 $\frac{1}{2}$ -universality<sub>1</sub>.  $\frac{1}{2}$ -universality<sub>1</sub> follows from the fact that for any  $\mathbf{pp} \leftarrow \mathsf{Setup}_{\lambda}$ ,  $\mathbf{x} \in X_{\lambda} \setminus L_{\lambda}$ ,  $\mathbf{pk} \in PK_{\lambda}$ , and  $\pi \in \Pi_{\lambda}$ , the number of secret keys such that  $\mathbf{pk} = \alpha_{\lambda}(\mathbf{sk})$  is 2 and the number of secret keys such that  $\mathbf{pk} = \alpha_{\lambda}(\mathbf{sk}) \land \pi =$   $\mathsf{Priv}_{\lambda}(\mathsf{pp}, \mathbf{sk}, \mathbf{x})$  is 1 as described above. Therefore, we have

$$\begin{aligned} \Pr[\mathsf{Priv}_{\lambda}(\mathsf{pp}, \mathbf{sk}, \mathbf{x}) &= \pi \wedge \alpha_{\lambda}(\mathbf{sk}) = \mathbf{pk}] &= \frac{1}{|SK_{\lambda}|} = \frac{1}{2} \frac{2}{|SK_{\lambda}|} \\ &= \frac{1}{2} \Pr[\alpha_{\lambda}(\mathbf{sk}) = \mathbf{pk}] \\ &\Leftrightarrow \Pr[\mathsf{Priv}_{\lambda}(\mathsf{pp}, \mathbf{sk}, \mathbf{x}) = \pi | \alpha_{\lambda}(\mathbf{sk}) = \mathbf{pk}] = \frac{1}{2}. \end{aligned}$$

Therefore, Construction 3 satisfies  $\frac{1}{2}$ -universality<sub>1</sub>.

Putting all the above together, Theorem 2 immediately follows.

Multi-bit NC<sup>1</sup>-HPS. Notice that the size of proof space of Construction 3 is only one-bit, which makes it less useful. However, we can extend this construction with multi-bit proofs by running multiple HPS in parallel. We refer the reader to Section B in the supplementary material for the multi-bit version of our HPS and the security proof.

**Universal**<sub>2</sub> NC<sup>1</sup>-HPS. By carefully adopting the technique by Cramer and Shoup [16], we achieve a universal<sub>2</sub> NC<sup>1</sup>-HPS. The resulting scheme can be computed in NC<sup>1</sup> and it is secure against NC<sup>1</sup> circuits under the assumption NC<sup>1</sup>  $\subseteq \oplus L/poly$ . We refer the reader to Section C in the supplementary material for the details.

# 4.2 Application: NC<sup>1</sup>-CCA Secure PKE

As one of the most important application of HPSs, Cramer and shoup [16] constructed a CCA secure PKE scheme. Interestingly, by instantiating the underlying HPS with our construction, we immediately achieve an  $NC^1$ -CCA secure PKE scheme against  $NC^1$  circuits restricted in the same way as the ones defined for verifiable computation schemes by Campanelli and Gennaro [14], i.e., ones allowed to make constant rounds of adaptive queries to the decryption oracle, while in each round, they can make arbitrary polynomial number of queries. We refer the reader to Section D in the supplementary material for the details on this application.

# 5 Construction of NC<sup>1</sup>-TDF against NC<sup>1</sup>

In this section, we give our construction of  $NC^1$ -TDF against  $NC^1$  under the assumption  $NC^1 \subsetneq \oplus L/poly$ .

**Construction 4** (NC<sup>1</sup>-**TDF**) Let  $\lambda$  be a security parameter and l be a polynomial in  $\lambda$ . Let  $\mathcal{F} = \{f_{\lambda} : \{0,1\}^{\lambda} \to \{0,1\}^{l(\lambda)}\}$  be a NC<sup>1</sup>-OWF against NC<sup>1</sup>. We define the families  $\mathcal{K}ey\mathcal{G}en = \{\mathsf{KeyGen}_{\lambda}\}$  with key spaces  $EK_{\lambda} = \{\mathbf{M} \mid \mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)\}$  and  $TK_{\lambda} = \mathrm{Ker}(\mathbf{M})$ ,  $\mathcal{E}val = \{\mathsf{Eval}_{\lambda}\}$  and  $\mathcal{I}nverse = \{\mathsf{Inverse}_{\lambda}\}$  as follows.

- KeyGen<sub> $\lambda$ </sub>:
  - 1. Run  $\mathbf{R} \leftarrow \mathsf{LSamp}(\lambda), \mathbf{R}' \leftarrow \mathsf{RSamp}(\lambda).$
  - 2. Set  $\mathbf{k} := (\mathbf{r} \ 1)^{\top}$  where  $(\mathbf{r} \ 1)^{\top}$  is the last column of  $\mathbf{R}'$ .
  - 3. Compute  $\mathbf{M} := \mathbf{R}\mathbf{M}_0^{\lambda}\mathbf{R}'$  where  $\mathbf{M}_0^{\lambda}$  is defined as Construction 1.
  - Set ek := M and tk := k, and output (ek, tk) (according to the proof of Lemma 3, it holds that k ∈ Ker(M) ).

The domain  $\mathcal{D}_{\lambda,ek}$  is defined as follows.

$$\mathcal{D}_{\lambda,ek} := \{0,1\}^{\lambda} \times \left( \operatorname{Im}(\mathbf{M}^{\top}) \times \{0,1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top}) \right)^{\lambda}.$$

The range  $\mathcal{R}_{\lambda,ek}$  is defined as follows.

$$\mathcal{R}_{\lambda,ek} := \{ Y \mid \exists X \in \mathcal{D}_{\lambda,ek} \ s.t. \ Y = \mathsf{Eval}_{\lambda}(ek, X) \}.$$

- $\mathsf{Eval}_{\lambda}(ek, X)$ :
  - 1. Parse  $X := (x, (\mathbf{c}_{1,0}, \mathbf{c}_{1,1}), (\mathbf{c}_{2,0}, \mathbf{c}_{2,1}), \cdots, (\mathbf{c}_{\lambda,0}, \mathbf{c}_{\lambda,1})) \in \mathcal{D}_{\lambda,ek}.$
  - 2. For  $x = x_1 x_2 \cdots x_\lambda \in \{0, 1\}^\lambda$  and all  $i \in [\lambda]$ , if  $x_i = 0$ , set  $(\mathbf{c}_i, \mathbf{c}'_i) := (\mathbf{c}_{i,0}, \mathbf{c}_{i,1})$ , otherwise set  $(\mathbf{c}_i, \mathbf{c}'_i) := (\mathbf{c}_{i,1}, \mathbf{c}_{i,0})$ .
  - 3. Compute  $y := f_{\lambda}(x)$ .
  - 4. Set  $Y := (y, (\mathbf{c}_1, \mathbf{c}'_1), (\mathbf{c}_2, \mathbf{c}'_2), \cdots, (\mathbf{c}_{\lambda}, \mathbf{c}'_{\lambda}))$  and output Y.
- Inverse<sub> $\lambda$ </sub>(tk, Y):
  - 1. Parse  $tk := \mathbf{k}$  and  $Y := (y, (\mathbf{c}_1, \mathbf{c}'_1), (\mathbf{c}_2, \mathbf{c}'_2), \cdots, (\mathbf{c}_{\lambda}, \mathbf{c}'_{\lambda})) \in \mathcal{R}_{\lambda, ek}$ .
  - 2. For all  $i \in [\lambda]$ , if  $\mathbf{k}^{\top} \mathbf{c}_i = 0 \land \mathbf{k}^{\top} \mathbf{c}'_i = 1$ , then set  $x_i := 0$  and  $(\mathbf{c}_{i,0}, \mathbf{c}_{i,1}) := (\mathbf{c}_i, \mathbf{c}'_i)$ .
  - 3. Else if  $\mathbf{k}^{\top}\mathbf{c}_i = 1 \wedge \mathbf{k}^{\top}\mathbf{c}'_i = 0$ , then set  $x_i := 1$  and  $(\mathbf{c}_{i,0}, \mathbf{c}_{i,1}) := (\mathbf{c}'_i, \mathbf{c}_i)$ .
  - 4. Else output  $\perp$  and halt.
  - 5. Set  $X = (x, (\mathbf{c}_{1,0}, \mathbf{c}_{1,1}), \cdots, (\mathbf{c}_{\lambda,0}, \mathbf{c}_{\lambda,1}))$  and output X.

**Theorem 3** If  $NC^1 \subsetneq \oplus L/poly$  and there exists an  $NC^1$ -OWF against  $NC^1$  circuits, Construction 4 is an  $NC^1$ -TDF against  $NC^1$ .

