# Nearly Optimal Robust Secret Sharing against Rushing Adversaries 

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#### Abstract

Robust secret sharing is a strengthening of standard secret sharing that allows the shared secret to be recovered even if some of the shares being used in the reconstruction have been adversarially modified. In this work, we study the setting where out of all the $n$ shares, the adversary is allowed to adaptively corrupt and modify up to $t$ shares, where $n=2 t+1 .{ }^{3}$ Further, we deal with rushing adversaries, meaning that the adversary is allowed to see the honest parties' shares before modifying its own shares. It is known that when $n=2 t+1$, to share a secret of length $m$ bits and recover it with error less than $2^{-\lambda}$, shares of size at least $m+\lambda$ bits are needed. Recently, Bishop, Pastro, Rajaraman, and Wichs (EUROCRYPT 2016) constructed a robust secret sharing scheme with shares of size $m+O(\lambda \cdot \operatorname{polylog}(n, m, \lambda))$ bits that is secure in this setting against non-rushing adversaries. Later, Fehr and Yuan (EUROCRYPT 2019) constructed a scheme that is secure against rushing adversaries, but has shares of size $m+O\left(\lambda \cdot n^{\varepsilon} \cdot \operatorname{polylog}(n, m, \lambda)\right)$ bits for an arbitrary constant $\varepsilon>0$. They also showed a variant of their construction with share size $m+O(\lambda \cdot \operatorname{polylog}(n, m, \lambda))$ bits, but with super-polynomial reconstruction time. We present a robust secret sharing scheme that is simultaneously close-to-optimal in all of these respects - it is secure against rushing adversaries, has shares of size $m+O(\lambda \log n(\log n+\log m))$ bits, and has polynomial-time sharing and reconstruction. Central to our construction is a polynomial-time algorithm for a problem on semi-random graphs that arises naturally in the paradigm of local authentication of shares used by us and in the aforementioned work.


## 1 Introduction

Secret sharing, first studied by Shamir [Sha79] and Blakley [Bla79], is a fundamental cryptographic primitive that allows a secret to be shared among several parties in such a way that certain authorized subsets of parties can reconstruct

[^0]the secret, while unauthorized subsets learn no information about the secret from their shares. Secret sharing has widespread applications across cryptography, ranging from secure multiparty computation [GMW87, BGW88, CCD88] to threshold cryptographic systems [DF90, Fra90, DDFY94].

Typically, threshold secret sharing schemes ${ }^{4}$ are required to satisfy two properties: correctness, which says that more than a certain number of parties can use their shares to reconstruct the secret, and privacy, which says that if there are fewer than this number of parties, then their shares together reveal nothing about the secret. A number of strengthenings of secret sharing have also been studied in the past owing to various applications, such as verifiable secret sharing [CGMA85], robust secret sharing [RB89], leakage-resilient secret sharing [BDIR18, GK18], etc.. In this work, we focus on robust secret sharing.

Robust Secret Sharing. In robust secret sharing, in addition to the standard correctness and privacy properties, we require the following robustness property: even if some of the shares are adversarially modified, there is a reconstruction procedure that can recover the original secret given all the shares (among which it does not know which have been modified). In this sense, robust secret sharing is to standard threshold secret sharing as decoding from errors is to decoding from erasures in coding theory.

To be more specific, suppose a secret of length $m$ bits is to be shared among $n$ parties with threshold $t$ - meaning the adversary is allowed to (adaptively) corrupt up to $t$ of the parties. The properties we ask of a robust secret sharing scheme are:

- Correctness - given $(t+1)$ shares, it is possible to reconstruct the secret,
- Privacy - given $t$ shares, the secret is hidden, and,
- Robustness - even if the adversary arbitrarily modifies up to $t$ shares belonging to the parties it corrupts, the secret should be recoverable given all $n$ shares.

If $t<n / 3$, it may be seen that Shamir secret sharing [Sha79] with threshold $t$ satisfies the robustness requirement, owing to the error correcting properties of the Reed-Solomon code. On the other hand, if $t \geqslant n / 2$, robustness is impossible as the adversary could modify a majority of the shares. In addition, it is known that for $n / 3 \leqslant t<n / 2$, it is not possible to achieve perfect robustness, and any construction will necessarily have a small probability of failure of reconstruction [Cev11], which we will call the robustness error. Further, any robust secret sharing scheme for $n=2 t+1$ that has robustness error at most $2^{-\lambda}$ has shares of length at least $(m+\lambda)$ bits [CDV94].

In this work, we are interested in schemes that are robust in this extreme case of $n=2 t+1$. And the quantity we are most interested in is the size of the shares as a function of the robustness error and the number of parties.

[^1]Prior Work. There has been significant past work [RB89,CDV94,CDF01, CDF ${ }^{+}$08, CFOR12, Che15, BP16, BPRW16,HO18,FY19b] in studying and constructing robust secret sharing schemes, both in the setting where $n=2 t+1$, and where $t<(1-\delta) n / 2$ for some constant $\delta>0$. We discuss here only the former line of work, which is what leads up to our own, and refer the reader to the paper by Bishop et al. [BPRW16] for discussions of the rest.

The first construction of robust secret sharing was by Rabin and Ben-Or [RB89], and had a share size of $m+\tilde{O}(n \lambda)$ bits. ${ }^{5}$ This was done by giving each party a set of $(n-1)$ keys of a message authentication code (MAC) that it could use to authenticate the shares held by each other party, and a set of $(n-1)$ MAC tags that other parties could use to authenticate its share. This was improved by Cramer, Damgard and Fehr [CDF01] to $m+\tilde{O}(n+\lambda)$ but with inefficient reconstruction, and later Cevallos et al. [CFOR12] achieved the same overhead with efficient reconstruction. The approach of the latter was to reduce the size of the authentication keys used in the MAC at the expense of a more complicated reconstruction procedure.

Recently, Bishop et al [BPRW16] improved the overhead in the share size to $m+\tilde{O}(\lambda)$. The central idea in their work is to not authenticate each share to every other party, but instead, for each party to have $d$ (roughly $O(\log n)$ ) authentication keys/tags corresponding to $d$ other randomly chosen parties. They further showed that even though some of the keys/tags themselves could be adversarially modified, enough information can be recovered after such corruptions to reconstruct the secret. However, it was pointed out later by Fehr and Yuan [FY19b] that the proof of robustness of this scheme relies on the adversary being non-rushing.

Rushing Adversaries. A rushing adversary, in our context, is one that decides how to change the shares of the parties it has corrupted after seeing the honest parties' shares. In the case of interactive reconstruction (which is what is used in our construction and in prior work), in each round the parties corrupted by the adversary may wait till they see all the honest parties' messages and then decide what to send. Robustness against such adversaries becomes relevant, for instance, if the parties are conducting the reconstruction amongst themselves as happens in multiparty computation protocols.

As mentioned above, Fehr and Yuan [FY19b] noted that the proof of robustness of Bishop et al [BPRW16] does not work if the adversary is rushing, though it is not known whether their construction is actually non-robust in this case. Fehr and Yuan then presented a construction of robust secret sharing, using the local authentication approach of Bishop et al that was robust against rushing adversaries, but with a share size of $m+\tilde{O}\left(\lambda \cdot n^{\varepsilon}\right)$ for an arbitrary constant $\varepsilon>0$. They also showed how to improve this to $m+\tilde{O}(\lambda)$ if the reconstruction procedure was allowed to run in super-polynomial time.

[^2]Our Results. We construct robust secret sharing against rushing adversaries in the setting of $n=2 t+1$, with a share size of $m+O(\lambda \cdot \log n(\log n+\log m))$ for secrets of size $m$ and robustness error $2^{-\lambda}$. (Note that the overhead here is only polylogarithmically larger than in the lower bound of $m+\lambda$ for share size shown in [CDV94].) Our reconstruction procedure is interactive, with two rounds of interaction, and both sharing and reconstruction are polynomial-time. Our approach is similar to those of Bishop et al [BPRW16] and Fehr and Yuan [FY19b], though our construction is simpler than either and does not use some of the sophisticated tools used there.

### 1.1 Technical Overview

In this subsection, we give a high-level overview of our construction and the techniques we use. Recall that we wish to share secrets of length $m$ among $n$ parties, with a threshold of $t$ (the adversary is allowed to corrupt up to $t$ parties), with a robustness error smaller than $2^{-\lambda}$, and with $n=2 t+1$.

Sharing and Reconstruction We follow the local authentication paradigm used in [BPRW16, FY19b]. In this approach, each party is given authentication information about the shares of a small set of parties, which we will call its "watchlist." We will set the size of the watchlist to be roughly $O(\log n)$, thus this contributes only a $\operatorname{polylog}(n)$ factor to the overhead in the size of each share. This is to be contrasted with Rabin and Ben-Or's approach [RB89] where each party stores authentication information about every other party, thus leading to a linear blow-up in the overhead. We now give some more details on how we use the local authentication paradigm.

Sharing. In the sharing phase, we first compute a set of Shamir shares $\left(\mathrm{Sh}_{1}\right.$, $\ldots, \mathrm{Sh}_{n}$ ) for the given secret. Then, for each $i \in[n]$, we pick a random multiset $S^{i}$ of size $d$ (where $d$ is roughly $O(\log n)$ ) from $[n] \backslash\{i\}$. $S^{i}$ will be the set of parties in the watchlist of party $i$. For every $j \in S^{i}$, we pick a random MAC key $k_{i \rightarrow j}$ and compute the tag $\sigma_{i \rightarrow j}$ of the $j^{\text {th }}$ Shamir share $\mathrm{Sh}_{j}$ using $k_{i \rightarrow j}$. The tuple ( $k_{i \rightarrow j}, \sigma_{i \rightarrow j}$ ) constitutes authentication information for $\mathrm{Sh}_{j}$. The share corresponding to the $i^{\text {th }}$ party includes the Shamir share $\mathrm{Sh}_{i}$, the watchlist $S^{i}$, and the authentication information $\left\{k_{i \rightarrow j}, \sigma_{i \rightarrow j}\right\}_{j \in S^{i}}$ of the parties in its watchlist.

There is, however, a concern about privacy as we are storing both the key $k_{i \rightarrow j}$ and the tag $\sigma_{i \rightarrow j}$ together and this might leak some information about the share $\mathrm{Sh}_{j}$ that is being authenticated. In order to deal with this issue, we use a tool called private (randomized) MAC introduced in [BPRW16]. This private MAC has the property that for any key $k_{i \rightarrow j}$, the pair ( $k_{i \rightarrow j}, \sigma_{i \rightarrow j}$ ) does not reveal any information about $\mathrm{Sh}_{j}$. This allows us to argue that even when the key is stored together with the tag, the privacy is still preserved.

Reconstruction. Recontruction is performed by a two-round interactive protocol that proceeds as follows.

