# **Upgrading Fuzzy Extractors**

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Abstract. Fuzzy extractors derive stable keys from noisy sources non-interactively (Dodis et al., SIAM Journal of Computing 2008). Since their introduction, research has focused on two tasks: 1) showing security for as many distributions as possible and 2) providing stronger security guarantees including allowing one to enroll the same value multiple times (reusability), security against an active attacker (robustness), and preventing leakage about the enrolled value (privacy).

Existing constructions of reusable fuzzy extractors are direct and do not support as many distributions as the best non-reusable constructions. Constructions of robust fuzzy extractors require strong assumptions even in the CRS model.

Given the need for progress on the basic fuzzy extractor primitive, it is prudent to seek generic mechanisms to transform a fuzzy extractor into one that is robust, private, and reusable so that it can inherit further improvements.

This work asks if one can generically upgrade fuzzy extractors to achieve robustness, privacy, and reusability. We show positive and negative results: we show upgrades for robustness and privacy, but we provide a negative result on reuse.

- We upgrade (private) fuzzy extractors to be robust under weaker assumptions than previously known in the common reference string model.
- 2. We show a generic upgrade for a private fuzzy extractor using multibit compute and compare (MBCC) obfuscation (Wichs and Zirdelis, FOCS 2017) that requires less entropy than prior work.
- 3. We show one cannot arbitrarily compose private fuzzy extractors. It is known one cannot reuse an arbitrary fuzzy extractor; each enrollment can leak a constant fraction of the input entropy.

We show that one cannot build a reusable private fuzzy extractor by considering other enrollments as auxiliary input. In particular, we show that assuming MBCC obfuscation and collision-resistant hash functions, there does not exist a private fuzzy extractor secure against unpredictable auxiliary inputs strengthening a negative result of Brzuska et al. (Crypto 2014).

Keywords: Fuzzy extractors, obfuscation, biometrics, key derivation.

### 1 Introduction

Fuzzy extractors [BBR88, DRS04, DORS08, ŠTO05, HAD06, DKRS06, FMR13, CFP<sup>+</sup>16,FRS16,ACEK17,ABC<sup>+</sup>18,FP19,WLH18,WL18,DFR21,ACF<sup>+</sup>22,FRS20,Ful23] derive stable keys from noisy sources. They are used on devices to derive keys from biometrics, physical unclonable functions and quantum information. They also are used in interactive protocols such as distributed key agreement and password-authenticated key exchange. [BBCS91,DKRS06,BG11,EHKM11, DKK<sup>+</sup>12,BCP13,BDCG13,DCH<sup>+</sup>16,DHP<sup>+</sup>18]. A fuzzy extractor is a pair of algorithms called generate (Gen) and reproduce (Rep) with two properties:

**Correctness** For all  $w, w' \in \mathcal{M}$  such that  $\operatorname{dist}(w, w') \leq t$ , let  $(\text{key}, \text{pub}) \leftarrow \operatorname{\mathsf{Gen}}(w)$  where  $\operatorname{\mathsf{pub}}$  is a public *helper* value used to provide correctness. Then it should be the case that  $\operatorname{\mathsf{key}} \leftarrow \operatorname{\mathsf{Rep}}(w', \operatorname{\mathsf{pub}})$ .

**Security** Let W be a probability distribution of noisy values. For  $(\text{key}, \text{pub}) \leftarrow \text{Gen}(W)$  it should be the case that key is indistinguishable from uniform even knowing pub.

Security is defined relative to the statistics of the probability distribution. The most common and useful of which is fuzzy min-entropy [FRS16,WCD+17,FRS20] which measures the adversary's success when provided with the functionality of reproduce. For security to be possible, it must be the case that a negligible fraction of the weight of W lies within any ball of radius t. Fuzzy min-entropy measures the adversary's success when providing the fixed "best" point  $w^*$  to the reproduce functionality. Even after 25 years of research, the design of fuzzy extractors for distributions with fuzzy min-entropy is an unsettled problem with advancements yet to be made.

There are constructions for distributions with high entropy, where bits are independent, or display additional statistical properties (see [DFR21] for an overview of considered properties). There are two known methods to build a fuzzy extractor for all such distributions, using virtual grey box obfuscation for NC1 evasive circuits [BCKP17]<sup>4</sup> or with a new subset-product assumption [GZ19]. Both of these assumptions require additional study before deployment.

Fuzzy extractor security is also insufficient for many applications. There are three primary augmentations to the definition that exist in the literature:

**Reusability** [Boy04] One can enroll the noisy source multiple times with different devices. Crucially, the multiple enrollments are subject to noise. In prior work [Boy04, CFP<sup>+</sup>16], this noise is controlled by an adversary.

Robustness [BDK<sup>+</sup>05] If an attacker modifies pub to a related value pub', this behavior is detectable. That is, Rep(w', pub') should only output the original key or  $\bot$ .

<sup>&</sup>lt;sup>4</sup> Virtual gray box obfuscation of all evasive programs implies virtual gray box obfuscation for all programs [BBC<sup>+</sup>14]. Virtual gray box and virtual black box obfuscation are equivalent in the distributional setting for evasive circuit families [BCKP17].

Scheme	Model	Distribution	Reuse	Robust	Private
[Boy04]	RO*	High entropy	Shift	✓	Х
[DS05]	Plain	High entropy	X	X	✓
$[CFP^+16]$		Avg. Subsets Entropy	Correlation	×	Х
[ACEK17]		Independent	Shift	X	Х
[BCKP17]	Plain	All	×	×	✓
[WLH18]	CRS	High entropy	Shift	X	Х
[WL18]	CRS	High entropy	Shift	✓	Х
[WLG19]	CRS	High entropy	Shift	✓	Х
[GZ19]		All	X	X	✓
[DFR21]		MIPURS	Correlation	X	X

**Table 1.** Previous constructions of fuzzy extractors that are reusable, robust, or private. See Demarest, Fuller, and Russell [DFR21] for descriptions of distributional properties. "High entropy" is used when the construction relies on an information-theoretic error correction component. Such constructions usually require the input source W to have entropy of at least  $h_2(t/n)$ , see [CFP<sup>+</sup>16, Proposition 1]. For Boyen's [Boy04] work, the RO model is only required for insider security, when keys are seen from other enrollments.

**Privacy** [**DS05**] Privacy ensures no information is leaked about the enrolled value. More specifically, it ensures that no predicate of the enrollment value can be guessed better after seeing pub.

Table 1 summarizes prior constructions of fuzzy extractors with at least one of these additional properties. No previous construction that is reusable or robust supports all distributions with fuzzy min-entropy. The prior gray-box obfuscation [BCKP17] and subset product constructions [GZ19] are obfuscations of the fuzzy extractor functionality; they are by definition private. Given the unsettled nature of constructing fuzzy extractors for distributions with fuzzy min-entropy, it is prudent to seek generic mechanisms to transform a fuzzy extractor into a reusable, robust, and private one.

### 1.1 Our Contribution

We present three contributions, 1) an upgrade for privacy 2) an upgrade for robustness that preserves privacy, and 3) a negative result for reusability.

Privacy We show how to construct a private fuzzy extractor from either a secure sketch or a non-private fuzzy extractor. Our contribution is a strengthening of the previous upgrade from a secure sketch to a private secure sketch using multibit compute and compare obfuscation (MBCC) [WZ17]. We support a wider family of distributions with lower entropy than prior work [WZ17]. We first introduce Wichs and Zirdelis' [WZ17] construction and then the advantages of our construction.

Both our work and prior work is based on the the notion of a secure sketch. A secure sketch recovers the original value w rather than deriving a random key. It is a pair of algorithms (Sketch, Rec) such that

Correctness For all  $w, w' \in \mathcal{M}$  such that  $\operatorname{dist}(w, w') \leq t$ , then  $\operatorname{Rec}(w', \operatorname{Sketch}(w)) = w$ .

**Security** Let W be a probability distribution of noisy values. Given  $\mathsf{Sketch}(W)$ , W has high min-entropy. (One can also use computational notions of security using pseudoentropy or unpredictability [FMR20].)

Wichs and Zirdelis use MBCC program obfuscation at the core of their scheme. MBCC program has three values f, y, z. On input x it computes f(x), if f(x) = y then it outputs z otherwise it outputs  $\bot$ . In their prior construction they showed how to obfuscate a family of such programs where y has pseudoentropy [HILL99] conditioned on f and z. They show how to upgrade a secure sketch into a private one by obfuscating  $h(\text{Rec}(\cdot, ss))$  where h is a pairwise independent hash function. They have to choose the output length of the hash function based on the entropy of the input, which keeps the construction from working for all distributions with sufficient entropy for the MBCC obfuscation to be secure.

In our work, we use a secure sketch to construct a private fuzzy extractor to directly analyze the construction without h, making the same construction work for any distribution where the secure sketch retains (a super-logarithmic amount of) min-entropy. By removing the hash function, we also reduce the amount of entropy required as one doesn't "leak" the hash value in the security analysis. We also show a similar upgrade from fuzzy extractors to private fuzzy extractors. There are stronger negative results on constructing secure sketches [DORS08, FRS16, Ful23], so direct constructions from fuzzy extractors may yield better parameters.

The privacy definition of Wichs and Zirdelis [WZ17] considers predicting predicates of the source w in contrast to Dodis and Smith [DS05] who consider functions. We call this weak-privacy to distinguish from Dodis and Smith's definition. Restriction to predicates is standard in the obfuscation literature as an obfuscation itself is a function that a simulator cannot hope to reproduce.

Robustness We provide a simpler construction of robust fuzzy extractors than prior work. Our result only requires the existence of true simulation extractable NIZKs [DHLAW10]. Prior work of Feng and Tang [FT21] also required the existence of extremely lossy functions or ELFs [Zha19].<sup>5</sup> Additionally, our result shows that this transform preserves privacy, which was not considered by any prior robustness upgrade.

Reuse One cannot expect reuse of arbitrary fuzzy extractors. Each value of pub can leak a constant fraction of the entropy in the source w while remaining secure. However, a (weakly) private fuzzy extractor cannot "leak" on input w. There are multiple private fuzzy extractors (see Table 1) that are not known to be reusable. The most natural approach for a reusable private fuzzy extractor

 $<sup>^5</sup>$  Feng and Tang's primary goal was to construct robust extractors, not robust fuzzy extractors. Unlike Feng and Tang we work in the standard CRS model, they allow the source W to depend on the CRS.

