# Near-Optimal Secret Sharing and Error Correcting Codes in AC $^{0}$ 

Kuan Cheng ${ }^{1(\boxtimes)}$, Yuval Ishai ${ }^{2,3}$, and Xin $\mathrm{Li}^{1}$<br>${ }^{1}$ Department of Computer Science, Johns Hopkins University, Baltimore, USA<br>kcheng17@jhu.edu,lixints@cs.jhu.edu<br>${ }^{2}$ Department of Computer Science, Technion, Haifa, Israel<br>yuvali@cs.technion.ac.il<br>3 UCLA, Los Angeles, USA


#### Abstract

We study the question of minimizing the computational complexity of (robust) secret sharing schemes and error correcting codes. In standard instances of these objects, both encoding and decoding involve linear algebra, and thus cannot be implemented in the class $A C^{0}$. The feasibility of non-trivial secret sharing schemes in $\mathrm{AC}^{0}$ was recently shown by Bogdanov et al. (Crypto 2016) and that of (locally) decoding errors in $\mathrm{AC}^{0}$ by Goldwasser et al. (STOC 2007).

In this paper, we show that by allowing some slight relaxation such as a small error probability, we can construct much better secret sharing schemes and error correcting codes in the class $\mathrm{AC}^{0}$. In some cases, our parameters are close to optimal and would be impossible to achieve without the relaxation. Our results significantly improve previous constructions in various parameters.

Our constructions combine several ingredients in pseudorandomness and combinatorics in an innovative way. Specifically, we develop a general technique to simultaneously amplify security threshold and reduce alphabet size, using a two-level concatenation of protocols together with a random permutation. We demonstrate the broader usefulness of this technique by applying it in the context of a variant of secure broadcast.


## 1 Introduction

The motivation for this paper comes from two different sources. The first is the general theme of improving performance at the price of allowing some small probability of error or failure. This is evident throughout computer science. For example, randomized algorithms tend to be much more efficient than their deterministic counterparts. In cryptography and coding theory, randomization with small failure probability can often be used to amplify security or improve efficiency. This is arguably a good tradeoff in practice.

The second source of motivation is the goal of minimizing the computational complexity of cryptographic primitives and related combinatorial objects. For example, a line of work on the parallel complexity of cryptography [2,3,16,18,29]

[^0]successfully constructed one way functions and other cryptographic primitives in the complexity class $\mathrm{NC}^{0}$ based on different kinds of assumptions, including very standard cryptographic assumptions. Works along this line have found several unexpected applications, most recently in the context of general-purpose obfuscation [24]. The study of low-complexity cryptography is also motivated by the goal of obtaining stronger negative results. For instance, low-complexity pseudo-random functions imply stronger hardness results for learning [30] and stronger natural proof barriers [27], and low-complexity decryption [8] implies a barrier for function secret sharing [10].

In this paper, we address the question of minimizing the complexity of secret sharing schemes and error correcting codes by introducing additional randomization and allowing for a small failure probability. We focus on the complexity class $A C^{0}$, which is the lowest class for which a secret can be reconstructed or a message be decoded with negligible error probability. We show that the randomization approach can be used towards obtaining much better parameters than previous constructions. In some cases, our parameters are close to optimal and would be impossible to achieve without randomization.

We now give a more detailed account of our results, starting with some relevant background.

## 1.1 (Robust) Secret Sharing in $\mathrm{AC}^{0}$

A secret sharing scheme allows a dealer to randomly split a secret between $n$ parties so that qualified subsets of parties can reconstruct the secret from their shares while unqualified subsets learn nothing about the secret. We consider here a variant of threshold secret sharing (also known as a "ramp scheme"), where any $k$ parties can learn nothing about the secret, whereas all $n$ parties together can recover the secret from their shares. We also consider a robust variant where the secret should be correctly reconstructed even if at most $d$ shares are corrupted by an adversary, possibly in an adaptive fashion. We formalize this below.

Definition 1 (secret sharing). An $(n, k)$ secret sharing scheme with message alphabet $\Sigma_{0}$, message length $m$, and share alphabet $\Sigma$ is a pair of functions (Share, Rec), where Share : $\Sigma_{0}^{m} \rightarrow \Sigma^{n}$ is probabilistic and Rec: $\Sigma^{n} \rightarrow \Sigma_{0}^{m}$ is deterministic, which satisfy the following properties.

- Privacy: For a privacy threshold $k$, the adversary can choose a sequence $W=$ $\left(w_{1}, \ldots, w_{k}\right) \in[n]^{k}$ of share indices to observe, either adaptively (where each $w_{i}$ depends on previously observed shares Share $\left.(x)_{w_{1}}, \ldots, \operatorname{Share}(x)_{w_{i-1}}\right)$ or non-adaptively (where $W$ is picked in one shot). We say that the scheme is $\epsilon$-private if for every such strategy, there is a share distribution $\mathcal{D}$ over $\Sigma^{k}$ such that for every secret message $x \in \Sigma_{0}^{m}$, Share $(x)_{W}$ is $\epsilon$-close (in statistical distance) to $\mathcal{D}$. We refer to $\epsilon$ as the privacy error and say that the scheme has perfect privacy if $\epsilon=0$.
- Reconstruction: We say that the scheme has reconstruction error $\eta$ if for every $x \in \Sigma_{0}^{m}$,

$$
\operatorname{Pr}[\operatorname{Rec}(\operatorname{Share}(x))=x] \geq 1-\eta
$$

We say the scheme has perfect reconstruction if $\eta=0$.
We are also interested in robust secret sharing, where an adversary is allowed to modify at most d shares.

- Robustness: For any secret $x \in \Sigma_{0}^{m}$, let $Y=\operatorname{Share}(x)$. Consider an arbitrary adversary who (adaptively or non-adaptively) observes d shares and can then arbitrarily change these $d$ shares, transforming $Y$ to $Y^{\prime}$. The scheme is $d$ robust if for every such adversary,

$$
\operatorname{Pr}\left[\operatorname{Rec}\left(Y^{\prime}\right)=x\right] \geq 1-\eta .
$$

If the share alphabet and the message alphabet are both $\Sigma$, then we simply say the alphabet of the scheme is $\Sigma$. By saying that a secret sharing scheme is in $\mathrm{AC}^{0}$, we mean that both the sharing function and the reconstruction function can be computed by (uniform) $\mathrm{AC}^{0}$ circuits.

A recent work of Bogdanov et al. [7] considers the complexity of sharing and reconstructing secrets. The question is motivated by the observation that almost all known secret sharing schemes, including the well known Shamir's scheme [31], require the computation of linear functions over finite fields, and thus cannot be implemented in the class $A C^{0}$ (i.e., constant depth circuits). Thus a natural question is whether there exist secret sharing schemes in $A C^{0}$ with good parameters. In the case of threshold secret sharing, Bogdanov et al. [7] showed a relation between the approximate degree ${ }^{1}$ of a function and the privacy threshold of a secret sharing scheme. Using this and known approximate degree lower bounds, they obtained several secret sharing schemes with sharing and reconstruction functions computable in $\mathrm{AC}^{0}$. However, to achieve a large privacy threshold (e.g., $k=\Omega(n)$ ) their construction needs to use a large alphabet (e.g., size $2^{\text {poly }(n)}$ ). In the case of binary alphabet, they can only achieve privacy threshold $\Omega(\sqrt{n})$ with perfect reconstruction and privacy threshold $\Omega\left((n / \log n)^{2 / 3}\right)$ with constant reconstruction error $\eta<1 / 2$. This limit is inherent without improving the best known approximate degree of an $\mathrm{AC}^{0}$ function [11]. Furthermore, their schemes only share one bit, and a naive approach of sharing more bits by repeating the scheme multiple times will lead to a bad information rate. This leaves open the question of improving these parameters. Ideally, we would like to share many bits (e.g., $\Omega(n)$ ), obtain a large privacy threshold (e.g., $\Omega(n)$ ), and achieve perfect reconstruction and small alphabet size at the same time.

In order to improve the $A C^{0}$ secret sharing schemes from [7], we relax their perfect privacy requirement and settle for the notion of $\epsilon$-privacy from Definition 1. (This relaxation was recently considered in [9], see discussion below.) Note that this relaxation is necessary to improve the privacy threshold of $A C^{0}$ secret sharing schemes, unless one can obtain better approximate degree lower bounds of an explicit $A C^{0}$ function (as [7] showed that an explicit $A C^{0}$ secret

[^1]sharing scheme with privacy threshold $k$ and perfect privacy also implies an explicit function in $\mathrm{AC}^{0}$ with approximate degree at least $k$ ). Like most schemes in [7], we only require that the secret can be reconstructed by all $n$ parties. On the other hand, we always require perfect reconstruction. We show that under this slight relaxation, we can obtain much better secret sharing schemes in $\mathrm{AC}^{0}$. For an adaptive adversary, we can achieve both a constant information rate and a large privacy threshold $(k=\Omega(n))$ over a binary alphabet. In addition, our privacy error is exponentially small. Specifically, we have the following theorem.

Theorem 1 (adaptive adversary). For every $n \in \mathbb{N}$ and constant $\gamma \in$ $(0,1 / 4)$, there exists an explicit $(n, \Omega(n))$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, secret length $m=\Omega(n)$, adaptive privacy error $2^{-\Omega\left(n^{\frac{1}{4}-\gamma}\right)}$ and perfect reconstruction.

Note that again, by using randomization and allowing for a small privacy error, we can significantly improve both the privacy threshold and the information rate, while also making the scheme much more efficient by using a smaller alphabet.

Remark 1. We note that a recent paper by Bun and Thaler [11] gave improved lower bounds for the approximate degree of $\mathrm{AC}^{0}$ functions. Specifically, for any constant $\alpha>0$ they showed an explicit $\mathrm{AC}^{0}$ function with approximate degree at least $n^{1-\alpha}$, and by the relation established in [7] this also gives a secret sharing scheme in $\mathrm{AC}^{0}$ with privacy threshold $n^{1-\alpha}$. However, our results are stronger in the sense that we can achieve threshold $\Omega(n)$, and furthermore we can achieve perfect reconstruction while the secret sharing scheme in [11] only has constant reconstruction error.

Remark 2. Our construction of $\mathrm{AC}^{0}$ secret sharing schemes is actually a general transformation and can take any such scheme in [7] or [11] as the starting point. The error $2^{-\Omega\left(n^{\frac{1}{4}-\gamma}\right)}$ in Theorem 1 comes from our use of the one-in-a-box function [28], which has approximate degree $n^{1 / 3}$. We can also use the new $\mathrm{AC}^{0}$ function of [11] with approximate degree $n^{1-\alpha}$, which will give us an error of $2^{-\Omega\left(n^{\frac{1}{2}-\gamma}\right)}$ but the reconstruction error will become a constant. We note that the privacy error of our construction is also close to optimal, without further improvement on the lower bounds of approximate degree of $A C^{0}$ functions. This is because a privacy error of $2^{-s}$ will imply an $\mathrm{AC}^{0}$ function of approximate degree $\Omega(s / \log n)$. Thus if one can achieve a sufficiently small privacy error (e.g., $2^{-\Omega(n)}$ ), then this will give an improved approximate degree lower bound for an $\mathrm{AC}^{0}$ function. See Appendix A of the full version for a more detailed explanation.

A very recent paper by Bogdanov and Williamson [9] considered a similar relaxation as ours. Specifically, they showed how to construct two distributions over $n$ bits that are ( $k, \epsilon$ )-wise indistinguishable, but can be distinguished with advantage $1-\eta$ by some $\mathrm{AC}^{0}$ function. Here $(k, \epsilon)$-wise indistinguishable means
that if looking at any subset of $k$ bits, the two distributions have statistical distance at most $\epsilon$. Translating into the secret sharing model, this roughly implies an $\mathrm{AC}^{0}$ secret sharing scheme with binary alphabet, privacy threshold $k$, privacy error $\epsilon$ and reconstruction error $\eta$. Bogdanov and Williamson [9] obtained several results in this case. Specifically, they showed a pair of such distributions for any $k \leq n / 2$ with $\epsilon=2^{-\Omega(n / k)}$, that can be distinguished with $\eta=\Omega(1)$ by the OR function; or for any $k$ with $\epsilon=2^{-\Omega\left((n / k)^{1-1 / d}\right)}$, that can be distinguished with $\eta=0$ by a depth- $d$ AND-OR tree.

We note the following important differences between our results and the corresponding results by Bogdanov and Williamson [9]: first, the results in [9], in the language of secret sharing, only consider a 1-bit secret, while our results can share $\Omega(n)$ bits with the same share size. Thus our information rate is much larger than theirs. Second, we can achieve a privacy threshold of $k=\Omega(n)$ while simultaneously achieving an exponentially small privacy error of $\epsilon=2^{-n^{\Omega(1)}}$ and perfect reconstruction $(\eta=0)$. In contrast, the results in [9], when going into the range of $k=\Omega(n)$, only have constant privacy error. In short, our results are better than the results in [9], in the sense that we can simultaneously achieve asymptotically optimal information rate and privacy threshold, exponentially small privacy error and perfect reconstruction. As a direct corollary, we have the following result, which is incomparable to the results in [9].

Corollary 1. There exists a constant $\alpha>0$ such that for every $n$ and $k \leq \alpha n$, there exists a pair of $\left(k, 2^{-n^{\Omega(1)}}\right)$-wise indistinguishable distributions $X, Y$ over $\{0,1\}^{n}$ and an $\mathrm{AC}^{0}$ function $D$ such that $\operatorname{Pr}[D(X)]-\operatorname{Pr}[D(Y)]=1$.

Next, we extend our $A C^{0}$ secret sharing schemes to the robust case, where the adversary can tamper with several parties' shares. Our goal is to simultaneously achieve a large privacy threshold, a large tolerance to errors, a large information rate and a small alphabet size. We can achieve a constant information rate with privacy threshold and error tolerance both $\Omega(n)$, with constant size alphabet, exponentially small privacy error and polynomially small reconstruction error. However, here we can only handle a non-adaptive adversary. Specifically, we have the following theorem.

Theorem 2 (non-adaptive adversary). For every $n \in \mathbb{N}$, every $\eta=\frac{1}{\operatorname{poly}(n)}$, there exists an explicit $(n, \Omega(n))$ robust secret sharing scheme in $\mathrm{AC}^{0}$ with share alphabet $\{0,1\}^{O(1)}$, message alphabet $\{0,1\}$, message length $m=\Omega(n)$, nonadaptive privacy error $2^{-n^{\Omega(1)}}$, non-adaptive robustness $\Omega(n)$ and reconstruction error $\eta$.

### 1.2 Error Correcting Codes for Additive Channels in AC $^{0}$

Robust secret sharing schemes are natural generalizations of error correcting codes. Thus our robust secret sharing schemes in $\mathrm{AC}^{0}$ also give error correcting codes with randomized $\mathrm{AC}^{0}$ encoding and deterministic $\mathrm{AC}^{0}$ decoding. The model of our error correcting codes is the same as that considered by Guruswami and

Smith [20]: stochastic error correcting codes for additive channels. Here, the code has a randomized encoding function and a deterministic decoding function, while the channel can add an arbitrary error vector $e \in\{0,1\}^{n}$ of Hamming weight at most $\rho n$ to the transmitted codeword of length $n$. As in [20], the error may depend on the message but crucially does not depend on the randomness used by the encoder. Formally, we have the following definition.

Definition 2. For any $n, m \in \mathbb{N}$, any $\rho, \epsilon>0$, an ( $n, m, \rho$ ) stochastic binary error correcting code (Enc, Dec) with randomized encoding function Enc : $\{0,1\}^{m} \rightarrow\{0,1\}^{n}$, deterministic decoding function Dec : $\{0,1\}^{n} \rightarrow\{0,1\}^{m}$ and decoding error $\epsilon$, is such that for every $x \in\{0,1\}^{m}$, every $e=\left(e_{1}, \ldots, e_{m}\right) \in$ $\{0,1\}^{m}$ with hamming weight at most $\rho n$,

$$
\operatorname{Pr}[\operatorname{Dec}(\operatorname{Enc}(x)+e)=x] \geq 1-\epsilon
$$

An ( $n, m, \rho$ ) stochastic error correcting code (Enc, Dec) can be computed by $\mathrm{AC}^{0}$ circuits if both Enc and Dec can be computed by $\mathrm{AC}^{0}$ circuits.

Previously, Guruswami and Smith [20] constructed such codes that approach the Shannon capacity $1-H(\rho)$. Their encoder and decoder run in polynomial time and have exponentially small decoding error. Here, we aim at constructing such codes with $\mathrm{AC}^{0}$ encoder and decoder. In a different setting, Goldwasser et al. [19] gave a construction of locally decodable codes that can tolerate a constant fraction of errors and have $\mathrm{AC}^{0}$ decoding. Their code has deterministic encoding but randomized decoding. By repeating the local decoder for each bit for $O(\log n)$ times and taking majority, one can decode each bit in $\mathrm{AC}^{0}$ with error probability $1 / \operatorname{poly}(n)$ and thus by a union bound the original message can also be decoded with error probability $1 / \operatorname{poly}(n)$. However we note that the encoding function of [19] is not in $\mathrm{AC}^{0}$, and moreover their message rate is only polynomially small. In contrast, our code has constant message rate and can tolerate a constant fraction of errors (albeit in a weaker model) when the decoding error is $1 / \operatorname{poly}(n)$ or even $2^{-\operatorname{poly} \log (n)}$. The rate and tolerance are asymptotically optimal. We can achieve even smaller error $\left(2^{-\Omega(r / \log n)}\right)$ with message rate $1 / r$. Furthermore both our encoding and decoding are in $A C^{0}$. Specifically, we have the following theorems.