- In the first round, the $i$-th party broadcasts its Shamir share $\mathrm{Sh}_{i}$. The honest parties will broadcast the correct shares, whereas for the adversarial parties, the broadcasted shares could either be the original share or some modified (even empty) share. At this point, we may partition the set of parties into three sets - the set $H$ of honest parties, the set $P$ of "passive" corrupted parties $i$ that are corrupted but broadcast the correct share $\mathrm{Sh}_{i}$, and the set $A$ of "active" corrupt parties that broadcast a modified share.
At the end of the first round, all the parties can determine if the shares of the parties in its watchlist have been modified or not by checking if the corresponding MAC tag verifies under the respective key. Specifically, for every $j \in S^{i}$, such that the $\sigma_{i \rightarrow j}$ verifies, party $i$ labels $j$ as "good." Similarly, if the tag does not verify, it labels $j$ as "bad." Thus, at the end of the first round, the parties can obtain the labels for each $j \in S^{i}$. Note that the honest parties will always label a party $j \in H \cup P$ as "good" and with overwhelming probability, will label a party in $A$ as "bad." Furthermore, at the end of the first round, the adversarial parties do not learn any information about the watchlist of the honest parties. This will be crucially used to argue robustness.
- In the second round, each party $i$ broadcasts $S^{i}$ along with the labels it computed as above for each $j \in S^{i}$. Again, the honest parties will broadcast the correct information whereas the adversarial parties, including the parties in $P$, can broadcast incorrect information. In particular, an adversarial party might modify its watchlist, and also incorrectly accuse some honest party as being "bad" or label a party in $A$ as "good."
The action of any adversary in this protocol effectively induces a (labelled directed) graph on the vertex set $V=[n]$ (with a vertex representing each party) that is generated by the following process:
- The adversary partitions $V$ into sets $H, P$, and $A$, where $|H| \geqslant t+1$.
- For every $i \in H$, we choose a random multiset $S^{i}$ of size $d$ from $[n] \backslash\{i\}$. For each $j \in S^{i}$, we add an edge $(i, j)$ to the graph. We label the edge $(i, j)$ to be "good" if $j \in H \cup P$, and "bad" if $j \in A$.
- The outgoing edges from $P \cup A$ and their labels are generated adversarially after seeing the edges and labels from $H$.
(In the above process, we sample the watchlists $S^{i}$ of honest parties after the adversary has partitioned the set into $H, P, A$. We note that this is fine, in spite of the fact that the watchlists are actually sampled during the sharing, since the adversary does not learn any information about the watchlist of honest parties at the end of the first round of the reconstruction protocol.) Suppose we have an algorithm that on input the above graph, outputs the set $S=H \cup P$. In this case, we are done since we can use the shares of these parties to reconstruct the correct secret by the correctness of Shamir sharing. We give an algorithm that finds an $S$ that has a large intersection with $H \cup P$ and a small intersection with $A$. With such an $S$, we can use the error correction properties of Shamir secret sharing (a.k.a. Reed-Solomon codes) to recover the correct secret.

We next briefly describe how the above graph algorithm works.

Vertex Identification Algorithm Our vertex identification algorithm crucially uses the connection between the problem at hand and the independent set problem. Recall that a subset of vertices in an undirected graph is said to be an independent set if there is no edge between any pair of vertices in the set. While our graph $G$ is a directed graph, there is a natural way to view it as an undirected graph: by simply keeping each edge with a "bad" label as an undirected edge and discarding all edges with "good" labels. It is not hard to show that a maximum independent set in this undirected version of $G$ would give the desired set $S$.

Unfortunately, computing the maximum independent set is NP-hard in the worst case [Kar72]. On the other hand, our graph is not a worst case graph since all edges from $H$ are random, although the edges from $A \cup P$ are worst case (i.e., adversarially generated). Finding independent sets in such "semi-random" graphs has long been a topic of study in literature, starting with the work of Feige and Kilian [FK01] (see also [FK00, CSV17,MMT18]). Unfortunately, these works do not apply to our scenario because of the following two reasons. First, the guarantees from this line of work do not suffice for us; a typical guarantee there is that an independent set found has a large size relative to the maximum independent set, whereas we need the fact that the independent set has a large intersection with $H \cup P$ and a small intersection with $A$. Second, the distribution of our graph is unlike those considered in [FK01,FK00, CSV17,MMT18]. Specifically, the distributions considered in literature are often the following: pick a set $I$ of vertices (i.e., the "planted independent set") and add random edges between $I$ and the remaining vertices. Then, the adversary is allowed to add arbritrary edges that are not within $I$. However, this is not the case for us since the edges from $P$ to $H$ are not random.

Despite the challenges mentioned in the previous paragraph, several things go in our favor. First, our directed graph $G$ actually contains more information than its undirected variant considered in the previous paragraph. For instance, if we have two vertices $u_{1}, u_{2}$ each having a directed edge pointing to $v$ but with different labels, then we know that either $u_{1}$ or $u_{2}$ must be corrupted. Such information is not included when we just consider finding an independent set in the trivial undirected version of $G$. This motivates us to look instead to what we call the conflict graph $G^{\text {conf }}$, where we add an edge between every pair of vertices $u_{1}, u_{2}$ that label a common neighbor differently. Clearly, $H$ remains an independent set in $G^{\text {conf }}$. Moreover, from the definition of $G^{\text {conf }}$, any independent set $I$ of $G^{\text {conf }}$ has a "consistent opinion" on all vertices in the following sense: every vertex $v$ is labelled with the same label by all its in-neighbors that lie in $I$. This leads us to the overall structure of our algorithm: (1) find a large independent set $I$ in $G^{\text {conf }}$ and (2) output the set of all vertices labelled "good" by (at least one vertex in) $I$.

Of course, we have not yet specified how we find a large independent set $I$ in $G^{\text {conf }}$. This is indeed where a second advantage of our scenario comes in: we are guaranteed to have an independent set $H$ of size more than half of the graph, unlike previous works that place weaker assumptions on the size of the "planted"
independent set. It turns out that this $1 / 2$ threshold makes the problem "easier." For instance, Nemhauser and Trotter [NT74] show that any extremal solution to the linear program (LP) relaxation of the independent set problem is half integral (i.e., every variable is assigned either $0,1 / 2$, or 1 ), which means that at least one vertex in $H$ is assigned 1 in the solution. In our proof, we use a more specific structural lemma from $\left[\mathrm{ACF}^{+} 04\right]$ (Lemma 3) together with the expansion property of the random part of our graph (Lemma 4) to argue that, if we let $I$ be the set of all vertices assigned 1 by the LP solution, then it contains sufficiently many vertices from $H$. This in turn implies that $I$ labels most of the vertices in $H \cup P$ as "good" and most of those in $A$ as "bad." A more quantitative version of this argument shows that the output set satisfies the desired properties.

### 1.2 Comparison with Prior Work

As mentioned earlier, our construction follows the paradigm of local authentication of shares introduced by Bishop et al [BPRW16] and used also by Fehr and Yuan [FY19b]. There are a number of similarities and differences between how we proceed in this paradigm and how these papers do, and we briefly explain these below.

Bishop et al [BPRW16]. The authors here use local authentication of shares to reduce the robust reconstruction to a graph theoretic problem called graph bisection, which when solved gives a set $S$ of trustworthy parties whose shares are used to reconstruct the secret.

Their reduction also involves partitioning the corrupted parties into a set $P$ of "passive" corruptions and a set $A$ of "active" corruptions according to whether a certain part of the shares are reported correctly during reconstruction. But their notion of passive corruption is stronger than the one we use here - they also require that parties in $P$ never falsely label a party in $H$ as being "bad." This required them to store the authentication information in a distributed manner, using a primitive they call robust distributed storage. Additionally, they had to authenticate not only the Shamir shares (as we do), but also the MAC keys themselves. As pointed out by Fehr and Yuan [FY19b], authenticating the keys in this way is what causes their proof of robustness to not include rushing adversaries. In the end, the set $S$ that is the solution to the graph bisection problem instance generated by this reduction is the set $H \cup P$.

We use a weaker notion of passive corruptions, where parties in $P$ are allowed to label parties in $H$ as "bad." This allows us leave out the distributed storage and the authentication of the MAC keys, enabling proofs of security against rushing adversaries. And solving the graph problem that we reduce to does not require recovering the entire set $H \cup P$, but only a set $S$ that has a large intersection with $H \cup P$ and a small intersection with $A$. Such a set is easier to find, which is what lets us relax the definition of passive corruptions, and is still sufficient to recover the shared secret due to the error correction properties of Shamir sharing.

Fehr and Yuan [FY19b]. The approach of Fehr and Yuan is more similar to ours. They also partition the parties into sets $H, P$ and $A$, and their definitions for these sets are the same as ours. However, they still use robust distributed storage to store authentication information. Our construction is simpler, using only private MACs, which both of these papers also use.

Their approach is also to recover a set $S$ of vertices that has a large intersection with $H \cup P$ and a small intersection with $A$, and then use the error correction properties of Shamir sharing to recover the secret. Further, similarly to us, they do this by reducing robust reconstruction to the vertex identification problem in the model of random graphs that comes up in our work.

But their algorithm to solve this problem over $n$ vertices with out-degree $d$ only works when the number of passive parties $(|P|)$ is at most roughly $n \cdot(\log d / \log n)^{2}$. And so, when they find that $|P|$ is more than this and their algorithm fails, they fall back to list-decoding of all the shares together to recover a list of possible sharings and then iterate over this list to find the actual secret. The size of this list is roughly $(\log n / \log d)^{\tilde{O}\left((\log n / \log d)^{2}\right)}$, leading to the restriction of $d \geqslant n^{\varepsilon}$ for some constant $\varepsilon$ for the list-decoding to run in polynomial-time.

Our algorithm, on the other hand, solves the same graph problem without any such restriction on $|P|$. To be more precise, it first solves the problem when $|P|$ is at most roughly $0.84 \cdot n$, and then observes that if $|P|$ is more than this, then $S=[n]$ is already a solution. As $|P|$ can be efficiently estimated from the graph, this solves the problem. This releases us from their restriction on $d$, which we can set to be $O(\log n)$, leading to our shares being significantly smaller.

### 1.3 Concurrent Work

In concurrent and independent work, Fehr and Yuan [FY19a] also give a construction of a robust secret sharing scheme secure against rushing adversaries in the setting of $n=2 t+1$ with near-optimal parameters. Their construction has shares of size $m+O\left(\lambda \cdot \log ^{3} n(\log n+\log m)\right)$ for secrets of size $m$ and robustness error $2^{-\lambda}$, and polynomial-time sharing and reconstruction, where the latter involves five rounds of interaction. We obtain the slightly better share size of $m+O(\lambda \cdot \log n(\log n+\log m)$ ), and our reconstruction procedure has two rounds of interaction.

In terms of techniques, while both papers have as their starting point the ideas underlying the construction in [FY19b], the papers differ significantly in how they proceed from there. Our improvements come from designing a better algorithm for the vertex identification problem described in Section 1.2. [FY19a] use the same algorithm for this problem that [FY19b] did, and develop new techniques to deal with the case where the algorithm fails. Specifically, they add additional instances of the consistency graph that are revealed over the course of the reconstruction, and use this information to do better list-decoding, thus eliminating the restrictions described above that the construction in [FY19b] was subject to.