Upgrade	Scheme	Model	Any FE	Tools	Err.
Reusability	[ABC <sup>+</sup> 18]	Plain	✓	composable DL [BC10]	t
Robustness	[BDK <sup>+</sup> 05]	RO	Х	RO	t
	$[CDF^+08]$	CRS	Х	$\operatorname{IT}$	2t
	[DKK <sup>+</sup> 12]	Plain	Х	$\operatorname{IT}$	t
	[FT21]	CRS*	✓	ELFS [Zha19] +	2t
				true sim. extract NIZK [DHLAW10]	
	$[ACF^+22]$	Plain	Х	comp.* DL [BLMZ19, Assumption 3]+	2t
				true sim. extract NIZK [DHLAW10]	
	This work	CRS	✓	true sim. extract NIZK [DHLAW10]	2t
Private	[WZ17]	Plain	✓	LWE + ELFS [Zha19]	t
	This work	Plain	✓	LWE + ELFS [Zha19]	t

Table 2. Previous upgrades of fuzzy extractors. If there is an X in the Any FE. column the construction requires the use of the syndrome secure sketch. As mentioned in Table 1 this places a lower bound on entropy of the distribution W. CRS\* is the CRS model where the distribution W being enrolled can depend on the CRS. Err. column describes how many errors the underlying primitive is required to correct. Multiple robust constructions require a secure sketch or fuzzy extractor that corrects 2t errors to be able to extract a value from the adversary.

is to construct a fuzzy extractor for all sources W that are unpredictable in the presence of auxiliary input available to the adversary. The security analysis would follow by including other enrollments of the same source in the auxiliary input.

We show this proof technique is not possible. Namely, we show that the existence of MBCC obfuscation and collision-resistant hash functions imply that one cannot construct private fuzzy extractors for all W that are unpredictable conditioned on auxiliary input. We do this by showing that auxiliary-input secure digital lockers cannot be secure in the presence of MBCC obfuscation. We then show that a variant of private fuzzy extractors imply digital lockers. Brzuska et al. [BFM14] proved an analogous result where auxiliary-input secure digital lockers were incompatible with indistinguishability obfuscation [GGH+13b, GGH13a]. Since MBCC obfuscation implies auxiliary-input secure digital lockers, this shows MBCC obfuscation cannot be safely composed either.

### 1.2 Related Work

Reusability Alamelou et al. [ABC<sup>+</sup>18] show how to create reuse for the Hamming and set difference metrics when the source has symbols that are super polynomial size. However, most natural sources consider small, often binary, alphabets.

<sup>&</sup>lt;sup>6</sup> Their actual result showed the impossibility of auxiliary input universal computational extractors. This object implies auxiliary-input secure digital lockers

Alamelou et al.'s technique cannot work in this setting.<sup>7</sup> We note this technique is applied before the source is input to the fuzzy extractor.

Robustness In the random oracle model, for a fuzzy extractor with output key, pub and random oracle h one can split key =  $(\text{key}_1, \text{key}_2)$  and include  $h(\text{key}_2||\text{pub})$  as part of pub. As needed the random oracle can expand the amount of available keying material.<sup>8</sup> Without resorting to random oracles, one can use algebraic manipulation detection codes [CDF+08] and pairwise independent hashes as one-time MACs. In the CRS model, Feng and Tang [FT21] codify the security required from the MAC, and show how to generically lift a secure sketch into a robust fuzzy extractor using a primitive they call a  $\kappa$ -MAC that is secure for low-entropy keys that can be manipulated by an adversary. This upgrade is agnostic in the underlying secure sketch. Apon et al. [ACF+22] propose a standard model upgrade that requires the syndrome secure sketch.

Privacy For privacy, if one has a secure sketch that retains superlogarithmic entropy, one can upgrade it to a private secure sketch using multi-bit compute and compare (MBCC) obfuscation [WZ17]. Roughly, MBCC allows one to compute a function on some input and compare the result with a target value. If the output of the function matches the target, the MBCC circuit returns a fixed value.

### 1.3 Discussion and Future Work

The privacy upgrade is this work is not yet of practical efficiency. MBCC obfuscation has nearly as much overhead as indistinguishability obfuscation. A natural question is whether an upgrade that preserves privacy must use a type of obfuscation and if so, can one use a obfuscation of a weaker class of obfuscation?

Our negative result for reuse does leave open the possibility of upgrading fuzzy extractors to be reusable. It does not rule out techniques that transform w to a new metric space [ABC<sup>+</sup>18]. Furthermore, one may able to use a more fine grained argument for reuse. As a reminder, our negative result only rules out private fuzzy extractors secure in the presence of unpredictable auxiliary input. One may be able to sidestep the result by only showing security when the auxiliary input is a fuzzy extractor enrollment. We tried to extend our negative result to this setting but were not successful.

<sup>&</sup>lt;sup>7</sup> Alamelou et al. use a pseudoentropic isometry that maps points to a new metric space while 1) preserving distance and 2) the value in the new metric space doesn't reveal the value on the original metric space. For the Hamming metric, the only such transforms are equivalent to a per-symbol permutation and a permutation of symbol order. Such a transform can only be one-way if symbols are super-polynomial size. No pseudoentropic isometric exists for the Hamming metric with polynomial size symbols.

<sup>&</sup>lt;sup>8</sup> Boyen [BDK<sup>+</sup>05] considers a secure sketch, the same idea works for a fuzzy extractor.

Organization The rest of this work is organized as follows: Section 2 covers mathematical preliminaries, Section 3 shows our privacy upgrade, Section 4 covers robustness, and Section 5 covers reuse. Appendix A shows that weak-privacy does not imply fuzzy extractor security and Appendix B shows that a composable MBCC obfuscation would yield a reusable upgrade (but is ruled out by our negative result).

### 2 Preliminaries

Let  $\lambda$  be the security parameter throughout this paper. A function  $\operatorname{ngl}(\lambda)$  is negligible in  $\lambda$  if for all  $a \in \mathbb{Z}^+$  we have  $\operatorname{ngl}(\lambda) = o(\frac{1}{\lambda^a})$ . A function  $\operatorname{poly}(\lambda)$  is polynomial in  $\lambda$  if there exists some constant  $a \in \mathbb{Z}^+$  such that  $\operatorname{poly}(\lambda) = O(\lambda^a)$ . We use  $\operatorname{poly}(\lambda)$  and  $\operatorname{ngl}(\lambda)$  to denote unspecified functions that are polynomial and negligible in  $\lambda$ , respectively. The notation id is used to denote the identity function:  $\forall x, \operatorname{id}(x) = x$ . For some  $n \in \mathbb{N}$ , [n] denotes the set  $\{1, \cdots, n\}$ . Let  $x \stackrel{\$}{\leftarrow} S$  denote sampling x uniformly at random from the finite set S. We say that distributions X and Y are computationally indistinguishable if for all PPT (in  $\lambda$ ) adversaries A,  $|\operatorname{Pr}[A(X) = 1] - \operatorname{Pr}[A(Y) = 1]| \leq \operatorname{ngl}(\lambda)$ .

### 2.1 Entropy definitions

**Definition 1 (Min-entropy).** For a discrete random variable X, the min-entropy of X is

$$H_{\infty}(X) = -\log\left(\max_{x} \Pr[X = x]\right)$$

**Definition 2 (Average conditional min-entropy [DORS08]).** For a pair of discrete random variables X, Y, the average min-entropy of X|Y is

$$\tilde{\mathrm{H}}_{\infty}(X\mid Y) = -\log\left(\mathit{Exp}_{y\in Y}\left(2^{-\mathrm{H}_{\infty}(X\mid Y)}\right)\right).$$

**Definition 3 (Conditional HILL entropy [HILL99, HLR07]).** Let X, Y be ensembles of jointly distributed random variables. The conditional pseudo-entropy of X conditioned on Y, denoted as  $H_{\mathsf{HILL}}(X \mid Y)$ , is greater or equal to  $\ell(\lambda)$  if there exists some ensemble X' such that (X,Y) and (X',Y) are computationally indistinguishable and  $H_{\infty}(X' \mid Y) \geq \ell(\lambda)$ .

#### 2.2 Obfuscation definitions

Definition 4 (Distributional Virtual Black Box (dist-VBB) obfuscation). Let  $\mathcal{P}$  be a family of programs and Obf be a PPT algorithm that takes as input a program  $P \in \mathcal{P}$ , a security parameter  $\lambda \in \mathbb{N}$  and outputs a program  $\tilde{P} \leftarrow \mathsf{Obf}(1^{\lambda}, P)$ . Let  $\mathcal{D}$  be a class of distribution ensembles  $D = \{D_{\lambda}\}_{{\lambda} \in \mathbb{N}}$  which samples  $(P, \mathsf{aux}) \leftarrow D_{\lambda}$  with  $P \in \mathcal{P}$ . Then Obf is an obfuscator for the distribution class  $\mathcal{D}$  over the program family  $\mathcal{P}$  if it satisfies the following:

- Functionality preserving: For all  $P \in \mathcal{P}$  and for all inputs  $x \in \{0,1\}^n$ , we have

$$\Pr[P(x) = \tilde{P}(x)] \ge 1 - \mathsf{ngl}(\lambda)$$

- **Polynomial slowdown:** For all sufficiently large  $\lambda \in \mathbb{N}$  and for all  $P \in \mathcal{P}_{\lambda}$ ,

$$|\tilde{P}| \leq \mathsf{poly}(|P|)$$

- **Distributional Virtual Black-Box:** For every PPT adversary  $\mathcal{A}$  there exists a non-uniform polynomial size simulator Sim, such that for every distribution ensemble  $D = \{D_{\lambda}\} \in \mathcal{D}$ , and every predicate  $\phi : \mathcal{P} \to \{0,1\}$ , we have

$$\begin{split} & \Big| \Pr_{(P,\mathsf{aux}) \leftarrow D_{\lambda}} [\mathcal{A}(\mathsf{Obf}(1^{\lambda},P),\mathsf{aux}) = \phi(P)] \\ & - \Pr_{(P,\mathsf{aux}) \leftarrow D_{\lambda}} [\mathsf{Sim}^P(1^{\lambda},1^{|P|},\mathsf{aux}) = \phi(P)] \Big| \leq \mathsf{ngl}(\lambda) \end{split}$$

where  $Sim^P$  has black-box access to the program P.