Theorem 3 (error-correcting codes). For any $n \in \mathbb{N}$ and $\epsilon=2^{-\operatorname{poly} \log (n)}$, there exists an $(n, \Omega(n), \Omega(1))$ stochastic binary error correcting code with decoding error $\epsilon$, which can be computed by $\mathrm{AC}^{0}$ circuits.

Theorem 4 (error-correcting codes with smaller decoding error). For any $n, r \in \mathbb{N}$, there exists an ( $n, m=\Omega(n / r), \Omega(1))$ stochastic binary error correcting code with decoding error $2^{-\Omega(r / \log n)}$, which can be computed by $\mathrm{AC}^{0}$ circuits.

Note that Theorem 4 is interesting mainly in the case where $r$ is at least poly $\log n$.

Remark 3. We note that, without randomization, it is well known that deterministic $A C^{0}$ circuits cannot compute asymptotically good codes [25]. Thus the randomization in our $\mathrm{AC}^{0}$ encoding is necessary here. For deterministic $\mathrm{AC}^{0}$ decoding, only very weak lower bounds are known. In particular, Lee and Viola [23] showed that any depth-c $\mathrm{AC}^{0}$ circuit with parity gates cannot decode beyond error $\left(1 / 2-1 / O(\log n)^{c+2}\right) d$, where $d$ is the distance of the code. While the repetition code can be decoded in $\mathrm{AC}^{0}$ with a near-optimal fraction of errors by using approximate majority, obtaining a similar positive result for codes with a significantly better rate is open.

### 1.3 Secure Broadcasting with an External Adversary

We apply our ideas and technical approach to the following flavor of secure broadcasting in the presence of an adversary. The problem can be viewed as a generalization of a one-time pad encryption. In a one-time pad encryption, two parties share a secret key which can be used to transmit messages with information-theoretic security. Suppose that each party wants to transmit an $m$-bit string to the other party. If an external adversary can see the entire communication, then it is well known that to keep both messages secret, the parties must share a secret key of length at least 2 m . This can be generalized to the case of $n$ parties, where we assume that they have access to a public broadcast channel, and each party wants to securely communicate an $m$-bit string to all other parties. This problem can be useful, for example, when $n$ collaborating parties want to compute a function of their secret inputs without revealing the inputs to an external adversary. Again, if the adversary can see the entire communication, then the parties need to share a secret key of length at least $n m$.

Now, what if we relax the problem by restricting the adversary's power? Suppose that instead of seeing the entire communication, the adversary can only see some fraction of the communicated messages. Can we get more efficient solutions? We formally define this model below, requiring not only the secrecy of the inputs but also correctness of the outputs in the presence of adaptive tampering with a bounded fraction of messages.

Definition 3. Let $n, m \in \mathbb{N}$ and $\alpha, \epsilon>0$. An ( $n, m, \alpha, \epsilon, \eta$ )-secure broadcasting protocol is an n-party protocol with the following properties. Initially, each party $i$ has a local input $x_{i} \in\{0,1\}^{m}$ and the parties share a secret key. The parties can then communicate over a public broadcast channel. At the end of the communication, each party computes a local output. We require the protocol to satisfy the following security properties.

- (Privacy) For any adaptive adversarial observation W which observes at most $1-\alpha$ fraction of the messages, there is a distribution $\mathcal{D}$, such that for any inputs $x=\left(x_{1}, \ldots, x_{n}\right) \in\left(\{0,1\}^{m}\right)^{n}$ leading to a sequence of messages $Y$, the distribution $Y_{W}$ of observed messages is $\epsilon$-close to $\mathcal{D}$.
- (Robustness) For any adaptive adversary that corrupts at most $1-\alpha$ fraction of the messages, and any $n$-tuple of inputs $x=\left(x_{1}, \ldots, x_{n}\right) \in\left(\{0,1\}^{m}\right)^{n}$, all
$n$ parties can reconstruct $x$ correctly with probability at least $1-\eta$ after the communication.

The naive solution of applying one-time pad still requires a shared secret key of length at least $n m$, since otherwise even if the adversary only sees part of the communication, he may learn some information about the inputs. However, by using randomization and allowing for a small error, we can achieve much better performance. Specifically, we have the following theorem.

Theorem 5 (secure broadcasting). For any $n, m, r \in \mathbb{N}$ with $r \leq m$, there exists an explicit ( $\left.n, m, \alpha=\Omega(1), n 2^{-\Omega(r)}, n 2^{-\Omega(r)}+n m 2^{-\Omega(m / r)}\right)$ secure broadcasting protocol with communication complexity $O(n m)$ and shared secret key of length $O(r \log (n r))$.

### 1.4 Overview of the Techniques

Secret sharing. Here we give an overview of the techniques used in our constructions of $\mathrm{AC}^{0}$ secret sharing schemes and error correcting codes. Our constructions combine several ingredients in pseudorandomness and combinatorics in an innovative way, so before describing our constructions, we will first describe the important ingredients used.

The secret sharing scheme in [7]. As mentioned before, Bogdanov et al. [7] were the first to consider secret sharing schemes in $\mathrm{AC}^{0}$. Our constructions will use one of their schemes as the starting point. Specifically, since we aim at perfect reconstruction, we will use the secret sharing scheme in [28] based on the so called "one-in-a-box function" or Minsky-Papert CNF function. This scheme can share one bit among $n$ parties, with binary alphabet, privacy threshold $\Omega\left(n^{1 / 3}\right)$ and perfect reconstruction.

Random permutation. Another important ingredient, as mentioned before, is random permutation. Applying a random permutation, in many cases, reduces worst case errors to random errors, and the latter is much more convenient to handle. This property has been exploited in several previous work, such as the error correcting codes by Guruswami and Smith [20]. We note that a random permutation from $[n]$ to $[n]$ can be computed in $\mathrm{AC}^{0}[22,26,33]$.
$K$-wise independent generators. The third ingredient of our construction is the notion of $k$-wise independent pseudorandom generators. This is a function that stretches some $r$ uniform random bits to $n$ bits such that any subset of $k$ bits is uniform. Such generators are well studied, while for our constructions we need such generators which can be computed by $\mathrm{AC}^{0}$ circuits. This requirement is met by using $k$-wise independent generators based on unique neighbor expander graphs, such as those constructed by Guruswami et al. [21] which use seed length $r=k$ poly $\log (n)$.

Secret sharing schemes based on error correcting codes. Using asymptotically good linear error correcting codes, one can construct secret sharing schemes that
simultaneously achieve constant information rate and privacy threshold $\Omega(n)$ (e.g., [12]). However, certainly in general these schemes are not in $A C^{0}$ since they need to compute linear functions such as parity. For our constructions, we will use these schemes with a small block length (e.g., $O(\log n)$ or poly $\log (n))$ such that parity with such input length can be computed by constant depth circuits. For robust secret sharing, we will also be using robust secret sharing schemes based on codes, with constant information rate, privacy threshold and tolerance $\Omega(n)$ (e.g., [15]), with a small block length.

The constructions. We can now give an informal description of our constructions. As mentioned before, our construction is a general transformation and can take any scheme in [7] or [11] as the starting point. A specific scheme of interest is the one in [7] based on the one-in-a-box function, which has perfect reconstruction. Our goal then is to keep the property of perfect reconstruction, while increasing the information rate and privacy threshold. One naive way to share more bits is to repeat the scheme several times, one for each bit. Of course, this does not help much in boosting the information rate. Our approach, on the other hand, is to use this naive repeated scheme to share a short random seed $R$. Suppose this gives us $n$ parties with privacy threshold $k_{0}$. We then use $R$ and the $k$-wise independent generator $G$ mentioned above to generate an $n$-bit string $Y$, and use $Y$ to share a secret $X$ by computing $Y \oplus X$.

Note that now the length of the secret $X$ can be as large as $n$ and thus the information rate is increased to $1 / 2$. To reconstruct the secret, we can use the first $n$ parties to reconstruct $R$, then compute $Y$ and finally $X$. Note that the whole computation can be done in $\mathrm{AC}^{0}$ since the $k$-wise independent generator $G$ is computable in $\mathrm{AC}^{0}$. The privacy threshold, on the other hand, is the minimum of $k_{0}$ and $k$. This is because if an adversary learns nothing about $R$, then $Y$ is $k$-wise independent and thus by looking at any $k$ shares in $Y \oplus X$, the adversary learns nothing about $X$. This is the first step of our construction.

In the next step, we would like to boost the privacy threshold to $\Omega(n)$ while decreasing the information rate by at most a constant factor. Our approach for this purpose can be viewed as concatenating a larger outer protocol with a smaller inner protocol, which boosts the privacy threshold while keeping the information rate and the complexity of the whole protocol. More specifically, we first divide the parties obtained from the first step into small blocks, and then for each small block we use a good secret sharing scheme based on error correcting codes. Suppose the adversary gets to see a constant fraction of the shares, then on average for each small bock the adversary also gets to see only a constant fraction of the shares. Thus, by Markov's inequality and adjusting the parameters, the adversary only gets to learn the information from a constant fraction of the blocks. However, this is still not enough for us, since the outer protocol only has threshold $n^{\Omega(1)}$.

We solve this problem by using a threshold amplification technique. This is one of our main innovations, and a key step towards achieving both constant information rate and privacy threshold $\Omega(n)$ without sacrificing the error. On a high level, we turn the inner protocol itself into another concatenated protocol
(i.e., a larger outer protocol combined with a smaller inner protocol), and then apply a random permutation. Specifically, we choose the size of the block mentioned above to be something like $O\left(\log ^{2} n\right)$, apply a secret sharing scheme based on asymptotically good error correcting codes and obtain $O\left(\log ^{2} n\right)$ shares. We then divide these shares further into $O(\log n)$ smaller blocks each of size $O(\log n)$ (alternatively, this can be viewed as a secret sharing scheme using alphabet $\{0,1\}^{O(\log n)}$ ), and now we apply a random permutation of these smaller blocks. If we are to use a slightly larger alphabet, we can now store each block together with its index before the permutation as one share. Note that we need the index information when we try to reconstruct the secret, and the reconstruction can be done in $\mathrm{AC}^{0}$.

Now, suppose again that the adversary gets to see some small constant fraction of the final shares, then since we applied a random permutation, we can argue that each smaller block gets learned by the adversary only with some constant probability. Thus, in the larger block of size $O\left(\log ^{2} n\right)$, by a Chernoeff type bound, except with probability $1 / \operatorname{poly}(n)$, we have that only some constant fraction of the shares are learned by the adversary. Note that here by using two levels of blocks, we have reduced the probability that the adversary learns some constant fraction of the shares from a constant to $1 /$ poly $(n)$, which is much better for the outer protocol as we shall see soon. By adjusting the parameters we can ensure that the number of shares that the adversary may learn is below the privacy threshold of the larger block and thus the adversary actually learns nothing. Now, going back to the outer protocol, we know that the expected number of large blocks the adversary can learn is only $n /$ poly $(n)$; and again by a Chernoff type bound, except with probability $2^{-n^{\Omega(1)}}$, the outer protocol guarantees that the adversary learns nothing. This gives us a secret sharing scheme with privacy threshold $\Omega(n)$ while the information rate is still constant since we only increased the number of shares by a constant factor. With the $O(\log n)$ size alphabet, we can actually achieve privacy threshold $(1-\alpha) n^{\prime}$ for any constant $0<\alpha<1$, where $n^{\prime}$ is the total number of final parties.

To reduce to the binary alphabet, we can apply another secret sharing scheme based on error correcting codes to each share of length $O(\log n)$. In this case then we won't be able to achieve privacy threshold $(1-\alpha) n^{\prime}$, but we can achieve $\beta n^{\prime}$ for some constant $\beta>0$. This is because if the adversary gets to see a small constant fraction of the shares, then by Markov's inequality only for some constant fraction of the smaller blocks the adversary can learn some useful information. Thus the previous argument still holds.

As described above, our general construction uses two levels of concatenated protocols, which corresponds to two levels of blocks. The first level has larger blocks of size $O\left(\log ^{2} n\right)$, where each larger block consists of $O(\log n)$ smaller blocks of size $O(\log n)$. We use this two-level structure to reduce the probability that an adversary can learn some constant fraction of shares, and this enables us to amplify the privacy threshold to $\Omega(n)$. We choose the smaller block to have size $O(\log n)$ so that both a share from the larger block with length $O(\log n)$ and its index information can be stored in a smaller block. This ensures that the
information rate is still a constant even if we add the index information. Finally, the blocks in the second level are actually the blocks that go into the random permutation. This general strategy is one of our main contributions and we hope that it can find other applications.

The above construction gives an $\mathrm{AC}^{0}$ secret sharing scheme with good parameters. However, it is not a priori clear that it works for an adaptive adversary. In standard secret sharing schemes, a non-adaptive adversary and an adaptive adversary are almost equivalent since usually we have privacy error 0. More specifically, a secret sharing scheme for a non-adaptive adversary with privacy error $\epsilon$ and privacy threshold $k$ is also a secret sharing scheme for an adaptive adversary with privacy error $n^{k} \epsilon$ and privacy threshold $k$. However in our $\mathrm{AC}^{0}$ secret sharing scheme the error $\epsilon$ is not small enough to kill the $n^{k}$ factor. Instead, we use the property of the random permutation to argue that our final distribution is essentially symmetric; and thus informally no matter how the adversary picks the shares to observe adaptively, he will not gain any advantage. This will show that our $A C^{0}$ secret sharing scheme also works for an adaptive adversary.

To extend to robust secret sharing, we need to use robust secret sharing schemes instead of normal schemes for the first and second level of blocks. Here we use the nearly optimal robust secret sharing schemes based on various codes by Cheraghchi [15]. Unfortunately since we need to use it on a small block length of $O(\log n)$, the reconstruction error becomes $1 /$ poly $(n)$. Another tricky issue here is that an adversary may modify some of the indices. Note that we need the correct index information in order to know which block is which before the random permutation. Suppose the adversary does not modify any of the indices, but only modify the shares, then the previous argument can go through exactly when we change the secret sharing schemes based on error correcting codes into robust secret sharing schemes. However, if the adversary modifies some indices, then we could run into situations where more than one block have the same index and thus we cannot tell which one is correct (and it's possible they are all wrong). To overcome this difficulty, we store every index multiple times among the blocks in the second level. Specifically, after we apply the random permutation, for every original index we randomly choose $O(\log n)$ blocks in the second level to store it. As the adversary can only corrupt a small constant fraction of the blocks in the second level, for each such block, we can correctly recover its original index with probability $1-1 / \operatorname{poly}(n)$ by taking the majority of the backups of its index. Thus by a union bound with probability $1-1 / \operatorname{poly}(n)$ all original indices can be correctly recovered. In addition, we use the same randomness for each block to pick the $O(\log n)$ blocks, except we add a different shift to the selected blocks. This way, we can ensure that for each block the $O(\log n)$ blocks are randomly selected and thus the union bound still holds. Furthermore the randomness used here is also stored in every block in the second level, so that we can take the majority to reconstruct it correctly. In the above description, we sometimes need to take majority for $n$ inputs, which is not computable in $\mathrm{AC}^{0}$. However, we note that by adjusting parameters we can ensure that at least say $2 / 3$ fraction of the
inputs are the same, and in this case it suffices to take approximate majority, which can be computed in $\mathrm{AC}^{0}$ [32].

For our error correcting codes, the construction is a simplified version of the robust secret sharing construction. Specifically, we first divide the message itself into blocks of the first level, and then encode every block using an asymptotically good code and divide the obtained codeword into blocks of the second level. Then we apply a random permutation to the blocks of the second level as before, and we encode every second level block by another asymptotically good code. In short, we replace the above mentioned robust secret sharing schemes by asymptotically good error correcting codes. We use the same strategy as in robust secret sharing to identify corrupted indices. Using a size of $O\left(\log ^{2} n\right)$ for blocks in the first level will result in decoding error $1 /$ poly $(n)$, while using larger block size (e.g., poly $\log (n)$ ) will result in decoding error $2^{- \text {poly } \log (n)}$. This gives Theorem 3. To achieve even smaller error, we can first repeat each bit of the message $r$ times for some parameter $r$. This serves as an outer error correcting code, which can tolerate up to $r / 3$ errors, and can be decoded in in $\mathrm{AC}^{0}$ by taking approximate majority. The two-level block structure and the argument we described before can now be used to show a smaller decoding error of $2^{-\Omega\left(r / \log ^{2} n\right)}$. This gives Theorem 4 .