Outline of paper. In Section 2, we define robust secret sharing and private MACs, and state some known facts and theorems that will be useful later. In Section 3, we present our vertex identification algorithm that will be used in our reconstruction procedure. For readers who are only interested in our cryptographic constructions, we suggest to look at Theorem 3 in Section 3 and skip to Section 4, where we present our robust secret sharing scheme.

## 2 Preliminaries

Notation. We use capital letters to denote distributions and their support, and corresponding lowercase letters to denote a sample from the same. Let [ $n$ ] denote the set $\{1,2, \ldots, n\}$ and $U_{r}$ denote the uniform distribution over $\{0,1\}^{r}$. For a finite set $S$, we denote $x \stackrel{\$}{\leftarrow} S$ as sampling $x$ uniformly at random from the set $S$. For any $i \in[n]$, let $x_{i}$ denote the symbol at the $i$-th co-ordinate of $x$, and for any $T \subseteq[n]$, let $x_{T} \in\{0,1\}^{|T|}$ denote the projection of $x$ to the co-ordinates indexed by $T$. We write $\circ$ to denote concatenation.

Multisets. Let $S$ be a multiset and for any element $a$, we define the multiplicity $m_{a}(S)$ to be the number of times $a$ occurs in the multiset $S$. For any two multisets $S_{1}, S_{2}$, we define $S_{1} \| S_{2}$ to be the multiset such that for any element $a, m_{a}\left(S_{1} \| S_{2}\right)=m_{a}\left(S_{1}\right)+m_{a}\left(S_{2}\right)$.

We assume the reader's familiarity with the definition of statistical distance.

### 2.1 Private MAC

In this subsection, we recall the definition of a private message authentication code (MAC) used by Bishop et al [BPRW16] and Fehr and Yuan [FY19b], but using different terminology. A private MAC for message space $\{0,1\}^{\eta}$ for some $\eta \in \mathbb{N}$ consists of the following algorithms, all of them running in time poly $(\eta)$.

- KeyGen : A randomized algorithm that outputs a key $k$.
- $\operatorname{Tag}(k,(m, r)):$ A deterministic algorithm that takes a key $k$, a "message tuple" $(m, r) \in\{0,1\}^{\eta} \times\{0,1\}^{\kappa}$ for some $\kappa \in \mathbb{N}$ (called the randomness length), and outputs a tag $\sigma$.
- Verify $(k,(m, r), \sigma):$ A deterministic algorithm that takes a key $k$, a message tuple $(m, r)$, and a tag $\sigma$, and outputs 1 or 0 .

Definition 1. For an $\eta, \ell \in \mathbb{N}$ and $\epsilon \in[0,1]$, a triple of algorithms (KeyGen, Tag, Verify) is an ( $\ell, \epsilon$ )-private $M A C$ for a message space $\{0,1\}^{\eta}$ if the following properties are satisfied for some $\kappa \in \mathbb{N}$.

- Correctness. For every message tuple $(m, r) \in\{0,1\}^{\eta} \times\{0,1\}^{\kappa}$, with $k \leftarrow$ KeyGen, and $\sigma \leftarrow \operatorname{Tag}(k,(m, r))$,

$$
\operatorname{Pr}[\operatorname{Verify}(k,(m, r), \sigma)=1]=1
$$

- Unforgeability. For any message tuple $(m, r) \in\{0,1\}^{\eta} \times\{0,1\}^{\kappa}$, and any adversary $\mathcal{A}$, with $k \leftarrow$ KeyGen, $\sigma \leftarrow \operatorname{Tag}(k,(m, r))$, and $\left(m^{\prime}, r^{\prime}, \sigma^{\prime}\right) \leftarrow \mathcal{A}(m$, $r, \sigma)$,

$$
\operatorname{Pr}\left[(m, r) \neq\left(m^{\prime}, r^{\prime}\right) \wedge \operatorname{Verify}\left(k,\left(m^{\prime}, r^{\prime}\right), \sigma^{\prime}\right)=1\right] \leqslant \epsilon .
$$

- Privacy. For every $m_{0}, m_{1} \in\{0,1\}^{\eta}$, any arbitrary set of $\ell$ keys $\left\{k_{1}, \ldots\right.$, $\left.k_{\ell}\right\}$, and any adversary $\mathcal{A}$, with $k \leftarrow$ KeyGen, $r \leftarrow\{0,1\}^{\kappa}$, and $\sigma_{i}^{b} \leftarrow \operatorname{Tag}\left(k_{i}\right.$, $\left.\left(m_{b}, r\right)\right)$ for $i \in[\ell]$ and $b \in\{0,1\}$,

$$
\operatorname{Pr}\left[\mathcal{A}\left(\sigma_{1}^{0}, \sigma_{2}^{0}, \ldots, \sigma_{\ell}^{0}\right)=1\right]=\operatorname{Pr}\left[\mathcal{A}\left(\sigma_{1}^{1}, \sigma_{2}^{1}, \ldots, \sigma_{\ell}^{1}\right)=1\right]
$$

- Uniformity. There is an $s \in \mathbb{N}$ such that for every $(m, r) \in\{0,1\}^{\eta} \times\{0,1\}^{\kappa}$, with $k \leftarrow$ KeyGen, the distribution of $\sigma \leftarrow \operatorname{Tag}(k,(m, r))$ is uniform over $\{0,1\}^{s}$.
The following theorem follows from the construction of a private MAC presented in [BPRW16], using GF[ $2^{\lambda}$ ] as the field there.
Theorem 1 ([BPRW16]). For any $\eta, \ell \in \mathbb{N}$ and $\varepsilon \in[0,1]$, there exists an $(\ell, \varepsilon)$-private MAC for message space $\{0,1\}^{\eta}$, with randomness length $\ell \lambda$, key length $2 \lambda$ and tag length $\lambda$, where $\lambda=\lceil\log ((\eta+\ell) / \varepsilon)\rceil$.


### 2.2 Secret Sharing Scheme

We start with the definition of the sharing function and then give the definition of a threshold secret sharing scheme.

Definition 2 (Sharing Function [Bei11]). Let $[n]=\{1,2, \ldots, n\}$ be a set of identities of $n$ parties. Let $\mathcal{M}$ be the domain of secrets. A sharing function Share is a randomized mapping from $\mathcal{M}$ to $\mathcal{S}_{1} \times \mathcal{S}_{2} \times \ldots \times \mathcal{S}_{n}$, where $\mathcal{S}_{i}$ is called the domain of shares of party with identity $i$. A dealer distributes a secret $m \in \mathcal{M}$ by computing the vector $\operatorname{Share}(m)=\left(\mathrm{S}_{1}, \ldots, \mathrm{~S}_{n}\right)$, and privately communicating each share $\mathrm{S}_{i}$ to the party $i$. For a set $T \subseteq[n]$, we denote $\operatorname{Share}(m)_{T}$ to be a restriction of Share $(m)$ to its $T$ entries.
Definition 3 ( $\left(t, n, \epsilon_{c}, \epsilon_{s}\right)$-Secret Sharing Scheme). Let $\mathcal{M}$ be a finite set of secrets, where $|\mathcal{M}| \geqslant 2$. Let $\epsilon_{c}, \epsilon_{s} \in[0,1], t, n \in \mathbb{N}$ such that $t \leqslant n$, and $[n]=\{1$, $2, \ldots, n\}$ be a set of identities (indices) of $n$ parties. A sharing function Share with domain of secrets $\mathcal{M}$ is a $\left(t, n, \epsilon_{c}, \epsilon_{s}\right)$-secret sharing scheme if the following two properties hold:

- Correctness: The secret can be reconstructed by anyt-out-of-n parties. That is, for any set $T \subseteq[n]$ such that $|T| \geqslant t$, there exists a deterministic, interactive reconstruction protocol Rec between the parties in $T$ with the input of $i \in T$ being Share $(m)_{i}$ such that for every $m \in \mathcal{M}$,

$$
\operatorname{Pr}\left[\operatorname{Rec}\left(\operatorname{Share}(m)_{T}\right)=m\right]=1-\epsilon_{c}
$$

where the probability is over the randomness of the Share function. We will slightly abuse the notation and denote Rec as the reconstruction protocol that takes in $T$ and $\operatorname{Share}(m)_{T}$ where $T$ is of size at least $t$ and outputs the secret.

- Statistical Privacy: Any collusion of less than t parties should have "almost" no information about the underlying secret. More formally, for any unauthorized set $U \subseteq[n]$ such that $|U|<t$, and for every pair of secrets $m_{0}, m_{1} \in M$, for any distinguisher $D$ with output in $\{0,1\}$, the following holds :

$$
\left|\operatorname{Pr}\left[D\left(\operatorname{Share}\left(m_{0}\right)_{U}\right)=1\right]-\operatorname{Pr}\left[D\left(\operatorname{Share}\left(m_{1}\right)_{U}\right)=1\right]\right| \leqslant \epsilon_{s}
$$

We define the rate of the secret sharing scheme as

$$
\lim _{|m| \rightarrow \infty} \frac{|m|}{\max _{i \in[n]}\left|\operatorname{Share}(m)_{i}\right|}
$$

Remark 1. The above definition of privacy considers a weaker notion where the unauthorized set $U$ is specified upfront. We can also consider a stronger variant where the adversary adaptively specifies this set $U$ one party at a time, seeing the share of each party as it is specified. We note that for the case of perfect privacy (i.e., $\epsilon_{s}=0$ ), the above two variants are equivalent.

### 2.3 Robust Secret Sharing

We now give the definition of robust secret sharing scheme.
Definition 4. Let $\epsilon_{c}, \epsilon_{s}, \delta \in[0,1], t, n, \kappa, \tau \in \mathbb{N}$ such that $t \leqslant n$, and $\tau \leqslant \kappa \leqslant n$. An $\left(t, n, \epsilon_{c}, \epsilon_{s}\right)$ secret sharing scheme (Share, Rec) for message space $\mathcal{M}$ is said to be $(\delta, \kappa, \tau)$-robust if for every interactive adversary $\mathcal{A}$ and message $m \in \mathcal{M}$,

$$
\operatorname{Pr}\left[\operatorname{Expt}_{\mathcal{A}, m, \kappa, \tau}=1\right] \leqslant \delta
$$

where $\operatorname{Expt}_{\mathcal{A}, m, \kappa, \tau}$ is defined below.
$-\left(\right.$ share $_{1}, \ldots$, share $\left._{n}\right) \leftarrow \operatorname{Share}(m)$.