**Proof sketch.** Let **k** and **M** be the trapdoor key and evaluation key generated by KeyGen<sub> $\lambda$ </sub> respectively. For any  $\mathbf{c} \in \text{Im}(\mathbf{M}^{\top})$ , we must have  $\mathbf{k}^{\top}\mathbf{c} = 0$  since  $\mathbf{k} \in \text{Ker}(\mathbf{M})$ . Also, we prove that for any  $\mathbf{c} \in \{0, 1\}^{\lambda} \setminus \text{Im}(\mathbf{M}^{\top})$ , there must

exists  $\mathbf{w}$  such that  $\mathbf{w} \in 1 \times \{0,1\}^{\lambda-1}$  and  $\mathbf{c} = \begin{pmatrix} \mathbf{M} + \begin{pmatrix} \mathbf{0} \cdots & \mathbf{0} & 1 \\ \mathbf{0} & \mathbf{0} \cdots & \mathbf{0} \\ \vdots & \ddots & \vdots \\ \mathbf{0} \cdots & \mathbf{0} & \mathbf{0} \end{pmatrix} \end{pmatrix} \mathbf{w}$ . Since

 $\mathbf{k}^{\top}\mathbf{M}^{\top}\mathbf{w} = 0 \text{ and } \mathbf{k}^{\top} \begin{pmatrix} 0 \cdots 0 & 1\\ 0 & 0 & \cdots & 0\\ \vdots & \ddots & \vdots\\ 0 \cdots & 0 & 0 \end{pmatrix}^{\top} \mathbf{w} = 1, \text{ we must have } \mathbf{k}^{\top}\mathbf{c} = 1 \text{ in this case.}$ 

Therefore, **k**, which is samplable in NC<sup>1</sup>, can be used to determine whether **c** is in Im( $\mathbf{M}^{\top}$ ) or  $\{0,1\}^{\lambda} \setminus \text{Im}(\mathbf{M}^{\top})$  and recover  $x_i$  by checking whether  $\mathbf{c}_i$  and  $\mathbf{c}'_i$  are swapped in the inversing procedure, i.e., correctness holds.

Moreover, due to the subset membership problem for  $L = \text{Im}(\mathbf{M})$ , the uniform distributions over  $\text{Im}(\mathbf{M}^{\top})$  and  $\{0,1\}^{\lambda} \setminus \text{Im}(\mathbf{M}^{\top})$  are indistinguishable when  $\mathbf{M}$  is a correctly generated evaluation key, i.e., the distributions  $(\mathbf{c}_{i,0}, \mathbf{c}_{i,1}) \leftarrow \text{Im}(\mathbf{M}^{\top}) \times \{0,1\}^{\lambda} \setminus \text{Im}(\mathbf{M}^{\top})$  and  $(\mathbf{c}_{i,0}, \mathbf{c}_{i,1}) \leftarrow \text{Im}(\mathbf{M}^{\top}) \times \text{Im}(\mathbf{M}^{\top})$ are indistinguishable. Therefore, the adversary in the one-way game can only obtain information on  $f_{\lambda}(x)$  (which is one-way), and the additional pairs  $(\mathbf{c}_{i,0}, \mathbf{c}_{i,1})$ can be simulated by just sampling them from  $\text{Im}(\mathbf{M}^{\top}) \times \text{Im}(\mathbf{M}^{\top})$ , i.e., they reveal little information on x.

The formal proof is as follows.

*Proof.* Note that  $\mathcal{K}ey\mathcal{G}en$ ,  $\mathcal{E}val$ , and  $\mathcal{I}nverse$  only involve operations including multiplications of the constant number of matrices, inner products and sampling random bits. Since these operation can be performed in NC<sup>1</sup>, we have  $\mathcal{K}ey\mathcal{G}en$ ,  $\mathcal{E}val$ , and  $\mathcal{I}nverse$  can be computed in NC<sup>1</sup>.

Next, we prove that Construction 4 satisfies correctness and one-wayness.

**Correctness.** For any  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and any  $\mathbf{c} \in \mathrm{Im}(\mathbf{M}^{\top})$ , we have

$$\mathbf{k}^{\top} \boldsymbol{c} = \mathbf{k}^{\top} \mathbf{M}^{\top} \mathbf{w} = (\mathbf{M} \mathbf{k})^{\top} \mathbf{w} = \mathbf{0}^{\top} \mathbf{w} = 0$$

where  $\mathbf{k} \in \text{Ker}(\mathbf{M})$  and  $\mathbf{w}$  is a vector such that  $\mathbf{c} = \mathbf{M}\mathbf{w}$ .

Next we show that when  $c \in \{0, 1\}^{\lambda} \setminus \text{Im}(\mathbf{M}^{\top})$  then  $\mathbf{k}^{\top} \mathbf{c} = 1$ . Before showing this, we first propose the following lemma, which is straightforwardly implied by Proposition 2 in Theorem 2.

**Lemma 16** For any  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ , it holds that

$$\{0,1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top}) = \{\mathbf{x} \mid \exists \mathbf{w} \in 1 \times \{0,1\}^{\lambda-1}, \mathbf{x} = \mathbf{M}'^{\top}\mathbf{w}\}$$

where

$$\mathbf{M}' = \mathbf{M} + \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix}$$

According to Lemma 16, for any  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and any  $\mathbf{c} \in \{0,1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$ , we have

$$\mathbf{k}^{\top} \mathbf{c} = \mathbf{k}^{\top} \mathbf{M}'^{\top} \mathbf{w} = \mathbf{k}^{\top} \left( \mathbf{M} + \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix} \right)^{\top} \mathbf{w}$$
$$= (\mathbf{M} \mathbf{k})^{\top} \mathbf{w} + \left( \begin{pmatrix} 0 \cdots & 0 & 1 \\ 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 \cdots & 0 & 0 \end{pmatrix} \mathbf{k} \right)^{\top} \mathbf{w}$$
$$= \mathbf{0}^{\top} \mathbf{w} + (1 \ 0 \ \cdots \ 0) \mathbf{w} = 0 + 1 = 1$$

where **k** is a vector in the kernel of **M** and **w** is a vector such that  $\mathbf{w} \in 1 \times \{0,1\}^{\lambda-1} \wedge \mathbf{c} = \mathbf{M}^{\prime \top} \mathbf{w}$ .

As a result, for all  $i \in [\lambda]$ , **k** generated by  $\mathsf{KeyGen}_{\lambda}$  can be used to determine whether  $c_i$  (resp.,  $c'_i$ ) generated by  $\mathsf{Eval}_{\lambda}$  are in  $\mathrm{Im}(\mathbf{M}^{\top})$  or  $\{0,1\}^{\lambda} \setminus \mathrm{Im}(\mathbf{M}^{\top})$ , and hence recover  $x_i$ .

**One-wayness.** Let  $\mathcal{A} = \{a_{\lambda}\}$  be any adversary in NC<sup>1</sup>. We give hybrid games to show that the advantage of  $\mathcal{A}$  in breaking the one-wayness of Construction 4 is negligible.

**Game 0:** This is the original one-wayness game for  $\mathcal{A} = \{a_{\lambda}\}$ .  $\mathcal{CH}$  runs  $(ek, tk) \leftarrow \mathsf{KeyGen}_{\lambda}$ , samples  $X \leftarrow \mathcal{D}_{\lambda, ek}$ , and runs  $Y = \mathsf{Eval}_{\lambda}(ek, X)$ . Then it sends (ek, Y) to  $a_{\lambda}$ .  $a_{\lambda}$  succeeds if  $a_{\lambda}$  outputs  $X^*$  such that  $\mathsf{Eval}_{\lambda}(ek, X^*) = Y$ . Otherwise it fails.

Game 1 ~ Game  $\lambda$ : For  $i \in [\lambda]$ , Game i is the same as Game i-1 except that  $\mathcal{CH}$  samples  $(\mathbf{c}_{i,0}, \mathbf{c}_{i,1}) \leftarrow \operatorname{Im}(\mathbf{M}^{\top}) \times \operatorname{Im}(\mathbf{M}^{\top})$ .

**Lemma 17** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_{i-1}$  (resp.,  $\epsilon_i$ ) in **Game i-1** (resp., **Game i**), then  $|\epsilon_{i-1} - \epsilon_i| = \operatorname{negl}(\lambda)$ .

*Proof.* According to the proof of the part of subset membership problem in Theorem 2, we have the following lemma.

**Lemma 18** For any  $\mathcal{G} = \{g_{\lambda}\} \in \mathsf{NC}^1$ ,

$$|\Pr[g_{\lambda}(\mathbf{M}, \mathbf{c}_0) = 1] - \Pr[g_{\lambda}(\mathbf{M}, \mathbf{c}_1) = 1]| \le \mathsf{negl}(\lambda),$$

where  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ ,  $\mathbf{c}_0 \leftarrow \mathrm{Im}(\mathbf{M}^{\top})$ , and  $\mathbf{c}_1 \leftarrow \{0,1\}^{\lambda} \setminus \mathrm{Im}(\mathbf{M}^{\top})$ .