Secure broadcasting. Rather than use the naive approach of one-time pad, here a more clever solution is to use secret sharing (assuming that each party also has access to local private random bits). By first applying a secret sharing scheme to the input and then broadcasting the shares, a party can ensure that if the adversary only gets to see part of the messages (below the secrecy threshold), then the adversary learns nothing. In this case the parties do not even need shared secret key. However, one problem with this solution is that the adversary cannot be allowed to see more than $1 / n$ fraction of the messages, since otherwise he can just choose the messages broadcasted from one particular party, and then the adversary learns the input of that party. This is the place where randomization comes into play. If in addition, we allow the parties to share a small number of secret random bits, then the parties can use this secret key to randomly permute the order in which the they broadcast their messages (after applying the secret sharing scheme). Since the adversary does not know the secret key, we can argue that with high probability only a small fraction of each party's secret shares are observed. Therefore, by the properties of secret sharing we can say that the adversary learns almost nothing about each party's input. The crucial features of this solution are that first, the adversary can see some fixed fraction of messages, which is independent of the number of parties $n$ (and thus can be much larger than $1 / n$ ). Second, the number of shared secret random bits is much smaller than the naive approach of one-time pad. Indeed, as we show in Theorem 23, to achieve security parameter roughly $r$ it is enough for the parties to share $O(r(\log n+\log r))$ random bits. Finally, by using an appropriate secret sharing scheme, the communication complexity of our protocol for each party is $O(m)$, which is optimal up to a constant factor. Note that here, by applying random permutation and allowing for a small probability of error, we simulta-
neously improve the security threshold (from $1 / n$ to $\Omega(1)$ ) and the length of the shared secret key (from $n m$ to $O(r(\log n+\log r)))$.

Discussions and open problems. In this paper we continue the line of work on applying randomization and allowing a small failure probability for minimizing the computational complexity of cryptographic primitives and related combinatorial objects while maximizing the level of achievable security. In the context of secret sharing in $A C^{0}$, we show how to get much better parameters by allowing an (exponentially) small privacy error. We note that achieving exponentially small error here is non-trivial. In fact, if we allow for a larger error then (for a non-adaptive adversary) there is a simple protocol for $A C^{0}$ secret sharing: one can first take a random seed $R$ of length $\Omega(n)$, and then apply a deterministic $\mathrm{AC}^{0}$ extractor for bit-fixing sources to obtain an output $Y$ of length $\Omega(n)$. The secret $X$ can then be shared by computing the parity of $Y$ and $X$. This way, one can still share $\Omega(n)$ bits of secret, and if the adversary only learns some small fraction of the seed, then the output $Y$ is close to uniform by the property of the extractor, and thus $X$ remains secret. However, by the lower bound of [14], the error of such $A C^{0}$ extractors (or even for the stronger seeded $A C^{0}$ extractors) is at least $2^{- \text {poly } \log (n)}$. Therefore, one has to use additional techniques to achieve exponentially small error. We also extended our techniques to robust $\mathrm{AC}^{0}$ secret sharing schemes, stochastic error correcting codes for additive channels, and secure broadcasting. Several intriguing open problems remain.

First, in our robust $\mathrm{AC}^{0}$ secret sharing schemes, we only achieve reconstruction error $1 / \operatorname{poly}(n)$. This is because we need to use existing robust secret sharing schemes on a block of $\operatorname{size} O(\log n)$. Is it possible to avoid this and make the error exponentially small? Also, again in this case we can only handle non-adaptive adversaries, and it would be interesting to obtain a robust $\mathrm{AC}^{0}$ secret sharing scheme that can handle adaptive adversaries. These questions are open also for $\mathrm{AC}^{0}$ stochastic error correcting codes.

Second, as we mentioned in Remark 2 (see also [9]), a sufficiently small privacy error in an $A C^{0}$ secret sharing scheme would imply an improved approximate degree lower bound for $A C^{0}$ functions. Is it possible to improve our $A C^{0}$ secret sharing scheme, and use this approach to obtain better approximate degree lower bound for $\mathrm{AC}^{0}$ functions? This seems like an interesting direction.

In addition, the privacy threshold amplification technique we developed, by using two levels of concatenated protocols together with a random permutation, is quite general and we feel that it should have applications elsewhere. We note that the approach of combining an "outer scheme" with an "inner scheme" to obtain the best features of both has been applied in many previous contexts. For instance, it was used to construct better codes $[1,20]$ or better secure multiparty computation protocols [17]. However, in almost all of these previous applications, one starts with an outer scheme with a very good threshold (e.g., the Reed-Solomon code which has a large distance) and the goal is to use the inner scheme to inherit this good threshold while improving some other parameters (such as alphabet size). Thus, one only needs one level of concatenation. In our case, instead, we start with an outer scheme with a very weak threshold (e.g.,
the one-in-a-box function which only has privacy threshold $n^{1 / 3}$ ). By using two levels of concatenated protocols together with a random permutation, we can actually amplify this threshold to $\Omega(n)$ while simultaneously reducing the alphabet size. This is an important difference to previous constructions and one of our main contributions. We hope that these techniques can find other applications in similar situations.

Finally, since secret sharing schemes are building blocks of many other important cryptographic applications, it is an interesting question to see if the lowcomplexity secret sharing schemes we developed here can be used to reduce the computational complexity of other cryptographic primitives.
Paper organization. We introduce some notation and useful results in Sect. 2. In Sect. 3 we give our privacy threshold amplification techniques. In Sect.4, we show how to increase the information rate using k-wise independent generators. Combining all the above techniques, our final construction of $\mathrm{AC}^{0}$ secret sharing schemes is given in Sect.5. Instantiations appear in Sect.6. Finally, we give our constructions of robust $A C^{0}$ secret sharing schemes, $A C^{0}$ error correcting codes, and secure broadcast protocols in Sect. 7. Some proofs and additional details have been deferred to the full version [13].

## 2 Preliminaries

Let $|\cdot|$ denote the size of the input set or the absolute value of an input real number, based on contexts.

For any set $I$ of integers, for any $r \in \mathbb{Z}$, we denote $r+I$ or $I+r$ to be $\left\{i^{\prime}: i^{\prime}=i+r, i \in I\right\}$.

We use $\Sigma$ to denote the alphabet. Readers can simply regard $\Sigma$ as $\{0,1\}^{l}$ for some $l \in \mathbb{N}$. For $\sigma \in \Sigma$, let $\sigma^{n}=(\sigma, \sigma, \ldots, \sigma) \in \Sigma^{n}$. For any sequence $s=\left(s_{1}, s_{2}, \ldots, s_{n}\right) \in \Sigma^{n}$ and sequence of indices $W=\left(w_{1}, \ldots, w_{t}\right) \in[n]^{t}$ with $t \leq n$, let $s_{W}$ be the subsequence $\left(s_{w_{1}}, s_{w_{2}}, \ldots, s_{w_{t}}\right)$.

For any two sequences $a \in \Sigma^{n}, b \in \Sigma^{\prime n^{\prime}}$ where $a=\left(a_{1}, a_{2}, \ldots, a_{n}\right), b=$ $\left(b_{1}, b_{2}, \ldots, b_{n^{\prime}}\right)$, let $a \circ b=\left(a_{1}, \ldots, a_{n}, b_{1}, \ldots, b_{n^{\prime}}\right) \in \Sigma^{n} \times \Sigma^{\prime n^{\prime}}$.

Let $\operatorname{supp}(\cdot)$ denote the support of the input random variable. Let $I(\cdot)$ be the indicator function.

Definition 4 (Statistical Distance). The statistical distance between two random variables $X$ and $Y$ over $\Sigma^{n}$ for some alphabet $\Sigma$, is $\operatorname{SD}(X, Y)$ which is defined as follows,

$$
\mathrm{SD}(X, Y)=1 / 2 \sum_{a \in \Sigma^{n}}|\operatorname{Pr}[X=a]-\operatorname{Pr}[Y=a]|
$$

Here we also say that $X$ is $\mathrm{SD}(X, Y)$-close to $Y$.

## Lemma 1 (Folklore Properties of Statistical Distance [4]).

1. (Triangle Inequality) For any random variables $X, Y, Z$ over $\Sigma^{n}$, we have

$$
\mathrm{SD}(X, Y) \leq \mathrm{SD}(X, Z)+\mathrm{SD}(Y, Z)
$$

2. $\forall n, m \in \mathbb{N}$, any deterministic function $f:\{0,1\}^{n} \rightarrow\{0,1\}^{m}$ and any random variables $X, Y$ over $\Sigma^{n}$,

$$
\mathrm{SD}(f(X), f(Y)) \leq \mathrm{SD}(X, Y)
$$

We will use the following well known perfect XOR secret sharing scheme.
Theorem 6 (Folklore XOR secret sharing). For any finite field $\mathbb{F}$, define Share $_{+}: \mathbb{F} \rightarrow \mathbb{F}^{n}$ and $\operatorname{Rec}_{+}: \mathbb{F}^{n} \rightarrow \mathbb{F}$, such that for any secret $x \in \mathbb{F}$, $\operatorname{Share}_{+}(x)=$ $y$ such that $y$ is uniformly chosen in $\mathbb{F}^{n}$ conditioned on $\sum_{i \in[n]} y_{i}=x$ and $\operatorname{Rec}_{+}$ is taking the sum of its input.
(Share ${ }_{+}, \operatorname{Rec}_{+}$) is an $(n, n-1)$ secret sharing scheme with share alphabet and message alphabet both being $\mathbb{F}$, message length 1, perfect privacy and reconstruction.

Definition 5 (Permutation). For any $n \in \mathbb{N}$, a permutation over $[n]$ is defined to be a bijective function $\pi:[n] \rightarrow[n]$.

Definition 6 (k-wise independence). For any set $S$, let $X_{1}, \ldots, X_{n}$ be random variables over $S$. They are $k$-wise independent (and uniform) if any $k$ of them are independent (and uniformly distributed).

For any $r, n, k \in \mathbb{N}$, a function $g:\{0,1\}^{r} \rightarrow \Sigma^{n}$ is a $k$-wise (uniform) independent generator, if for $g(U)=\left(Y_{1}, \ldots, Y_{n}\right), Y_{1}, \ldots, Y_{n}$ are $k$-wise independent (and uniform). Here $U$ is the uniform distribution over $\{0,1\}^{r}$.

Definition 7 [21]. A bipartite graph with $N$ left vertices, $M$ right vertices and left degree $D$ is a $(K, A)$ expander if for every set of left vertices $S \subseteq[N]$ of size $K$, we have $|\Gamma(S)|>A K$. It is a $\left(\leq K_{\max }, A\right)$ expander if it is a $(K, A)$ expander for all $K \leq K_{\text {max }}$.

Here $\forall x \in[N], \Gamma(x)$ outputs the set of all neighbours of $x$. It is also a set function which is defined accordingly. Also $\forall x \in[N], d \in[D]$, the function $\Gamma:[N] \times[D] \rightarrow[M]$ is such that $\Gamma(x, d)$ is the $d$ th neighbour of $x$.

Theorem 7 [21]. For all constants $\alpha>0$, for every $N \in \mathbb{N}, K_{\max } \leq N$, and $\epsilon>0$, there exists an explicit $\left(\leq K_{\max },(1-\epsilon) D\right)$ expander with $N$ left vertices, $M$ right vertices, left degree $D=O\left((\log N)\left(\log K_{\max }\right) / \epsilon\right)^{1+1 / \alpha}$ and $M \leq D^{2} K_{\max }^{1+\alpha}$. Here $D$ is a power of 2 .

For any circuit $C$, the size of $C$ is denoted as size $(C)$. The depth of $C$ is denoted as depth $(C)$. Usually when we talk about computations computable by $A C^{0}$ circuits, we mean uniform $A C^{0}$ circuits, if not stated specifically.

Lemma 2 (Folklore properties of $\mathrm{AC}^{0}$ circuits [4,19]). For every $n \in \mathbb{N}$,

1. ([4] Folklore) every Boolean function $f:\{0,1\}^{l=\Theta(\log n)} \rightarrow\{0,1\}$ can be computed by an $\mathrm{AC}^{0}$ circuit of size poly $(n)$ and depth 2 .
2. [19] for every $c \in \mathbb{N}$, every integer $l=\Theta\left(\log ^{c} n\right)$, if the function $f_{l}:\{0,1\}^{l} \rightarrow$ $\{0,1\}$ can be computed by a circuit with depth $O(\log l)$ and size poly $(l)$, then it can be computed by a circuit with depth $c+1$ and size poly $(n)$.

Remark 4. We briefly describe the proof implied in [19] for the second property of our Lemma 2. As there exists an $N C^{1}$ complete problem which is downward self-reducible, the function $f_{l}$ can be reduced to ( $\mathrm{AC}^{0}$ reduction) a function with input length $O(\log n)$. By Lemma 2 part 1, and noting that the reduction here is an $\mathrm{AC}^{0}$ reduction, $f_{l}$ can be computed by an $\mathrm{AC}^{0}$ circuit.

## 3 Random Permutation

### 3.1 Increasing the Privacy Threshold

The main technique we use here is random permutation.
Lemma $3[22,26,33]$. For any constant $c \geq 1$, there exists an explicit $\mathrm{AC}^{0}$ circuit $C:\{0,1\}^{r} \rightarrow[n]^{n}$ with size poly $(n)$, depth $O(1)$ and $r=O\left(n^{c+1} \log n\right)$ such that with probability $1-2^{-n^{c}}, C\left(U_{r}\right)$ gives a uniform random permutation of $[n]$; When this fails the outputs are not distinct.

In the following we give a black box $\mathrm{AC}^{0}$ transformation of secret sharing schemes increasing the privacy threshold.

Construction 1. For any $n, k, m \in \mathbb{N}$ with $k \leq n$, any alphabet $\Sigma, \Sigma_{0}$, let (Share, Rec) be an ( $n, k$ ) secret sharing scheme with share alphabet $\Sigma$, message alphabet $\Sigma_{0}$, message length $m$.

Let $\left(\right.$ Share $\left._{+}, \operatorname{Rec}_{+}\right)$be a $(t, t-1)$ secret sharing scheme with alphabet $\Sigma$ by Theorem 6 .

For any constant $a \geq 1, \alpha>0$, large enough $b \geq 1$, we can construct the following $\left(n^{\prime}=t n \bar{n}, k^{\prime}=(1-\alpha) n^{\prime}\right)$ secret sharing scheme (Share' ${ }^{\prime} \operatorname{Rec}^{\prime}$ ) with share alphabet $\Sigma \times\left[n^{\prime}\right]$, message alphabet $\Sigma_{0}$, message length $m^{\prime}=m \bar{n}$, where $t=O(\log n), \bar{n}=b n^{a-1}$.

Function Share ${ }^{\prime}: \Sigma_{0}^{m^{\prime}} \rightarrow\left(\Sigma \times\left[n^{\prime}\right]\right)^{n^{\prime}}$ is as follows.

1. On input secret $x \in \Sigma_{0}^{m \bar{n}}$, parse $x$ to be $\left(x_{1}, x_{2}, \ldots, x_{\bar{n}}\right) \in\left(\Sigma_{0}^{m}\right)^{\bar{n}}$.
2. Compute $y=\left(y_{1}, \ldots, y_{\bar{n}}\right)=\left(\operatorname{Share}\left(x_{1}\right), \ldots, \operatorname{Share}\left(x_{\bar{n}}\right)\right)$ and parse it to be $\hat{y}=\left(\hat{y}_{1}, \ldots, \hat{y}_{n \bar{n}}\right) \in \Sigma^{n \bar{n}}$. Note that Share is from $\Sigma_{0}^{m}$ to $\Sigma^{n}$.
3. Compute $\left(\operatorname{Share}_{+}\left(\hat{y}_{1}\right), \ldots\right.$, Share $\left._{+}\left(\hat{y}_{n \bar{n}}\right)\right) \in\left(\Sigma^{t}\right)^{n \bar{n}}$ and split every entry to be $t$ elements in $\Sigma$ to get $y^{\prime}=\left(y_{1}^{\prime}, \ldots, y_{n^{\prime}}^{\prime}\right) \in \Sigma^{n^{\prime}}$. Note that Share ${ }_{+}$is from $\Sigma$ to $\Sigma^{t}$.
4. Generate $\pi$ by Lemma 3 which is uniformly random over permutations of $\left[n^{\prime}\right]$. If it fails, which can be detected by checking element distinctness, set $\pi$ to be such that $\forall i \in\left[n^{\prime}\right], \pi(i)=i$.
5. Let

$$
\operatorname{Share}^{\prime}(x)=\left(y_{\pi^{-1}(1)}^{\prime} \circ \pi^{-1}(1), \ldots, y_{\pi^{-1}\left(n^{\prime}\right)}^{\prime} \circ \pi^{-1}\left(n^{\prime}\right)\right) \in\left(\Sigma \times\left[n^{\prime}\right]\right)^{n^{\prime}}
$$

Function $\operatorname{Rec}^{\prime}:\left(\Sigma \times\left[n^{\prime}\right]\right)^{n^{\prime}} \rightarrow \Sigma_{0}^{m^{\prime}}$ is as follows.