Wichs and Zirdelis [WZ17] build a dist-VBB obfuscator for  $\alpha$ -pseudo entropy distributions (see Definition 6) over multi-bit-compute-and-compare circuits.<sup>9</sup>

**Definition 5 (Multi-bit compute-and-compare circuit).** Let  $n, \ell, \kappa \in \mathbb{N}$  and consider a function  $f: \{0,1\}^n \to \{0,1\}^\ell$ , a target value  $y \in \{0,1\}^\ell$  and some value  $z \in \{0,1\}^\kappa$ . A multi-bit compute-and-compare circuit is defined for all inputs  $x \in \{0,1\}^n$  as

$$\mathsf{MBCC}_{f,y,z}(x) = \begin{cases} z & \textit{if } f(x) = y \\ \bot & \textit{otherwise}. \end{cases}$$

Wich s and Zirdelis [WZ17] also define  $\alpha$ -pseudo entropy, a specific case of HILL entropy:

**Definition 6** ( $\alpha$ -pseudo entropy). For function  $\alpha(\lambda)$ , the class of  $\alpha$ -pseudo-entropy distributions consists of ensembles  $D = \{D_{\lambda}\}$  such that  $(\mathsf{MBCC}[f, y, z], \mathsf{aux}) \leftarrow D_{\lambda}$  satisfies  $H_{\mathsf{HILL}}(y \mid f, z, \mathsf{aux}) \geq \alpha(\lambda)$ .

#### 2.3 Fuzzy extractors

Fuzzy extractors allow to generate stable cryptographic keys from noisy sources. We focus on computational fuzzy extractors.

Definition 7 (Computational Fuzzy Extractor [FMR13, FMR20]). An  $(\mathcal{M}, \mathcal{W}, \ell, t, \epsilon)$ -fuzzy extractor with error  $\delta$  is a pair of PPT algorithms (Gen, Rep) where for all  $w, w' \in \mathcal{M}$ ,

<sup>&</sup>lt;sup>9</sup> In an independent and concurrent work, Goyal et al. [GKW17] proposed a similar object they called *lockable obfuscation*.

- (key, pub)  $\leftarrow$  Gen(w), where key  $\in \{0,1\}^{\ell}$  and pub  $\in \{0,1\}^*$  - key'  $\leftarrow$  Rep(pub, w')

the following properties are true:

1. Correctness: For all  $w, w' \in \mathcal{M}$  such that  $dist(w, w') \leq t$ ,

$$\Pr\left[\mathsf{key'} = \mathsf{key} \ \big| \ (\mathsf{key}, \mathsf{pub}) \leftarrow \mathsf{Gen}(w), \mathsf{key'} \leftarrow \mathsf{Rep}(\mathsf{pub}, w')\right] \geq 1 - \delta$$

2. **Security**: For any PPT distinguisher A and distribution  $W \in W$ ,

$$|\Pr[\mathcal{A}(\mathsf{key},\mathsf{pub})=1] - \Pr[\mathcal{A}(U_{\ell},\mathsf{pub})=1]| \leq \epsilon$$

where (key, pub)  $\leftarrow$  Gen(W) and  $U_{\ell}$  is a uniformly distributed random variable over  $\{0,1\}^{\ell}$ .

### 3 Weakly-private fuzzy extractors

Fuzzy extractor security does not prevent leaking information about the value w, called a template. For example, consider a fuzzy extractor where the public value leaks a random bit of the template. This can be problematic, especially if the biometric source is used in different contexts. Preventing such leakage, although not mandatory to achieve fuzzy extractor security, is thus desirable. Constructions that prevent such leakage are said to be private [DS05]. We adapt Wichs and Zirdelis [WZ17] privacy definition for secure sketches to fuzzy extractors, we call this weak privacy. This definition differs from Dodis and Smith's [DS05] in that the adversary is restricted to predicting predicates about the value W (in place of general functions). We start by introducing the definition of a weakly private fuzzy extractor.

**Definition 8 (Weakly Private Fuzzy Extractor).** Let FE = (Gen, Rep) satisfy the correctness condition of Definition 7 for parameters t and  $\delta$ . We say that FE is  $\eta$ -weakly-private if for all adversary  $\mathcal{A}$ , there exists a simulator Sim such that for every source W over W and every predicate  $\phi: \{0,1\}^* \to \{0,1\}$ , we have

$$\label{eq:prediction} \begin{split} \Big| & \Pr[\mathcal{A}(\mathsf{pub}, \mathsf{key}) = \phi(W) \mid (\mathsf{key}, \mathsf{pub}) \leftarrow \mathsf{FE}.\mathsf{Gen}(W)] \\ & - \Pr[\mathsf{Sim}(1^{\lambda}, 1^{|\mathsf{pub}|}, 1^{|\mathsf{key}|}) = \phi(W)] \ \Big| \leq \eta \end{split}$$

Fuzzy extractors were originally built following a *sketch-then-extract* approach. First, a secure sketch [DRS04] is used to recover the enrolled w from a close value w', then a randomness extractor is used to derive the secret key. We add the definition for secure sketches:

**Definition 9 (Secure sketch).** Let  $\lambda$  be a security parameter. Let  $W = W_{\lambda}$  be a family of random variables over the metric space  $(\mathcal{M}, \mathsf{dist}) = (\mathcal{M}_{\lambda}, \mathsf{dist}_{\lambda})$ . Then (Sketch, Rec) is a  $(\mathcal{M}, \mathcal{W}, \ell, t, \delta)$ -secure sketch if the following hold:

- Correctness: For all  $w, w' \in \mathcal{M}$  such that  $\operatorname{dist}(w, w') \leq t$ ,

$$\Pr[\mathsf{Rec}(w',\mathsf{Sketch}(w)) = w] \ge 1 - \delta.$$

- **Security:** For all distributions  $W \in \mathcal{W}$  it is true that

$$H_{\infty}(W \mid \mathsf{Sketch}(W)) \geq \ell$$
.

We propose two weakly private fuzzy extractors constructions using dist-VBB obfuscation for multi-bit-compute-and-compare (MBCC) circuits. The first construction builds on a non-private fuzzy extractor whereas the second builds on a non-private secure sketch.

Intuitively, we can build weakly private fuzzy extractors as follows: we first build an MBCC circuit for function  $f_{w,y,t}$  that outputs target value y on input w' only when  $\operatorname{dist}(w,w') \leq t$  and we set the output value z to be sampled uniformly at random. We then set pub to be the obfuscated MBCC program and key to be z. Note that in this case, since z is sampled independently from all other values, the entropy requirement for the MBCC circuit to be obfuscable can be simplified to

$$H_{\mathsf{HILL}}(y \mid f, \mathsf{aux}) \geq \alpha.$$

### 3.1 Weakly private FE from FE and MBCC obfuscation

Our first construction builds weakly-private fuzzy extractors from non-private fuzzy extractors and MBCC obfuscation.

Construction 1 (Weakly Private FE from MBCC obfuscation and FE) Let FE be an  $(\mathcal{M}, \mathcal{W}, \ell, t, s, \epsilon)$ -fuzzy extractor and Obf be an obfuscator for  $\ell$ pseudo-entropy distributions over multi-bit compute-and-compare circuits. We
can build an  $(\mathcal{M}, \mathcal{W}, \kappa, t, s, \epsilon)$ -fuzzy extractor PFE as follows:

- $(\text{key}', \text{pub}') \leftarrow \text{PFE.Gen}(w)$ :
  - 1.  $Compute (key, pub) \leftarrow FE.Gen(w)$ .
  - 2. Sample  $\ker' \stackrel{\$}{\leftarrow} \{0,1\}^{\kappa}$ .
  - 3. Define the circuit  $f_{\mathsf{pub}}(\cdot) := \mathsf{FE}.\mathsf{Rep}(\cdot,\mathsf{pub})$ .
  - 4. Compute  $pub' \leftarrow Obf(1^{\lambda}, MBCC_{f_{pub}, key, key'})$ .
  - 5. Output (key', pub').
- $\text{key}' \leftarrow \text{PFE.Rep}(\text{pub}', w')$ : Interpret pub' as an obfuscated program and return  $\text{key}' \leftarrow \text{pub}'(w')$ .

**Theorem 1.** Construction 1 is a secure and weakly-private  $(\mathcal{M}, \mathcal{W}, \kappa, t, s, \epsilon)$ -fuzzy extractor.

Proof (Theorem 1).

Correctness: Recall that pub' is an obfuscated MBCC circuit such that

$$\begin{split} \mathsf{pub'}(w') &= \mathsf{Obf} \big( 1^\lambda, \mathsf{MBCC}_{f_{\mathsf{pub}}, \mathsf{key}, \mathsf{key'}} \big)(w') \\ &= \mathsf{MBCC}_{f_{\mathsf{pub}}, \mathsf{key}, \mathsf{key'}}(w') \\ &= \begin{cases} \mathsf{key'} & \text{if} \quad f_{\mathsf{pub}}(w') = \mathsf{key} \\ \bot & \text{otherwise}. \end{cases} \\ &= \begin{cases} \mathsf{key'} & \text{if} \quad \mathsf{FE.Rep}(w', \mathsf{pub}) = \mathsf{key} \\ \bot & \text{otherwise}. \end{cases} \end{split}$$

Then since FE is a fuzzy extractor, it is true that

$$\begin{split} & \Pr\left[\mathsf{PFE}.\mathsf{Rep}(\mathsf{pub}',w') = \mathsf{key}' \mid (\mathsf{pub}',\mathsf{key}') \leftarrow \mathsf{PFE}.\mathsf{Gen}(w) \text{ and } \mathsf{dist}(w,w') \leq t\right] \\ & = \Pr\left[\mathsf{FE}.\mathsf{Rep}(\mathsf{pub},w') = \mathsf{key} \mid (\mathsf{pub},\mathsf{key}) \leftarrow \mathsf{FE}.\mathsf{Gen}(w) \text{ and } \mathsf{dist}(w,w') \leq t\right] \\ & \geq 1 - \delta \end{split}$$

and PFE is thus correct.