*Proof.* Let  $(Setup, Samp \mathcal{Y}es, Samp \mathcal{N}o, \mathcal{K}ey \mathcal{G}en, \mathcal{P}riv, \mathcal{P}ub)$  be a strong smooth HPS defined as Construction 3. According to Propositions 1, 2, and 3, we have

- the distributions of  $\mathbf{x}_0$  and  $\mathbf{c}_0$  are identical where  $\mathsf{pp} \leftarrow \mathsf{Setup}_{\lambda}$ ,  $(\mathbf{x}_0, w) \leftarrow \mathsf{SampYes}_{\lambda}(\mathsf{pp})$ , and  $\mathbf{c} \leftarrow \mathrm{Im}(\mathbf{M}^{\top})$  ( $\because$  Proposition 1).
- the distributions of  $\mathbf{x}_1$  and  $\mathbf{c}_1$  are identical where  $\mathsf{pp} \leftarrow \mathsf{Setup}_{\lambda}$ ,  $(\mathbf{x}_0, w) \leftarrow \mathsf{SampNo}_{\lambda}(\mathsf{pp})$ , and  $\mathbf{c} \leftarrow \{0, 1\}^{\lambda} \setminus \mathrm{Im}(\mathbf{M}^{\top})$  (: Proposition 2).

- it holds that for any  $\mathcal{G} = \{g_{\lambda}\} \in \mathcal{C}_2$ ,

$$|\Pr[g_{\lambda}(\mathsf{pp}, \mathbf{x}_0) = 1] - \Pr[g_{\lambda}(\mathsf{pp}, \mathbf{x}_1) = 1]| \le \mathsf{negl}(\lambda)$$

where  $pp \leftarrow Setup_{\lambda}$ ,  $(\mathbf{x}_0, w) \leftarrow SampYes_{\lambda}(pp)$ , and  $\mathbf{x}_1 \leftarrow SampNo_{\lambda}(pp)$  (: Proposition 3).

Moreover, the distribution of  $pp \leftarrow Setup_{\lambda}$  depends only on the distribution of  $\mathbf{M} \leftarrow ZeroSamp(\lambda)$ . Therefore, for any  $\mathcal{G} = \{g_{\lambda}\} \in \mathcal{C}_2$ , we have

$$|\Pr[g_{\lambda}(\mathbf{M}, \mathbf{c}_0) = 1] - \Pr[g_{\lambda}(\mathbf{M}, \mathbf{c}_1) = 1]| \le \mathsf{negl}(\lambda),$$

where  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$ ,  $\mathbf{c}_0 \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$ , and  $\mathbf{c}_1 \leftarrow \{0,1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$ , completing the proof of Lemma 18.

We now construct  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$  that distinguishes  $\mathbf{c} \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$  and  $\mathbf{c} \leftarrow \{0,1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$  with advantage  $|\epsilon_{i-1} - \epsilon_i|$ , which contradicts to Lemma 18.

 $b_{\lambda}$  takes as input  $(\mathbf{M}, \mathbf{c})$ , which is generated as  $\mathbf{M} \leftarrow \mathsf{ZeroSamp}(\lambda)$  and  $\mathbf{c}$ sampled as  $\mathbf{c} \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$  or  $\mathbf{c} \leftarrow \{0, 1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$  from its challenger. Then, it sets  $ek := \mathbf{M}$  and  $\mathbf{c}_{i,1} := \mathbf{c}$ . Next,  $b_{\lambda}$  samples  $x \leftarrow \{0, 1\}^{\lambda}$ ,  $(\mathbf{c}_{j,0}, \mathbf{c}_{j,1}) \leftarrow$  $\operatorname{Im}(\mathbf{M}^{\top}) \times \operatorname{Im}(\mathbf{M}^{\top})$  for all  $j \in [i - 1]$ ,  $\mathbf{c}_{i,0} \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$ , and  $(\mathbf{c}_{j,0}, \mathbf{c}_{j,1}) \leftarrow$  $\operatorname{Im}(\mathbf{M}^{\top}) \times \{0, 1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$  for all  $j \in \{i + 1, \dots, \lambda\}$ . Next,  $b_{\lambda}$  sets X := $(x, (\mathbf{c}_{1,0}, \mathbf{c}_{1,1}), (\mathbf{c}_{2,0}, \mathbf{c}_{2,1}), \dots, (\mathbf{c}_{\lambda,0}, \mathbf{c}_{\lambda,1}))$  and computes  $Y = \operatorname{Eval}_{\lambda}(ek, X)$ . Finally,  $b_{\lambda}$  gives (ek, Y) to  $a_{\lambda}$ . When  $a_{\lambda}$  output  $X^*$ , if  $Y = \operatorname{Eval}_{\lambda}(ek, X^*), b_{\lambda}$ outputs 1. Otherwise, it outputs 0.

Since all operations in  $b_{\lambda}$  are performed in  $\mathsf{NC}^1$ , we have  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$ .

One can see that when  $\mathbf{c} \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$  (resp.,  $\mathbf{c} \leftarrow \{0, 1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$ ), the view of  $a_{\lambda}$  is identical to its view in **Game i-1** (resp., **Game i**), i.e.,  $b_{\lambda}$  outputs 1 with probability  $\epsilon_{i-1}$  (resp.,  $\epsilon_i$ ). Therefore,  $\mathcal{B} = \{b_{\lambda}\}$  distinguishes  $\mathbf{c} \leftarrow \operatorname{Im}(\mathbf{M}^{\top})$ and  $\mathbf{c} \leftarrow \{0, 1\}^{\lambda} \setminus \operatorname{Im}(\mathbf{M}^{\top})$  with advantage  $|\epsilon_{i-1} - \epsilon_i|$ , which should be negligible according to Lemma 18, completing the proof of Lemma 17.  $\Box$ 

**Lemma 19** If  $\mathcal{A} = \{a_{\lambda}\}$  succeeds with advantage  $\epsilon_{\lambda}$  in **Game**  $\lambda$ , then  $\epsilon_{\lambda} = \operatorname{negl}(\lambda)$ .

*Proof.* We now construct  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$  that breaks the one-wayness of  $\mathcal{F} = \{f_{\lambda}\}$  with advantage  $\epsilon_n$ .

 $b_{\lambda}$  takes as input y, which is generated as  $x \leftarrow \{0,1\}^{\lambda}$  and  $y = f_{\lambda}(y)$  from its challenger. Then,  $b_{\lambda}$  runs  $(ek, tk) \leftarrow \mathsf{KeyGen}_{\lambda}$ , parses  $ek := \mathbf{M}$ , and samples  $((\mathbf{c}_{0,0}, \mathbf{c}_{0,1}), \cdots, (\mathbf{c}_{\lambda,0}, \mathbf{c}_{\lambda,1})) \leftarrow \mathrm{Im}(\mathbf{M}^{\top})^{2\lambda}$ . Next  $b_{\lambda}$  gives  $(y, (\mathbf{c}_{0,0}, \mathbf{c}_{0,1}), \cdots, (\mathbf{c}_{\lambda,0}, \mathbf{c}_{\lambda,1}))$  to  $a_{\lambda}$ . When  $a_{\lambda}$  outputs  $X^*$ ,  $b_{\lambda}$  parses  $X^* := (x^*(\mathbf{c}_{0,0}^*, \mathbf{c}_{0,1}^*), \cdots, (\mathbf{c}_{\lambda,0}^*, \mathbf{c}_{\lambda,1}^*))$  and outputs  $x^*$ .

Since all operations in  $b_{\lambda}$  are performed in NC<sup>1</sup>, we have  $\mathcal{B} = \{b_{\lambda}\} \in \mathsf{NC}^1$ .

One can see that the view of  $a_{\lambda}$  is identical to its view in **Game**  $\lambda$ , i.e.,  $b_{\lambda}$  outputs  $x^*$  such that  $y = f_{\lambda}(x^*)$  with probability  $\epsilon_{\lambda}$ . Therefore,  $\mathcal{B} = \{b_{\lambda}\}$  breaks the one-wayness of  $\mathcal{F} = \{f_{\lambda}\}$  with advantage  $\epsilon_{\lambda}$ , which should be negligible, completing the proof of Lemma 19.

Since for  $i \in [\lambda]$ ,  $|\epsilon_{i-1} - \epsilon_i| = \mathsf{negl}(\lambda)$ ,  $\epsilon_{\lambda} = \mathsf{negl}(\lambda)$ , we have

$$\epsilon_0 \leq \sum_{i=1}^{\lambda} |\epsilon_{i-1} - \epsilon_i| + \epsilon_{\lambda} = \operatorname{negl}(\lambda).$$

Therefore, Construction 4 satisfies one-wayness.

Putting all the above together, Theorem 3 immediately follows.