- $\mathcal{A}$ outputs a set $\Gamma \subseteq[n]$ such that $|\Gamma|=\kappa$.
- Set $T=\emptyset$. Repeat until $|T|=\tau$ :
- $\mathcal{A}$ chooses $i \in \Gamma \backslash T$.
- Update $T=T \cup\{i\}$ and give share $_{i}$ to $\mathcal{A}$.
- Run the reconstruction protocol among the parties in $\Gamma$ with every party $i \in \Gamma \backslash T$ behaving honestly using its share share ${ }_{i}$ and the adversary $\mathcal{A}$ taking control the parties in $T . \mathcal{A}$ is allowed to behave maliciously (possibly using a different share) and can deviate arbitrarily from the specification of the reconstruction protocol. Here, we assume that in every round of the reconstruction protocol, $\mathcal{A}$ can send its outgoing messages after seeing all its incoming messages from the honest parties (a.k.a. rushing adversary). For every $i \in \Gamma \backslash T$, let $m_{i}^{\prime}$ be the output of the $i$-th party at the end of the reconstruction algorithm.
- Output 1 if and only if there exists an $i \in \Gamma \backslash T$ such that $m \neq m_{i}^{\prime}$.

We call $\log \left(1 / \epsilon_{s}\right)$ as the privacy parameter and $\log (1 / \delta)$ as the robustness parameter.

Fact 2 Fix $t \in \mathbb{N}, n, \kappa \geqslant t$. There is a robust reconstruction protocol such that Shamir secret sharing is a $(t, n, 0,0)$ secret sharing scheme that is $(0, \kappa$, $\lfloor(\kappa-t) / 2\rfloor)$-robust.

## 3 Vertex Identification Algorithm

In this section, we give a polynomial time graph algorithm that we call the vertex identification algorithm, which will be used as a building block in the construction of robust secret sharing. We start with the description of the model of semi-random graphs the algorithm works for.

Model. Our (directed) $\operatorname{graph}^{6} G$ where $|V|=n$ and a labeling $L: E_{G} \rightarrow$ good, bad \} is generated as follows:

- First, the adversary partitions $V$ into three parts $H, P, A$ such that $|H| \geqslant$ $\frac{n+1}{2}$.
- For every $u \in H$ and every $i \in\{1, \ldots, d\}$, select a vertex $v_{i} \in(V \backslash\{u\})$ uniformly at random and add an edge $\left(u, v_{i}\right)$ to the graph. The label for $L\left(u, v_{i}\right)$ of the edge is "good" if $v_{i}$ lies in $H \cup P$, and is "bad" if $v_{i} \in A$.
- The outgoing edges of $A \cup P$ and their labels are generated by the adversary after seeing the edges and the labels from $H$.

For notational convenience, we will always think of each vertex $v \in V$ as having a self-loop with label $L(v, v)=$ good. Our main theorem in this section is the following:

Theorem 3. There is a polynomial-time (deterministic) algorithm VertexID that, given $G, L$ generated as above with $d \geqslant C \log n$, outputs a set $S \subseteq V$ that satisfies

$$
\begin{equation*}
|S \cap(H \cup P)| \geqslant \frac{n+1}{2}+2 \cdot|S \cap A| \tag{1}
\end{equation*}
$$

with probability $1-O\left(e^{-\beta d}\right)$, where $C>1$ and $\beta>0$ are some constants.
The Algorithm. Our algorithm relies on the linear programming (LP) relaxation of the Independent Set problem. To state the LP relaxation for Independent Set, let first recall that we may reformulate Independent Set on an undirected graph $F=\left(V, E_{F}\right)$ as the following integer program (IP):

$$
\begin{array}{cr}
\max \sum_{v \in V} x_{v} & \\
\text { subject to } x_{v} \in\{0,1\} & \forall v \in V \\
x_{u}+x_{v} \leqslant 1 & \forall\{u, v\} \in E_{F}
\end{array}
$$

Notice that a solution $\left(x_{v}\right)_{v \in V}$ to the above IP corresponds to an independent set $\left\{v \in V \mid x_{v}=1\right\}$. Since solving the above IP is equivalent to finding the maximum indepent set of the graph $F$, the problem remains NP-hard. As a result, we have to resort to the LP relaxation of the above IP where the condition $x_{v} \in\{0,1\}$ is relaxed to $0 \leqslant x_{v} \leqslant 1$.

[^3]More specifically, the LP relaxation of Independent Set for an undirected graph $F=\left(V, E_{F}\right)$ can be stated as:

$$
\begin{array}{rr}
\max \sum_{v \in V} x_{v} & \\
\text { subject to } 0 \leqslant x_{v} \leqslant 1 & \forall v \in V \\
x_{u}+x_{v} \leqslant 1 & \forall\{u, v\} \in E_{F} \tag{4}
\end{array}
$$

We refer to the above relaxation as LP-IS of $F$.
As the reader might have noticed, the Independent Set (IS) problem is defined on undirected graphs, whereas our input graph $G$ is a directed graph. To turn this into an instance of Independent Set, we create what we will call the conflict graph of $G, L$ :

Definition 4 Given a directed graph $G=\left(V, E_{G}\right)$ and a labeling $L: E_{G} \rightarrow$ $\left\{\right.$ good, bad\}, their conflict graph is denoted ${ }^{7}$ by $G^{\text {conf }}=\left(V, E^{\text {conf }}\right)$. This is an undirected graph on the same vertex set of $G$, and there is an edge between two vertices $u, v \in V$ iff there exists a common out-neighbor $w$ of $u, v$ such that $L(u, w) \neq L(v, w)$.

Recall that we always add a self-loop with "good" label to every vertex in $G$, which means that $\{u, v\}$ will always be an edge in the conflict graph if (u, $v) \in E_{G}$ and $L(u, v)=\mathrm{bad}$.

There are a couple (straightforward) observations that will be useful to keep in mind. The first one is that $H$ is an independent set in this conflict graph:

Observation $5 H$ is an independent set in the conflict graph of $G, L$.
The second is that, for any independent set $I$ of the conflict graph, its vertices never label a vertex inconsistently. This follows from the definition of the conflict graph, as such an inconsistency would create an edge in the graph.

Observation 6 Let $I$ be any independent set of a conflict graph of $G, L$. Then, for any $v \in V$, it must fall into one of the following three categories:
$-v$ has no in-neighbor (w.r.t. G) in I.

- $v$ has at least one in-neighbor (w.r.t. $G$ ) in $I$ and each of $v$ 's neighbors in $I$ labels it with good.
- v has at least one in-neighbor (w.r.t. G) in I and each of v's neighbors in I labels it with bad.

Moreover, we have to recall the concept of extreme point solutions of linear programs. A feasible solution is said to be an extreme point solution if it cannot be written as a convex combination of other feasible solutions ${ }^{8}$. For any LP with

[^4]a finite feasible region, it is known that an extreme point optimum solution exists and can be found in polynomial time (see e.g. [Jai98]). Extreme point solutions are widely used in approximation algorithms; interested readers may refer to the survey [LRS11] for more details.

With these ready, we can now describe our algorithm, which we call ConflictLP:
$\operatorname{Conflict-LP}\left(G=\left(V, E_{G}\right), L\right)$
0 . For each vertex $v \in V$, let $b_{v}$ denote the number of vertices labelled bad by it. If the median of $b_{v}$ 's is at most $0.16 d$, then output $V$ and terminate.

1. Construct the conflict graph $G^{\mathrm{conf}}$ of $G, L$ (as in Definition 4).
2. Solve for an extreme point optimum solution $\left\{x_{v}^{*}\right\}_{v \in V}$ of LP-IS of $G^{\text {conf }}$. Let $I$ denote the set of all vertices $u \in V$ such that $x_{u}^{*}=1$.
3. Let $S$ be the set of all vertices $v \in V$ that has at least one in-neighbor (w.r.t. $G$ ) in $I$ and is labelled good by the neighbor(s). Output $S$.

It is obvious to see that the algorithm runs in polynomial time.
Correctness Intuition. Before we proceed with the formal proof of correctness, let us briefly give an informal intuition behind the proof. First, Step 0 simply helps us deal with the "trivial" case where $|A|$ is less than ${ }^{9}$ say $0.15(n-1)$. In this case, we can simply output the whole vertex set $V$; this is indeed what Step 0 does. Thus, from this point onward, we may assume that $|A| \geqslant 0.15(n-1)$.

Next, a priori, it is not even clear that $I$ must be non-empty. To see this, let us first recall a classic result of Nemhauser and Trotter that an extreme point solution of LP-IS is always half-integral, meaning that $x_{v} \in\{0,1 / 2,1\}$ for every vertex $v$.

Theorem 7 ( [NT74]). In any extreme point solution of LP-IS, $x_{v} \in\{0,1 / 2$, $1\}$ for all $v \in V$.

Now, if $I$ were empty, then we would have $x_{v} \in\{0,1 / 2\}$, which would imply that the optimum of LP-IS is at most $n / 2<|H|$. This would be a contradiction to Observation 5.

Next, let us consider the set $T=\left\{v \mid x_{v}^{*}=0\right\}$. It is well-known that $|I| \geqslant|T|$; this can be easily seen because otherwise we can instead assign 1 to $T$ and 0 to $I$ and obtain a valid LP solution with larger objective value. (In fact, we will use a stronger property between the two sets below.)

For simplicity of exposition, let us assume for now that $I$ only contains honest players, i.e., $I \subseteq H$, and that $T$ only contains active adversary, i.e.,

[^5]$T \subseteq A$. Observe that, due to condition (4) of the relaxation, we must have $N_{G^{\text {conf }}}(I) \subseteq T$. Notice also that, since every honest player labels its out-neighbor in $A$ as "bad", we must have $N_{G^{\text {conf }}}(I) \supseteq\left(N_{G}^{\text {out }}(I) \cap A\right)$. From this and from the bound $|I| \geqslant|T|$ in the previous paragraph, it must be that $|I| \geqslant\left|N_{G}^{\text {out }}(I) \cap A\right|$. However, recall that the edges from $I$ to $A$ are ("essentially") random of degree $\Omega(d) \geqslant 100 \log n$. Such a "non-expansion" condition can only hold when $N_{G}^{\text {out }}(I) \cap A$ already contains all but $o(1)$ fraction of $A$. In other words, we have $|I| \geqslant\left|N_{G}^{\text {out }}(I) \cap A\right| \geqslant(1-o(1))|A| \geqslant(0.15-o(1)) n$, where the last inequality comes from our assumption that $|A| \geqslant 0.15(n-1)$.

Now, note that we never output vertices from $N_{G}^{o u t}(I) \cap A$ because they are already labelled as bad by at least one vertex in $I$. This means that $S \cap A$ is small (i.e. $o(n)$ ). On the other hand, we always output all vertices in $N_{G}^{\text {out }}(I) \cap(H \cup P)$ because they are labelled good by at least one vertex in $I$; since we concluded that $|I| \geqslant(0.15-o(1)) n$ in the previous paragraph, it follows from vertex expansion of random graphs that $\left|N_{G}^{\text {out }}(I) \cap(H \cap P)\right| \geqslant|H \cup P|-o(n) \geqslant \frac{n+1}{2}+|P|-o(n)$. By a more careful calculation of the terms $o(n)$, it is then possible to show that (1) holds.