1. Parse the input to be $\left(y_{\pi^{-1}(1)}^{\prime} \circ \pi^{-1}(1), \ldots, y_{\pi^{-1}\left(n^{\prime}\right)}^{\prime} \circ \pi^{-1}\left(n^{\prime}\right)\right)$.
2. Compute $y^{\prime}=\left(y_{1}^{\prime}, \ldots, y_{n^{\prime}}^{\prime}\right)$ according to the permutation.
3. Apply $\operatorname{Rec}_{+}$on $y^{\prime}$ for every successive $t$ entries to get $\hat{y}$.
4. Parse $\hat{y}$ to be $y$.
5. Compute $x$ by applying Rec on every entry of $\hat{y}$.
6. Output $x$.

Lemma 4. If Share and Rec can be computed by $\mathrm{AC}^{0}$ circuits, then Share' and Rec' can also be computed by $\mathrm{AC}^{0}$ circuits.
Proof. As Share can be computed by an $\mathrm{AC}^{0}$ circuit, $y$ can be computed by an $\mathrm{AC}^{0}$ circuit (uniform). By Lemma 2 part 1, we know that (Share ${ }_{+}, \mathrm{Rec}_{+}$) both can be computed by $\mathrm{AC}^{0}$ circuits. By Lemma 3, $\left(\pi^{-1}(1), \pi^{-1}(2), \ldots, \pi^{-1}\left(n^{\prime}\right)\right)$ can be computed by an $\mathrm{AC}^{0}$ circuit. Also

$$
\begin{equation*}
\forall i \in\left[n^{\prime}\right], y_{\pi^{-1}(i)}^{\prime}=\bigvee_{j \in\left[n^{\prime}\right]}\left(y_{j}^{\prime} \wedge\left(j=\pi^{-1}(i)\right)\right) \tag{1}
\end{equation*}
$$

Thus Share ${ }^{\prime}$ can be computed by an $\mathrm{AC}^{0}$ circuit.
For $\operatorname{Rec}^{\prime}, \forall i \in\left[n^{\prime}\right], y_{i}^{\prime}=\bigvee_{j \in\left[n^{\prime}\right]}\left(y_{\pi^{-1}(j)}^{\prime} \wedge\left(\pi^{-1}(j)=i\right)\right)$. As Rec ${ }_{+}$can be computed by an $\mathrm{AC}^{0}$ circuit, $y$ can be computed by an $\mathrm{AC}^{0}$ circuit. As Rec can be computed by an $\mathrm{AC}^{0}$ circuit, Rec can be computed by an $\mathrm{AC}^{0}$ circuit.
Lemma 5. If the reconstruction error of (Share, Rec) is $\eta$, then the reconstruction error of (Share ${ }^{\prime}$, $\operatorname{Rec}^{\prime}$ ) is $\eta^{\prime}=\bar{n} \eta$.

Proof. According to the construction, as (Share ${ }_{+}$, Rec $_{+}$) has perfect reconstruction by Lemma 6, the $y$ computed in $\operatorname{Rec}^{\prime}$ is exactly $\left(\operatorname{Share}\left(x_{1}\right), \ldots, \operatorname{Share}\left(x_{\bar{n}}\right)\right)$. As $\forall i \in[\bar{n}], \operatorname{Pr}\left[\operatorname{Rec}\left(\operatorname{Share}\left(x_{i}\right)\right)=x_{i}\right] \geq 1-\eta$,

$$
\begin{equation*}
\operatorname{Pr}\left[\operatorname{Rec}^{\prime}\left(\operatorname{Share}^{\prime}(x)\right)=x\right]=\operatorname{Pr}\left[\bigwedge_{i \in[\bar{n}]}\left(\operatorname{Rec}\left(\operatorname{Share}\left(x_{i}\right)\right)=x_{i}\right)\right] \geq 1-\bar{n} \eta \tag{2}
\end{equation*}
$$

by the union bound.
In order to show privacy, we need the following Chernoff Bound.
Definition 8 (Negative Correlation [5,6]). Binary random variables $X_{1}, X_{2}, \ldots, X_{n}$ are negative correlated, if $\forall I \subseteq[n]$,

$$
\operatorname{Pr}\left[\bigwedge_{i \in I}\left(X_{i}=1\right)\right] \leq \prod_{i \in I} \operatorname{Pr}\left[X_{i}=1\right] \text { and } \operatorname{Pr}\left[\bigwedge_{i \in I}\left(X_{i}=0\right)\right] \leq \prod_{i \in I} \operatorname{Pr}\left[X_{i}=0\right]
$$

Theorem 8 (Negative Correlation Chernoff Bound [5,6]). Let $X_{1}, \ldots, X_{n}$ be negatively correlated random variables with $X=\sum_{i=1}^{n} X_{i}, \mu=$ $\mathbb{E}[X]$.

- For any $\delta \in(0,1)$,

$$
\operatorname{Pr}[X \leq(1-\delta) \mu] \leq e^{-\delta^{2} \mu / 2} \text { and } \operatorname{Pr}[X \geq(1+\delta) \mu] \leq e^{-\delta^{2} \mu / 3}
$$

- For any $d \geq 6 \mu, \operatorname{Pr}[X \geq d] \leq 2^{-d}$.

Lemma 6. Let $\pi:[n] \rightarrow[n]$ be a random permutation. For any set $S, W \subseteq[n]$, let $u=\frac{|W|}{n}|S|$. Then the following holds.

- for any constant $\delta \in(0,1)$,

$$
\begin{aligned}
& \operatorname{Pr}[|\pi(S) \cap W| \leq(1-\delta) \mu] \leq e^{-\delta^{2} \mu / 2} \\
& \operatorname{Pr}[|\pi(S) \cap W| \geq(1+\delta) \mu] \leq e^{-\delta^{2} \mu / 3}
\end{aligned}
$$

- for any $d \geq 6 \mu, \operatorname{Pr}[|\pi(S) \cap W| \geq d] \leq 2^{-d}$.

Proof. For every $s \in S$, let $X_{s}$ be the indicator such that $X_{s}=1$ is the event that $\pi(s)$ is in $W$. Let $X=\sum_{s \in S} X_{s}$. So $|\pi(S) \cap W|=X$. Note that $\operatorname{Pr}\left[X_{s}=\right.$ $1]=|W| / n$. So $\mu=\mathbb{E}(X)=\frac{|W|}{n}|S|$.

For any $I \subseteq S$,

$$
\begin{equation*}
\operatorname{Pr}\left[\bigwedge_{i \in I}\left(X_{i}=1\right)\right]=\frac{|W|}{n} \cdot \frac{|W|-1}{n-1} \cdots \frac{|W|-|I|}{n-|I|} \tag{3}
\end{equation*}
$$

(if $|W|<|I|$, it is 0 ). This is because the random permutation can be viewed as throwing elements $1, \ldots, n$ into $n$ boxes uniformly one by one, where every box can have at most one element. We know that for $j=1, \ldots,|I|, \frac{|W|-j}{n-j} \leq \frac{|W|}{n}$ as $|W| \leq n$. So $\operatorname{Pr}\left[\bigwedge_{i \in I}\left(X_{i}=1\right)\right] \leq \prod_{i \in I} \operatorname{Pr}\left[X_{i}=1\right]$. In the same way, for any $I \subseteq[n]$,

$$
\begin{equation*}
\operatorname{Pr}\left[\bigwedge_{i \in I}\left(X_{i}=0\right)\right]=\frac{n-|W|}{n} \cdot \frac{n-|W|-1}{n-1} \cdots \frac{n-|W|-|I|}{n-|I|} \tag{4}
\end{equation*}
$$

(if $n-|W|<|I|$, it is 0 ). Thus $\forall I \subseteq[n], \operatorname{Pr}\left[\bigwedge_{i \in I}\left(X_{i}=0\right)\right] \leq \prod_{i \in I} \operatorname{Pr}\left[X_{i}=0\right]$. By Theorem 8, the conclusion follows.

We can get the following more general result by using Lemma 6 .
Lemma 7. Let $\pi:[n] \rightarrow[n]$ be a random permutation. For any $W \subseteq[n]$ with $|W|=\gamma n$, any constant $\delta \in(0,1)$, any $t, l \in \mathbb{N}^{+}$such that $t l \leq \frac{0.9 \delta}{1+0.9 \delta} \gamma n$, any $\mathcal{S}=\left\{S_{1}, \ldots, S_{l}\right\}$ such that $\forall i \in[l], S_{i} \subseteq[n]$ are disjoint sets and $\left|S_{i}\right|=t$, let $X_{i}$ be the indicator such that $X_{i}=1$ is the event $\left|\pi\left(S_{i}\right) \cap W\right| \geq(1+\delta) \gamma t$. Let $X=\sum_{i \in[l]} X_{i}$. Then for any $d \geq 0$,

$$
\operatorname{Pr}[X \geq d] \leq e^{-2 d+\left(e^{2}-1\right) e^{-\Omega(\gamma t)} l}
$$

Proof. For any $s>0, \operatorname{Pr}[X \geq d]=\operatorname{Pr}\left[e^{s X} \geq e^{s d}\right] \leq \frac{\mathbb{E}\left[e^{s X}\right]}{e^{s d}}$ by Markov's inequality. For every $i \in[l], \forall x_{1}, \ldots, x_{i-1} \in\{0,1\}$, consider $p=\operatorname{Pr}\left[X_{i}=1 \mid \forall j<\right.$ $\left.i, X_{j}=x_{j}\right]$. Let $\bar{S}_{i}=\bigcup_{j=1}^{i} S_{j}$ for $i \in[l]$. Note that the event $\forall j<i, X_{j}=x_{j}$ is the union of exclusive events $\pi\left(\bar{S}_{i-1}\right)=V, \forall j<i, X_{j}=x_{j}$ for $V \subseteq[n]$ with $|V|=(j-1) t$ and $\pi\left(\bar{S}_{i-1}\right)=V$ does not contradict $\forall j<i, X_{j}=x_{j}$. Conditioned on any one of those events, saying $\pi\left(\bar{S}_{i-1}\right)=V, \forall j<i, X_{j}=x_{j}, \pi$ is
a random bijective mapping from $[n]-\bar{S}_{i}$ to $[n]-V$. Note that $\frac{|W \cap([n]-V)|}{n-(i-1) t} \leq$ $\frac{\gamma n}{n-\frac{0.9 \delta}{1+0.9 \delta} \gamma n} \leq \frac{\gamma n}{n-\frac{0.9 \delta}{1+0.9 \delta} n} \leq(1+0.9 \delta) \gamma n$, since $(i-1) t \leq l t \leq \frac{0.9 \delta}{1+0.9 \delta} \gamma n$. So $\mathbb{E}\left[\pi\left(S_{i}\right) \cap W \| \pi\left(\bar{S}_{i-1}\right)=V, \forall j<i, X_{j}=x_{j}\right] \leq(1+0.9 \delta) \gamma t$. By Lemma 6, $\operatorname{Pr}\left[X_{i}=1 \mid \pi\left(\bar{S}_{i-1}\right)=V, \forall j<i, X_{j}=x_{j}\right]=\operatorname{Pr}\left[\left|\pi\left(S_{i}\right) \cap W\right| \geq(1+\delta) \gamma t \mid \pi\left(\bar{S}_{i-1}\right)=\right.$ $\left.V, \forall j<i, X_{j}=x_{j}\right] \leq e^{-\Omega(\gamma t)}$. Thus $p \leq e^{-\Omega(\gamma t)}$. Next note that

$$
\begin{align*}
& \mathbb{E}\left[e^{s \sum_{k=i}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}\right] \\
= & p e^{s} \mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=1\right] \\
& +(1-p) \mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=0\right] \\
\leq & \left(p e^{s}+1-p\right) \max \left(\mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=1\right],\right. \\
& \left.\mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=0\right]\right)  \tag{5}\\
\leq & e^{p\left(e^{s}-1\right)} \max \left(\mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=1\right],\right. \\
& \left.\mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=0\right]\right) \\
\leq & e^{e^{\Omega(\gamma t)}\left(e^{s}-1\right)} \max \left(\mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=1\right],\right. \\
& \left.\mathbb{E}\left[e^{s \sum_{k=i+1}^{l} X_{k}} \mid \forall j<i, X_{j}=x_{j}, X_{i}=0\right]\right) .
\end{align*}
$$

As this holds for every $i \in[l]$ and every $x_{1}, \ldots, x_{i-1} \in\{0,1\}$, we can iteratively apply the inequality and get the result that there exists $x_{1}^{\prime}, \ldots, x_{l}^{\prime} \in\{0,1\}$ such that

$$
\begin{align*}
\mathbb{E}\left[e^{s X}\right] & \leq e^{e^{-\Omega(\gamma t)}\left(e^{s}-1\right)} \mathbb{E}\left[e^{s \sum_{k=2}^{l} X_{k}} \mid X_{1}=x_{1}^{\prime}\right] \\
& \leq e^{2 e^{-\Omega(\gamma t)}\left(e^{s}-1\right)} \mathbb{E}\left[e^{s \sum_{k=3}^{l} X_{k}} \mid X_{1}=x_{1}^{\prime}, X_{2}=x_{2}^{\prime}\right] \\
& \leq \cdots \leq e^{(l-1) e^{-\Omega(\gamma t)}\left(e^{s}-1\right)} \mathbb{E}\left[e^{s X_{l}} \mid X_{1}=x_{1}^{\prime}, X_{2}=x_{2}^{\prime}, \ldots, X_{l-1}=x_{l-1}^{\prime}\right] \\
& \leq e^{e^{-\Omega(\gamma t)}\left(e^{s}-1\right) l} . \tag{6}
\end{align*}
$$

Let's take $s=2$. So $\operatorname{Pr}[X \geq d] \leq \frac{\mathbb{E}\left[e^{s X}\right]}{e^{s d}} \leq e^{-2 d+\left(e^{2}-1\right) e^{-\Omega(\gamma t)} l}$.
Let's first show the non-adaptive privacy of this scheme.
Lemma 8. If the non-adaptive privacy error of (Share, Rec) is $\epsilon$, then the nonadaptive privacy error of (Share' $\left.{ }^{\prime} \operatorname{Rec}^{\prime}\right)$ is $\bar{n}\left(\epsilon+2^{-\Omega(k)}\right)$.

Proof. We show that there exists a distribution $\mathcal{D}$ such that for any string $x \in$ $\Sigma_{0}^{m^{\prime}}$, for any sequence of distinct indices $W=\left(w_{1}, w_{2}, \ldots, w_{k^{\prime}}\right) \in\left[n^{\prime}\right]^{k^{\prime}}$ (chosen before observation),

$$
\operatorname{SD}\left(\operatorname{Share}^{\prime}(x)_{W}, \mathcal{D}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega(k)}\right)
$$

For every $i \in[n \bar{n}]$, the block Share $_{+}\left(\hat{y}_{i}\right)$ has length $t$. Let the indices of shares in Share $_{+}\left(\hat{y}_{i}\right)$ be $S_{i}=\{(i-1) t+1, \ldots, i t\}$.

For every $i \in[\bar{n}]$, let $E_{i}$ be the event that for at most $k$ of $j \in\{(i-1) n+$ $1, \ldots, i n\}, \pi\left(S_{j}\right) \subseteq W$. Let $E=\bigcap_{i \in[\bar{n}]} E_{i}$. We choose $b$ to be such that $t n \leq$ $\frac{0.9 \alpha}{1+0.9 \alpha}|W|$. So by Lemma $7, \operatorname{Pr}\left[E_{i}\right] \geq 1-e^{-\Omega(k)+\left(e^{2}-1\right) e^{-\Omega((1-\alpha) t)} n}$. We choose a large enough $t=O(\log n)$ such that $\operatorname{Pr}\left[E_{i}\right] \geq 1-e^{-\Omega(k)}$. So $\operatorname{Pr}[E] \geq 1-\bar{n} e^{-\Omega(k)}$ by the union bound.

Let's define the distribution $\mathcal{D}$ to be Share' $(\sigma)_{W}$ for some $\sigma \in \Sigma_{0}^{m^{\prime}}$. We claim that Share ${ }^{\prime}(x)_{W} \mid E$ and $\mathcal{D} \mid E$ have statistical distance at most $\bar{n} \epsilon$. The reason is as follows.