# 6 Conclusion

In this paper, we formalize fine-grained OWPs, HPSs (which in turn derives a CCA-secure PKE), and TDFs, and show how to construct the  $NC^1$  versions of them secure against  $NC^1$  adversaries. Compared with traditional cryptographic

primitives, our schemes treat restricted class of adversaries, while they can be run more efficiently and are only based on the mild worst case assumption  $NC^1 \subsetneq \oplus L/poly$ . It remains open how to construct more fine-grained primitives not implied by our results, such as pseudo-random functions and signature schemes, in the same model.

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# Supplementary Material

# A Transformation from Collection of OWPs to OWP

We now give our construction of OWP from a collection of OWPs as follows.

**Construction 5 (OWP from collection of OWPs)** Let  $(KeyGen = \{KeyGen_{\lambda}\}, \mathcal{E}val = \{Eval_{\lambda}\})$  be a collection of  $C_1$ -OWPs against  $C_2$  with the key space  $K_{\lambda}$  and  $Eval_{\lambda}(k, \cdot) : D_{\lambda,k} \to D_{\lambda,k}$  for  $k \in K_{\lambda}$ , where it holds that the distributions  $k \leftarrow KeyGen_{\lambda}$  and  $k \leftarrow K_{\lambda}$  are identical. Define a function family  $\mathcal{G} = \{g_{\lambda} : D_{\lambda} \to D_{\lambda}\}$  as follows.

 $\begin{array}{l} - \ D_{\lambda} := \bigcup_{k \in K_{\lambda}} (\{k\} \times D_{\lambda,k}). \\ - \ For \ (k,x) \in D_{\lambda}, \ g_{\lambda}(k,x) := (k, \mathsf{Eval}_{\lambda}(k,x)). \end{array}$ 

**Lemma 20** If ( $\mathcal{K}ey\mathcal{G}en, \mathcal{E}val$ ) is a collection of  $\mathcal{C}_1$ -OWPs against  $\mathcal{C}_2$  which satisfies that the distributions  $k \leftarrow \mathsf{KeyGen}_{\lambda}$  and  $k \leftarrow K_{\lambda}$  are identical, the function family  $\mathcal{G}$  is a  $\mathcal{C}_1$ -OWP against  $\mathcal{C}_2$ .

*Proof.* Since  $\mathsf{Eval}_{\lambda} \in \mathcal{C}_1$ , we have  $g_{\lambda} \in \mathcal{C}_1$  immediately.

**Permutation.** Since  $\mathsf{Eval}_{\lambda}$  is a permutation, we have

- for any  $(k, x) \in D_{\lambda}$ ,  $g_{\lambda}(k, x) \in D_{\lambda}$  holds since  $\mathsf{Eval}_{\lambda}(k, x) \in D_{\lambda}$ .
- for any  $(k,x) \neq (k',x') \in D_{\lambda}, g_{\lambda}(k,x) = (k, \mathsf{Eval}_{\lambda}(k,x)) \neq 0$  $(k', \mathsf{Eval}_{\lambda}(k', x')) = g_{\lambda}(k', x')$  holds, i.e.,  $g_{\lambda}$  is injective, and
- for any  $(k, y) \in D_{\lambda}$ , there must exist x such that  $x = \mathsf{Eval}_{\lambda}(k, y)$ , i.e.,  $g_{\lambda}$  is surjective.

Therefore,  $g_{\lambda}$  is a permutation.

**One-wayness.** Let  $\mathcal{A} = \{a_{\lambda}\}$  be any adversary in  $\mathcal{C}_2$  with probability  $\epsilon$ . We now construct  $\mathcal{B} = \{b_{\lambda}\} \in \mathcal{C}_2$  that breaks the one-wayness of  $(\mathcal{K}ey\mathcal{G}en, \mathcal{E}val)$ .

 $b_{\lambda}$  takes as input (k, y) which is generated as  $k \leftarrow \mathsf{KeyGen}_{\lambda}, x \leftarrow D_{\lambda,k}$ , and  $y = \mathsf{Eval}_{\lambda}(k, x)$ . Then,  $b_{\lambda}$  gives (k, y) to  $a_{\lambda}$ . When  $a_{\lambda}$  outputs (k', x'),  $b_{\lambda}$  outputs x'.

Since the operations of  $b_{\lambda}$  only runs  $a_{\lambda}$  and returns the output of  $a_{\lambda}$ , we have  $b_{\lambda} \in \mathcal{C}_2$ . Furthermore, since the distributions  $k \leftarrow \mathsf{KeyGen}_{\lambda}$  and  $k \leftarrow K_{\lambda}$  are identical,  $a_{\lambda}$  outputs (k', x) such that  $g_{\lambda}(k', x) = (k', \mathsf{Eval}_{\lambda}(k', x)) = (k, y)$  with probability  $\epsilon$ . Therefore,  $b_{\lambda}$  outputs x such that  $\mathsf{Eval}_{\lambda}(k, x) = y$  with probability  $\epsilon$ , completing the proof of one-wayness.

Putting all the above together, Lemma 20 immediately follows.

#### В Perfectly Smooth and Strong Universal<sub>1</sub> for Multi-bit

We now construct a multi-bit perfectly smooth and strong universal<sub>1</sub>  $NC^{1}$ -HPS against  $NC^1$  as follows.

Construction 6 (Multi-bit NC<sup>1</sup>-HPS) Let  $\mathcal{P}$ = $(Setup, Samp \mathcal{Y}es,$  $\mathcal{S}amp\mathcal{N}o, \mathcal{K}ey\mathcal{G}en, \mathcal{P}riv, \mathcal{P}ub)$  be perfectly smooth and  $\frac{1}{2}$ -universal<sub>1</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup> in Construction 3 (see Section 4) and let n be a polynomial in  $\lambda$ .  $Define \ \mathcal{S}etup' = \{\mathsf{Setup}'_{\lambda}\}, \ \mathcal{S}amp \mathcal{Y}es' = \{\mathsf{Samp}\mathsf{Yes}'_{\lambda}\}, \ \mathcal{S}amp \mathcal{N}o' = \{\mathsf{Samp}\mathsf{No}'_{\lambda}\},$  $\mathcal{K}ey\mathcal{G}en' = \{\mathsf{KeyGen}_{\lambda}\}, \ \mathcal{P}riv' = \{\mathsf{Priv}_{\lambda}\} \ and \ \mathcal{P}ub' = \{\mathsf{Pub}_{\lambda}\} \ as \ follows.$ 

- Setup<sub>3</sub>:
  - 1. Run pp  $\leftarrow$  Setup<sub> $\lambda$ </sub>.
  - 2. Parse  $pp = (X_{\lambda}, L_{\lambda}, W_{\lambda}, R_{\lambda}, SK_{\lambda}, PK_{\lambda}, \Pi_{\lambda}, H_{\lambda}, \alpha_{\lambda}, aux_{\lambda})$  and define  $X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, H'_{\lambda}, \Pi'_{\lambda}, \alpha'_{\lambda}, and aux'_{\lambda} as follows.$ 
    - $X'_{\lambda} := X_{\lambda}.$

    - $L'_{\lambda} := L_{\lambda}$ .  $W'_{\lambda} := W_{\lambda}$ .  $R'_{\lambda} := R_{\lambda}$ .
    - $SK'_{\lambda} := SK^{n(\lambda)}_{\lambda}$
    - $PK'_{\lambda} := PK^{n(\lambda)}_{\lambda}.$

    - $\Pi'_{\lambda} := \Pi^{n(\lambda)}_{\lambda} = \{0, 1\}^{n(\lambda)}.$
    - $H'_{\lambda}(\mathbf{sk}', \mathbf{x}) := (H_{\lambda}(\mathbf{sk}_1, \mathbf{x}) \cdots H_{\lambda}(\mathbf{sk}_{n(\lambda)}, \mathbf{x})).$
    - $\alpha'_{\lambda}(\mathbf{sk}') = (\alpha_{\lambda}(\mathbf{sk}_1) \cdots \alpha_{\lambda}(\mathbf{sk}_{n(\lambda)})).$

•  $aux'_{\lambda} := pp.$ 

- Here,  $\mathbf{sk}' = (\mathbf{sk}_1 \cdots \mathbf{sk}_{n(\lambda)}) \in SK'_{\lambda}$ .
- 3. Set  $pp' := (X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, \Pi'_{\lambda}, H'_{\lambda}, \alpha'_{\lambda}, aux'_{\lambda})$  and output pp'.
- SampYes'<sub> $\lambda$ </sub>(pp'):
  - 1. Parse pp' :=  $(X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, \Pi'_{\lambda}, H'_{\lambda}, \alpha'_{\lambda}, \mathsf{aux}'_{\lambda})$  and let  $\mathsf{aux}'_{\lambda} = \mathsf{pp}.$
  - 2. Run  $(\mathbf{x}, \mathbf{w}) \leftarrow \mathsf{SampYes}_{\lambda}(\mathsf{pp})$  and  $output (\mathbf{x}, \mathbf{w})$ .
- SampNo'<sub> $\lambda$ </sub>(pp'):