To turn the above intuition into a formal proof, we not only have to make the calculations more precise, but we also need to deal with the case where $I \nsubseteq H$ ( or $T \nsubseteq A$ ). Nevertheless, we can still show, using a more general structural result (see Lemma 3), that $I \cap H$ still satisfies "non-expansion". This allows the proof to go through in a similar manner.

### 3.1 Proof of Correctness of the Algorithm

We now give a formal proof of correctness of our algorithm. We will need several additional notations:

- Once again, let $I=\left\{v \in V \mid x_{v}^{*}=1\right\}$ and $T=\left\{v \in V \mid x_{v}^{*}=0\right\}$. Moreover, let $R=\left\{v \in V \mid x_{v}^{*}=1 / 2\right\}$.
- Let $I_{H}, T_{H}$ and $R_{H}$ denote $I \cap H, T \cap H$ and $R \cap H$ respectively. Similarly, let $I_{\bar{H}}, T_{\bar{H}}$ and $R_{\bar{H}}$ denote $I \backslash H, T \backslash H$ and $R \backslash H$ respectively.

We will prove our main theorem for the constants $C=10^{10}$ and $\beta=10^{-10}$. It is henceforth assumed that $d \geqslant C \log n$, and this will not be explicitly stated. We remark that we make no attempt in optimizing these constants and it is likely that they can be reduced substantially.

Step 0: Dealing with the trivial case. As stated earlier, Step 0 in our algorithm helps us take care of the "trivial" case where $A$ is already small. In particular, we can show that, if $|A| \leqslant 0.15(n-1)$, then the algorithm w.h.p. simply outputs $V$, which is a correct output in this case. Moreover, it is not hard to see that, when $A$ is larger than $(n-1) / 6$ and $V$ is the wrong answer, then we do not terminate in this step and proceed to the remaining part of the algorithm. This is encapsulated in the following lemma.

Lemma 1. When $|A| \leqslant 0.15(n-1)$, our algorithm outputs $V$ and terminates at Step 0 with probability $1-O\left(e^{-\beta d}\right)$. On the other hand, when $|A|>(n-1) / 6$, our algorithm terminates at Step 0 with probability only $O\left(e^{-\beta d}\right)$.

It turns out that the above lemma follows easily from the concentration of the number of bad labels given by each honest vertex. This concentration is stated and proved below.

Observation 8 With probability $1-O\left(e^{-\beta d}\right)$, for all vertices $u \in H$, we have $(\mu-0.001) d \leqslant\left|N_{G}^{\text {out }}(u) \cap A\right| \leqslant(\mu+0.001) d$ where $\mu=\frac{|A|}{n-1}$.

Proof. We will only prove that $\left|N_{G}^{o u t}(u) \cap A\right| \geqslant(\mu-0.001) d$ for every $u \in H$ with high probability. The upper bound can be shown analogously. Note that our desired bound is obvious when $\mu \leqslant 0.001$. Hence, we may assume that $\mu \geqslant 0.001$, or equivalently $|A| \geqslant 0.001(n-1)$.

Let us fix $u \in H$ and a vertex $v \in A$. The probability that $v$ belongs to $N_{G}^{\text {out }}(u)$ is exactly $\frac{d}{n-1}$. Moreover, from how the graph is generated, the events $v \in N_{G}^{o u t}(u)$ for different $v$ 's are independent. Hence, Chernoff bounds implies that

$$
\begin{aligned}
& \operatorname{Pr}\left[\left|N_{G}^{\text {out }}(u) \cap A\right|<0.999 \cdot\left(\frac{d}{n-1} \cdot|A|\right)\right] \\
& \leqslant \exp \left(-\frac{10^{-6}}{2} \cdot \frac{d}{n-1} \cdot|A|\right) \\
& \leqslant \exp \left(-\frac{10^{-6}}{2} \cdot \frac{d}{n-1} \cdot 0.001(n-1)\right) \\
& \leqslant n^{-2} \cdot e^{-\beta d}
\end{aligned}
$$

where the second and third inequalities follow from $|A| \geqslant 0.001(n-1)$ and $d \geqslant$ $10^{10} \log n$ respectively. Furthermore, observe that $0.999 \cdot\left(\frac{d}{n-1} \cdot|A|\right)=0.999 \mu d \geqslant$ $(\mu-0.001) d$. Hence, we have $\operatorname{Pr}\left[\left|N_{G}^{o u t}(u) \cap A\right|<(\mu-0.001) d\right] \leqslant n^{-2} \cdot e^{-\beta d}$. Using union bound over all $u \in H$ concludes our proof.

Now that we have proved the concentration, we can prove Lemma 1 simply as follows.
Proof of Lemma 1: Suppose that $|A| \leqslant 0.15(n-1)$. Then, from Observation 8, w.p. $1-O\left(e^{-\beta d}\right)$ we have $\left|N_{G}^{\text {out }}(u) \cap A\right| \leqslant\left(\frac{|A|}{n-1}+0.001\right) d<0.16 d$ for all $u \in H$. Notice that $N_{G}^{\text {out }}(u) \cap A$ is exactly the set of vertices for which $u$ labels bad. As a result, for these $|H| \geqslant t+1$ vertices, they label bad to at most $0.16 d$ vertices. This means that the condition in Step 0 is satisfied and the algorithm outputs $V$.

On the other hand, if $|A|>(n-1) / 6$, then Observation 8 gives the bound $\left|N_{G}^{\text {out }}(u) \cap A\right| \geqslant\left(\frac{|A|}{n-1}-0.001\right) d>0.16 d$ with probability $1-O\left(e^{\beta d}\right)$. When this event occurs, each $u \in H$ labels more than $0.16 d$ vertices as bad. Thus, the condition in Step 0 is not satisfied in this case.

Step I: Non-Expansion of $I_{H}$ to $A$. The first step of the remaining part of the proof is to show that the set $I_{H}$ does not (vertex-)expand in $A$ (w.r.t out-edges in $G)$, as stated below. We remark here that it also implicitly implies that $I_{H} \neq \emptyset$.

Lemma 2. $\left|I_{H}\right|>\left|N_{G}^{\text {out }}\left(I_{H}\right) \cap A\right|$.
To prove non-expansion of $I_{H}$, we will resort to a structural result regarding an extreme point LP solution. It is easiest to state in terms of crown as defined below [CFJ04, $\mathrm{ACF}^{+}$04]:

Definition 9 For any undirected graph $F=\left(V, E_{F}\right)$ and disjoint subsets $I$, $T \subseteq F,(I, T)$ is said to be a crown of $F$ if (i) $T=N(I)$ and (ii) there is a matching between $T$ and $I$ such that all vertices in $T$ are matched.

Lemma 3 ( [AFLS07]). For any undirected graph $F=\left(V, E_{F}\right)$, let $\left\{x_{v}^{*}\right\}_{v \in V}$ be an extreme point solution, and let $I=\left\{v \in V \mid x_{v}^{*}=1\right\}$ and $T=\{v \in V \mid$ $\left.x_{v}^{*}=0\right\}$. Then, $(I, T)$ forms a crown (w.r.t $F$ ).

Notice that if $(I, T)$ is a crown, then it must be that $|I| \geqslant|T|$ due to (ii). Hence, the above result is stronger than the one we used in the informal exposition. We are now ready to prove Lemma 2.
Proof of Lemma 2: First, we claim that

$$
\begin{equation*}
\left|R_{H}\right| \leqslant\left|R_{\bar{H}}\right| \tag{5}
\end{equation*}
$$

This is because otherwise we can instead set $x_{v}^{*}=1$ for $v \in R_{H}$ and $x_{v}^{*}=0$ for $v \in R_{\bar{H}}$, which would give an LP solution with higher value. (Note that this is a valid LP solution because $R_{H}$ is an independent set from Observation 5, and all neighbors of $I$ lie in $T$.)

From Lemma 3, $(I, T)$ forms a crown. Consider a matching from $T$ to $I$ such that all vertices in $T$ are matched (which is guaranteed to exist from the definition of a crown). Notice that there is no edge from $T_{H}$ to $I_{H}$ in the graph $G^{\text {conf. }}$ hence, all vertices in $T_{H}$ must be matched to vertices in $I_{\bar{H}}$. In other words, we have

$$
\begin{equation*}
\left|T_{H}\right| \leqslant\left|I_{\bar{H}}\right| \tag{6}
\end{equation*}
$$

From (5) and (6), we have

$$
\left|I_{H}\right|=|H|-\left|T_{H}\right|-\left|R_{H}\right|>|A \cup P|-\left|R_{\bar{H}}\right|-\left|I_{\bar{H}}\right|=\left|T_{\bar{H}}\right| .
$$

Finally, observe that $\left(N_{G}^{\text {out }}\left(I_{H}\right) \cap A\right) \subseteq\left(N_{G^{\text {conf }}}\left(I_{H}\right) \cap A\right) \subseteq T_{\bar{H}}$, which yields the desired bound.

Step II: Expansion of Subsets of $H$ in $G$. Similar to the outline, the second step of the proof is to observe that most subsets $X \subseteq H$ expands very well into $A$ (or $V$ ), with respect to out-edges in $G$. The reason is simply that the graph from $H$ to these sets are (essentially) random bipartite graphs of out-degree $\Omega(d)$. Due to technical reasons, we will also state the vertex expansion properties in terms of the in-degree graphs to $V$. The formal statement is as follows.

Lemma 4. Suppose that $|A| \geqslant 0.15(n-1)$. Then, with probability $1-O\left(e^{-\beta d}\right)$, the following holds:

1. For any set $X \subseteq H$ such that $|X| \leqslant 0.05(n-1)$, we have

$$
\begin{equation*}
\left|A \cap N_{G}^{\text {out }}(X)\right| \geqslant|X| \tag{7}
\end{equation*}
$$

2. For any set $W \subseteq V$, we have

$$
\begin{equation*}
\left|H \cap N_{G}^{i n}(W)\right| \geqslant \min \{|H|-0.05(n-1), 10|W|\} . \tag{8}
\end{equation*}
$$

We remark here that the above lemma is the main place we use $|A| \geqslant \Omega(n)$ as guaranteed from Step 0; otherwise the inequality (7) may not be true (some vertex in $H$ might not even have an outgoing edge to $A$ at all if $A$ is too small).