Let's fix a permutation $\pi$ for which $E$ happens. We claim that $\operatorname{Share}^{\prime}(x)_{W}$ is a deterministic function of at most $k$ entries of each $y_{i}$ for $i \in[\bar{n}]$ and some extra uniform random bits. This is because, as $E$ happens, for those $i \in[n \bar{n}]$ with $\pi\left(S_{i}\right) \nsubseteq W$, the shares in $\pi\left(S_{i}\right) \cap W$ are independent of the secret by the privacy of $\left(\mathrm{Share}_{+}, \mathrm{Rec}_{+}\right)$. Note that they are also independent of other shares since the construction uses independent randomness for Share ${ }_{+}\left(\hat{y}_{i}\right), i \in[n \bar{n}]$. For those $i \in[n \bar{n}]$ with $\pi\left(S_{i}\right) \subseteq W$, the total number of them is at most $k$. So the claim holds. Hence by the privacy of (Share, Rec) with noting that $y_{i}, i \in[\bar{n}]$ are generated using independent randomness,

$$
\begin{equation*}
\mathrm{SD}\left(\operatorname{Share}^{\prime}(x)_{W}, \mathcal{D}\right) \leq \bar{n} \epsilon \tag{7}
\end{equation*}
$$

So with probability at least $1-\bar{n} e^{-\Omega(k)}$ over the fixing of $\pi$, Share $^{\prime}(x)_{W}$ and $\mathcal{D}$ have statistical distance at most $\bar{n} \epsilon$, which means that

$$
\begin{equation*}
\mathrm{SD}\left(\operatorname{Share}^{\prime}(x)_{W}, \mathcal{D}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega(k)}\right) \tag{8}
\end{equation*}
$$

Next we show the adaptive privacy.
Lemma 9. For any alphabet $\Sigma$, any $n, k \in \mathbb{N}$ with $k \leq n$, for any distribution $X=\left(X_{1}, \ldots, X_{n}\right)$ over $\Sigma^{n}$, let $Y=\left(\left(X_{\pi^{-1}(1)} \circ \pi^{-1}(1)\right), \ldots,\left(X_{\pi^{-1}(n)} \circ \pi^{-1}(n)\right)\right)$ where $\pi$ is a random permutation over $[n] \rightarrow[n]$. For any adaptive observation $W$ with $|W|=k, Y_{W}$ is the same distribution as $Y_{[k]}$.
Proof. Let $W=\left(w_{1}, \ldots, w_{k}\right)$.
We use induction.
For the base step, for any $x \in \Sigma$, any $i \in[n]$,

$$
\begin{equation*}
\operatorname{Pr}\left[Y_{w_{1}}=(x, i)\right]=\operatorname{Pr}\left[X_{i}=x\right] / n \tag{9}
\end{equation*}
$$

while

$$
\begin{equation*}
\operatorname{Pr}\left[Y_{1}=(x, i)\right]=\operatorname{Pr}\left[X_{i}=x\right] / n \tag{10}
\end{equation*}
$$

So $Y_{w_{1}}$ and $Y_{1}$ are the same distributions.
For the inductive step, assume that $Y_{W_{[i]}}$ and $Y_{[i]}$ are the same distributions. We know that for any $u \in(\Sigma \times[n])^{i}$,

$$
\begin{equation*}
\operatorname{Pr}\left[Y_{W_{[i]}}=u\right]=\operatorname{Pr}\left[Y_{[i]}=u\right] . \tag{11}
\end{equation*}
$$

Fix a $u \in(\Sigma \times[n])^{i}$. For any $v=\left(v_{1}, v_{2}\right) \in(\Sigma \times[n])$, where $v_{1} \in \Sigma, v_{2} \in[n]$, $\operatorname{Pr}\left[Y_{w_{i+1}}=v \mid Y_{W_{[i]}}=u\right]=0$ if $v_{2}$ has already been observed in the previous $i$
observations; otherwise $\operatorname{Pr}\left[Y_{w_{i+1}}=v \mid Y_{W_{[i]}}=u\right]=\frac{\operatorname{Pr}\left[X_{v_{2}}=v_{1}\right]}{n-i}$. Also $\operatorname{Pr}\left[Y_{i+1}=\right.$ $\left.v \mid Y_{[i]}=u\right]=0$ if $v_{2}$ has already been observed in the previous $i$ observations; otherwise $\operatorname{Pr}\left[Y_{i+1}=v \mid Y_{[i]}=u\right]=\frac{\operatorname{Pr}\left[X_{v_{2}}=v_{1}\right]}{n-i}$.

Thus $Y_{W_{[i+1]}}$ and $Y_{[i+1]}$ are the same distributions. This finishes the proof.
Lemma 10. If (Share, Rec) has non-adaptive privacy error $\epsilon$, then (Share', Rec') has adaptive privacy error $\bar{n}\left(\epsilon+2^{-\Omega(k)}\right)$.

Proof. First we assume that the adaptive observer always observes $k^{\prime}$ shares. For every observer $M$ which does not observe $k^{\prime}$ shares, there exists another observer $M^{\prime}$ which can observe the same shares as $M$ and then observe some more shares. That is to say that if the number of observed shares is less than $k^{\prime}$, $M^{\prime}$ will choose more unobserved shares (sequentially in a fixed order) to observe until $k^{\prime}$ shares are observed. Since we can use a deterministic function to throw away the extra observes of $M^{\prime}$ to get what $M$ should observe, by Lemma 1 part 2, if the privacy holds for $M^{\prime}$ then the privacy holds for $M$. As a result, we always consider observers which observe $k^{\prime}$ shares.

By Lemma 9, for any $s \in \Sigma_{0}^{m^{\prime}}$, any adaptive observation $W$, $\operatorname{Share}^{\prime}(s)_{W}$ is the same distribution as $\operatorname{Share}^{\prime}(s)_{W^{\prime}}$ where $W=\left\{w_{1}, w_{2}, \ldots, w_{k^{\prime}}\right\}, W^{\prime}=\left[k^{\prime}\right]$. As $W^{\prime}$ is actually a non-adaptive observation, by Lemma 8 , for distinct $s, s^{\prime} \in\{0,1\}^{m^{\prime}}$, SD $\left(\right.$ Share $^{\prime}(s)_{W^{\prime}}$, Share $\left.^{\prime}\left(s^{\prime}\right)_{W^{\prime}}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega(k)}\right)$. So

$$
\begin{equation*}
\operatorname{SD}\left(\operatorname{Share}^{\prime}(s)_{W}, \operatorname{Share}^{\prime}\left(s^{\prime}\right)_{W}\right)=\operatorname{SD}\left(\operatorname{Share}^{\prime}(s)_{W^{\prime}}, \text { Share }^{\prime}\left(s^{\prime}\right)_{W^{\prime}}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega(k)}\right) \tag{12}
\end{equation*}
$$

The theorem below now follows from Construction 1, Lemmas 4, 5 and 10.
Theorem 9. For any $n, m \in \mathbb{N}, m \leq n$, any $\epsilon, \eta \in[0,1]$ and any constant $a \geq 1, \alpha \in(0,1]$, if there exists an explicit $(n, k)$ secret sharing scheme in $\mathrm{AC}^{0}$ with share alphabet $\Sigma$, message alphabet $\Sigma_{0}$, message length $m$, non-adaptive privacy error $\epsilon$ and reconstruction error $\eta$, then there exists an explicit ( $n^{\prime}=$ $\left.O\left(n^{a} \log n\right),(1-\alpha) n^{\prime}\right)$ secret sharing scheme in $\mathrm{AC}^{0}$ with share alphabet $\Sigma \times$ [ $n^{\prime}$ ], message alphabet $\Sigma_{0}$, message length $\Omega\left(m n^{a-1}\right)$, adaptive privacy error $O\left(n^{a-1}\left(\epsilon+2^{-\Omega(k)}\right)\right)$ and reconstruction error $O\left(n^{a-1} \eta\right)$.

### 3.2 Binary Alphabet

In this subsection, we construct $\mathrm{AC}^{0}$ secret sharing schemes with binary alphabet based on some existing schemes with binary alphabets, enlarging the privacy threshold.

We use some coding techniques and secret sharing for small blocks.
Lemma 11 ([12] Sect.4). For any $n \in \mathbb{N}$, any constant $\delta_{0}, \delta_{1} \in(0,1)$, let $C \subseteq \mathbb{F}_{2}^{n}$ be an asymptotically good $\left(n, k=\delta_{0} n, d=\delta_{1} n\right)$ linear code .

1. There exists an $(n, d)$ secret sharing scheme (Share, Rec) with alphabet $\{0,1\}$, message length $k$, perfect privacy and reconstruction. Here $\forall x \in\{0,1\}^{k}$,

Share $(x)=f(x)+c$ with $c$ drawn uniform randomly from $C^{\perp}$ (the dual code of $C$ ) and $f$ is the encoding function from $\{0,1\}^{k}$ to $C$. For $y \in\{0,1\}^{n}$, $\operatorname{Rec}(y)$ is to find $x$ such that there exists a $c \in C^{\perp}$ with $f(x)+c=y$.
2. For any $p=\operatorname{poly}(n)$, there exists an explicit $(n, d)$ secret sharing scheme (Share, Rec) with alphabet $\{0,1\}^{p}$, message length $k$, perfect privacy and reconstruction.
3. If the codeword length is logarithmic (say $n=O(\log N)$ for some $N \in \mathbb{N})$, then both schemes can be constructed explicitly in $\mathrm{AC}^{0}$ (in $N$ ).

Proof. The first assertion is proved in [12].
The second assertion follows by applying the construction of the first assertion in parallel $p$ times.

The third assertion holds because, when the codeword length is $O(\log N)$, both encoding and decoding functions have input length $O(\log N)$. For encoding, we can use any classic methods for generating asymptotically good binary codes. For decoding, we can try all possible messages to uniquely find the correct one. By Lemma 2, both functions can be computed by $\mathrm{AC}^{0}$ circuits.

Now we give the secret sharing scheme in $\mathrm{AC}^{0}$ with a constant privacy rate while having binary alphabet.

Construction 2. For any $n, k, m \in \mathbb{N}$ with $k, m \leq n$, let (Share, Rec) be an $(n, k)$ secret sharing scheme with alphabet $\{0,1\}$, message length $m$.

Let $\left(\mathrm{Share}_{C}, \operatorname{Rec}_{C}\right)$ be an $\left(n_{C}, k_{C}\right)$ secret sharing scheme with alphabet $\{0,1\}^{p}, p=O(\log n)$, message length $m_{C}$ by Lemma 11, where $m_{C}=\delta_{0} n_{C}$, $k_{C}=\delta_{1} n_{C}, n_{C}=O(\log n)$ for some constants $\delta_{0}, \delta_{1}$.

Let $\left(\mathrm{Share}_{0}, \mathrm{Rec}_{0}\right)$ be an $\left(n_{0}, k_{0}\right)$ secret sharing scheme with alphabet $\{0,1\}$, message length $m_{0}$ by Lemma 11, where $m_{0}=\delta_{0} n_{0}=p+O(\log n), k_{0}=\delta_{1} n_{0}$.

For any constant $a \geq 1$, we can construct the following $\left(n^{\prime}=O\left(n^{a}\right), k^{\prime}=\right.$ $\left.\Omega\left(n^{\prime}\right)\right)$ secret sharing scheme (Share' ${ }^{\prime}$ Rec') with alphabet $\{0,1\}$, message length $m^{\prime}=m \bar{n}$, where $\bar{n}=\Theta\left(n^{a-1}\right)$ is large enough.

Function Share ${ }^{\prime}:\{0,1\}^{m^{\prime}} \rightarrow\{0,1\}^{n^{\prime}}$ is as follows.

1. On input $x \in\{0,1\}^{m \bar{n}}$, parse it to be $\left(x_{1}, x_{2}, \ldots, x_{\bar{n}}\right) \in\left(\{0,1\}^{m}\right)^{\bar{n}}$.
2. Compute $y=\left(y_{1}, \ldots, y_{\bar{n}}\right)=\left(\operatorname{Share}\left(x_{1}\right), \ldots\right.$, Share $\left.\left(x_{\bar{n}}\right)\right) \in\left(\{0,1\}^{n}\right)^{\bar{n}}$. Split each entry to be blocks each has length pme to get $\hat{y}=\left(\hat{y}_{1}, \ldots, \hat{y}_{\tilde{n}}\right) \in$ $\left(\{0,1\}^{p m_{C}}\right)^{\tilde{n}}$, where $\tilde{n}=\bar{n}\left\lceil\frac{n}{p m_{C}}\right\rceil$.
3. Let $y^{*}=\left(\operatorname{Share}_{C}\left(\hat{y}_{1}\right), \ldots, \operatorname{Share}_{C}\left(\hat{y}_{\tilde{n}}\right)\right)$. Parse it to be $y^{*}=\left(y_{1}^{*}, \ldots, y_{n^{*}}^{*}\right) \in$ $\left(\{0,1\}^{p}\right)^{n^{*}}, n^{*}=\tilde{n} n_{C}$.
4. Generate $\pi$ by Lemma 3 which is uniform random over permutations of $\left[n^{*}\right]$. If it failed, which can be detected by checking element distinctness, set $\pi$ to be such that $\forall i \in\left[n^{*}\right], \pi(i)=i$.
5. Compute

$$
\begin{aligned}
z(x) & =\left(\operatorname{Share}_{0}\left(y_{\pi^{-1}(1)}^{*} \circ \pi^{-1}(1)\right), \ldots, \operatorname{Share}_{0}\left(y_{\pi^{-1}\left(n^{*}\right)}^{*} \circ \pi^{-1}\left(n^{*}\right)\right)\right) \\
& \in\left(\{0,1\}^{n_{0}}\right)^{n^{*}}
\end{aligned}
$$

6. Parse $z(x)$ to be bits and output.

Function $\operatorname{Rec}^{\prime}:\{0,1\}^{n^{\prime}=n_{0} n^{*}} \rightarrow\{0,1\}^{m^{\prime}}$ is as follows.

1. Parse the input bits to be $z \in\left(\{0,1\}^{n_{0}}\right)^{n^{*}}$ and compute

$$
\left(y_{\pi^{-1}(1)}^{*} \circ \pi^{-1}(1), \ldots, y_{\pi^{-1}\left(n^{*}\right)}^{*} \circ \pi^{-1}\left(n^{*}\right)\right)=\left(\operatorname{Rec}_{0}\left(z_{1}\right), \ldots, \operatorname{Rec}_{0}\left(z_{n^{*}}\right)\right)
$$

2. Compute $y^{*}=\left(y_{1}^{*}, \ldots, y_{n^{*}}^{*}\right)$.
3. Compute $\hat{y}$ by applying $\operatorname{Rec}_{C}$ on $y^{*}$ for every successive $n_{C}$ entries.
4. Parse $\hat{y}$ to be $y$.
5. Compute $x$ by applying Rec on every entry of $y$.

We have the following two lemmas, whose proofs are deferred to the full version.
Lemma 12. If Share and Rec can be computed by $\mathrm{AC}^{0}$ circuits, then Share' and $\mathrm{Rec}^{\prime}$ can be computed by $\mathrm{AC}^{0}$ circuits.

Lemma 13. If the reconstruction error of (Share, Rec) is $\eta$, then the reconstruction error of (Share ${ }^{\prime}$, $\mathrm{Rec}^{\prime}$ ) is $\eta^{\prime}=\bar{n} \eta$.

Lemma 14. If the non-adaptive privacy error of (Share, Rec) is $\epsilon$, then the nonadaptive privacy error of (Share ${ }^{\prime}$, $\left.\mathrm{Rec}^{\prime}\right)$ is $\bar{n}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right)$.

Proof. Let $k^{\prime}=0.9 \delta_{1}^{2} n^{\prime}$. We show that there exists a distribution $\mathcal{D}$ such that for any string $x \in\{0,1\}^{m}$, for any $W \subseteq\left[n^{\prime}\right]$ with $|W| \leq k^{\prime}$,

$$
\begin{equation*}
\mathrm{SD}\left(\operatorname{Share}^{\prime}(x)_{W}, \mathcal{D}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right) \tag{13}
\end{equation*}
$$

Let $\mathcal{D}$ be Share $^{\prime}(\sigma)_{W}$ for some $\sigma \in\{0,1\}^{m^{\prime}}$.
Consider an arbitrary observation $W \subseteq\left[n^{\prime}\right]$, with $|W| \leq k^{\prime}$. Note that for at least $1-0.9 \delta_{1}$ fraction of all blocks $z_{i} \in\{0,1\}^{n_{0}}, i=1, \ldots, n^{*}$, at most $\delta_{1}$ fraction of the bits in the block can be observed. Otherwise the number of observed bits is more than $0.9 \delta_{1} \times \delta_{1} n^{\prime}$. Let $W^{*}$ be the index set of those blocks which have more than $\delta_{1}$ fraction of bits being observed.

For every $i \in\left[n^{*}\right] \backslash W^{*}, z_{i}$ is independent of $y_{\pi^{-1}(i)}^{*} \circ \pi^{-1}(i)$ by the privacy of $\left(\right.$ Share $\left._{0}, \operatorname{Rec}_{0}\right)$. Note that $z_{i}$ is also independent of $z_{i^{\prime}}, i^{\prime} \in\left[n^{*}\right], i^{\prime} \neq i$ since it is independent of $y_{\pi^{-1}(i)}^{*} \circ \pi^{-1}(i)$ (its randomness is only from the randomness of the Share ${ }_{0}$ function) and every Share $_{0}$ function uses independent randomness. So we only have to show that

$$
\begin{equation*}
\mathrm{SD}\left(z_{W^{*}}(x), z_{W^{*}}(\sigma)\right) \leq \bar{n}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right) \tag{14}
\end{equation*}
$$

For every $i \in[\tilde{n}]$, let $S_{i}=\left\{(i-1) n_{C}+1, \ldots, i n_{C}\right\}$. Let $X_{i}$ be the indicator that $\left|\pi\left(S_{i}\right) \cap W^{*}\right|>k_{C}, i \in[\tilde{n}]$. Note that $\mathbb{E}\left[\left|\pi\left(S_{i}\right) \cap W^{*}\right|\right] \leq 0.9 \delta_{1} n_{C}=0.9 k_{C}$.