  - 2. Run  $\mathbf{x} \leftarrow \mathsf{SampNo}_{\lambda}(\mathsf{pp})$  and output  $\mathbf{x}$ .
- KeyGen $'_{\lambda}(pp')$ :
  - 1. Parse  $pp' := (X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, \Pi'_{\lambda}, H'_{\lambda}, \alpha'_{\lambda}, \mathsf{aux}'_{\lambda}).$
  - 2. Sample  $\mathbf{sk}' \leftarrow SK'_{\lambda}$ .
  - 3. Compute  $\mathbf{pk}' := \alpha'_{\lambda}(\mathbf{sk}')$  and output  $(\mathbf{pk}', \mathbf{sk}')$ .
- $\mathsf{Priv}'_{\lambda}(\mathsf{pp}', \mathbf{sk}', \mathbf{x})$ :
  - 1. Parse  $pp' := (X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, \Pi'_{\lambda}, H'_{\lambda}, \alpha'_{\lambda}, \mathsf{aux}'_{\lambda}).$
  - 2. Compute  $\pi := H'_{\lambda}(\mathbf{sk}', \mathbf{x})$  and output  $\pi$ .
- $\mathsf{Pub}'_{\lambda}(\mathsf{pp}', \mathbf{pk}', \mathbf{x}, \mathbf{w})$ :
  - 1. Parse  $\mathbf{pk}' := (\mathbf{pk}_1 \cdots \mathbf{pk}_{n(\lambda)}).$
  - 2. Compute  $\boldsymbol{\pi} := (\mathsf{Pub}_{\lambda}(\mathbf{pk}_1, \mathbf{x}, \mathbf{w}) \cdots \mathsf{Pub}_{\lambda}(\mathbf{pk}_{n(\lambda)}, \mathbf{x}, \mathbf{w}))^{\top}$  and output  $\boldsymbol{\pi}$ .

**Lemma 21** Let  $\mathcal{P}' = (Setup', Samp \mathcal{Y}es', Samp \mathcal{N}o', \mathcal{K}ey \mathcal{G}en', \mathcal{P}riv', \mathcal{P}ub')$  be the construction defined as Construction 6, then,  $\mathcal{P}'$  is perfectly smooth and strong universal<sub>1</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup>.

*Proof.* Since  $\mathcal{P} \in \mathsf{NC}^1$  and there are sampling random elements and computing  $n(\lambda)$  sets of algorithms in  $\mathcal{P}$  which is computable in parallel, we have  $\mathcal{P}' \in \mathsf{NC}^1$ .

Correctness and subset membership problem follow from the fact that  ${\cal P}$  is the perfectly smooth  $NC^1\text{-}{\rm HPS}$  against  $NC^1$ .

**Perfect smoothness.** We now show that for any  $\mathbf{pp}' \leftarrow \mathsf{Setup}'_{\lambda}$ , the random variables  $(\mathbf{x}, \mathbf{pk}', \pi)$  and  $(\mathbf{x}, \mathbf{pk}', \pi^*)$  are identical where  $\mathbf{x} \leftarrow X'_{\lambda} \setminus L'_{\lambda}$ ,  $(\mathbf{pk}', \mathbf{sk}') \leftarrow \mathsf{KeyGen}'_{\lambda}(\mathbf{pp}')$ , and  $\pi^* \leftarrow \Pi'_{\lambda}$ , i.e., let  $S_{\lambda}$  be a finite set  $(X'_{\lambda} \setminus L'_{\lambda} \times PK'_{\lambda} \times \Pi'_{\lambda})$ , we show that for any  $(\tilde{\mathbf{x}}, \mathbf{pk}', \tilde{\pi}) \in S_{\lambda}$ , we have  $\Pr[(\mathbf{x}, \mathbf{pk}', \pi) = (\tilde{\mathbf{x}}, \mathbf{pk}', \tilde{\pi})] = \Pr[(\mathbf{x}, \mathbf{pk}', \pi^*) = (\tilde{\mathbf{x}}, \mathbf{pk}', \tilde{\pi})]$ . Since we have

$$\begin{aligned} \Pr[(\mathbf{x}, \mathbf{p}\mathbf{k}', \boldsymbol{\pi}) &= (\tilde{\mathbf{x}}, \mathbf{p}\mathbf{k}', \tilde{\boldsymbol{\pi}})] \\ &= \Pr\left[ \begin{array}{c} \mathbf{p}\mathbf{k}' = \mathbf{p}\tilde{\mathbf{k}}' \land \\ \boldsymbol{\pi} = \tilde{\boldsymbol{\pi}} \end{array} \right] \Pr[\mathbf{x} = \tilde{\mathbf{x}}] \\ &= \Pr\left[ \begin{array}{c} \mathbf{p}\tilde{\mathbf{k}}' = \alpha_{\lambda}'(\mathbf{s}\mathbf{k}') \land \\ \tilde{\boldsymbol{\pi}} = \mathsf{Priv}_{\lambda}'(\mathbf{s}\mathbf{k}', \tilde{\mathbf{x}}) \end{array} \right] \Pr[\mathbf{x} = \tilde{\mathbf{x}}] \\ &= \left( \prod_{i=1}^{n(\lambda)} \Pr\left[ \begin{array}{c} \mathbf{p}\tilde{\mathbf{k}}_i = \alpha_{\lambda}(\mathbf{s}\mathbf{k}_i) \land \\ \tilde{\boldsymbol{\pi}}_i = \mathsf{Priv}_{\lambda}(\mathbf{s}\mathbf{k}, \mathbf{x}) \end{array} \right] \right) \Pr[\mathbf{x} = \tilde{\mathbf{x}}] \\ &= \left( \prod_{i=1}^{n(\lambda)} \Pr\left[ \begin{array}{c} \mathbf{p}\tilde{\mathbf{k}}_i = \alpha_{\lambda}(\mathbf{s}\mathbf{k}_i) \land \\ \tilde{\boldsymbol{\pi}}_i = \pi_i^* \end{array} \right] \right) \Pr[\mathbf{x} = \tilde{\mathbf{x}}] \\ &= \Pr\left[ \left[ \begin{array}{c} \mathbf{p}\tilde{\mathbf{k}}' = \alpha_{\lambda}'(\mathbf{s}\mathbf{k}') \land \\ \pi^* = \tilde{\boldsymbol{\pi}} \end{array} \right] \Pr[\mathbf{x} = \tilde{\mathbf{x}}] \\ &= \Pr[(\mathbf{x}, \mathbf{p}\mathbf{k}', \boldsymbol{\pi}^*) = (\tilde{\mathbf{x}}, \mathbf{p}\tilde{\mathbf{k}}', \tilde{\boldsymbol{\pi}})] \end{aligned} \end{aligned}$$

where  $\mathbf{sk}' = (\mathbf{sk}_1 \cdots \mathbf{sk}_{n(\lambda)})$  and  $\tilde{\mathbf{pk}}' = (\tilde{\mathbf{pk}}_1 \cdots \tilde{\mathbf{pk}}_{n(\lambda)})$ ,  $\mathcal{P}'$  satisfies perfect smoothness.

Strong universality<sub>1</sub>. For any  $pp' \leftarrow \text{Setup}'_{\lambda}$ ,  $pk' \in PK'_{\lambda}$ ,  $\mathbf{x} \in X'_{\lambda} \setminus L'_{\lambda}$ , and  $\pi \in \Pi'_{\lambda}$ , we have

$$\begin{split} \Pr[\mathsf{Priv}_{\lambda}'(\mathsf{pp}', \mathbf{sk}', \mathbf{x}) &= \boldsymbol{\pi} | \alpha_{\lambda}'(\mathbf{sk}') = \mathbf{pk}'] = \prod_{i=1}^{n(\lambda)} \Pr[\mathsf{Priv}_{\lambda}(\mathsf{pp}', \mathbf{sk}_i, \mathbf{x}) = \pi_i | \alpha_{\lambda}(\mathbf{sk}_i) = \mathbf{pk}_i] \\ &\leq \prod_{i=1}^{n(\lambda)} \left(\frac{1}{2}\right) \quad (\because \frac{1}{2}\text{-}universality_1 \text{ of } \mathcal{P}) \\ &= \frac{1}{2^{n(\lambda)}} = \mathsf{negl}(\lambda) \end{split}$$

where  $\mathbf{sk}' = (\mathbf{sk}_1 \cdots \mathbf{sk}_{n(\lambda)})$  and  $\mathbf{pk}' = (\mathbf{pk}_1 \cdots \mathbf{pk}_{n(\lambda)})$ , and it follows that  $\mathcal{P}'$  satisfies strong universality<sub>1</sub>.