Security: We proceed by contradiction. Suppose PFE is not a secure fuzzy extractor, then there exists some PPT adversary  $\mathcal{A}$  and polynomial  $p(\lambda)$  such that

$$\left|\Pr[\mathcal{A}(\mathsf{key}',\mathsf{pub}')=1] - \Pr[\mathcal{A}(U_\kappa,\mathsf{pub}')=1]\right| > 1/p(\lambda)$$

where  $(\text{key}', \text{pub}') \leftarrow \text{PFE.Gen}(W)$  and  $U_{\kappa} \stackrel{\$}{\leftarrow} \{0, 1\}^{\kappa}$ . Now note that  $\text{pub}' = \text{Obf}(1^{\lambda}, \text{MBCC}_{f_{\text{pub}}, \text{key}, \text{key}'})$  is distributional VBB secure. Define  $r(\lambda) = 3p(\lambda)$  and let Sim be the simulator of  $\mathcal{A}$  for polynomial  $r(\lambda)$ . Then we have

$$\left|\Pr[\mathcal{A}(\mathsf{pub}',\mathsf{key}) = 1] - \Pr[\mathsf{Sim}^{\mathsf{pub}'}(1^{\lambda},1^{|\mathsf{pub}'|},\mathsf{key}) = 1]\right| \leq \frac{1}{3p(\lambda)}. \tag{1}$$

Note that the above is also true if key is replaced by  $U_{\kappa}$ , a uniform random variable over  $\{0,1\}^{\kappa}$ . In other words, we have

$$\left| \Pr[\mathcal{A}(\mathsf{pub}', U_{\kappa}) = 1] - \Pr[\mathsf{Sim}^{\mathsf{pub}'}(1^{\lambda}, 1^{|\mathsf{pub}'|}, U_{\kappa}) = 1] \right| \le \frac{1}{3p(\lambda)}. \tag{2}$$

We adapt Canetti et al.'s lemma [CFP+16, Lemma 2]:

**Lemma 1.** Let  $U_{\kappa}$  denote the uniform distribution over  $\{0,1\}^{\kappa}$ , then

$$\begin{split} &\left|\Pr[\mathsf{Sim}^{\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},\mathsf{key},\mathsf{key}']}\left(1^{\lambda},\left|\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},\mathsf{key},\mathsf{key}']\right|,\mathsf{key}'\right) = 1\right] \\ &-\Pr[\mathsf{Sim}^{\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},\mathsf{key},\mathsf{key}']}\left(1^{\lambda},\left|\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},\mathsf{key},\mathsf{key}']\right|,U_{\kappa}\right) = 1]\right| \\ &\leq \frac{1}{3p(\lambda)} \end{split}$$

Proof (Lemma 1). Fix any  $u \in \{0,1\}^{\kappa}$ , the lemma will follow by averaging over all u. The information about whether the key value, denoted V, is key or u can only be obtained by Sim through the query responses. First, we modify Sim to quit immediately when it gets a response not equal to  $\bot$ . Such Sim is equally successful at distinguishing between key and u since the first non- $\bot$  response tells Sim if its input is equal to key. Subsequent responses add nothing to this knowledge. Since Sim can make at most q queries, there are q+1 possible values for the view of Sim on a given input. Of those, q views consist of some number of non- $\bot$  responses followed by a  $\bot$  response, and one view consists of all q responses equal to  $\bot$ . Then by [DRS04, Lemma 2.2b],

$$\tilde{H}_{\infty}(V|View(\mathsf{Sim}),\mathsf{aux}) \geq \tilde{H}_{\infty}(V) - \log(q+1)$$
  
  $\geq \alpha - \log(q+1).$ 

where  $aux = (|MBCC[Rep_{pub}, key, key']|).$ 

Thus, at each query, the probability that Sim gets a non- $\perp$  response and guesses V is at most  $(q+1)/2^{\alpha}$ . Since there are q queries of Sim, the overall probability is at most  $q(q+1)/2^{\alpha}$ . Then since  $2^{\alpha}$  is negligible in  $\lambda$ , there exists some  $\lambda_0$  such that for all  $\lambda \geq \lambda_0$ ,  $q(q+1)/2^{\alpha} \leq 1/(3p(\lambda))$ .

We know continue the proof of Theorem 1, from Lemma 1, we have

$$\big|\Pr[\mathsf{Sim}^{\mathsf{pub}'}(1^{\lambda},1^{|\mathsf{pub}'|},\mathsf{key}')=1] - \Pr[\mathsf{Sim}^{\mathsf{pub}'}(1^{\lambda},1^{|\mathsf{pub}'|},U_{\kappa})=1]\big| \leq \frac{1}{3p(\lambda)}$$

Using the triangle inequality on Equations 1, 2 and 3 we obtain

$$|\Pr[\mathcal{A}(\mathsf{pub}',\mathsf{key}')=1] - \Pr[\mathcal{A}(\mathsf{pub}',U_{\kappa})=1]| \leq \frac{1}{p(\lambda)}$$

which is a contradiction and ends the proof of security.

Weak privacy: Let FE be an  $(\mathcal{M}, \mathcal{W}, \ell, t, \epsilon)$ -computational fuzzy extractor. Consider random variables W, aux, and  $U_{\ell}$ , the uniform distribution over  $\ell$  bit strings, and  $(\text{key}, \text{pub}) \leftarrow \text{FE.Gen}(W)$ . Then by definition, for any PPT adversary  $\mathcal{A}$ , we have

$$|\Pr[\mathcal{A}(\mathsf{key},\mathsf{pub})=1]-\Pr[\mathcal{A}(U_{\ell},\mathsf{pub})=1]| < \epsilon$$

which implies

$$H_{\mathsf{HILL}}(\mathsf{key} \mid \mathsf{pub}) = \ell.$$

Since key' is sampled independently from key,

$$\begin{split} H_{\mathsf{HILL}}\big(\mathsf{key} \mid (\mathsf{FE}.\mathsf{Rep}(\cdot,\mathsf{pub}),\mathsf{key}',\mathsf{aux})\big) &= H_{\mathsf{HILL}}\big(\mathsf{key} \mid (\mathsf{FE}.\mathsf{Rep}(\cdot,\mathsf{pub}),\mathsf{aux})\big) \\ &= H_{\mathsf{HILL}}\big(\mathsf{key} \mid \mathsf{pub}\big) \\ &\geq \ell \end{split}$$

Let  $\mathsf{Obf}$  be a distributional VBB secure obfuscator for  $\ell\text{-pseudo-entropy}$  distributions. Then

$$\mathsf{pub'} = \mathsf{Obf}\big(1^\lambda, \mathsf{MBCC}_{\mathsf{FE}.\mathsf{Rep}(\cdot, \mathsf{pub}), \mathsf{key}, \mathsf{key'}}\big)$$

can be simulated and for every A, there exists a simulator Sim such that for every predicate  $\phi$  we have:

$$\begin{split} & \left| \Pr[\mathcal{A}(\mathsf{pub'},\mathsf{aux}) = \phi(\mathsf{pub'})] - \Pr[\mathsf{Sim}^{\mathsf{MBCC}_{\mathsf{FE.Rep}(\cdot,\mathsf{pub}),\mathsf{key},\mathsf{key'}}}(1^{\lambda},\mathsf{param},\mathsf{aux}) = \phi(\mathsf{pub'})] \right| \\ & \leq \mathsf{ngl}(\lambda). \end{split}$$

Note that if we set  $\mathsf{aux} = \mathsf{key}'$  and since  $\mathsf{key}'$  is drawn randomly and independently, it is true that

$$\begin{split} &\left|\Pr[\mathcal{A}(\mathsf{pub}',\mathsf{key}') = \phi(\mathsf{pub}')] - \Pr[\mathsf{Sim}(1^{\lambda},\mathsf{param},\mathsf{key}') = \phi(\mathsf{pub}')]\right| \\ &= \left|\Pr[\mathcal{A}(\mathsf{pub}',\mathsf{key}') = \phi(W)] - \Pr[\mathsf{Sim}(1^{\lambda},\mathsf{param}) = \phi(W)]\right| \\ &\leq \mathsf{ngl}(\lambda) \end{split}$$

which concludes the proof that PFE is a weakly private fuzzy extractor.

### 3.2 Weakly private FE from secure sketch and MBCC obfuscation

Our second construction builds weakly-private fuzzy extractors from non-private secure sketches and MBCC obfuscation. Although this construction relies on a secure sketch, like Wichs and Zirdelis's private secure sketch scheme, we show that the pairwise independent hash function they use isn't necessary. This reduces the amount of entropy required and allows support of a wider family of distributions. However, we build a fuzzy extractor not a secure sketch, some constructions may rely on the functionality of a secure sketch.

Construction 2 (Weakly Private Fuzzy Extractor from SS and MBCC) Let (Sketch, Rec) be an  $(\mathcal{M}, \mathcal{W}, \ell, t, \delta)$ -secure sketch and Obf be an obfuscator for  $\ell$ -pseudo-entropy distributions over multi-bit compute-and-compare circuits. Then we can build an  $(\mathcal{M}, \mathcal{W}, \kappa, t, s, \epsilon)$ -fuzzy extractor PFE as follows:

- (key, pub) ← PFE.Gen(w):
  - 1. Compute  $SS \leftarrow Sketch(w)$ .
  - 2. Sample key  $\stackrel{\$}{\leftarrow} \{0,1\}^{\kappa}$ .
  - 3. Define the circuit  $f_{SS}(\cdot) := Rec(\cdot, SS)$ .
  - 4. Compute pub  $\leftarrow \mathsf{Obf}(1^{\lambda}, \mathsf{MBCC}_{fss, w, \mathsf{key}})$ .
  - 5. Output (key, pub).
- key  $\leftarrow$  PFE.Rep(pub, w'): Interpret pub as an obfuscated program and return key  $\leftarrow$  pub(w').

**Theorem 2.** Construction 2 is a secure and weakly private  $(\mathcal{M}, \mathcal{W}, \kappa, t, s, \epsilon)$ -fuzzy extractor.

Proof (Theorem 2).

Correctness: Recall that pub' is an obfuscated MBCC circuit such that

$$\begin{split} \mathsf{pub}(w') &= \mathsf{Obf} \big( 1^\lambda, \mathsf{MBCC}_{f\mathsf{ss}, w, \mathsf{key}} \big)(w') \\ &= \mathsf{MBCC}_{f\mathsf{ss}, w, \mathsf{key}}(w') \\ &= \begin{cases} \mathsf{key} & \text{if } f\mathsf{ss}(w') = w \\ \bot & \text{otherwise.} \end{cases} \\ &= \begin{cases} \mathsf{key} & \text{if } \mathsf{Rec}(w', \mathsf{SS}) = w \\ \bot & \text{otherwise.} \end{cases} \end{split}$$

Then since (Sketch, Rec) is a secure sketch, it is true that

$$\begin{split} &\Pr\left[\mathsf{PFE}.\mathsf{Rep}(\mathsf{pub},w') = \mathsf{key} \mid (\mathsf{pub},\mathsf{key}) \leftarrow \mathsf{PFE}.\mathsf{Gen}(w) \text{ and } \mathsf{dist}(w,w') \leq t\right] \\ &= \Pr\left[\mathsf{Rec}(w',\mathsf{SS}) = w \mid \mathsf{SS} \leftarrow \mathsf{Sketch}(w) \text{ and } \mathsf{dist}(w,w') \leq t\right] \\ &\geq 1 - \delta \end{split}$$

and PFE is thus correct.