For every $i \in[\bar{n}]$, let $E_{i}$ be the event that $\sum_{j=(i-1)\left\lceil\frac{n}{p m_{C}}\right\rceil+1}^{i\left\lceil\frac{n}{p m_{C}}\right\rceil} X_{j} \leq \frac{k}{p m_{C}}$. Let $E=\bigcap_{i \in[\bar{n}]} E_{i}$. We take $\bar{n}$ to be large enough such that $n_{C}\left\lceil\frac{n}{p m_{C}}\right\rceil \leq$ $\frac{0.9 \times 0.1}{1+0.9 \times 0.1}\left|W^{*}\right|$. For every $i \in[\bar{n}]$, by Lemma 7 ,

$$
\begin{equation*}
1-\operatorname{Pr}\left[E_{i}\right] \leq e^{-2 k /\left(p m_{C}\right)+\left(e^{2}-1\right) e^{-\Omega\left(0.9 \delta_{1}^{2} n_{C}\right)}\left\lceil\frac{n}{p m_{C}}\right\rceil} \tag{15}
\end{equation*}
$$

We take $n_{C}=O(\log n)$ to be large enough such that the probability is at most

$$
\begin{equation*}
e^{-\Omega\left(k /\left(p m_{C}\right)\right)} \leq e^{-\Omega\left(k / \log ^{2} n\right)} \tag{16}
\end{equation*}
$$

Next we do a similar argument as that in the proof of Lemma 8. We know that $\operatorname{Pr}[E] \geq 1-\bar{n} e^{-\Omega\left(k / \log ^{2} n\right)}$. We claim that $z_{W^{*}}(x) \mid E$ and $z_{W^{*}}(\sigma) \mid E$ have statistical distance at most $\bar{n} \epsilon$. The reason follows.

Let's fix a permutation $\pi$ for which $E$ happens. We claim that $z_{W^{*}}(x)$ is a deterministic function of at most $k$ bits of each $y_{i}$ for $i \in[\bar{n}]$ and some extra uniform random bits. This is because, as $E$ happens, for those $i \in[\tilde{n}]$ with $\left|\pi\left(S_{i}\right) \cap W^{*}\right| \leq k_{C}$, the shares in $\pi\left(S_{i}\right) \cap W^{*}$ are independent of the secret by the privacy of $\left(S_{h a r e}^{C}, \operatorname{Rec}_{C}\right)$. Note that they are also independent of other shares since the construction uses independent randomness for Share $_{C}\left(\hat{y}_{i}\right), i \in[\tilde{n}]$. For those $i \in[\tilde{n}]$ with $\left|\pi\left(S_{i}\right) \cap W^{*}\right|>k_{C}$, the total number of them is at most $\frac{k}{p m_{C}}$. By the construction, Share ${ }^{\prime}(x)_{W^{*}}$ is computed from at most $\frac{k}{p m_{C}} \times p m_{C}=k$ bits of each $y_{i}$ for $i \in[\bar{n}]$ and some extra uniform random bits. Hence by the privacy of (Share, Rec) and noting that $y_{i}, \in[\bar{n}]$ are generated using independent randomness,

$$
\begin{equation*}
\mathrm{SD}\left(z_{W^{*}}(x), z_{W^{*}}(\sigma)\right) \leq \bar{n} \epsilon \tag{17}
\end{equation*}
$$

Thus with probability at least $1-\bar{n} e^{-\Omega\left(k / \log ^{2} n\right)}$ over the fixing of $\pi, z_{W^{*}}(x)$ and $z_{W^{*}}(\sigma)$ have statistical distance at most $\bar{n} \epsilon$, which means that

$$
\begin{equation*}
\mathrm{SD}\left(z_{W^{*}}(x), z_{W^{*}}(\sigma)\right) \leq \bar{n}\left(\epsilon+e^{-\Omega\left(k / \log ^{2} n\right)}\right) \tag{18}
\end{equation*}
$$

Lemma 15. For any alphabet $\Sigma$, any $n \in \mathbb{N}$, Let $X=\left(X_{1}, \ldots, X_{n}\right)$ be an arbitrary distribution over $\Sigma^{n}$. For any $n_{0}, k_{0} \in \mathbb{N}$ with $k_{0} \leq n_{0}$, let $\left(\right.$ Share $\left._{0}, \operatorname{Rec}_{0}\right)$ be an arbitrary $\left(n_{0}, k_{0}\right)$-secret sharing scheme with binary alphabet, message length $m_{0}=\log |\Sigma|+O(\log n)$, perfect privacy. Let

$$
Y=\left(\operatorname{Share}_{0}\left(X_{\pi^{-1}(1)} \circ \pi^{-1}(1)\right), \ldots, \operatorname{Share}_{0}\left(X_{\pi^{-1}(n)} \circ \pi^{-1}(n)\right)\right)
$$

where $\pi$ is a random permutation over $[n] \rightarrow[n]$. For any $t \leq n \cdot k_{0}$, let $W$ be an any adaptive observation which observes $t$ shares. Then there exists a deterministic function $f:\{0,1\}^{\mathrm{poly}(n)} \rightarrow\{0,1\}^{t}$ such that $Y_{W}$ has the same distribution as $f\left(Y_{W^{\prime}} \circ S\right)$, where $S$ is uniform over $\{0,1\}^{\text {poly }(n)}$ and $W^{\prime}=\left[t^{\prime} n_{0}\right], t^{\prime}=\left\lceil\frac{t}{k_{0}}\right\rceil$.

Proof. For every $i \in[n]$, Let $B_{i}=\left\{(i-1) n_{0}+1, \ldots, i n_{0}\right\}$. Assume the adaptive adversary is $M$.

Let $f$ be defined as in Algorithm 1 .
Let $W=\left(w_{1}, \ldots, w_{t}\right) \in\left[n \cdot n_{0}\right]^{t}, Z=f\left(Y_{W^{\prime}} \circ S\right)$. Let $R \in\{0,1\}^{n n_{0}}$ be the random variable corresponds to $r$.

We use induction to show that $Y_{W}$ has the same distribution as $Z$.
For the base case, the first bits of both random variables have the same distributions by the perfect privacy of $\left(\mathrm{Share}_{0}, \mathrm{Rec}_{0}\right)$.

For the inductive step, assume that, projected on the first $d$ bits, the two distributions are the same. Fix the first $d$ observed bits for both $Y_{W}$ and $Z$ to

```
Algorithm 1. \(f(\cdot)\)
    Input: \(y \in\{0,1\}^{t^{\prime} n_{0}}, s \in\{0,1\}^{\text {poly }(n)}\)
    Let \(c=1\);
    \(\forall i \in[n], l_{i} \in[n] \cup\{\) null \(\}\) is assigned to be null;
    Compute the secrets for the \(t^{\prime}\) blocks \(y\), which are
        \(\left(x_{1}, \ldots, x_{t^{\prime}}\right) \in\left(\{0,1\}^{m_{0}}\right)^{t^{\prime}} ;\)
    Compute \(\left(\operatorname{Share}_{0}(\sigma), \ldots, \operatorname{Share}_{0}(\sigma)\right) \in\left(\{0,1\}^{n_{0}}\right)^{n}\) and parse it to be
    \(r \in\{0,1\}^{n_{0} n}\), for an arbitrary \(\sigma \in \Sigma\). Here for each Share \({ }_{0}\) function, we take
    some unused bits from \(s\) as the random bits used in that function.
    Next \(f\) does the following computation by calling \(M\);
    while \(M\) wants to observe the ith bit which is not observed previously do
        Find \(j \in[n]\) such that \(i \in B_{j}\);
        if the number of observed bits in the \(j\) th block is less than \(k_{0}\) then
            Let \(M\) observe \(r_{i}\);
        else
            Let \(I_{j}\) be the indices of the observed bits in the \(j\) th block. (The indices
            here are the relative indices in the \(j\) th block)
            if \(l_{j}=\) null then
                    \(l_{j}=c ;\)
                    \(c=c+1 ;\)
                            Draw a string \(v^{j}\) from \(\operatorname{Share}_{0}\left(x_{c}\right) \mid \operatorname{Share}_{0}\left(x_{c}\right)_{I_{j}}=r_{(j-1) n_{0}+I_{j}}\) by using
                    some unused bits of \(s\);
            end
            Let \(M\) observe \(v_{i-(j-1) n_{0}}^{j}\);
        end
    end
```

be $\bar{y} \in\{0,1\}^{d}$. Assume that the $(d+1)$ th observation is to observe the $w_{d}$ th bit where $w_{d}$ is in $B_{j}$ for some $j$.

If the number of observed bits in the $j$ th block is less than $k_{0}$ then $Y_{\left\{w_{1}, \ldots, w_{d+1}\right\} \cap B_{j}}$ has the same distribution as $R_{\left\{w_{1}, \ldots, w_{d+1}\right\} \cap B_{j}}$, following the privacy of $\left(\right.$ Share $\left._{0}, \operatorname{Rec}_{0}\right)$. Note that the blocks $Y_{\left\{w_{1}, \ldots, w_{d+1}\right\} \cap B_{i}}, i \in[n]$ are independent. The blocks $R_{\left\{w_{1}, \ldots, w_{d+1}\right\} \cap B_{i}}, i \in[n]$ are also independent. As $f$ will output $R_{w_{d+1}}$, the conclusion holds for $d+1$.

Else, if the number of observed bits in the $j$ th block is at least $k_{0}$, it is sufficient to show that $\left.Y_{w_{d+1}}\right|_{Y_{\left\{w_{1}, \ldots, w_{d}\right\}}=\bar{y}}$ has the same distribution as that of $\left.Z_{d+1}\right|_{\{1, \ldots, d\}}=\bar{y}$. Note that there are $c$ blocks such that $W$ observes more than $k_{0}$ bits for each of them. Let $q_{1}, \ldots, q_{c}$ denote those blocks. Let $I=\left(\left(q_{1}-1\right) n_{0}+\right.$ $\left.I_{q_{1}}, \ldots,\left(q_{c}-1\right) n_{0}+I_{q_{c}}\right)$, which is the set of indices of all observed bits. Note that $I \subseteq\left\{w_{1}, \ldots, w_{d}\right\}$.

By the privacy of the secret sharing scheme, for those blocks which have at most $k_{0}$ bits being observed, they are independent of the secret and hence independent of other blocks. So $\left.Y_{w_{d+1}}\right|_{Y_{\left\{w_{1}, \ldots, w_{d}\right\}}=\bar{y}}$ is in fact $\left.Y_{w_{d+1}}\right|_{Y_{I}=y^{*}}$ where $y^{*}$ are the corresponding bits from $\bar{y}$ with a proper rearrangement according to
$I$. From the definition of $f$ we know that for $i \in[c]$, the observed bits in the $q_{i}$ th block is exactly the same distribution as $\left(Y_{B_{l_{q_{i}}}}\right)_{I_{q_{i}}}=\operatorname{Share}_{0}\left(x_{l_{q_{i}}}\right)_{I_{q_{i}}}$. So for $\left.Z_{d+1}\right|_{\{1, \ldots, d\}}=\bar{y}$, it is the same distribution as

$$
\begin{align*}
T & =\left.\left(Y_{B_{l_{j}}}\right)_{w_{d}-(j-1) n_{0}}\right|_{i=1} ^{c}\left(\left(Y_{B_{l_{q_{i}}}}\right)_{I_{q_{i}}}=y_{\left(q_{i}-1\right) n_{0}+I_{q_{i}}}^{*}\right. \\
& =\left.\operatorname{Share}_{0}\left(x_{l_{j}}\right)_{w_{d}-(j-1) n_{0}}\right|_{\bigwedge_{i=1}^{c}} ^{c}\left(\operatorname{Share}_{0}\left(x_{l_{q_{i}}}\right)_{I_{q_{i}}}=y_{\left(q_{i}-1\right) n_{0}+I_{q_{i}}}^{*}\right) \tag{19}
\end{align*}
$$

By Lemma $9,\left(Y_{B_{q_{1}}}, \ldots, Y_{B_{q_{c}}}\right)$ has the same distribution as $\left(Y_{B_{l_{q_{1}}}}, \ldots, Y_{B_{l_{q_{c}}}}\right)$ as they both are the same distribution as $\left(\operatorname{Share}_{0}\left(x_{1}\right), \ldots, \operatorname{Share}_{0}\left(x_{c}\right)\right)$. Thus $\left.Y_{w_{d+1}}\right|_{Y_{I}=y^{*}}$ has the same distribution as $T$, as $\left.Y_{w_{d+1}}\right|_{Y_{I}=y^{*}}$ is the distribution of some bits in $\left(Y_{B_{q_{1}}}, \ldots, Y_{B_{q_{c}}}\right)$ and $T$ is the distribution of the corresponding bits (same indices) in $\left(Y_{B_{l_{q_{1}}}}, \ldots, Y_{B_{l_{q_{c}}}}\right)$. So we know that $\left.Y_{w_{d+1}}\right|_{Y_{\left\{w_{1}, \ldots, w_{d}\right\}}=\bar{y}}$ has the same distribution as $\left.Z_{d+1}\right|_{\{1, \ldots, d\}}=\bar{y}$ and this shows our conclusion.

Lemma 16. If the non-adaptive privacy error of (Share, Rec) is $\epsilon$, then the adaptive privacy error of $\left(\mathrm{Share}^{\prime}, \operatorname{Rec}^{\prime}\right)$ is $\bar{n}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right)$.

Proof. Let $W$ be an adaptive observation. Let $W^{\prime}=\left[\left\lceil|W| / k_{0}\right\rceil n_{0}\right]$. Let $|W|=$ $\Omega\left(n^{\prime}\right)$ be small enough such that $\left|W^{\prime}\right| \leq 0.9 \delta_{1}^{2} n^{\prime}$. By Lemma 15 , there exists a deterministic function $f$ such that for any $x, x^{\prime} \in\{0,1\}^{m^{\prime}}$,

$$
\begin{equation*}
\mathrm{SD}\left(\operatorname{Share}^{\prime}(x)_{W}, \operatorname{Share}\left(x^{\prime}\right)_{W}\right)=\operatorname{SD}\left(f\left(\operatorname{Share}^{\prime}(x)_{W^{\prime}} \circ S\right), f\left(\operatorname{Share}^{\prime}\left(x^{\prime}\right)_{W^{\prime}} \circ S\right)\right) \tag{20}
\end{equation*}
$$

where $S$ is the uniform distribution as defined in Lemma 15 which is independent of Share ${ }^{\prime}(x)_{W^{\prime}}$ or Share ${ }^{\prime}\left(x^{\prime}\right)_{W^{\prime}}$. By Lemma 1, we know that

$$
\begin{equation*}
\operatorname{SD}\left(f\left(\operatorname{Share}^{\prime}(x)_{W^{\prime}} \circ S\right), f\left(\operatorname{Share}^{\prime}\left(x^{\prime}\right)_{W^{\prime}} \circ S\right)\right) \leq \operatorname{SD}\left(\text { Share }^{\prime}(x)_{W^{\prime}}, \text { Share }^{\prime}\left(x^{\prime}\right)_{W^{\prime}}\right) \tag{21}
\end{equation*}
$$

By Lemma 14 we know that

$$
\begin{equation*}
\mathrm{SD}\left(\text { Share }^{\prime}(x)_{W^{\prime}}, \text { Share }^{\prime}\left(x^{\prime}\right)_{W^{\prime}}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right) \tag{22}
\end{equation*}
$$

Hence

$$
\begin{equation*}
\mathrm{SD}\left(\text { Share }^{\prime}(x)_{W}, \text { Share }^{\prime}\left(x^{\prime}\right)_{W}\right) \leq \bar{n}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right) \tag{23}
\end{equation*}
$$

Theorem 10. For any $n, m \in \mathbb{N}, m \leq n$, any $\epsilon, \eta \in[0,1]$ and any constant $a \geq 1$, if there exists an explicit $(n, k)$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, message length $m$, non-adaptive privacy error $\epsilon$ and reconstruction error $\eta$, then there exists an explicit $\left(n^{\prime}=O\left(n^{a}\right), k^{\prime}=\Omega\left(n^{\prime}\right)\right)$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, message length $\Omega\left(m n^{a-1}\right)$, adaptive privacy error $O\left(n^{a-1}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right)\right)$ and reconstruction error $O\left(n^{a-1} \eta\right)$.

Proof. It follows from Construction 2, Lemmas 12, 13 and 16.

## 4 k-Wise Independent Generator in $\mathrm{AC}^{0}$

In this section we focus on increasing the secret length to be linear of the number of shares while keeping the construction in $\mathrm{AC}^{0}$. The privacy rate is not as good as the previous section. The main technique is to use the following well known $k$-wise independent generator which is constructed from expander graphs.

Theorem 11 [29]. For any $N, D, M \in \mathbb{N}$, any $\epsilon>0$, if there exists a $\left(\leq K_{\max },\left(\frac{1}{2}+\epsilon\right) D\right)$ expander with left set of vertices $[N]$, right set of vertices $[M]$, left degree $D$, then the function $g:\{0,1\}^{M} \rightarrow\{0,1\}^{N}$, defined by $g(x)_{i}=\bigoplus_{j \in[D]} x_{\Gamma(i, j)}, i=1,2, \ldots, N$, is a $K_{\max }$-wise uniform independent generator.