The proof is via a standard approach to prove vertex/edge expansion of graph: we bound the probability that each neighbor is a subset of a too-small set and then use union bound in the end.
Proof. 1. Let us consider any sets $X \subseteq H$ and $Y \subseteq A$. For any vertex $u \in X$, the probability that $\left(N_{G}^{\text {out }}(u) \cap A\right) \subseteq Y$ is exactly $\left(\frac{|H|-1+|Y|}{n-1}\right)^{d}$. Since the events $N_{G}^{o u t}(u) \subseteq Y$ are independent for all $u \in X$, we have

$$
\operatorname{Pr}\left[\left(N_{G}^{o u t}(X) \cap A\right) \subseteq Y\right]=\left(\frac{|H|-1+|Y|}{n-1}\right)^{d|X|} \leqslant\left(1-\frac{|A \backslash Y|}{n-1}\right)^{d|X|}
$$

Hence, the undesired event happens with probability at most

$$
\begin{aligned}
& \sum_{\substack{X \subseteq H, Y \subseteq A \\
|X| \leqslant 0.05(n-1),|Y|=|X|-1}} \operatorname{Pr}\left[\left(N_{G}^{\text {out }}(X) \cap A\right) \subseteq Y\right] \\
\leqslant & \sum_{\substack{X \subseteq H, Y \subseteq A \\
|X| \leqslant 0.05(n-1),|Y|=|X|-1}}\left(1-\frac{|A \backslash Y|}{n-1}\right)^{d|X|} \\
\leqslant & \sum_{\substack{X \subseteq H, Y \subseteq A \\
|X| \leqslant 0.05(n-1),|Y|=|X|-1}}(0.9)^{d|X|} \\
= & \sum_{i=1}^{\lfloor 0.05(n-1)\rfloor}\binom{|H|}{i}(0.9)^{d|X|} \\
\leqslant & \sum_{i=1}^{\lfloor 0.05(n-1)\rfloor} n^{i}(0.9)^{d i} \\
\leqslant & \sum_{i=1}^{\lfloor 0.05(n-1)\rfloor} n^{-2} \cdot e^{-\beta d} \\
= & O\left(e^{-\beta d}\right),
\end{aligned}
$$

where the second inequality follows from $|A| \geqslant 0.15(n-1)$ and $|Y| \leqslant 0.05(n-$ $1)$ and we use our choice of $d \geqslant 10^{10} \log n$ in the last inequality.
2. Consider any set $W \subseteq V$ and $X \subseteq H$. We will bound the probability that $\left(H \cap N_{G}^{i n}(W)\right)$ is a subset of $X$. Recall from our definition that every vertex has a self-loop. Hence, if $(W \cap H) \nsubseteq X$, it is immediate that $\operatorname{Pr}\left[\left(H \cap N_{G}^{i n}(W)\right) \subseteq X\right]=0$.
Now, for the case $(W \cap H) \subseteq X$, we can bound $\operatorname{Pr}\left[\left(H \cap N_{G}^{i n}(W)\right) \subseteq X\right]$ as follows. First, notice that each vertex $u \in(H \backslash X), u$ does not belongs to $N_{G}^{i n}(W)$ (or equivalently $N_{G}^{o u t}(u) \cap W=\emptyset$ ) with probability exactly $\left(1-\frac{|W|}{n-1}\right)^{d}$. Since the events $u \notin N_{G}^{i n}(W)$ are independent for all $u \in(H \backslash X)$, we have

$$
\operatorname{Pr}\left[\left(H \cap N_{G}^{i n}(W)\right) \subseteq X\right]=\left(1-\frac{|W|}{n-1}\right)^{d \cdot|H \backslash X|} \leqslant e^{-\frac{d|W| \cdot(|H|-|X|)}{n-1}}
$$

For convenience, let $\mu(|W|)$ denote $\lceil\min \{|H|-0.05(n-1), 10|W|\}\rceil-1$. From union bound and the previous inequality, the probability of the undesired event is at most

$$
\begin{aligned}
\sum_{\substack{W \subseteq V, X \subseteq H \\
|X|=\mu(|\bar{W}|)}} \operatorname{Pr}\left[\left(H \cap N_{G}^{i n}(W)\right) \subseteq X\right] & \leqslant \sum_{\substack{W \subseteq V, X \subseteq H \\
|X|=\mu(|\bar{W}|)}} e^{-\frac{d \cdot|W| \cdot(| | H|-|X|)}{n-1}} \\
& \leqslant \sum_{\substack{W \subseteq V, X \subseteq H \\
|X|=\mu| | \bar{W} \mid)}} e^{-0.05 d|W|} \\
& =\sum_{i=1}^{n}\binom{n}{i}\binom{|H|}{\mu(i)} e^{-0.05 d i} \\
& \leqslant \sum_{i=1}^{n} n^{11 i} e^{-0.05 d i} \\
& \leqslant \sum_{i=1}^{n} n^{-2} \cdot e^{-\beta d} \\
& =O\left(e^{-\beta d}\right)
\end{aligned}
$$

where the second inequality is due to $\mu(|W|) \leqslant|H|-0.05(n-1)$ and the last inequality follows from $d \geqslant 10^{10} \log n$. This completes our proof.

We can deduce from the above lemma the following corollary, which will be more convenient to use in the main proof.

Corollary 1. Suppose that $|A| \geqslant 0.15(n-1)$. Then, with probability $1-O\left(e^{-\beta d}\right)$, for any set $X \subseteq H$, at least one of the following must hold: (i) $|X| \leqslant \mid N_{G}^{o u t}(X) \cap$ $A \mid$ or $($ ii $)\left|N_{G}^{\text {out }}(X) \cap(H \cup P)\right| \geqslant \frac{n+1}{2}+2 \cdot\left|A \backslash N_{G}^{\text {out }}(X)\right|$.
Proof. Suppose for the sake of contradiction that there exists $X \subseteq H$ that violates both inequalities. Since $X$ violates (i) and from the first item (i.e. (7)) of Lemma 4, we must have $|X|>0.05(n-1)$, which means

$$
\begin{equation*}
|H \backslash X|<|H|-0.05(n-1) \tag{9}
\end{equation*}
$$

Let $Y=\left(A \backslash N_{G}^{\text {out }}(X)\right)$ and $k=|Y|$. From the violation of (i), we have

$$
|X| \geqslant\left|N_{G}^{\text {out }}(X) \cap A\right|+1=|A|-k+1=(n-|H|-|P|)-k+1
$$

From the above, we have

$$
|H \backslash X|=|H|-|X| \leqslant|P|+k+2\left(|H|-\frac{n+1}{2}\right) .
$$

For convenience, we let $\Gamma=|H|-\frac{n+1}{2}$. We may write the above inequality as

$$
\begin{equation*}
|H \backslash X| \leqslant|P|+k+2 \Gamma \tag{10}
\end{equation*}
$$

Now, from (8) in Lemma 4 with $W=Y$, we have

$$
|H \backslash X| \geqslant\left|N_{G}^{i n}(Y) \cap H\right| \geqslant \min \{|H|-0.05(n-1), 10 k\} .
$$

From (9), it cannot be that $|H \backslash X| \geqslant|H|-0.05(n-1)$. As a result, we have

$$
\begin{equation*}
|H \backslash X| \geqslant 10 k \tag{11}
\end{equation*}
$$

Similarly, let $Z=(H \cup P) \backslash N_{G}^{o u t}(X)$. From the violation of (ii), we have

$$
\begin{align*}
|Z|=|H|+|P|-\left|N_{G}^{\text {out }}(X) \cap(H \cup P)\right| & \geqslant|H|+|P|-\frac{n-1}{2}-2 k \\
& =|P|-2 k+1+\Gamma \tag{12}
\end{align*}
$$

Moreover, from (8) in Lemma 4 with $W=Z$, we have

$$
|H \backslash X| \geqslant\left|N_{G}^{i n}(Z) \cap H\right| \geqslant \min \{|H|-0.05(n-1), 10|Z|\}
$$

Once again, (9) implies that $|H \backslash X|$ cannot be at least $|H|-0.05(n-1)$. Hence, we have

$$
\begin{equation*}
|H \backslash X| \geqslant 10|Z| \stackrel{(12)}{\geqslant} 10(|P|-2 k+1+\Gamma) . \tag{13}
\end{equation*}
$$

By combining (10), (11) and (13), we arrive at

$$
\begin{aligned}
|P|+k+2 \Gamma \stackrel{(10)}{\geqslant}|H \backslash X| & =0.8|H \backslash X|+0.2|H \backslash X| \\
& \left.\begin{array}{l}
(11),(13) \\
\\
\\
\end{array}\right)=2|P|+4 k+2(|P|-2 k+1+\Gamma) \\
& =2 \Gamma
\end{aligned}
$$

a contradiction.

Step III: Putting things together. With the above lemmas ready, we now prove our main theorem by simply plugging them together.
Proof of Theorem 3: From Lemma 1, if $|A|<0.15(n-1)$, then we output the entire vertex set $V$ and terminates with probability $1-O\left(e^{-\beta d}\right)$; this is a correct output. Moreover, Lemma 1 also ensures that we w.p. $1-O\left(e^{\beta d}\right)$ do not terminate here with an incorrect output.

We may now assume for the rest of the proof that $|A| \geqslant 0.15(n-1)$. From Lemma 2, we have $\left|I_{H}\right|>\left|N_{G}^{\text {out }}\left(I_{H}\right) \cap A\right|$. As a result, from Corollary 1, we with probability $1-O\left(e^{-\beta d}\right)$ must have

$$
\begin{equation*}
\left|N_{G}^{\text {out }}\left(I_{H}\right) \cap(H \cup P)\right| \geqslant \frac{n+1}{2}+2 \cdot\left|A \backslash N_{G}^{\text {out }}\left(I_{H}\right)\right| \tag{14}
\end{equation*}
$$

Now, observe that, every vertex in $N_{G}^{o u t}\left(I_{H}\right) \cap(H \cup P)$ is labelled good by at least one vertex in $I_{H} \subseteq I$; hence, they will be included in $S$. In other words, $S \cap(H \cup P) \supseteq N_{G}^{\text {out }}\left(I_{H}\right) \cap(H \cup P)$.

On the other hand, all vertices in $N_{G}^{o u t}\left(I_{H}\right) \cap A$ are labelled bad by at least one vertex in $I_{H} \subseteq I$; hence, they will not be included in $S$. In other words, we have $S \cap A \subseteq\left(A \backslash N_{G}^{o u t}\left(I_{H}\right)\right)$.

As a result, we arrive at

$$
\begin{aligned}
|S \cap(H \cup P)| \geqslant\left|N_{G}^{\text {out }}\left(I_{H}\right) \cap(H \cup P)\right| & \stackrel{(14)}{\geqslant} \frac{n+1}{2}+2 \cdot\left|A \backslash N_{G}^{\text {out }}\left(I_{H}\right)\right| \\
& \geqslant \frac{n+1}{2}+2 \cdot|S \cap A|
\end{aligned}
$$

which concludes our proof.

## 4 Construction of Robust Secret Sharing

Let $\lambda$ be the robustness parameter. In this section, we give a construction of robust secret sharing for messages of length $m$ with share size $m+O(\lambda \log n(\log n+$ $\log m)$ ). In Section 4.1, we first give a construction of a basic robust secret sharing scheme with share size $m+O\left(\lambda^{2}+\lambda(\log n+\log m)+\log ^{2} n+\log n \log m\right)$. In the Section 4.2, we use parallel repetition (as was done by Bishop et al [BPRW16]) to improve the share size of our construction to $m+O(\lambda \log n(\log n+\log m))$.