Security: This proof is the same as the security proof of Theorem 1.

Weak privacy: Let (Sketch, Rec) be an  $(\mathcal{M}, \mathcal{W}, \ell, t, \delta)$ -secure sketch. Then for random variables W, aux, and  $SS \leftarrow Sketch(W)$  we have

$$H_{\mathsf{HILL}}(W \mid \mathsf{SS}) \geq \ell$$

Since key is sampled independently from all other values,

$$\begin{split} H_{\mathsf{HILL}}\big(w \mid \mathsf{Rec}(\cdot, \mathsf{SS}), \mathsf{key}, \mathsf{aux})\big) &= H_{\mathsf{HILL}}\big(w \mid \mathsf{Rec}(\cdot, \mathsf{SS})\big) \\ &= H_{\mathsf{HILL}}\big(w \mid \mathsf{SS}\big) \geq \ell \end{split}$$

Let  $\mathsf{Obf}$  be a distributional VBB secure obfuscator for  $\ell\text{-pseudo-entropy}$  distributions. Then

$$\mathsf{pub}' = \mathsf{Obf}\big(1^{\lambda}, \mathsf{MBCC}_{\mathsf{Rec}(\cdot, \mathsf{SS}), w, \mathsf{key}}\big)$$

can be simulated and for every A, there exists a simulator Sim such that for every predicate  $\phi$  we have:

$$\left|\Pr[\mathcal{A}(\mathsf{pub},\mathsf{aux}) = \phi(\mathsf{pub})] - \Pr[\mathsf{Sim}^{\mathsf{MBCC}_{\mathsf{Rec}(\cdot,\mathsf{SS}),w,\mathsf{key}}}(1^{\lambda},\mathsf{param},\mathsf{aux}) = \phi(\mathsf{pub})]\right| \leq \mathsf{ngl}(\lambda)$$

Note that if we set aux = key and since key is drawn randomly and independently, it is true that

$$\begin{split} & \left| \Pr[\mathcal{A}(\mathsf{pub}, \mathsf{key}) = \phi(\mathsf{pub})] - \Pr[\mathsf{Sim}(1^{\lambda}, \mathsf{param}, \mathsf{key}) = \phi(\mathsf{pub})] \right| \\ & = \left| \Pr[\mathcal{A}(\mathsf{pub}, \mathsf{key}') = \phi(W)] - \Pr[\mathsf{Sim}(1^{\lambda}, \mathsf{param}) = \phi(W)] \right| \leq \mathsf{ngl}(\lambda) \end{split}$$

which concludes the proof that PFE is a weakly private fuzzy extractor.

### 4 Robustness

We first define robustness of a fuzzy extractor.

**Definition 10 (Robust Fuzzy extractor).** Let FE be an  $(\mathcal{M}, \mathcal{W}, \ell, t, s, \epsilon)$ -fuzzy extractor with error  $\delta$  as defined above. FE is a robust fuzzy extractor if for all  $W, W' \in \mathcal{W}$ , such that

$$\Pr_{(w,w')\leftarrow(W,W')}[\mathsf{dist}(w,w')\leq t]=1,$$

and for all adversaries A, the advantage of A in the following experiment is at most  $ngl(\lambda)$ :

- 1. Sample  $(w, w') \leftarrow (W, W')$ .
- 2. Compute (key, pub)  $\leftarrow \mathsf{FE}.\mathsf{Gen}(w)$  and send it to  $\mathcal{A}$ .
- 3. A outputs pub' and wins if pub'  $\neq$  pub and FE.Rep(pub', w')  $\notin \{\bot, \text{key}\}$ .

We propose a generic technique to upgrade a fuzzy extractor and achieve robustness. This method relies on non-interactive zero-knowledge (NIZK) [DHLAW10]. We also show that this technique preserves privacy of the underlying fuzzy extractor. This yields a robust, weakly-private fuzzy extractor construction in the common reference string (CRS) model.

**Definition 11 (True simulation extractable NIZK).** Let R be an NP relation on pairs (x, w) with corresponding language  $L_R = \{x : \exists w \text{ such that } (x, w) \in R\}$ . A true-simulation extractable non-interactive zero-knowledge (NIZK) argument for a relation R consists of three algorithms (Setup, Prove, Verify) with the following syntax:

- (crs, TK, EK)  $\leftarrow$  Setup(1 $^{\lambda}$ ): creates a common reference string crs, a trapdoor TK, and an extraction key EK.
- $-\pi \leftarrow \mathsf{Prove}(\mathsf{crs}, x, w) \colon creates \ an \ argument \ \pi \ that \ R(x, w) = 1.$
- $-0/1 \leftarrow \mathsf{Verify}(\mathsf{crs}, x, \pi)$ : verifies whether or not the argument  $\pi$  is correct.

For presentation simplicity, we omit crs in the Prove and Verify. We require that the following three properties hold:

- *Completeness.* For any  $(x, w) \in R$ , if  $(crs, TK, EK) \leftarrow Setup(1^{\lambda})$ ,  $\pi \leftarrow Prove(x, w)$ , then  $Verify(x, \pi) = 1$ .
- **Soundness.** For any PPT adversary  $\mathcal{A}$ , the following probability is negligible: for (crs, TK, EK)  $\leftarrow$  Setup( $1^{\lambda}$ ),  $(x^*, \pi^*) \leftarrow \mathcal{A}(\text{crs})$  such that  $x^* \notin L_R$  but  $\text{Verify}(x^*, \pi^*) = 1$ .
- Composable Zero-knowledge. There exists a PPT simulator Sim such that for any PPT A, the advantage (the probability A wins minus one half) is negligible in the following game.
  - The challenger samples (crs, TK, EK)  $\leftarrow$  Setup(1 $^{\lambda}$ ) and sends (crs, TK) to  $\mathcal{A}$
  - $\mathcal{A}$  chooses  $(x, w) \in R$  and sends to the challenger.

- The challenger generates  $\pi_0 \leftarrow \text{Prove}(x, w)$ ,  $\pi_1 \leftarrow \text{Sim}(x, \mathsf{TK})$ , and then samples a random bit  $b \leftarrow \{0, 1\}$ . Then he sends  $\pi_b$  to  $\mathcal{A}$ .
- A outputs a guess bit b', and wins if b' = b.
- Extractibility. Additionally, true simulation extractability requires that there
  exists a PPT extractor Ext such that for any PPT adversary A, the probability A wins is negligible in the following game:
  - The challenger picks (crs, TK, EK)  $\leftarrow$  Setup(1 $^{\lambda}$ ) and sends crs to  $\mathcal{A}$ .
  - A is allowed to make oracle queries to the simulation algorithm  $\mathsf{Sim}'((x,w),\mathsf{TK})$  adaptively.  $\mathsf{Sim}'$  first checks if  $(x,w) \in R$  and returns  $\mathsf{Sim}(x,TK)$  if that is the case.
  - A outputs a tuple  $x^*, L^*, \pi^*$ .
  - The challenger runs the extractor  $w^* \leftarrow \text{Ext}(L^*, (x^*, \pi^*), \text{EK})$ .
  - A wins if 1) the pair  $(x^*, L^*)$  was not part of the simulator query, 2) the proof  $\pi^*$  verifies, and (3)  $R(x^*, w^*) = 0$ .

Construction 3 (Robust, weakly-private fuzzy extractor) Let FE be a weakly-private fuzzy extractor and (Setup, Prove, Verify) be a NIZK system for language  $\mathcal{L} = \{ \mathsf{pub} \mid \mathsf{FE}.\mathsf{Gen}(w;r) = (\mathsf{pub},\mathsf{key}) \}$ . Here, the statement is  $\mathsf{pub} = \mathsf{FE}.\mathsf{Gen}(w;r)$  and the witness is the pair of values (w,r), where w is the original reading and r the internal randomness of  $\mathsf{Gen}$ .

```
 \begin{array}{l} -\ (\mathsf{key},\mathsf{pub}^*) \leftarrow \mathsf{FE}'.\mathsf{Gen}(w)\colon \\ 1.\ Sample\ (\mathsf{crs},\mathsf{TK},\mathsf{EK}) \leftarrow \mathsf{Setup}(1^\lambda). \\ 2.\ Compute\ (\mathsf{key},\mathsf{pub}) \leftarrow \mathsf{FE}.\mathsf{Gen}(w;r). \\ 3.\ Compute\ \pi \leftarrow \mathsf{Prove}(\mathsf{crs},\mathsf{pub},w,r)\ and\ set\ \mathsf{pub}^* = (\mathsf{pub},\pi). \\ 4.\ Output\ (\mathsf{key},\mathsf{pub}^*). \\ -\ \mathsf{key}' \leftarrow \mathsf{FE}'.\mathsf{Rep}(\mathsf{pub}^*,w')\colon \\ 1.\ Run\ b \leftarrow \mathsf{Verify}(\mathsf{crs},\mathsf{pub},\pi)\ and\ output\ \bot\ if\ b = 0. \\ 2.\ Output\ \mathsf{key}' \leftarrow \mathsf{FE}.\mathsf{Rep}(\mathsf{pub},w'). \end{array}
```

**Theorem 3.** Let FE be an weakly-private,  $(\mathcal{M}, \mathcal{W}, \ell, 2t, s, \epsilon)$ -fuzzy extractor and (Setup, Prove, Verify) be a NIZK system. Then FE' as described in Construction 3 is a weakly-private, robust,  $(\mathcal{M}, \mathcal{W}, \ell, t, s, \epsilon)$ -fuzzy extractor.

Note that in this theorem the underlying fuzzy extractor  $\mathsf{FE}$  corrects  $\mathsf{2t}$  errors while the resulting fuzzy extractor  $\mathsf{FE}'$  corrects only  $\mathsf{t}$  errors. This is important for the corresponding proof to work. This requirement was present in some prior robustness upgrades for fuzzy extractors, see Table 2.

Proof (Theorem 3).

Correctness: Correctness is straightforward from the correctness of the underlying fuzzy extractor and the completeness of the NIZK system.

Security: Security is straightforward from the security of the underlying fuzzy extractor and the zero-knowledge property of the NIZK system.

*Privacy:* Privacy is straightforward from the privacy of the underlying fuzzy extractor and the zero-knowledge property of the NIZK system. We provide a short sketch below.