Next we directly give the main results for this section. Detailed proofs are deferred to the full version.

Theorem 12. For any $M \in \mathbb{N}, N=\operatorname{poly}(M)$, any alphabets $\Sigma_{0}, \Sigma$, any constant $\gamma \in(0,1]$, there exists an explicit $K$-wise independent generator $g: \Sigma_{0}^{M} \rightarrow$ $\Sigma^{N}$ in $\mathrm{AC}^{0}$, where $K=\left(\frac{M \log \left|\Sigma_{0}\right|}{\log |\Sigma|}\right)^{1-\gamma}$.

Now we give the construction of secret sharing schemes in $\mathrm{AC}^{0}$ with large message rate (saying $1-1 / \operatorname{poly}(n))$.

Construction 3. For any $n, k, m \in \mathbb{N}$ with $k \leq n$, any alphabets $\Sigma_{0}, \Sigma$, let (Share, Rec) be an ( $n, k$ ) secret sharing scheme with share alphabet $\Sigma$, message alphabet $\Sigma_{0}$, message length $m$.

For any constant $a>1, \gamma \in(0,1]$, we construct the following $\left(n^{\prime}=n+\right.$ $\left.m^{\prime}, k^{\prime}=\min (k, l)\right)$ secret sharing scheme $\left(\right.$ Share $\left.^{\prime}, \operatorname{Rec}^{\prime}\right)$ with alphabet $\Sigma$, message length $m^{\prime}=\Omega\left(n^{a}\right)$, where $l=\Theta\left(\frac{m \log \left|\Sigma_{0}\right|}{\log |\Sigma|}\right)^{1-\gamma}$.

The function Share' : $\Sigma^{m^{\prime}} \rightarrow \Sigma^{n^{\prime}}$ is as follows.

1. Let $g_{\Gamma}: \Sigma_{0}^{m} \rightarrow \Sigma^{m^{\prime}}$ be the l-wise independent generator by Theorem 12.
2. For secret $x \in \Sigma^{m^{\prime}}$, we draw $r$ uniform randomly from $\Sigma_{0}^{m}$ let

$$
\operatorname{Share}^{\prime}(x)=\left(\operatorname{Share}(r), g_{\Gamma}(r) \oplus x\right)
$$

The function $\operatorname{Rec}^{\prime}: \Sigma^{n^{\prime}} \rightarrow \Sigma^{m^{\prime}}$ is as follows.

1. The input is $y=\left(y_{1}, y_{2}\right)$ where $y_{1} \in \Sigma^{n}, y_{2} \in \Sigma^{m^{\prime}}$.
2. Let

$$
\operatorname{Rec}^{\prime}(y)=g_{\Gamma}\left(\operatorname{Rec}\left(y_{1}\right)\right) \oplus y_{2}
$$

Theorem 13. For any $n, m \in \mathbb{N}, m \leq n$, any $\epsilon, \eta \in[0,1]$, any constant $\gamma \in$ $(0,1]$, any $m^{\prime}=\operatorname{poly}(n)$ and any alphabets $\Sigma_{0}, \Sigma$, if there exists an explicit $(n, k)$ secret sharing scheme in $\mathrm{AC}^{0}$ with share alphabet $\Sigma$, message alphabet $\Sigma_{0}$, message length $m$, non-adaptive privacy error $\epsilon$ and reconstruction error $\eta$, then there exists an explicit $\left(n+m^{\prime}, \min \left(k,\left(\frac{m \log \left|\Sigma_{0}\right|}{\log |\Sigma|}\right)^{1-\gamma}\right)\right)$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\Sigma$, message length $m^{\prime}$, non-adaptive privacy error $\epsilon$ and reconstruction error $\eta$.

## 5 Final Construction

In this section we give our final $\mathrm{AC}^{0}$ construction of secret sharing schemes which has constant message rate and constant privacy rate.

Our construction will use both random permutation and k-wise independent generator proposed in the previous sections.

Construction 4. For any $n, k, m \in \mathbb{N}$ with $k, m \leq n$, let (Share, Rec) be an $(n, k)$ secret sharing scheme with alphabet $\{0,1\}$, message length $m$.

Let $\left(\mathrm{Share}_{C}, \operatorname{Rec}_{C}\right)$ be an $\left(n_{C}, k_{C}\right)$ secret sharing scheme from Lemma 11 with alphabet $\{0,1\}^{p=O(\log n)}$, message length $m_{C}$, where $m_{C}=\delta_{0} n_{C}, k_{C}=\delta_{1} n_{C}$, $n_{C}=O(\log n)$ for some constants $\delta_{0}, \delta_{1}$.

Let $\left(\operatorname{Share}_{C}^{*}, \operatorname{Rec}_{C}^{*}\right)$ be an $\left(n_{C}^{*}, k_{C}^{*}\right)$ secret sharing scheme from Lemma 11 with alphabet $\{0,1\}$, message length large enough $m_{C}^{*}$, where $m_{C}^{*}=\delta_{0} n_{C}^{*}=p+$ $O(\log n), n_{C}^{*}=\delta_{1} n_{C}^{*}$.

For any constant $a>1, \gamma>0$, we can construct the following $\left(n^{\prime}=\right.$ $O\left(n^{a}\right), k^{\prime}=\Omega\left(n^{\prime}\right)$ secret sharing scheme (Share' $\left.{ }^{\prime} \operatorname{Rec}^{\prime}\right)$ with alphabet $\{0,1\}$, message length $m^{\prime}=\Omega\left(n^{\prime}\right)$.

The function Share $:\{0,1\}^{m^{\prime}} \rightarrow\{0,1\}^{n^{\prime}}$ is as follows.

1. Let $\bar{n}=\Theta\left(n^{a-1}\right)$ where the constant factor is large enough.
2. Let $g_{\Gamma}:\{0,1\}^{m \bar{n}} \rightarrow\{0,1\}^{m^{\prime}}$ be the $l$-wise independent generator by Theorem 12, where $l=\Omega\left(m n^{a-1}\right)^{1-\gamma}$.
3. For secret $x \in\{0,1\}^{m^{\prime}}$, we draw a string $r=\left(r_{1}, \ldots, r_{\bar{n}}\right)$ uniform randomly from $\left(\{0,1\}^{m}\right)^{\bar{n}}$.
4. Let $y=\left(y_{s}, y_{g}\right)$, where

$$
\begin{gather*}
y_{s}=\left(y_{s, 1}, \ldots, y_{s, \bar{n}}\right)=\left(\operatorname{Share}\left(r_{1}\right), \ldots, \operatorname{Share}\left(r_{\bar{n}}\right)\right) \in\left(\{0,1\}^{n}\right)^{\bar{n}},  \tag{24}\\
y_{g}=\left(y_{g, 1}, \ldots, y_{g, m^{\prime}}\right)=g_{\Gamma}(r) \oplus x \in\{0,1\}^{m^{\prime}} . \tag{25}
\end{gather*}
$$

5. Compute $\hat{y}_{s} \in\left(\left(\{0,1\}^{p}\right)^{m_{C}}\right)^{n_{s}}$ from $y_{s}$ by parsing $y_{s, i}$ to be blocks over $\left(\{0,1\}^{p}\right)^{m_{C}}$ for every $i \in[\bar{n}]$, where $n_{s}=\left\lceil\frac{n}{p m_{C}}\right\rceil \bar{n}$.
6. Compute $\hat{y}_{g} \in\left(\left(\{0,1\}^{p}\right)^{m_{C}}\right)^{n_{g}}$ from $y_{g}$ by parsing $y_{g}$ to be blocks over $\left(\{0,1\}^{p}\right)^{m_{C}}$, where $n_{g}=\left\lceil\frac{m^{\prime}}{p m_{C}}\right\rceil$.
7. Let

$$
y^{\prime}=\left(\operatorname{Share}_{C}\left(\hat{y}_{s, 1}\right), \ldots, \operatorname{Share}_{C}\left(\hat{y}_{s, n_{s}}\right), \operatorname{Share}_{C}\left(\hat{y}_{g, 1}\right), \ldots, \operatorname{Share}_{C}\left(\hat{y}_{g, n_{g}}\right)\right) .
$$

Parse $y^{\prime}$ as $\left(y_{1}^{\prime}, \ldots, y_{n^{*}}^{\prime}\right) \in\left(\{0,1\}^{p}\right)^{n^{*}}$, where $n^{*}=\left(n_{s}+n_{g}\right) n_{C}$.
8. Generate a random permutation $\pi:\left[n^{*}\right] \rightarrow\left[n^{*}\right]$ and compute

$$
\begin{aligned}
& z(x)=\left(\operatorname{Share}_{C}^{*}\left(y_{\pi^{-1}(1)}^{\prime} \circ \pi^{-1}(1)\right), \ldots, \text { Share }_{C}^{*}\left(y_{\pi^{-1}\left(n^{*}\right)}^{\prime} \circ \pi^{-1}\left(n^{*}\right)\right)\right) \\
& \quad \in\left(\{0,1\}^{n_{C}^{*}}\right)^{n^{*}} .
\end{aligned}
$$

9. Parse $z(x)$ to be bits and output.

The function $\operatorname{Rec}^{\prime}:\{0,1\}^{n^{\prime}} \rightarrow\{0,1\}^{m^{\prime}}$ is as follows.

1. Parse the input bits to be $z=\left(z_{1}, \ldots, z_{n^{*}}\right) \in\left(\{0,1\}^{n_{C}^{*}}\right)^{n^{*}}$.
2. For every $i \in\left[n^{*}\right]$, let $\left(y_{\pi^{-1}(i)}^{\prime} \circ \pi^{-1}(i)\right)=\operatorname{Rec}_{C}^{*}\left(z_{i}\right)$ to get $y^{\prime}$.
3. Parse $y^{\prime}=\left(y_{s}^{\prime}, y_{g}^{\prime}\right)$, where $y_{s}^{\prime}=\left(y_{s, 1}^{\prime}, \ldots, y_{s, n_{s}}^{\prime}\right) \in\left(\{0,1\}^{p_{C}}\right)^{n_{s}}$ and $y_{g}^{\prime}=$ $\left(y_{g, 1}^{\prime}, \ldots, y_{g, n_{g}}^{\prime}\right) \in\left(\{0,1\}^{p n_{C}}\right)^{n_{g}}$.
4. Let

$$
\hat{y}_{s}=\left(\operatorname{Rec}_{C}\left(y_{s, 1}^{\prime}\right), \ldots, \operatorname{Rec}_{C}\left(y_{s, n_{s}}^{\prime}\right)\right), \hat{y}_{g}=\left(\operatorname{Rec}_{C}\left(y_{g, 1}^{\prime}\right), \ldots, \operatorname{Rec}_{C}\left(y_{g, n_{g}}^{\prime}\right)\right) .
$$

5. Parse $\hat{y}_{s}$ to get $y_{s}$.
6. Parse $\hat{y}_{g}$ to get $y_{g}$
7. Let $r=\left(\operatorname{Rec}\left(y_{s, 1}\right), \ldots, \operatorname{Rec}\left(y_{s, n}\right)\right)$.
8. Output

$$
\operatorname{Rec}^{\prime}(z)=g_{\Gamma}(r) \oplus y_{g} .
$$

We have the following lemmas, whose proofs are similar to previous ones and deferred to the full version.

Lemma 17. If (Share, Rec) can be computed by $\mathrm{AC}^{0}$ circuits, then (Share', Rec') can be computed by $\mathrm{AC}^{0}$ circuits.

Lemma 18. If the reconstruction error of (Share, Rec) is $\eta$, then the reconstruction error of (Share', Rec') is $\eta^{\prime}=\bar{n} \eta$.

Lemma 19. If the non-adaptive privacy error of (Share, Rec) is $\epsilon$, then the nonadaptive privacy error of (Share', $\left.\operatorname{Rec}^{\prime}\right)$ is $\bar{n}\left(\epsilon+e^{-\Omega\left(k / \log ^{2} n\right)}\right)+e^{-\Omega\left(l / \log ^{2} n\right)}$.

Lemma 20. If the non-adaptive privacy error of (Share, Rec) is $\epsilon$, then the adaptive privacy error of (Share' ${ }^{\prime} \operatorname{Rec}^{\prime}$ ) is $\bar{n}\left(\epsilon+e^{-\Omega\left(k / \log ^{2} n\right)}\right)+e^{-\Omega\left(l / \log ^{2} n\right)}$.

Theorem 14. For any $\epsilon, \eta \in[0,1]$, any $n, m \in \mathbb{N}, m \leq n$ and any constant $a>1, \gamma>0$, if there exists an explicit $(n, k)$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, message length $m$, non-adaptive privacy error $\epsilon$ and reconstruction error $\eta$, then there exists an explicit $\left(n^{\prime}=O\left(n^{a}\right), \Omega\left(n^{\prime}\right)\right)$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, message length $\Omega\left(n^{\prime}\right)$, adaptive privacy error $O\left(n^{a-1}\left(\epsilon+2^{-\Omega\left(k / \log ^{2} n\right)}\right)+2^{-\Omega\left(\left(m n^{a-1}\right)^{1-\gamma} / \log ^{2} n\right)}\right)$ and reconstruction error $O\left(n^{a-1} \eta\right)$.

Proof. It follows from Construction 4, Lemmas 17, 18, 20.

## 6 Instantiation

The Minsky-Papert function [28] gives a secret sharing scheme in $\mathrm{AC}^{0}$ with perfect privacy.

Theorem $15[7,28]$. For any $n \in \mathbb{N}$, there exists an explicit ( $n, n^{\frac{1}{3}}$ ) secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, message length 1 , perfect privacy and reconstruction.

Combining our techniques with Theorem 15, we have the following results. The detailed proofs are deferred to the full version.

Theorem 16. For any $n \in \mathbb{N}$, any constant $\alpha \in(0,1], \beta \in[0,1)$, there exists an explicit $(n,(1-\alpha) n)$ secret sharing scheme in $\mathrm{AC}^{0}$ with share alphabet $\{0,1\}^{O(\log n)}$, message alphabet $\{0,1\}$, message length $m=n^{\beta}$, adaptive privacy error $2^{-\Omega\left(\left(\frac{n}{m \log n}\right)^{1 / 3}\right)}$ and perfect reconstruction.

Theorem 17. For any $n \in \mathbb{N}$, for any constant $\gamma \in(0,1 / 4)$, there exists an explicit $(n, \Omega(n))$ secret sharing scheme in $\mathrm{AC}^{0}$ with alphabet $\{0,1\}$, message length $m=\Omega(n)$, adaptive privacy error $2^{-\Omega\left(n^{\frac{1}{4}-\gamma}\right)}$ and perfect reconstruction.

## 7 Extensions and Other Applications

The detailed constructions and proofs in this section appear in the full version.

### 7.1 Robust Secret Sharing

Our secret sharing schemes can be made robust by using robust secret sharing schemes and authentication techniques in small blocks.

We first recall the following robust secret sharing scheme.
Theorem 18 [15]. For any $n \in \mathbb{N}$, any constant $\rho<1 / 2$, there is an $(n, \Omega(n))$ robust secret sharing scheme, with alphabet $\{0,1\}^{O(1)}$, message length $\Omega(n)$, perfect privacy, robustness parameter $d=\rho n$ and reconstruction error $2^{-\Omega(n)}$.

We use concatenations of the schemes from Theorem 18 to get the following robust secret sharing scheme in $\mathrm{AC}^{0}$ with poly-logarithmic number of shares.

Lemma 21. For any $n \in \mathbb{N}$, any constant $a \in \mathbb{N}$, any $\epsilon=1 /$ poly $(n)$, there exists an $\left(n_{0}=O\left(\log ^{a} n\right), k_{0}=\Omega\left(n_{0}\right)\right)$ robust secret sharing scheme in $\mathrm{AC}^{0}$ (in $\left.n\right)$, with share alphabet $\{0,1\}^{O(1)}$, message alphabet $\{0,1\}$, message length $\Omega\left(n_{0}\right)$, perfect privacy, robustness parameter $\Omega\left(n_{0}\right)$, reconstruction error $\epsilon$.

Next, we give our construction of robust secret sharing scheme with "asymptotically good" parameters.

Theorem 19. For any $n \in \mathbb{N}$, any $\eta=\frac{1}{\text { poly }(n)}$, there exists an explicit $(n, \Omega(n))$ robust secret sharing scheme in $\mathrm{AC}^{0}$ with share alphabet $\{0,1\}^{O(1)}$, message alphabet $\{0,1\}$, message length $m=\Omega(n)$, non-adaptive privacy error $2^{-n^{\Omega(1)}}$, non-adaptive robustness $\Omega(n)$ and reconstruction error $\eta$.

### 7.2 Stochastic Error Correcting Code

Using our general strategy, we can also construct stochastic error correcting codes in $\mathrm{AC}^{0}$ which can resist additive errors [20].

One important component of our construction is the following "tiny" codes. It is constructed by classic code concatenation techniques.