This completes the proof of Lemma 21.

# C Strong Universal<sub>2</sub> for $L \times \{0, 1\}^n$

Next, we give the construction of strong universal<sub>2</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup> such that the language  $L_{\lambda}$  supports  $\{0, 1\}^n$ . Our construction is as follows.

**Construction 7** Let  $\mathcal{P}' = (Setup', Samp\mathcal{Y}es', Samp\mathcal{N}o', \mathcal{K}ey\mathcal{G}en', \mathcal{P}riv', \mathcal{P}ub')$ be strong universal<sub>1</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup> in Construction 6 (see Section B) and let n be a polynomial in  $\lambda$ . Define  $Setup^{\dagger} = {Setup_{\lambda}^{\dagger}}$ ,  $\mathcal{S}amp\mathcal{Y}es^{\dagger} = \{\mathsf{SampYes}^{\dagger}_{\lambda}\}, \ \mathcal{S}amp\mathcal{N}o^{\dagger} = \{\mathsf{SampNo}^{\dagger}_{\lambda}\}, \ \mathcal{K}ey\mathcal{G}en^{\dagger} = \{\mathsf{KeyGen}^{\dagger}_{\lambda}\}, \ \mathcal{P}riv^{\dagger} = \{\mathsf{Priv}^{\dagger}_{\lambda}\} \ and \ \mathcal{P}ub^{\dagger} = \{\mathsf{Pub}^{\dagger}_{\lambda}\} \ as \ follows.$ 

- Setup<sup>†</sup>: 1.  $Run pp' \leftarrow Setup'_{\lambda}$ . 2. Parse  $pp' = (X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, \Pi'_{\lambda}, H'_{\lambda}, \alpha'_{\lambda}, aux'_{\lambda})$  and define  $X_{\lambda}^{\dagger}, L_{\lambda}^{\dagger}, W_{\lambda}^{\dagger}, R_{\lambda}^{\dagger}, SK_{\lambda}^{\dagger}, PK_{\lambda}^{\dagger}, H_{\lambda}^{\dagger}, \Pi_{\lambda}^{\dagger}, \alpha_{\lambda}^{\dagger}, and aux_{\lambda}^{\dagger} as follows.$ •  $X_{\lambda}^{\dagger} := X_{\lambda}' \times \{0,1\}^{n(\lambda)}$ . •  $L_{\lambda}^{\dagger} := L_{\lambda}' \times \{0,1\}^{n(\lambda)}$ . •  $W_{\lambda}^{\dagger} := W_{\lambda}'$ . •  $R_{\lambda}^{\dagger} := R_{\lambda}' \times \{0,1\}^{n(\lambda)}$ . •  $SK_{\lambda}^{\dagger} := (SK_{\lambda}')^{2(\lambda+n(\lambda))}$ . •  $PK_{\lambda}^{\dagger} := (PK_{\lambda}')^{2(\lambda+n(\lambda))}$ . •  $\Pi_{\lambda}^{\dagger} := \Pi_{\lambda}'$ . •  $H^{\widehat{\uparrow}}_{\lambda}(\mathbf{sk}^{\dagger}, (\mathbf{x}, c)) := \bigoplus_{i=1}^{\lambda+n(\lambda)} H'_{\lambda}(\mathbf{sk}'_{i, z_i}, \mathbf{x}) \text{ where } \mathbf{z} = \mathbf{x} ||\mathbf{c}|| |\mathbf{c} = \mathbf{x}||\mathbf{c}|| |\mathbf{c}|| |\mathbf{c}||$  $(\mathbf{x}^{\top} \ \boldsymbol{c}^{\top})^{\top}).$ •  $\alpha_{\lambda}^{\dagger}(\mathbf{sk}^{\dagger}) = (\alpha_{\lambda}'(\mathbf{sk}_{1,0}') \ \alpha_{\lambda}'(\mathbf{sk}_{1,1}') \cdots \alpha_{\lambda}'(\mathbf{sk}_{\lambda+n(\lambda),0}') \ \alpha_{\lambda}'(\mathbf{sk}_{\lambda+n(\lambda),1}')).$ •  $\operatorname{aux}_{\lambda}^{\dagger} := \operatorname{pp}_{\lambda}'$ . Here,  $\mathbf{sk}^{\dagger} = (\mathbf{sk}'_{1,0} \ \mathbf{sk}'_{1,1} \cdots \mathbf{sk}'_{\lambda+n(\lambda),0} \ \mathbf{sk}'_{\lambda+n(\lambda),1}) \in SK^{\dagger}_{\lambda}$ . 3. Set  $pp^{\dagger} := (X_{\lambda}^{\dagger}, L_{\lambda}^{\dagger}, W_{\lambda}^{\dagger}, R_{\lambda}^{\dagger}, SK_{\lambda}^{\dagger}, PK_{\lambda}^{\dagger}, \Pi_{\lambda}^{\dagger}, H_{\lambda}^{\dagger}, \alpha_{\lambda}^{\dagger}, aux_{\lambda}^{\dagger})$  and output pp<sup>†</sup>. - SampYes $^{\dagger}_{\lambda}(pp^{\dagger})$ : 1. Parse  $pp^{\dagger} = (X^{\dagger}_{\lambda}, L^{\dagger}_{\lambda}, W^{\dagger}_{\lambda}, R^{\dagger}_{\lambda}, SK^{\dagger}_{\lambda}, PK^{\dagger}_{\lambda}, \Pi^{\dagger}_{\lambda}, H^{\dagger}_{\lambda}, \alpha^{\dagger}_{\lambda}, aux^{\dagger}_{\lambda})$  and let  $aux_{\lambda}^{\dagger} = pp'$ . 2.  $Run(\mathbf{x}, \mathbf{w}) \leftarrow \mathsf{SampYes}'_{\lambda}(\mathsf{pp}') \text{ and sample } \mathbf{c} \leftarrow \{0, 1\}^{n(\lambda)}.$ 3. Output  $((\mathbf{x}, \mathbf{c}), \mathbf{w})$ . - SampNo<sup>†</sup><sub> $\lambda$ </sub>(pp<sup>†</sup>): 1. Parse  $pp^{\dagger} = (X^{\dagger}_{\lambda}, L^{\dagger}_{\lambda}, W^{\dagger}_{\lambda}, R^{\dagger}_{\lambda}, SK^{\dagger}_{\lambda}, PK^{\dagger}_{\lambda}, \Pi^{\dagger}_{\lambda}, H^{\dagger}_{\lambda}, \alpha^{\dagger}_{\lambda}, aux^{\dagger}_{\lambda})$  and let  $\operatorname{aux}_{\lambda}^{\dagger} = \operatorname{pp}'$ . 2. Run  $\mathbf{x} \leftarrow \mathsf{SampNo}'_{\lambda}(\mathsf{pp}')$  and sample  $\mathbf{c} \leftarrow \{0,1\}^{n(\lambda)}$ . 3. Output  $(\mathbf{x}, \mathbf{c})$ . - KeyGen $^{\dagger}_{\lambda}(pp^{\dagger})$ :  $1. \ Parse \ \mathsf{pp}^{\dagger} = (X_{\lambda}^{\dagger}, L_{\lambda}^{\dagger}, W_{\lambda}^{\dagger}, R_{\lambda}^{\dagger}, SK_{\lambda}^{\dagger}, PK_{\lambda}^{\dagger}, \Pi_{\lambda}^{\dagger}, H_{\lambda}^{\dagger}, \alpha_{\lambda}^{\dagger}, \mathsf{aux}_{\lambda}^{\dagger}).$ 2. Sample  $\mathbf{sk}^{\dagger} \leftarrow SK_{\lambda}^{\dagger}$ . 3. Compute  $\mathbf{pk}^{\dagger} := \alpha_{\lambda}^{\dagger}(\mathbf{sk}^{\dagger})$  and output  $(\mathbf{pk}^{\dagger}, \mathbf{sk}^{\dagger})$ . -  $\mathsf{Priv}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathbf{sk}^{\dagger}, (\mathbf{x}, \mathbf{c}))$ : 1. Parse  $pp^{\dagger} = (X_{\lambda}^{\dagger}, L_{\lambda}^{\dagger}, W_{\lambda}^{\dagger}, R_{\lambda}^{\dagger}, SK_{\lambda}^{\dagger}, PK_{\lambda}^{\dagger}, \Pi_{\lambda}^{\dagger}, H_{\lambda}^{\dagger}, \alpha_{\lambda}^{\dagger}, aux_{\lambda}^{\dagger}).$ 2. Compute  $\boldsymbol{\pi} := H_{\lambda}^{\dagger}(\mathbf{sk}^{\dagger}, (\mathbf{x}, \boldsymbol{c}))$  and output  $\boldsymbol{\pi}$ . -  $\mathsf{Pub}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathsf{pk}^{\dagger}, (\mathbf{x}, c), \mathbf{w})$ : 1. Parse  $\mathbf{p}\mathbf{k}^{\dagger} = (\mathbf{p}\mathbf{k}'_{1,0} \ \mathbf{p}\mathbf{k}'_{1,1}\cdots\mathbf{p}\mathbf{k}'_{\lambda+n(\lambda),0} \ \mathbf{p}\mathbf{k}'_{\lambda+n(\lambda),1}).$ 2. let  $\mathbf{z} = \mathbf{x} || \mathbf{c}$  and compute  $\pi := \bigoplus_{i=1}^{\lambda+n(\lambda)} \mathsf{Pub}'_{\lambda}(\mathbf{p}\mathbf{k}'_{i,z_i}, \mathbf{x}, \mathbf{w}).$ 
  - 3. Output  $\pi$ .