Let Sim denote a simulator for the underlying weakly private FE. Suppose FE' is not weakly private, then there exists an adversary  $\mathcal{A}'$ , such that for any simulators  $\mathsf{Sim}'(|\mathsf{pub}|, |\pi|, |\mathsf{key}|)$ , we have

$$|\Pr[\mathcal{A}'(\mathsf{pub}, \pi, \mathsf{key}) = 1] - \Pr[\mathsf{Sim}'(|\mathsf{pub}|, |\pi|, |\mathsf{key}|) = 1]| > \mathsf{ngl}(\lambda)$$

We note that  $\mathsf{Sim}'(|\mathsf{pub}|,|\pi|,|\mathsf{key}|) = \mathsf{Sim}(|\mathsf{pub}|,|\mathsf{key}|)$  is one such valid simulator.

Then we can build an adversary A for FE,

- 1. Receive inputs pub and key.
- 2. Run NIZK setup (TK, EK)  $\leftarrow$  Setup( $1^{\lambda}$ ).
- 3. Run the NIZK simulator  $\pi \leftarrow \mathsf{Sim}_{\mathsf{NIZK}}(\mathsf{pub},\mathsf{TK})$ .
- 4. Run the FE' adversary  $b \leftarrow \mathcal{A}'(\mathsf{pub}, \pi, \mathsf{key})$ .
- 5. Return b.

Then

$$|\Pr[\mathcal{A}(\mathsf{pub},\mathsf{key})=1] - \Pr[\mathsf{Sim}(|\mathsf{pub}|,|\mathsf{key}|)=1]| > \mathsf{ngl}(\lambda)$$

which is a contradiction of FE's weak privacy.

Robustness: We proceed by contradiction. Suppose  $\mathsf{FE}'$  is not a robust fuzzy extractor, that is, for distributions W, W' such that  $\mathsf{dist}(W, W') \leq t$ , there exists a PPT adversary  $\mathcal{A}'_{\mathsf{FE}}$  such that

$$\Pr_{\substack{(w,w')\leftarrow(W,W')}}\left[\begin{array}{c|c} \mathsf{Verify}(\mathsf{crs},\mathsf{pub'},\pi')=1 & (\mathsf{key},\mathsf{pub})\leftarrow\mathsf{FE}.\mathsf{Gen}(w) \\ \land \mathsf{FE}.\mathsf{Rep}(\mathsf{pub'},w')\neq\{\mathsf{key},\bot\} & \pi\leftarrow\mathsf{Prove}(\mathsf{crs},\mathsf{pub},\mathsf{key},w) \\ \land (\mathsf{pub'}\neq\mathsf{pub}\vee\pi'\neq\pi) & (\mathsf{pub'},\pi')\leftarrow\mathcal{A}_{\mathsf{PFE}}(\mathsf{key},\mathsf{pub},\pi) \end{array}\right] > \mathsf{ngl}(\lambda).$$

We can then build a PPT distinguisher  $\mathcal{A}$  for the fuzzy extractor security game as follows:

- 1. Receive (pub, key<sub>b</sub>) from the challenger, where for  $w \leftarrow W$ , (pub, key)  $\leftarrow$  FE.Gen(w;r) and for  $b \in \{0,1\}$ , key<sub>1</sub> = key and key<sub>0</sub> =  $U_{\ell}$ .
- 2. Sample (crs, TK, EK)  $\leftarrow$  Setup(1 $^{\lambda}$ ) from the NIZK proof system.
- 3. Run the NIZK simulator  $\pi \leftarrow \mathsf{Sim}(\mathsf{pub}, \mathsf{TK})$ .
- 4. Send  $(\mathsf{key}_b, \mathsf{pub}, \pi)$  to  $\mathcal{A}'_\mathsf{FE}$  and receives back  $(\mathsf{pub}', \pi')$ .
- 5. Run the NIZK extractor  $(w^*, r^*) \leftarrow \mathsf{Ext}(\mathsf{pub}', \pi', \mathsf{EK})$ .
- 6. Run key'  $\leftarrow$  FE.Rep(pub,  $w^*$ ).
- 7. If  $key' = key_b$  return 1, otherwise return 0.

Recall that a break in robustness requires  $(\mathsf{pub}', \pi') \neq (\mathsf{pub}, \pi)$  and  $\pi'$  to be a valid proof. Suppose  $\mathsf{pub}' = \mathsf{pub}$ , then  $\mathsf{dist}(w, w^*) \leq t$  and  $\mathsf{FE}.\mathsf{Rep}(\mathsf{pub}, w^*) = \mathsf{key}$ . So  $\mathsf{FE}'.\mathsf{Rep}(\mathsf{pub}'||\pi', w^*) = \mathsf{key}$ , which does not count as a break of the robustness property.

So it must be true that  $\mathsf{pub}' \neq \mathsf{pub}$ . In this situation,  $\mathcal{A}'_\mathsf{FE}$  outputs the pair  $(\mathsf{pub}', \pi')$  such that for some w', with  $\mathsf{dist}(w, w') \leq t$ ,  $\mathsf{FE}'.\mathsf{Rep}(\mathsf{pub}' || \pi', w') = \mathsf{key}^* \neq \mathsf{key}$ . Then the NIZK extractor outputs point  $w^*$  such that  $\mathsf{FE}.\mathsf{Gen}(w^*; r^*) = (\mathsf{key}^*, \mathsf{pub}' || \pi')$ . So  $\mathsf{dist}(w, w') \leq t$  and  $\mathsf{dist}(w^*, w') \leq t$ , which means that  $\mathsf{dist}(w, w^*) \leq 2t$ . Finally, since  $\mathsf{FE}$  corrects 2t errors, when b = 1,  $\mathsf{key}' = \mathsf{FE}.\mathsf{Rep}(\mathsf{pub}, w^*) = \mathsf{key} = \mathsf{key}_1$  and

$$|\Pr[\mathcal{A}(\mathsf{pub}, \mathsf{key}) = 1] - \Pr[\mathcal{A}(\mathsf{pub}, U_\ell) = 1]| > \mathsf{ngl}(\lambda)$$

which concludes our proof.

### 5 Reuse

In this section, we show that one cannot hope to compose MBCC obfuscation with an auxiliary input secure digital locker. We then show this implies a impossibility of a variant of private fuzzy extractors that can be constructed from MBCC obfuscation. This variant never outputs a value outside of the ball of the enrolled value.

**Definition 12.** Let (Gen, Rep) be an  $(\mathcal{M}, \mathcal{W}, \ell, t, \epsilon)$ -fuzzy extractor with error  $\delta$  (Definition 7). The pair is perfectly correct if for all  $w, w' \in \mathcal{M}$  such that  $\operatorname{dist}(w, w') > t$ :

$$\Pr\left[\bot\leftarrow \mathsf{Rep}(\mathsf{pub}, w') \mid (\mathsf{key}, \mathsf{pub}) \leftarrow \mathsf{Gen}(w)\right] \geq 1 - \mathsf{ngl}(\lambda).$$

We assume that any randomness for Rep is included in the string pub so this probability statement is only over the randomness of Gen.

We now define digital lockers, which have the same functionality as perfectly correct fuzzy extractors for t=0. Digital lockers [CTKVW10] are also a specific case of MBCC obfuscation where the function is the identity function, f(x) = id(x) = x.

**Definition 13 (Digital Locker).** An (W, n)-digital locker is a pair of PPT algorithms (lock, unlock) where for all val  $\in D^{\lambda}$  and key  $\in \{0, 1\}^n$ ,

- unlock ← lock(val, key)
- key' ← unlock(val')

such that the following properties are true:

1. Completeness: For all val  $\in D^{\lambda}$ , key  $\in \{0,1\}^n$  it holds that

$$\Pr[\mathsf{unlock}(\cdot) \equiv I_{\mathsf{val},\mathsf{kev}}(\cdot) \mid \mathsf{unlock} \leftarrow \mathsf{lock}(\mathsf{val},\mathsf{key})] \geq 1 - \mathsf{ngl}(\lambda),$$

where the probability is over the randomness of lock. Here  $I_{\text{val,key}}$  is a function that returns key when provided input val, otherwise  $I_{\text{val,key}}$  returns  $\perp$ .

2. Virtual Black Box Security: For all PPT A and  $p = poly(\lambda)$ ,  $\exists Sim$  and  $q(\lambda) = poly(\lambda)$  such that for all large enough  $\lambda \in \mathbb{N}$ ,  $\forall val \in D_{\lambda}$ , key  $\in \{0,1\}^n, \mathcal{P} : D_{\lambda} \times \{0,1\}^n \mapsto \{0,1\}$ ,

$$\left|\Pr[\mathcal{A}(\mathsf{lock}(\mathsf{val},\mathsf{key})) = \mathcal{P}(\mathsf{val},\mathsf{key})] - \Pr[\mathsf{Sim}^{I_{\mathsf{val},\mathsf{key}}}(1^{\lambda}) = \mathcal{P}(\mathsf{val},\mathsf{key})]\right| \leq \frac{1}{p(\lambda)},$$

where Sim is allowed  $q(\lambda)$  oracle queries to  $I_{\mathsf{val},\mathsf{key}}$  and the probabilities are over the internal randomness of  $\mathcal{A}$  and lock, and of Sim, respectively.

Construction One can construct a perfectly correct private fuzzy extractor by applying Construction 2 on a well-formed secure sketch [BDK+05, Definition 4]. A well-formed secure sketch on input w' never outputs a value with distance  $\geq t$  from w'. One can always construct a well-formed secure sketch with no loss in parameters by adding a distance check before output.

Since the circuit being obfuscated in Construction 2 only has an output when the output of the secure sketch is equal to w, these two modifications suffice to form a (private) perfectly correct fuzzy extractor.

**Proposition 1.** Perfectly correct private fuzzy extractors with auxiliary input imply digital lockers with auxiliary inputs.

*Proof.* This proposition easily follows by setting the required distance t equal to 0.

**Definition 14 (Collision-resistant Hash function).** Consider function  $h: \{0,1\}^n \to \{0,1\}^m$ , h is a collision-resistant hash function if the following are true:

- 1. Compression: m < n.
- 2. Collision-resistance: For any PPT adversary A,

$$\Pr[(x_0, x_1) \leftarrow \mathcal{A}(1^n, h) \mid x_0 \neq x_1 \land h(x_0) = h(x_1)] < \mathsf{ngl}(n).$$

Theorem 4 (Private FE with auxiliary input impossibility). If dist-VBB obfuscation for MBCC programs with  $\alpha$ -pseudo entropy and collision-resistant hash functions exist, no perfectly-correct private fuzzy extractor can be secure in the presence of unpredictability auxiliary inputs.