Lemma 22. For any $n \in \mathbb{N}$, any constant $a \in \mathbb{N}$, there exists an asymptotically good binary $\left(n=O\left(\log ^{a} n\right), m, d\right)$ code $C$ such that the encoding and decoding can both be computed by $\mathrm{AC}^{0}$ circuits of size poly $(n)$.

Here we give the construction of stochastic error correcting codes in $\mathrm{AC}^{0}$ which are "asymptotically good".

Construction 5. For any $n \in \mathbb{N}$, we construct the following ( $n$, $m=\Omega(n), \rho=$ $\Omega(1))$ stochastic error correcting code.

Let $\delta_{0}, \delta_{1}$ be some proper constants in $(0,1)$.
Let $\left(\mathrm{Enc}_{0}, \mathrm{Dec}_{0}\right)$ be an asymptotically good $\left(n_{0}, m_{0}, d_{0}\right)$ error correcting code with alphabet $\{0,1\}^{p}$, $n_{0}=O(\log n), m_{0}=\delta_{0} n_{0}, d_{0}=\delta_{1} n_{0}$. In fact we can realize this code by applying an asymptotically good binary code, having the same rate, in parallel $p$ times.

Let $\left(\mathrm{Enc}_{1}, \mathrm{Dec}_{1}\right)$ be an asymptotically good $\left(n_{1}, m_{1}, d_{1}\right)$ error correcting code from Lemma 22 with alphabet $\{0,1\}, n_{1}=p+O(\log n), m_{1}=\delta_{0} n_{1}=O(p)$, $d_{1}=\delta_{1} n_{0}$.

Encoding function Enc : $\{0,1\}^{m=\Omega(n)} \rightarrow\{0,1\}^{n}$ is a random function which is as follows.

1. On input $x \in\{0,1\}^{m}$, split $x$ into blocks of length pmo such that $x=$ $\left(\bar{x}_{1}, \ldots, \bar{x}_{m /\left(p m_{0}\right)}\right) \in\left(\{0,1\}^{p m_{0}}\right)^{m /\left(p m_{0}\right)}$.
2. Let $y=\left(y_{1}, \ldots, y_{n^{\prime}}\right)=\left(\operatorname{Enc}_{0}\left(\bar{x}_{1}\right), \ldots, \operatorname{Enc}_{0}\left(\bar{x}_{m /\left(p m_{0}\right)}\right)\right) \in\left(\{0,1\}^{p}\right)^{n^{\prime}}, n^{\prime}=$ $m /\left(\delta_{0} p\right)$
3. Generate a random permutation $\pi:\left[n^{\prime}\right] \rightarrow\left[n^{\prime}\right]$.
4. Randomly pick $l=O(\log n)$ different indices $r_{1}, \ldots, r_{l} \in\left[n^{\prime}\right]$ and let $r=$ $\left(r_{1}, \ldots, r_{l}\right)$.
5. For every $i \in\left[n^{\prime}\right]$, let $\tilde{y}_{i}=\left(y_{\pi^{-1}(i)}, \pi^{-1}(i), \pi^{-1}\left(i \gg r_{1}\right), \ldots, \pi^{-1}\left(i \gg r_{l}\right), r\right)$.
6. Output $z=\left(\operatorname{Enc}_{1}\left(\tilde{y}_{1}\right), \ldots, \operatorname{Enc}_{1}\left(\tilde{y}_{n^{\prime}}\right)\right) \in\left(\{0,1\}^{n_{1}}\right)^{n^{\prime}}$.

Decoding function Dec : $\{0,1\}^{n=n_{1} n^{\prime}} \rightarrow\{0,1\}^{m}$ is as follows.

1. On the input z, apply $\operatorname{Dec}_{1}$ on every block of length $n_{0}$ to get $\tilde{y}$.
2. Take the majority of the $r$ in every $\tilde{y}_{i}, i \in\left[n^{\prime}\right]$ to get $r$.
3. $\forall i \in\left[n^{\prime}\right]$, we do the following. Check that for every $j \in[l]$, the corresponding backup of $\pi^{-1}(i)$ in the $\left(i \ll r_{j}\right)$ th block is equal to the one stored in the ith block. Take the approximate majority of these l tests, if the output is true then mark $\tilde{y}_{i}$ as good, otherwise mark it as bad.
4. Compute the entries of $y$ from shares that are marked as good. Other entries are set as blank.
5. Apply $\mathrm{Dec}_{0}$ on every block of $y$ of length pn $n_{0}$ to get $x$.

Theorem 20. For any $n \in \mathbb{N}$ and any $\epsilon=1 / \operatorname{poly}(n)$, there exists an explicit ( $n, m=\Omega(n), \rho=\Omega(1))$ stochastic binary error correcting code with decoding error $\epsilon$, which can be computed by $\mathrm{AC}^{0}$ circuits.

Note that if we set both levels of codes in our construction to be from Lemma 22 with length poly $\log n$ and $l$ to be also poly $\log n$, we can get quasipolynomially small decoding error following the same proof. The result is stated as the follows.

Theorem 21. For any $n \in \mathbb{N}$, any $\epsilon=2^{- \text {poly } \log n}$, there exists an explicit ( $n, m=\Omega(n), \rho=\Omega(1))$ stochastic binary error correcting code with decoding error $\epsilon$, which can be computed by $\mathrm{AC}^{0}$ circuits.

We can use duplicating techniques to make the decoding error to be even smaller, however with a smaller message rate.

Theorem 22. For any $n, r \in \mathbb{N}$, there exists an $(n, m=\Omega(n / r), \rho=\Omega(1))$ stochastic binary error correcting code with decoding error $2^{-\Omega(r / \log n)}$, which can be computed by $\mathrm{AC}^{0}$ circuits.

### 7.3 Secure Broadcasting

We give a protocol that allows $n$ parties to securely communication their secret inputs to each other using only a small amount of common secret randomness and communication over a public broadcast channel. The protocol should be secure against an external adversary who can (adaptively) observe and tamper with a constant fraction of the messages. This notion is formalized in Definition 3.

Protocol 1. For any $n, m \in \mathbb{N}$, for any $i \in[n]$, let $x_{i} \in\{0,1\}^{m}$ be the input of party $i$. Let the security parameter be $r \in \mathbb{N}$ with $r \leq m$.

Let $\left(\mathrm{RShare}_{0}, \mathrm{RRec}_{0}\right)$ be an $\left(n_{0}, k_{0}=\delta_{0} n_{0}\right)$ robust secret sharing scheme with share alphabet $\{0,1\}^{p=O(1)}$, secret length $m_{0}=m=\delta n_{0}$ and robustness parameter $d_{0}=\delta_{1} n_{0}$, as given by Theorem 18 for some constant $\delta, \delta_{0}, \delta_{1}$ with $\delta_{0} \geq \delta_{1}$.

Let ( $\mathrm{RShare}_{1}, \mathrm{RRec}_{1}$ ) be an ( $n_{1}, k_{1}=\delta_{0} n_{1}$ ) robust secret sharing scheme with share alphabet $\{0,1\}^{p=O(1)}$, secret length $m_{1}=p n_{0} / r=\delta n_{1}$ and robustness parameter $d_{1}=\delta_{1} n_{1}$, by Theorem 18 .

Assume that all parties have a common secret key $s \in\{0,1\}^{O(r \log (n r))}$.
The $i$-th party does the following.

1. Generate a $2^{-\Omega(r)}$-almostr-wise independent random permutation $\pi$ over $[n r]$ using $s$.
2. Compute the secret shares $y_{i}=\operatorname{RShare}_{0}\left(x_{i}\right) \in\left(\{0,1\}^{p}\right)^{n_{0}}$. Split $y_{i}$ into $r$ blocks each of length pno/r such that $y_{i}=\left(y_{i, 1}, \ldots, y_{i, r}\right)$.
3. View the communication procedure as having $[n r]$ time slots. For $j \in[r]$, on the $\pi((i-1) r+j)$ 's time slot, send message $z_{i, j}=\operatorname{RShare}_{1}\left(y_{i, j}\right)$.
4. For every $i \in[n], j \in[r]$, compute $y_{i, j}=\operatorname{RRec}_{1}\left(z_{i, j}\right)$, where $z_{i, j}$ is the message received in the $\pi((i-1) r+j)$ 's time slot.
5. For every $i \in[n]$ get $y_{i}=\left(y_{i, 1}, \ldots, y_{i, r}\right)$.
6. For every $i \in[n], x_{i}=\operatorname{RRec}_{0}\left(y_{i}\right)$.

Theorem 23. For any $n, m, r \in \mathbb{N}$ with $r \leq m$, there exists an explicit ( $n, m, \alpha=\Omega(1), n 2^{-\Omega(r)}, n 2^{-\Omega(r)}+n m 2^{-\Omega(m / r)}$ ) secure broadcasting protocol with communication complexity $O(n m)$ and secret key length $O(r \log (n r))$.

Acknowledgements. This collaboration started at the DIMACS Workshop on Cryptography and its Interactions. The first and third authors were supported by NSF Grant CCF-1617713. The second author was supported by a DARPA/ARL SAFEWARE award, NSF Frontier Award 1413955, NSF grants 1619348, 1228984, 1136174, and 1065276, NSF-BSF grant 2015782, ERC grant 742754, ISF grant 1709/14, BSF grant 2012378, a Xerox Faculty Research Award, a Google Faculty Research Award, an equipment grant from Intel, and an Okawa Foundation Research Grant. This material is based upon work supported by the Defense Advanced Research Projects Agency through the ARL under Contract W911NF-15-C-0205. The views expressed are those of the authors and do not reflect the official policy or position of the Department of Defense, the National Science Foundation, or the U.S. Government.

## References

1. Alon, N., Bruck, J., Naor, J., Naor, M., Roth, R.M.: Construction of asymptotically good low-rate error-correcting codes through pseudo-random graphs. IEEE Trans. Inf. Theory 38(2), 509-516 (1992)
2. Applebaum, B., Ishai, Y., Kushilevitz, E.: Cryptography in NC ${ }^{0}$. SIAM J. Comput. 36(4), 845-888 (2006)
3. Applebaum, B., Ishai, Y., Kushilevitz, E.: On pseudorandom generators with linear stretch in $\mathrm{NC}^{0}$. Comput. Complex. 17(1), 38-69 (2008)
4. Arora, S., Barak, B.: Computational Complexity: A Modern Approach. Cambridge University Press, Cambridge (2009)
5. Auger, A., Doerr, B.: Theory of Randomized Search Heuristics: Foundations and Recent Developments, vol. 1. World Scientific, Singapore (2011)
6. Bishop, A., Pastro, V., Rajaraman, R., Wichs, D.: Essentially optimal robust secret sharing with maximal corruptions. In: Fischlin, M., Coron, J.-S. (eds.) EUROCRYPT 2016. LNCS, vol. 9665, pp. 58-86. Springer, Heidelberg (2016). https:// doi.org/10.1007/978-3-662-49890-3_3
7. Bogdanov, A., Ishai, Y., Viola, E., Williamson, C.: Bounded indistinguishability and the complexity of recovering secrets. In: Robshaw, M., Katz, J. (eds.) CRYPTO 2016. LNCS, vol. 9816, pp. 593-618. Springer, Heidelberg (2016). https://doi.org/ 10.1007/978-3-662-53015-3_21
8. Bogdanov, A., Lee, C.H.: Homomorphic evaluation requires depth. In: Kushilevitz, E., Malkin, T. (eds.) TCC 2016. LNCS, vol. 9562, pp. 365-371. Springer, Heidelberg (2016). https://doi.org/10.1007/978-3-662-49096-9_15
9. Bogdanov, A., Williamson, C.: Approximate bounded indistinguishability. In: ICALP (2017)
10. Boyle, E., Gilboa, N., Ishai, Y.: Function secret sharing. In: Oswald, E., Fischlin, M. (eds.) EUROCRYPT 2015. LNCS, vol. 9057, pp. 337-367. Springer, Heidelberg (2015). https://doi.org/10.1007/978-3-662-46803-6_12
11. Bun, M., Thaler, J.: A nearly optimal lower bound on the approximate degree of $\mathrm{AC}^{0}$. In: FOCS (2017)
12. Chen, H., Cramer, R., Goldwasser, S., de Haan, R., Vaikuntanathan, V.: Secure computation from random error correcting codes. In: Naor, M. (ed.) EUROCRYPT 2007. LNCS, vol. 4515, pp. 291-310. Springer, Heidelberg (2007). https://doi.org/ 10.1007/978-3-540-72540-4_17
13. Cheng, K., Ishai, Y., Li, X.: Near-optimal secret sharing and error correcting codes in $\mathrm{AC}^{0}$. Cryptology ePrint Archive, Report 2017/927 (2017). http://eprint.iacr. org/2017/927
14. Cheng, K., Li, X.: Randomness extraction in $\mathrm{AC}^{0}$ and with small locality. arXiv preprint arXiv:1602.01530 (2016)
15. Cheraghchi, M.: Nearly optimal robust secret sharing. In: 2016 IEEE International Symposium on Information Theory (ISIT), pp. 2509-2513. IEEE (2016)
16. Cryan, M., Miltersen, P.B.: On pseudorandom generators in NC ${ }^{0}$. In: Sgall, J., Pultr, A., Kolman, P. (eds.) MFCS 2001. LNCS, vol. 2136, pp. 272-284. Springer, Heidelberg (2001). https://doi.org/10.1007/3-540-44683-4_24
17. Damgård, I., Ishai, Y., Krøigaard, M., Nielsen, J.B., Smith, A.: Scalable multiparty computation with nearly optimal work and resilience. In: Wagner, D. (ed.) CRYPTO 2008. LNCS, vol. 5157, pp. 241-261. Springer, Heidelberg (2008). https://doi.org/10.1007/978-3-540-85174-5_14
18. Goldreich, O.: Candidate one-way functions based on expander graphs. In: Goldreich, O. (ed.) Studies in Complexity and Cryptography. Miscellanea on the Interplay between Randomness and Computation. LNCS, vol. 6650, pp. 76-87. Springer, Heidelberg (2011). https://doi.org/10.1007/978-3-642-22670-0_10
19. Goldwasser, S., Gutfreund, D., Healy, A., Kaufman, T., Rothblum, G.N.: Verifying and decoding in constant depth. In: STOC, pp. 440-449. ACM (2007)
20. Guruswami, V., Smith, A.: Optimal rate code constructions for computationally simple channels. J. ACM (JACM) 63(4), 35 (2016)
21. Guruswami, V., Umans, C., Vadhan, S.: Unbalanced expanders and randomness extractors from Parvaresh-Vardy codes. J. ACM 56(4), 20 (2009)
22. Hagerup, T.: Fast parallel generation of random permutations. In: Albert, J.L., Monien, B., Artalejo, M.R. (eds.) ICALP 1991. LNCS, vol. 510, pp. 405-416. Springer, Heidelberg (1991). https://doi.org/10.1007/3-540-54233-7_151
23. Lee, C.H., Viola, E.: Some limitations of the sum of small-bias distributions. In: ECCC, vol. 22, p. 5. Citeseer (2015)
24. Lin, H.: Indistinguishability obfuscation from constant-degree graded encoding schemes. In: Fischlin, M., Coron, J.-S. (eds.) EUROCRYPT 2016. LNCS, vol. 9665, pp. 28-57. Springer, Heidelberg (2016). https://doi.org/10.1007/ 978-3-662-49890-3_2
25. Lovett, S., Viola, E.: Bounded-depth circuits cannot sample good codes. In: CCC, pp. 243-251. IEEE (2011)
26. Matias, Y., Vishkin, U.: Converting high probability into nearly-constant time with applications to parallel hashing. In: STOC, pp. 307-316. ACM (1991)
27. Miles, E., Viola, E.: Substitution-permutation networks, pseudorandom functions, and natural proofs. J. ACM 62(6), 46:1-46:29 (2015)
28. Minsky, M., Papert, S.: Perceptrons. MIT Press, Cambridge (1988)
29. Mossel, E., Shpilka, A., Trevisan, L.: On $\varepsilon$-biased generators in $\mathrm{NC}^{0}$. Random Struct. Algorithms 29(1), 56-81 (2006)
30. Naor, M., Reingold, O.: Synthesizers and their application to the parallel construction of pseudo-random functions. J. Comput. Syst. Sci. 58(2), 336-375 (1999)
31. Shamir, A.: How to share a secret. Commun. ACM $22(11), 612-613$ (1979)
32. Viola, E.: On approximate majority and probabilistic time. Comput. Complex. 18(3), 337 (2009)
33. Viola, E.: The complexity of distributions. SIAM J. Comput. 41, 191-218 (2012)

[^0]:    A full version of this paper appears in [13].
    (c) International Association for Cryptologic Research 2017
    Y. Kalai and L. Reyzin (Eds.): TCC 2017, Part II, LNCS 10678, pp. 424-458, 2017.
    https://doi.org/10.1007/978-3-319-70503-3_14

[^1]:    ${ }^{1}$ The approximate degree of a Boolean function is the lowest degree of a real polynomial that can approximate the function within, say, an additive difference of $1 / 3$ on every input.