**Lemma 22** Let  $\mathcal{P}^{\dagger} = (Setup^{\dagger}, Samp\mathcal{Y}es^{\dagger}, Samp\mathcal{N}o^{\dagger}, \mathcal{K}ey\mathcal{G}en^{\dagger}, \mathcal{P}riv^{\dagger}, \mathcal{P}ub^{\dagger})$ be the construction defined as Construction 7, then,  $\mathcal{P}^{\dagger}$  is strong universal<sub>2</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup>.

*Proof.* Since  $\mathcal{P}' \in \mathsf{NC}^1$  and there are sampling random elements, xoring  $\lambda + n(\lambda)$  strings of  $\lambda$  bits, and computing  $\lambda + n(\lambda)$  sets of algorithms in  $\mathcal{P}'$  which is computable in parallel, we have  $\mathcal{P}^{\dagger} \in \mathsf{NC}^1$ .

Correctness and subset membership problem are follows from the fact that  $\mathcal{P}'$  is the strong universal<sub>1</sub> NC<sup>1</sup>-HPS against NC<sup>1</sup>.

**Strong universality**<sub>2</sub>. We now show that  $\mathcal{P}^{\dagger}$  satisfies strong universal<sub>2</sub>, i.e. we have to show that for any  $\mathbf{pp}^{\dagger} \leftarrow \mathsf{Setup}^{\dagger}_{\lambda}, \mathbf{pk}^{\dagger} \in PK^{\dagger}_{\lambda},$  $(\mathbf{x}, \mathbf{c}), (\mathbf{x}^*, \mathbf{c}^*) \in X^{\dagger}_{\lambda}$  with  $(\mathbf{x}, \mathbf{c}) \notin L^{\dagger}_{\lambda} \cup \{(\mathbf{x}^*, \mathbf{c}^*)\}$  and  $\pi, \pi^* \in \Pi^{\dagger},$ conditioned on  $\pi^* = \mathsf{Priv}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathsf{sk}^{\dagger}, (\mathbf{x}^*, \mathbf{c}^*))$  and  $\mathbf{pk}^{\dagger} = \alpha^{\dagger}_{\lambda}(\mathsf{sk}^{\dagger})$ , the probability that  $\pi = \mathsf{Priv}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathsf{sk}^{\dagger}, (\mathbf{x}, \mathbf{c}))$  is  $\mathsf{negl}(\lambda)$ , where  $\mathsf{sk}^{\dagger} = (\mathsf{sk}'_{1,0} \ \mathsf{sk}'_{1,1} \cdots \mathsf{sk}'_{\lambda+n(\lambda),0} \ \mathsf{sk}'_{\lambda+n(\lambda),1}) \leftarrow SK^{\dagger}_{\lambda}.$ 

 $\begin{array}{ll} (\mathbf{sk}_{1,0}' \ \mathbf{sk}_{1,1}' \cdots \mathbf{sk}_{\lambda+n(\lambda),0}' \ \mathbf{sk}_{\lambda+n(\lambda),1}') \leftarrow SK_{\lambda}^{\dagger}.\\ \text{Since } (\mathbf{x}, \boldsymbol{c}) \neq (\mathbf{x}^{*}, \boldsymbol{c}^{*}), \text{ there is } i \in [\lambda] \text{ such that } z_{i} \neq z_{i}^{*} \text{ where } \boldsymbol{z} = \mathbf{x} || \boldsymbol{c} \text{ and } \boldsymbol{z}^{*} = \mathbf{x}^{*} || \boldsymbol{c}^{*}. \text{ Without loss of generality, we can assume that } i = 1. \text{ When we fix the values of } \mathbf{sk}_{2,0}', \mathbf{sk}_{2,1}', \cdots, \mathbf{sk}_{\lambda+n(\lambda),0}', \mathbf{sk}_{\lambda+n(\lambda),1}', \text{ the value of } \boldsymbol{\pi} \text{ depends only on } \operatorname{Priv}_{\lambda}'(\mathsf{pp}', \mathbf{sk}_{1,z_{\lambda}}', (\mathbf{x}, \boldsymbol{c})). \text{ Therefore, for all } \mathbf{pk}^{\dagger} = (\mathbf{pk}_{1,0}' \ \mathbf{pk}_{1,1}' \cdots \mathbf{pk}_{\lambda+n(\lambda),0}' \ \mathbf{pk}_{\lambda+n(\lambda),1}') \in PK_{\lambda}^{\dagger}, \ (\mathbf{x}, \boldsymbol{c}), (\mathbf{x}^{*}, \boldsymbol{c}^{*}) \in X_{\lambda}^{\dagger}, \text{ and } \boldsymbol{\pi}, \boldsymbol{\pi}^{*} \in \boldsymbol{\Pi}_{\lambda}^{\dagger} \text{ with } (\mathbf{x}, \boldsymbol{c}) \notin \boldsymbol{L}_{\lambda}^{\dagger} \cup \{(\mathbf{x}^{*}, \boldsymbol{c}^{*})\}, \text{ it holds that} \end{array}$ 

$$\begin{aligned} \Pr \ [ \ \mathsf{Priv}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathbf{sk}^{\dagger}, (\mathbf{x}, \boldsymbol{c})) &= \pi \mid \mathsf{Priv}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathbf{sk}^{\dagger}, (\mathbf{x}^{*}, \boldsymbol{c}^{*})) &= \pi^{*} \land \alpha^{\dagger}_{\lambda}(\mathbf{sk}^{\dagger}) = \mathbf{pk}^{\dagger} ] \\ &= \Pr[\mathsf{Priv}^{\prime}_{\lambda}(\mathsf{pp}^{\prime}, \mathbf{sk}^{\prime}_{1, z_{1}}, \mathbf{x}) = \tilde{\pi} \mid \mathsf{Priv}^{\dagger}_{\lambda}(\mathsf{pp}^{\dagger}, \mathbf{sk}^{\dagger}, (\mathbf{x}^{*}, \boldsymbol{c}^{*})) &= \pi^{*} \land \alpha^{\dagger}_{\lambda}(\mathbf{sk}^{\dagger}) = \mathbf{pk}^{\dagger} ] \end{aligned}$$

$$= \Pr[\mathsf{Priv}_{\lambda}'(\mathsf{pp}', \mathbf{sk}_{1,z_1}', \mathbf{x}) = \tilde{\pi} \mid \alpha_{\lambda}^{\dagger}(\mathbf{sk}^{\dagger}) = \mathbf{pk}^{\dagger}]$$
(1)

$$= \Pr[\mathsf{Priv}_{\lambda}'(\mathsf{pp}', \mathbf{sk}_{1,z_1}', \mathbf{x}) = \tilde{\pi} \mid \alpha_{\lambda}'(\mathbf{sk}_{1,z_1}') = \mathbf{pk}_{1,\mathbf{x}_{\lambda}}']$$
(2)

$$\mathsf{negl}(\lambda),\tag{3}$$

where  $\tilde{\boldsymbol{\pi}} = \boldsymbol{\pi} \oplus \left( \bigoplus_{i=2}^{\lambda+n(\lambda)} \mathsf{Pub}_{\lambda}'(\mathbf{pk}_{i,z_i}', \mathbf{x}, \mathbf{w}) \right)$ , and it follows that  $\mathcal{P}^{\dagger}$  satisfies strong universality<sub>2</sub>.

Equation (1) follows from the fact that  $\operatorname{Priv}_{\lambda}'(\operatorname{pp}', \operatorname{sk}_{1,z_1}', \mathbf{x})$  is independent of  $\operatorname{Priv}_{\lambda}^{\dagger}(\operatorname{pp}^{\dagger}, \operatorname{sk}^{\dagger}, (\mathbf{x}^*, \mathbf{c}^*))$ . Equation (2) is satisfied since  $\operatorname{Priv}_{\lambda}'(\operatorname{pp}', \operatorname{sk}_{1,z_1}', \mathbf{x})$  depends only on  $\operatorname{sk}_{1,z_1}'$ . Equation (3) immediately follows strong universality<sub>1</sub> of  $\mathcal{P}'$ .