*Proof (Theorem 4).* This proof is built from a main lemma (see Lemma 2) which is then combined with Proposition 1. Lemma 2 shows that digital lockers with auxiliary input for unpredictable sources cannot exist if dist-VBB obfuscation for MBCC programs with  $\alpha$ -pseudo entropy exists.

Lemma 2 (Digital locker with auxiliary input impossibility). If dist-VBB obfuscation for MBCC programs with  $\alpha$ -pseudo entropy and collision-resistant hash functions exist, then security for digital lockers with auxiliary inputs for unpredictable sources cannot be achieved. *Proof (Lemma 2).* Let  $U_x$  denote the universal circuit that takes as input circuit C and computes  $U_x(C) = C(x)$ . Define the following MBCC program

$$\mathsf{MBCC}[U_x, \mathsf{key}, x](C) = \begin{cases} x & \text{if } C \text{ is a well-formed unlock program and } C(x) = \mathsf{key}. \\ \bot & \text{otherwise}. \end{cases}$$

Let  $h: \{0,1\}^{|x|} \to \{0,1\}^m$ , with m < |x|, be a collision-resistant hash function. Suppose x and key are independent and let  $\mathsf{aux} = h(x)$ , then we have

$$H^{\mathsf{HILL}}(\mathsf{key} \mid U_x, x, \mathsf{aux}) \geq \alpha(\lambda)$$

which implies that there exists a dist-VBB obfuscator Obf for this MBCC circuit. We now need to show that X remains unpredictable given  $\mathsf{Obf}(\mathsf{MBCC}[U_X, \mathsf{key}, x])$ , that is

$$H^{\mathsf{unp}}(X \mid \mathsf{Obf}(\mathsf{MBCC}[U_X, \mathsf{key}, x]) \ge \omega(\log \lambda)$$

In other words, we want to show that if  $\mathsf{Obf}(\mathsf{MBCC}[U_x, \mathsf{key}, x])$  is dist-VBB secure, then for all PPT  $\mathcal{A}$ , we have

$$\Pr\left[\mathcal{A}\left(\mathsf{Obf}(\mathsf{MBCC}[U_x,\mathsf{key},x])\right)=x\right] \leq \mathsf{ngl}(\lambda).$$

We proceed by contradiction. Suppose the above is not true and there exists a PPT  $\mathcal{A}$  that can predict x from  $\mathsf{Obf}(\mathsf{MBCC}[U_x, \mathsf{key}, x])$  with non-negligible probability. Then we can build a distinguisher for the MBCC obfuscation that breaks dist-IND security (which is equivalent to dist-VBB for evasive functions such as MBCC [BC10]). The distinguisher works as follows:

- 1. Receive  $P^*$  and  $\mathsf{aux} = h(x)$  as inputs.
- 2. Run  $x^* \leftarrow \mathcal{A}(P^*)$ .
- 3. If  $h(x^*) = h(x)$ , return 1, otherwise return 0.

If  $P^* = \mathsf{Obf}(1^\lambda, P)$ , then  $\mathcal{A}$  should be able to extract  $x^* = x$  and  $h(x^*) = h(x)$ . However, if  $P^* \leftarrow \mathsf{Sim}(1^\lambda, P.\mathsf{params})$ ,  $\mathcal{A}$  should not be able to extract correct  $x^*$ . Then the probability that  $x^* = x$  is  $\frac{1}{2^n}$  and when  $x^* \neq x$ ,  $h(x^*) = h(x)$  with negligible probability. This is a contradiction of dist-IND security of the MBCC obfuscator so we conclude that X remains unpredictable.

We now need to show that that this construction breaks digital locker security. Recall that digital locker security is VBB, that is for any PPT adversary  $\mathcal{A}$  and any polynomial p, there exists a simulator Sim such that

$$\Pr\left[\mathcal{A}(\mathsf{unlock},\mathsf{aux})=1\right] - \Pr\left[\mathsf{Sim}^{\mathsf{unlock}(\cdot)}(1^{\lambda},\mathsf{aux})=1\right] \leq \frac{1}{n(\lambda)}$$

where  $unlock \leftarrow lock(val, key)$ .

It is obvious that this does not hold when we set  $aux = MBCC[U_{val}, key, val]$ . Indeed, A can then run aux(unlock) and retrieve the correct val (and then key by running unlock(val)), whereas Sim cannot.

By chaining Lemma 2 and the contrapositive of Proposition 1, we obtain that if dist-VBB MBCC obfuscation exists then private fuzzy extractors with auxiliary inputs cannot be achieved, which conclude this Theorem's proof.

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### A Privacy vs FE security

Showing that fuzzy extractor security does not imply privacy is straightforward. Let  $\mathsf{FE}'$  be a fuzzy extractor for which  $\mathsf{pub}' = w_1 || \mathsf{pub}$ , where  $w_1 \in \{0,1\}$  denotes the first bit of w and  $\mathsf{pub}$  is a valid public value such that  $\mathsf{key} \leftarrow \mathsf{FE}.\mathsf{Rep}(\mathsf{pub}, w^*)$  when  $\mathsf{dist}(w, w^*) \leq t$ . Then it is obvious that even though  $\mathsf{FE}'$  is a secure fuzzy extractor, it is not private.

We will now show that the reverse is also not true.

**Theorem 5.** Privacy (Definition 8) does not imply fuzzy extractor security (Definition 7).

*Proof (Proof of Theorem 5).* We will prove this by presenting a counter example. Consider the following construction:

- (pub, key)  $\leftarrow$  Gen(w): key is sampled uniformly at random and pub is an obfuscation of the program p such that, for inputs  $x \in \{0,1\}^*$  and  $b \in \{0,1\}$ ,

$$p(b,x) = \begin{cases} \mathsf{key} & \text{if } b = 1 \text{ and } \mathsf{dist}(w,x) \leq t \\ \top & \text{if } b = 0 \text{ and } x = \mathsf{key} \\ \bot & \text{otherwise.} \end{cases}$$

- key  $\leftarrow \text{Rep}(\text{pub}, b, w')$ : run pub(1, w') and return its output.

Notice that for  $w, w' \in \mathcal{W}$  such that  $dist(w, w') \leq t$ , we have

$$\Pr[\mathsf{key} \leftarrow \mathsf{Rep}(\mathsf{pub}, w') \mid (\mathsf{pub}, \mathsf{key}) \leftarrow \mathsf{Gen}(w)] \ge 1 - \mathsf{ngl}(\lambda)$$

which is the expected behavior of a fuzzy extractor. Furthermore, note that this construction is private since by the obfuscation definition, for any PPT adversary  $\mathcal{A}$ , there exists simulator Sim such that for any predicate  $\phi$ 

$$\left|\Pr[\mathcal{A}(\mathsf{pub},\mathsf{key}) = \phi(W)] - \Pr[\mathsf{Sim}(1^{\lambda},1^{|\mathsf{pub}|},1^{|\mathsf{key}|}) = \phi(W)]\right| \leq \mathsf{ngl}(\lambda)$$

Now let's check fuzzy extractor security. Consider the following experiment:

- 1. Run (key, pub)  $\leftarrow \text{Gen}(w)$ .
- 2. Draw  $b \leftarrow \{0, 1\}$ .
- 3. If b = 0, sample  $U_{\ell} \stackrel{\$}{\leftarrow} \{0,1\}^{\ell}$  and send  $(U_{\ell}, \mathsf{pub})$  to  $\mathcal{A}$ . Otherwise, send (key, pub) to  $\mathcal{A}$ .
- 4.  $\mathcal{A}$  outputs  $b' \in \{0,1\}$  and wins if b' = b.

 $\mathcal{A}$  has a straightforward way of winning this experiment by running  $\mathsf{pub}(0,x)$ , where  $x = \mathsf{key}$  or  $x = U_\ell$  depending on drawn b. Then  $\mathcal{A}$  outputs b' = 1 if  $\mathsf{pub}(0,x) = \top$  and b' = 0 if  $\mathsf{pub}(0,x) = \bot$ . Thus we have

$$|\Pr[\mathcal{A}(\mathsf{key},\mathsf{pub})=1]-\Pr[\mathcal{A}(U_{\ell},\mathsf{pub})=1]|>\mathsf{ngl}$$

and we can conclude that this construction, although private, is not a secure fuzzy extractor.

# B Reusability from Composable MBCC Obfuscation

Reusability for Constructions 1 and 2 is achievable when the MBCC obfuscator is composable. We start by defining reuse.

**Definition 15 (Reusable Fuzzy extractor [CFP**<sup>+</sup>**16]).** Let FE be an  $(\mathcal{M}, \mathcal{W}, \ell, t, s, \epsilon)$ -fuzzy extractor with error  $\delta$  as defined above. Let  $(W_1, \dots, W_{\rho})$  be  $\rho \in \mathbb{N}$  correlated variables such that  $W_i \in \mathcal{W}$ . Let adversary  $\mathcal{A}$  be a PPT adversary, then for all  $j \in [1, \rho]$ :

- 1. The challenger samples  $w_i \leftarrow W_i$  and computes  $(\text{key}_i, \text{pub}_i) \leftarrow \text{FE.Gen}(w)$ .
- 2. The challenger samples a uniform  $u \stackrel{\$}{\leftarrow} \{0,1\}^{\ell}$  and sets  $K_0 = \ker_i$  and  $K_1 = u$ .
- 3. The challenger draws  $b \stackrel{\$}{\leftarrow} \{0,1\}$  and sends to A

$$(\mathsf{key}_1, \cdots, \mathsf{key}_{i-1}, K_b, \mathsf{key}_{i+1}, \cdots, \mathsf{key}_o, \mathsf{pub}_1, \cdots, \mathsf{pub}_o)$$

4. A outputs  $b' \in \{0,1\}$  and wins if b' = b.

We denote the above experiment as  $\mathsf{Exp}^{\mathsf{reusable}}_{\mathcal{A},b}$ , the advantage of  $\mathcal{A}$  is

$$\mathsf{Adv}(\mathcal{A}) = \left| \Pr[ \mathsf{Exp}_{\mathcal{A},0}^{\mathit{reusable}} = 1 ] - \Pr[ \mathsf{Exp}_{\mathcal{A},1}^{\mathit{reusable}} = 1 ] \right|.$$

FE is a  $(\rho, \epsilon)$ -reusable fuzzy extractor if for all A, for all  $i \in [1, \rho]$  the advantage of A is at most  $\epsilon$ .

However, as we show in Section 5 this is not possible without restricting the class of circuits being obfuscated.