### 4.1 Basic Scheme

The construction is described in Figure 1 and we show the following theorem. ${ }^{10}$
Theorem 10. For some $t, d, \lambda, \rho \in \mathbb{N}, \epsilon_{2}, \epsilon_{3} \in[0,1], n \geqslant 2 t+1$, and message space $\mathcal{M}$, assume we have the following:
$-A(t+1, n, 0,0)$ secret sharing scheme (Share, Rec) for $\mathcal{M}$ that is $(0, s, L(s-$ $(t+1)) / 2\rfloor)$-robust for any $s \geqslant t+1$. Further, the shares are strings in $\{0,1\}^{\rho}$.

[^6]- A $\left(2 d, \epsilon_{2}\right)$-secure private MAC scheme (KeyGen, Tag, Verify) for message space $\{0,1\}^{\rho}$ with randomness length $\nu$, key length $\kappa$ and tag length $\lambda$.
- The vertex identification algorithm VertexID from Theorem 3, when run on graphs with out-degree d, has error probability at most $\epsilon_{3}$.

Then, the construction in Figure 1 is a $\left(t+1, n, n e^{-d / 3}, 0\right)$ secret sharing scheme for $\mathcal{M}$ that is $\left(n d \epsilon_{2}+\epsilon_{3}+n e^{-d / 3}, n, t\right)$-robust. The size of each share is $(\rho+\nu+$ $d\lceil\log n\rceil+d(\kappa+\lambda))$ bits.

BasRobShare $(m)$ : To share a secret $m \in \mathcal{M}$ :

1. For each $i \in[n]$, select a multiset $S^{i} \subseteq[n] \backslash\{i\}$ of size $d$ uniformly at random with replacement.
2. If there is an $i^{\prime} \in[n]$ such that the multiplicity of $i^{\prime}$ in $\left(S^{1}\left\|S^{2}\right\| \ldots \| S^{n}\right)$ is greater than $2 d$, select $m^{*} \leftarrow \mathcal{M}$. Else, set $m^{*}=m$.
3. Run Share $\left(m^{*}\right)$ to obtain the shares $\left(\mathrm{Sh}_{1}, \ldots, \mathrm{Sh}_{n}\right)$.
4. For each $i \in[n]$, choose $r_{i} \leftarrow\{0,1\}^{\nu}$.
5. For each $i \in[n]$ and each $j \in[d]$ do:
(a) Let $v_{j}^{i}$ be the $j$-th element in the ordered multiset $S^{i}$.
(b) Choose a private MAC key $k_{i \rightarrow v_{j}^{i}}^{j} \leftarrow \operatorname{KeyGen}\left(1^{\lambda}\right)$.
(c) Compute $\sigma_{i \rightarrow v_{j}^{i}}^{j} \leftarrow \operatorname{Tag}\left(k_{i \rightarrow v_{j}^{i}}^{j},\left(\mathrm{Sh}_{v_{j}^{i}}, r_{v_{j}^{i}}\right)\right)$.
6. Set $\operatorname{share}_{i}=\left(\mathrm{Sh}_{i}, r_{i}, S^{i},\left\{k_{i \rightarrow v_{j}^{i}}^{j}, \sigma_{i \rightarrow v_{j}^{i}}^{j}\right\}_{j \in[d]}\right)$.

BasRobRec: The reconstruction algorithm proceeds in two rounds.

## - Round-1:

1. For each $i \in[n]$, party $i$ broadcasts $\left(\mathrm{Sh}_{i}, r_{i}\right)$ to every other party and initializes an empty list $N_{i}$.
2. For every $j \in[d]$, party $i$ does:
(a) Let $v_{j}^{i}$ be the $j$-th element in the ordered multiset $S^{i}$.
(b) Check if $\operatorname{Verify}\left(k_{i \rightarrow v_{j}^{i}}^{j},\left(\mathrm{Sh}_{v_{j}^{i}}, r_{v_{j}^{i}}\right), \sigma_{i \rightarrow v_{j}^{i}}^{j}\right)=1$.
(c) If the verification passes, party $i$ adds $\left(\left(i, v_{j}^{i}\right)\right.$, good) to $N_{i}$. Else, it adds $\left(\left(i, v_{j}^{i}\right), \mathrm{bad}\right)$ to $N_{i}$.

- Round-2:

1. For each $i \in[n]$, party $i$ broadcasts $N_{i}$ to every other party and initializes an empty graph $G$ and an empty labeling $L$.
2. For every $i \in[n]$ and every entry $\left((i, v)\right.$, lab $\left._{i, v}\right) \in N_{i}$, the parties add the edge $(i, v)$ in $G$ and set $L(i, v)=\operatorname{lab}_{i, v}$.
3. The parties (locally) run the algorithm VertexID on $G$ and $L$ to obtain the vertex set $S$.
4. The parties then (locally) run $\operatorname{Rec}\left(\left\{\mathrm{Sh}_{i}\right\}_{i \in S}\right)$ to obtain the secret $m$.

Fig. 1. Basic Construction of Robust Secret Sharing (using terminology from Theorem 10)

Proof of Theorem 10: The share size may be verified by inspection. We first show the correctness and privacy properties of our construction and finally show its robustness.

Correctness. Note that in an honest execution of the sharing and the reconstruction algorithm, in the absence of any corruptions, every edge in the graph $G$ will be labeled as good. Thus, the algorithm VertexID will output $V=[n]$. It now follows from the perfect correctness of Rec that the secret output by BasRobRec will be equal to $m^{*}$ with probability 1 . We will now bound the probability that $m^{*}$ is not equal to $m$.

We estimate the probability that the chosen multisets $\left\{S^{i}\right\}_{i \in[n]}$ has the property such that for $v \in[n]$, the multiplicity of $v$ in $S^{1}\|\ldots\| S^{n}$ is at most $2 d$. Let us fix a party $v \in[n]$, and call the event that its multiplicity is more than $2 d$ as $\operatorname{Bad}_{v}$. For any $i \neq v, v$ might get selected (possibly multiple times) in the multiset $S^{i}$. Thus, there are totally $d(n-1)$ random draws where $v$ might get selected. For any $i \in[d(n-1)]$, let $X_{i}$ be the indicator random variable which is 1 if and only if $v$ is selected in the $i$-th draw. Now, for any $i \in[d(n-1)]$,

$$
\begin{equation*}
\operatorname{Pr}\left[X_{i}=1\right]=\frac{1}{n-1} \tag{15}
\end{equation*}
$$

Then, $\operatorname{Bad}_{v}$ occurs if $\sum_{i} X_{i}>2 d$. The variables $\left\{X_{i}\right\}_{i}$ are independent and hence from Chernoff bounds,

$$
\begin{equation*}
\operatorname{Pr}\left[\operatorname{Bad}_{v}\right]=\operatorname{Pr}\left[\sum_{i} X_{i}>2 d\right] \leqslant e^{-d / 3} \tag{16}
\end{equation*}
$$

Let Bad be the event that there exists at least one $v \in[n]$ such that $\operatorname{Bad}_{v}$ happens. By union bound,

$$
\begin{equation*}
\operatorname{Pr}[\mathrm{Bad}] \leqslant \sum_{v} \operatorname{Pr}\left[\operatorname{Bad}_{v}\right]=n e^{-d / 3} \tag{17}
\end{equation*}
$$

Thus, with probability at least $1-n e^{-d / 3}$, the chosen multisets $\left\{S^{i}\right\}_{i \in[n]}$ satisfies the property the multiplicity of every $v \in[n]$ in $S^{1}\|\ldots\| S^{n}$ is at most $2 d$. Notice that when this happens, $m^{*}=m$. Thus, the correctness error is at most $n e^{-d / 3}$. We will call $\left\{S^{i}\right\}_{i \in[n]}$ that satisfies the above property to be bounded.

Privacy. To show perfect privacy, we need to argue that for every set $U$ of size at most $t$, for every pair of secrets $m_{0}, m_{1} \in M$, the distributions of BasRobShare $\left(m_{0}\right)_{U}$ and BasRobShare $\left(m_{1}\right)_{U}$ are identical. ${ }^{11}$ It is sufficient to show that this is in the case when $\left\{S^{i}\right\}$ is bounded, and when it is not bounded, by design the output of BasRobShare is independent of the message being shared, and so the shares of $U$ are identical.

When $\left\{S^{i}\right\}$ is bounded, we show this by the following hybrid argument. Fix any choice of the multisets $\left\{S^{i}\right\}_{i \in[n]}$ such that it is bounded. For every $i \in[n]$,

[^7]we define $T^{i}$ to be the sequence of $(v, j) \in[n] \times[d]$ such that the $j$-th entry of the multiset $S^{v}$ is equal to $i$. From the choice of the fixed $\left\{S^{i}\right\}_{i \in[n]}$, each $\left|T^{i}\right| \leqslant 2 d$. We first fix all the MAC keys chosen during the share phase, and make the following argument for any such set of keys. We now argue that BasRobShare $\left(m_{0}\right)_{U}$ is identical to BasRobShare $\left(m_{1}\right)_{U}$, going through the following hybrid distributions over the shares of $U$ :
$-\mathrm{Hyb}_{1}$ : This is the same as BasRobShare $\left(m_{0}\right)_{U}$.
$-\mathrm{Hyb}_{2}$ : In this hybrid, during the share phase, for every $i \notin U$, we generate $\sigma_{v \rightarrow i}^{j}$ (for any $j \in[d]$ and $\left.v \in[n]\right)$ as $\operatorname{Tag}\left(k_{v \rightarrow i}^{j},\left(0^{\rho}, r_{i}\right)\right)$. We finally output the shares corresponding to $U$. Note that $\mathrm{Hyb}_{1}$ is identical to $\mathrm{Hyb}_{2}$ from the perfect privacy property of the private MAC scheme, which is $\left(2 d, \varepsilon_{2}\right)$-secure, since $\left|T^{i}\right| \leqslant 2 d$ for every $i \in[n]$.
$-\mathrm{Hyb}_{3}$ : In this hybrid, during the share phase, instead of running Share ( $m^{*}$ ) to get the $\mathrm{Sh}_{i}$ 's, we run Share(0). ${ }^{12}$ We run the rest of the sharing normally and output the shares corresponding to $U . \mathrm{Hyb}_{3}$ is identical to $\mathrm{Hyb}_{2}$ by the perfect privacy of the secret sharing scheme (Share, Rec).

Note that via a similar argument we can show that $\operatorname{BasRobShare}\left(m_{1}\right)_{U}$ is also identical to $\mathrm{Hyb}_{3}$, and thus to BasRobShare $\left(m_{0}\right)_{U}$, proving perfect privacy.

Robustness. We now argue the robustness of our construction. Consider an interactive adversary $\mathcal{A}$ that adaptively corrupts a set $T$ of parties. We assume without loss of generality that this adversary corrupts $\lfloor(n-1) / 2\rfloor$ number of parties. In the case where the adversary corrupts less than this many parties, we consider another adversary that corrupts all the parties corrupted by the original adversary and corrupts some additional parties such that the total number of corrupted parties is $\lfloor(n-1) / 2\rfloor$. For the parties corrupted by the original adversary, this new adversary behaves exactly as specified by the original adversary. For the additional corrupted parties, this new adversary follows the reconstruction protocol as specified. The adversary is given $\left\{\text { share }_{i}\right\}_{i \in T}$. Let $H$ be the set of honest parties. Consider the interactive reconstruction algorithm.