This completes the proof of Lemma 22.

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# D Fine-Grained CCA-Secure Public Key Encryption

In this section, we give the definition of  $NC^1$ -CCA-secure PKE against  $NC^1$  circuits and give a construction by instantiating the Cramer-Shoup PKE scheme [15] with our HPS in Section 4.

#### D.1 Definition

**Definition 9 (CCA-secure Public Key Encryption)** Let n be a polynomial in  $\lambda$ . Let  $\mathcal{K}ey\mathcal{G}en = \{\mathsf{KeyGen}_{\lambda} : \phi \to SK_{\lambda} \times PK_{\lambda}\}, \mathcal{E}nc = \{\mathsf{Enc}_{\lambda} : PK_{\lambda} \times \{0,1\}^{n(\lambda)} \to C_{\lambda}\}, and \mathcal{D}ec = \{\mathsf{Dec}_{\lambda} : SK_{\lambda} \times C_{\lambda} \to \{0,1\}^{n(\lambda)}\}$  be function families. ( $\mathcal{K}ey\mathcal{G}en, \mathcal{E}nc, \mathcal{D}ec$ ) is a (k,l)-CCA-secure  $\mathcal{C}_1$ -public key encryption scheme (PKE) against  $\mathcal{C}_2$  if:

- Correctness: For any  $\lambda \in \mathbb{N}$  and any  $m \in \{0,1\}^{n(\lambda)}$ :

$$\Pr\left[\mathsf{Dec}_{\lambda}(\mathsf{sk}, c) = m \middle| \begin{array}{c} (\mathsf{sk}, \mathsf{pk}) \leftarrow \mathsf{KeyGen}_{\lambda} \\ c \leftarrow \mathsf{Enc}_{\lambda}(\mathsf{pk}, m) \end{array} \right] \ge 1 - \mathsf{negl}(\lambda)$$

- (k, l)-CCA-security: For any  $\mathcal{G} = (\mathcal{G}_1, \mathcal{G}_2) = (\{g_{1,\lambda}\}, \{g_{2,\lambda}\}) \in \mathcal{C}_2$ , and any  $\lambda \in \mathbb{N}$ :

$$\Pr\left[b = \tilde{b} \begin{vmatrix} (\mathsf{sk}, \mathsf{pk}) \leftarrow \mathsf{KeyGen}_{\lambda} \\ b \leftarrow \{0, 1\} \\ (m_1, m_2, \mathsf{aux}) \leftarrow g_{1,\lambda}^{\mathsf{Dec}_{\lambda}(\mathsf{sk}, \cdot)}(\mathsf{pk}) \\ c \leftarrow \mathsf{Enc}_{\lambda}(\mathsf{pk}, m_b) \\ \tilde{b} \leftarrow g_{2,\lambda}^{\mathsf{Dec}_{\lambda}(\mathsf{sk}, \cdot)}(c, \mathsf{aux}) \end{vmatrix}\right] \leq \frac{1}{2} + \mathsf{negl}(\lambda)$$

where  $g_{1,\lambda}$  and  $g_{2,\lambda}$  are allowed to do k rounds of adaptive queries to  $\mathsf{Dec}_{\lambda}(\mathsf{sk},\cdot)$  and each round they query l inputs.

# D.2 Construction

We now give our construction of  $NC^1$ -CCA PKE against  $NC^1$  from perfectly smooth  $NC^1$ -HPS against  $NC^1$  and strong universal<sub>2</sub>  $NC^1$ -HPS against  $NC^1$ .

**Construction 8** Let  $\mathcal{P}' = (Setup', Samp \mathcal{Y}es', Samp \mathcal{N}o', Key \mathcal{G}en', \mathcal{P}riv', \mathcal{P}ub')$  be perfectly smooth  $\mathsf{NC}^1$ -HPS against  $\mathsf{NC}^1$  defined as Construction 6 and  $\mathcal{P}^{\dagger} = (Setup^{\dagger}, Samp \mathcal{Y}es^{\dagger}, Samp \mathcal{N}o^{\dagger}, Key \mathcal{G}en^{\dagger}, \mathcal{P}riv^{\dagger}, \mathcal{P}ub^{\dagger})$  be strong universal<sub>2</sub>  $\mathsf{NC}^1$ -HPS against  $\mathsf{NC}^1$  defined as Construction 7. Then, define (Key  $\mathcal{G}en, \mathcal{E}nc, \mathcal{D}ec$ ) as follows.

- KeyGen<sub> $\lambda$ </sub>:
  - 1. Run pp' :=  $(X'_{\lambda}, L'_{\lambda}, W'_{\lambda}, R'_{\lambda}, SK'_{\lambda}, PK'_{\lambda}, H'_{\lambda}, \Pi'_{\lambda}, \alpha'_{\lambda}) \leftarrow \text{Setup}'_{\lambda} and (pk', sk') \leftarrow \text{KeyGen}'_{\lambda}(pp').$
  - 2. Set  $pp^{\dagger} := (X_{\lambda}^{\dagger}, L_{\lambda}^{\dagger}, W_{\lambda}^{\dagger}, R_{\lambda}^{\dagger}, SK_{\lambda}^{\dagger}, PK_{\lambda}^{\dagger}, H_{\lambda}^{\dagger}, \Pi_{\lambda}^{\dagger}, \alpha_{\lambda}^{\dagger})$  based on the procedure of Setup<sup>1</sup><sub>\lambda</sub>.
  - 3. Set  $pk := (pp', pp^{\dagger}, pk', pk^{\dagger})$  and  $sk := (sk', sk^{\dagger})$ .
  - 4. *Output* (pk, sk).
- $Enc_{\lambda}(pk, m)$ :
  - 1. Parse  $pk := (pp', pp^{\dagger}, pk', pk^{\dagger}).$
  - 2. Run  $(\mathbf{x}, \mathbf{w}) \leftarrow \mathsf{SampYes}'_{\lambda}(\mathsf{pp'}).$
  - 3. Run  $\pi' := \mathsf{Pub}'_{\lambda}(\mathsf{pp}', \mathbf{pk}', \mathbf{x}, \mathbf{w})$  and compute  $\mathbf{c} := m \oplus \pi' \in \{0, 1\}^{n(\lambda)}$ .

4. Run  $\pi^{\dagger} := \mathsf{Pub}_{\lambda}^{\dagger}(\mathsf{pp}^{\dagger}, \mathbf{pk}^{\dagger}, (\mathbf{x}, c), \mathbf{w}).$ 5. Set  $c := (\mathbf{x}, c, \pi^{\dagger})$  and output c.–  $\mathsf{Dec}_{\lambda}(\mathsf{sk}, c):$ 1. Parse  $\mathsf{sk} := (\mathsf{sk}', \mathsf{sk}^{\dagger})$  and  $c := (\mathbf{x}, c, \pi^{\dagger}).$ 2. Run  $\tilde{\pi}^{\dagger} := \mathsf{Priv}_{\lambda}^{\dagger}(\mathsf{pp}^{\dagger}, \mathsf{sk}^{\dagger}, (\mathbf{x}, c)).$ 3. If  $\pi^{\dagger} \neq \tilde{\pi}^{\dagger}$ , then output  $\bot$  and halt. 4. Run  $\tilde{\pi}' := \mathsf{Priv}_{\lambda}'(\mathsf{pp}', \mathsf{sk}', \mathbf{x})$  and compute  $\tilde{m} := c \oplus \tilde{\pi}'.$ 5. Output  $\tilde{m}.$ 

**Theorem 4** Let k be a constant and l be a polynomial in  $\lambda$ . If  $NC^1 \subsetneq \oplus L/poly$ , then (KeyGen, Enc, Dec) defined as Construction 6 is a (k, l)-CCA-secure  $NC^1$ -PKE scheme against  $NC^1$ .

We note that the algorithms  $\mathcal{K}ey\mathcal{G}en$ ,  $\mathcal{E}nc$ , and  $\mathcal{D}ec$  compute XOR, equality comparison, and run  $\mathcal{P}'$  and  $\mathcal{P}^{\dagger}$  (see Section B and C for  $\mathcal{P}'$  and  $\mathcal{P}^{\dagger}$ ). Since all these operations can be performed in NC<sup>1</sup>, we have  $\mathcal{K}ey\mathcal{G}en$ ,  $\mathcal{E}nc$ ,  $\mathcal{D}ec \in NC^1$ .

The proofs of correctness and CCA security are the same as the proofs by Cramer and Shoup [15], and we thus omit the details. Note that the adversary makes constant rounds of queries, and in each round, the simulation for queries made by the reduction algorithm is performed in  $NC^1$ . Moreover, one can see that the rest operations in the reduction algorithm can also be performed in  $NC^1$ . Therefore, Construction 8 is a CCA-secure  $NC^1$ -PKE scheme against  $NC^1$ , completing the proof of Theorem 4.

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