Definition 16 ( $\ell$ -Composable Obfuscation with auxiliary input). Obf is a  $\ell$ -composable obfuscator for distribution class  $\mathcal{D}$  over the family of circuits  $\mathcal{P}_{\lambda}$  if for any PPT adversary  $\mathcal{A}$  and polynomial p, there exists a simulator Sim such that for every distribution ensemble  $D = \{D_{\lambda}\} \in \mathcal{D}$  and  $(P_1, \dots, P_{\ell}, \mathsf{aux}) \leftarrow D_{\lambda}$ , with  $\ell = \mathsf{poly}(\lambda)$ ,

$$\begin{split} & \Big| \Pr[\mathcal{A} \big( \mathsf{Obf}(P_1), \cdots, \mathsf{Obf}(P_\ell), \mathsf{aux} \big) = 1] \\ & - \Pr[\mathsf{Sim}^{P_1, \cdots, P_\ell} (1^{|P_1|}, \cdots, 1^{|P_\ell|}, \mathsf{aux}) = 1] \Big| \leq \frac{1}{p(\lambda)} \end{split}$$

**Theorem 6.** Let Obf be a composable dist-VBB obfuscator for MBCC circuits, then Constructions 1 and 2 are reusable.

*Proof (Proof of Theorem 6).* Suppose PFE is not a reusable fuzzy extractor, that is, there exists a PPT adversary  $\mathcal{A}$  and a polynomial  $p(\lambda)$  such that for all  $1 \leq j \leq \rho$ :

$$\left| \begin{array}{l} \Pr[\mathcal{A}(\mathsf{key}_1,\cdots,\mathsf{key}_\rho,\mathsf{pub}_1,\cdots,\mathsf{pub}_\rho) = 1] \\ -\Pr[\mathcal{A}(\mathsf{key}_1,\cdots,\mathsf{key}_{i-1},U_\ell,\mathsf{key}_{i+1},\cdots,\mathsf{key}_\rho,\mathsf{pub}_1,\cdots,\mathsf{pub}_\rho) = 1] \end{array} \right| > \frac{1}{n(\lambda)}$$

where  $U_{\ell}$  is a uniform random string in  $\{0,1\}^{\ell}$ .

Remember that Obf is a composable obfuscator for MBCC[Rep<sub>pub</sub>, k, key]. Let  $r(\lambda) = 3p(\lambda)$  and suppose Sim is the simulator for  $\mathcal{A}$  for  $r(\lambda)$ , then we have

$$\begin{split} & \left| \Pr[\mathcal{A}(\{\mathsf{Obf}(1^\lambda, \mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}}, k, \mathsf{key}_i])\}_{i=1}^\rho, \mathsf{aux}) = 1] \\ & - \Pr[\mathsf{Sim}^{\{\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}}, k, \mathsf{key}_i]\}_{i=1}^\rho} (1^\lambda, \{|\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}}, k, \mathsf{key}_i]|\}_{i=1}^\rho, \mathsf{aux}) = 1] \right| \leq \frac{1}{3n(\lambda)} \end{split}$$

Note that in Construction 1,  $\mathsf{pub}_i = \mathsf{Obf}(1^\lambda, \mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}'}, k, \mathsf{key}_i])$  and set  $\mathsf{aux} = \mathsf{key}_1, \cdots, \mathsf{key}_\rho$  so we have

$$\begin{split} & \left| \Pr[\mathcal{A}(\{\mathsf{pub}_i\}_{i=1}^{\rho}, \{\mathsf{key}_i\}_{i=1}^{\rho}) = 1] \right. \\ & \left. -\Pr[\mathsf{Sim}^{\{\mathsf{pub}_i\}_{i=1}^{\rho}}(1^{\lambda}, \{|\mathsf{pub}_i|\}_{i=1}^{\rho}, \{\mathsf{key}_i\}_{i=1}^{\rho}) = 1] \right| \leq \frac{1}{3p(\lambda)} \end{split} \tag{3}$$

Notice that this also holds if we replace  $\ker_j$  by an independent uniform random variable  $U_\ell$  over  $\{0,1\}^\ell$ . Then for any  $j \in \{1,\rho\}$  we have:

$$\begin{split} & \left| \Pr[\mathcal{A}(\{\mathsf{pub}_i\}_{i=1}^\rho, \mathsf{key}_1, \cdots, \mathsf{key}_{j-1}, U_\ell, \mathsf{key}_{j+1}, \cdots, \mathsf{key}_\rho) = 1] \\ & - \Pr[\mathsf{Sim}^{\{\mathsf{pub}_i\}_{i=1}^\rho}(1^\lambda, \{|\mathsf{pub}_i|\}_{i=1}^\rho, \mathsf{key}_1, \cdots, \mathsf{key}_{j-1}, U_\ell, \mathsf{key}_{j+1}, \cdots, \mathsf{key}_\rho) = 1] \right| \leq \frac{1}{3p(\lambda)} \end{split}$$

Again we adapt Canetti et al.'s lemma [CFP+16, Lemma 2]:

**Lemma 3.** Let  $U_{\ell}$  denote the uniform distribution over  $\{0,1\}^{\ell}$ , then for  $1 \leq j \leq \rho$ ,

$$\begin{split} & \left| \Pr[\mathsf{Sim}^{\{\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},k,\mathsf{key}_i]\}_{i=1}^{\rho}} \left( 1^{\lambda}, \left\{ \left| \mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},k,\mathsf{key}_i] \right| \right\}_{i=1}^{\rho}, \left\{ \mathsf{key} \right\}_{i=1}^{\rho} \right) = 1 \right] \\ & - \Pr[\mathsf{Sim}^{\{\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},k,\mathsf{key}_i]\}_{i=1}^{\rho}} \left( 1^{\lambda}, \left\{ \left| \mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}},k,\mathsf{key}_i] \right| \right\}_{i=1}^{\rho}, \left\{ \mathsf{key}_i \right\}_{i=1}^{j-1}, U_{\ell}, \left\{ \mathsf{key}_i \right\}_{i=j+1}^{\rho} \right) = 1 \right] \\ & \leq \frac{1}{3p(\lambda)} \end{split}$$

*Proof.* Fix any  $u \in \{0,1\}^\ell$ , the lemma will follow by averaging over all u. The information about whether the  $j^{th}$  key value, denoted  $V_j$ , is  $\ker_j$  or u can only be obtain by Sim through the query responses. First, we modify Sim to quit immediately when it gets a response not equal to  $\bot$ . Such Sim is equally successful at distinguishing between  $\ker_j$  and u since the first non- $\bot$  response tells Sim if its input is equal to  $\ker_j$ . Subsequent responses add nothing to this knowledge. Since Sim can make at most q queries, there are q+1 possible values for the view of Sim on a given input. Of those, q views consist of some number of non- $\bot$  responses followed by a  $\bot$  response, and one view consists of all q responses equal to  $\bot$ .

Then by [DRS04, Lemma 2.2b],

$$\begin{split} \tilde{H}_{\infty}(V_j|View(\mathsf{Sim}),\mathsf{aux}) &\geq \tilde{H}_{\infty}(V_j) - \log(q+1) \\ &\geq \alpha - \log(q+1). \end{split}$$

 $\text{where aux} = \big(\{|\mathsf{MBCC}[\mathsf{Rep}_{\mathsf{pub}}, k, \mathsf{key}_i]|\}_{i=1}^\rho, \mathsf{key}_1, \cdots, \mathsf{key}_{j-1}, \mathsf{key}_{j+1}, \cdots, \mathsf{key}_\rho\big).$ 

Thus, at each query, the probability that Sim gets a non- $\perp$  response and guesses  $V_j$  is at most  $(q+1)/2^{\alpha}$ . Since there are q queries of Sim, the overall probability is at most  $q(q+1)/2^{\alpha}$ . Then since  $2^{\alpha}$  is negligible in  $\lambda$ , there exists some  $\lambda_0$  such that for all  $\lambda \geq \lambda_0$ ,  $q(q+1)/2^{\alpha} \leq 1/(3p(\lambda))$ .

Then from Lemma 3, we have

$$\begin{split} & \left| \Pr[\mathsf{Sim}^{\{\mathsf{pub}_i\}_{i=1}^{\rho}}(1^{\lambda}, \{|\mathsf{pub}_i|\}_{i=1}^{\rho}, \{\mathsf{key}_i\}_{i=1}^{\rho}) = 1] \\ & - \Pr[\mathsf{Sim}^{\{\mathsf{pub}_i\}_{i=1}^{\rho}}(1^{\lambda}, \{|\mathsf{pub}_i|\}_{i=1}^{\rho}, \mathsf{key}_1, \cdots, \mathsf{key}_{j-1}, U_{\ell}, \mathsf{key}_{j+1}, \cdots, \mathsf{key}_{\rho}) = 1] \right| \leq \frac{1}{3p(\lambda)} \end{split}$$

Using the triangle inequality on Equations 3, 4 and 5 we obtain

$$\left| \begin{array}{l} \Pr[\mathcal{A}(\mathsf{key}_1,\cdots,\mathsf{key}_\rho,\mathsf{pub}_1,\cdots,\mathsf{pub}_\rho) = 1] \\ -\Pr[\mathcal{A}(\mathsf{key}_1,\cdots,\mathsf{key}_{i-1},U_\ell,\mathsf{key}_{i+1},\cdots,\mathsf{key}_\rho,\mathsf{pub}_1,\cdots,\mathsf{pub}_\rho) = 1] \end{array} \right| \leq \frac{1}{p(\lambda)}$$

which is a contradiction and completes this proof.

Composable MBCC obfuscation Wichs and Zirdelis [WZ17] build obfuscation for multi-bit compute-and-compare circuits from single bit compute-and-compare by composing the function f with a strongly injective PRG. By doing so they ensure that the target values  $(y_1, \dots, y_\ell)$  are indistinguishable from uniform, even when given f, z and aux. Their proof then relies on the security of the obfuscator for the  $i^{th}$  circuit by passing all remaining circuits as auxiliary information.

Unfortunately this technique cannot be directly applied to build *composable* MBCC obfuscation since it requires keeping track of which parts of the PRG output have already been used. This is reasonable for their MBCC obfuscation scheme, where all obfuscated compute-and-compare circuits will be generated at the same time. However this is not practical in the case of composable obfuscation, where the obfuscator will typically be run at different times and without a shared state. One could use a PRG with exponential stretch and select a random part of its output, then the probability of reuse should be low. Another issue is that in Wichs and Zirdelis's scheme, the function and the input to the PRG are always the same. For composability, especially with the goal of building reusable FE, it would need to handle distinct but possibly correlated functions and values. It then is unclear what the auxiliary information (i.e. the other obfuscated programs) may leak on the current obfuscated circuit.