- In the first round of the reconstruction algorithm, the party $i$ broadcasts $\left(\mathrm{Sh}_{i}, r_{i}\right)$ to every other party. Now, every party $i \in T$ might broadcast the correct $\left(\mathrm{Sh}_{i}, r_{i}\right)$ or a modified $\left(\mathrm{Sh}_{i}^{\prime}, r_{i}^{\prime}\right)$. Based on this, we partition $T$ into two sets $P$ and $A$. $P$ consists of the parties that send the unmodified $\left(\mathrm{Sh}_{i}\right.$, $\left.r_{i}\right)$ whereas the parties in $A$ modify the shares and send $\left(\mathrm{Sh}_{i}^{\prime}, r_{i}^{\prime}\right) \neq\left(\mathrm{Sh}_{i}, r_{i}\right)$. Note that at the end of the first round, the adversarial parties learn no information about the multisets $S^{i}$ of the honest parties and conditioned on the information available to the adversary at the end of the first round, these multisets $S^{i}$ are still random.
- At the end of the first round, the parties will verify the tags of the MAC. Every $i \in H$ will generate the set $N_{i}$ as follows:
- Let $S^{i}=\left(v_{1}^{i}, \ldots, v_{d}^{i}\right)$.

[^8]- For every $j \in[d]$ such that $v_{j}^{i} \in H \cup P$, party $i$ adds $\left(\left(i, v_{j}^{i}\right)\right.$, good) to $N_{i}$.
- For every $j \in[d]$ such that $v_{j}^{i} \in A$, party $i$ adds $\left(\left(i, v_{j}^{i}\right)\right.$, bad) to $N_{i}$ except with probability at most $\epsilon_{2}$. The $\epsilon_{2}$ error probability follows directly from the $\varepsilon_{2}$-unforgeability of the private MAC.
By standard union bound, the probability that there exists an $i \in H$ such that for some $v \in A$, party $i$ adds $\left((i, v)\right.$, good) to $N_{i}$ is at most $n d \epsilon_{2}$ since each multiset $S^{i}$ has size $d$.
- Conditioned on the above event not happening, the graph $G=(V, E)$ with $V=[n]$ and the edge labeling $L$ is effectively generated as follows.

1. The adversary partitions $V$ into $H, P, A$ where $|H|=n-\lfloor(n-1) / 2\rfloor \geqslant$ $(n+1) / 2$.
2. For every $u \in H$, choose a multiset $S^{u}$ uniformly at random from $[n] \backslash\{u\}$ with replacement and let $S^{u}=\left(v_{1}^{u}, \ldots, v_{d}^{u}\right)$. For every $j \in[d]$, add an edge $\left(u, v_{j}^{u}\right)$ and set $L\left(u, v_{j}^{u}\right)=$ good if and only if $v_{j}^{u} \in H \cup P$. This is identically distributed to the distribution where we choose $S^{u}$ uniformly at random during the sharing phase since at the end of the first round, the adversary learns no information about the multisets of the honest parties.
3. The outgoing edges and their labels of $A \cup P$ can be generated adversarily after looking at the outgoing edges and the labels of the vertices in $H$.

This is exactly same as the graph generation procedure given in Section 3.

- It now follows from the correctness of the VertexID algorithm (Theorem 3) that its output $S$ when run on this graph satisfies the property that $\mid S \cap(H \cup$ $P)|\geqslant(n+1) / 2+2 \cdot| S \cap A|\geqslant(t+1)+2 \cdot| S \cap A \mid$ except with probability $\epsilon_{3}$. The fact that $\operatorname{Rec}\left(\left\{\mathrm{Sh}_{j}\right\}_{j \in S}\right)=m^{*}$ follows from robustness of secret sharing (Fact 2).
- Finally, as in the correctness argument, $m^{*}$ is equal to the actual secret $m$ except with probability $n e^{-d / 3}$. Thus, by the union bound, the probability of error of the whole reconstruction procedure is at most $\left(n d \varepsilon_{2}+\varepsilon_{3}+n e^{-d / 3}\right)$.

This completes the proof of the theorem.

Remark 2. While they are not explicitly covered by the discussion so far, our construction extends to the case of aborting adversaries in a straightforward manner. If an adversary does not send a message in the reconstruction phase then it must be a corrupted party. In this case, all the honest parties which have this party in its watchlist will mark the corresponding edge as being bad. Further, the parties will consider some default set of $d$ parties, (say the first $d$ parties) as being part of this corrupted party's watchlist and consider some default labeling of the edges. Notice that our vertex identification algorithm allows the watchlist and the labeling from the malicious parties to be arbitrary.

Instantiation We now provide the following instantiation of the building blocks of our robust secret sharing scheme. Let us fix the robustness parameter $\lambda$, and the length $m$ of the secret to be shared (if $m<\lceil\log n\rceil$, replace it with $\lceil\log n\rceil$ in the following).

- We set $d=10 \lambda / \beta+3 C \log n / \beta$, where $C(>1)$ and $\beta(\in(0,1))$ are constants from Theorem 3.
- We instantiate the secret sharing with Shamir secret sharing over GF[ $\left.2^{m}\right]$. This gives $\rho=m$.
- We instantiate the private MAC with the $\left(\ell, \varepsilon_{2}\right)$-secure MAC for message space $\{0,1\}^{\rho}$ from Theorem 1 , with $\ell=2 d$ and $\epsilon_{2}=2^{-\lambda} /(2 n d)$. The randomness, tag and key lengths are, respectively, $2 d(\lambda+\log (2 n d(\rho+2 d)))$, $\lambda+\log (2 n d(\rho+2 d))$, and $2(\lambda+\log (2 n d(\rho+2 d)))$.
- The VertexID algorithm has error probability $\epsilon_{3}<2^{-10 \lambda}$.

The robustness error is $n d \epsilon_{2}+\epsilon_{3}+n e^{-d / 3}<2^{-\lambda} / 2+2^{-10 \lambda}+n 2^{-3 \lambda-\log n} \leqslant$ $2^{-\lambda}$. The correctness error is $n e^{-d / 3} \leqslant 2^{-\lambda}$. The size of each share is $\rho+2 d(\lambda+$ $\log (2 n d(\rho+2 d)))+d \log n+3 d(\lambda+\log (2 n d(\rho+2 d)))=m+5 d \lambda+O(d \log (2 n d(m+$ $2 d))=m+O\left(\lambda^{2}\right)+O(\lambda \log n)+O((\lambda+\log n)(\log n+\log \lambda+\log m))=m+$ $O\left(\lambda^{2}+\lambda(\log n+\log m)+\log ^{2} n+\log n \log m\right)$.

Corollary 2. For any $\lambda, t, m, n \in \mathbb{N}$ with $n \geqslant 2 t+1$, there exists a $(t+1, n$, $\left.2^{-\lambda}, 0\right)$-secret sharing scheme that is $\left(2^{-\lambda}, n, t\right)$-robust and, for secrets of length $m$ bits, has shares of size $m+O\left(\lambda^{2}+\lambda(\log n+\log m)+\log ^{2} n+\log n \log m\right)$ bits.

### 4.2 Improved Parameters via Parallel Repetition

In this subsection, we improve the share size of our basic construction to $m+$ $O(\lambda \log n(\log n+\log m))$ to achieve robustness error of $2^{-\lambda}$ via parallel repetition. This is similar to the ideas explained in [BPRW16]. Before we describe the construction, we start with some notation.

Notation. We split BasRobRec into two steps. The first step BasRobRec ${ }_{1}$ is an interactive protocol comprising of the first two rounds of BasRobRec and the output of the protocol is the set $S$ which is the output of VertexID algorithm on the constructed graph $G$ and the labeling $L$. The second step consists of running Rec on $\left\{\mathrm{Sh}_{i}\right\}_{i \in S}$ and outputting the message.

Construction. The construction of robust secret sharing (RobShare, RobRec) with improved parameters is described in Figure 2.

Theorem 11. For any $\lambda, t, m, n \in \mathbb{N}$, with $n \geqslant 2 t+1$, the construction in Figure 2 is a $(t+1, n, 0,0)$ secret sharing scheme (with expected polynomial time sharing algortithm) for secrets of length $m$ that is $\left(e^{-\lambda / 24}, n, t\right)$-robust. The size of each share is $m+O(\lambda \log n(\log n+\log m))$.

Due to page limits, we defer the proof of this theorem to the full version of the paper.

Set $\lambda^{\prime}=2$, and $d=10 \lambda^{\prime} / \beta+3 C \log n / \beta$, where $C$ and $\beta$ are the constants from Theorem 3. In the following, (Share, Rec) represents $(t+1, n, 0,0)$ Shamir secret sharing over GF $\left[2^{m}\right]$. The private MAC (KeyGen, Tag, Verify) used for the message space $\{0,1\}^{m}$ is the $(\ell, \varepsilon)$-secure MAC from Theorem 1 with $\ell=2 d$ and $\varepsilon=2^{-\lambda^{\prime}} / 2 n d$. The protocol BasRobRec is from the construction in Figure 1.

RobShare $(m)$ : To share a secret $m \in\{0,1\}^{m}$ :

1. Run Share $(m)$ to obtain the shares $\left(\mathrm{Sh}_{1}, \ldots, \mathrm{Sh}_{n}\right)$.
2. For each $q$ in 1 to $\lambda$ do:
(a) For each $i \in[n]$, select a multiset $S^{q, i} \subseteq[n] \backslash\{i\}$ of size $d$ uniformly at random with replacement.
(b) If there is an $i^{\prime} \in[n]$ such that the multiplicity of $i^{\prime}$ in $S^{q, 1}\left\|S^{q, 2}\right\| \ldots \| S^{q, n}$ is greater than $2 d$, then go back to step (2a).
3. For each $q \in[\lambda], i \in[n]$ and each $j \in[d]$ do:
(a) For each $i \in[n]$, choose $r_{i}^{q} \leftarrow\{0,1\}^{\nu}$.
(b) Let $v_{j}^{q, i}$ be the $j$-th element in the ordered multiset $S^{q, i}$.
(c) Choose a private MAC key $k_{i \rightarrow v_{j}^{q, i}}^{q, j} \leftarrow$ KeyGen.
(d) Compute $\sigma_{i \rightarrow v_{j}^{q, i}}^{q, j} \leftarrow \operatorname{Tag}\left(k_{i \rightarrow v_{j}^{q, i}}^{q, j},\left(\mathrm{Sh}_{v_{j}^{q, i}}, r_{v_{j}^{q, i}}^{q}\right)\right)$.
4. Set share ${ }_{i}=\left(\operatorname{Sh}_{i},\left\{r_{i}^{q}\right\}_{q \in[k]},\left\{S^{q, i}\right\}_{q \in[k]},\left\{k_{i \rightarrow v_{j}^{q, i}}^{q, j}, \sigma_{i \rightarrow v_{j}^{q, i}}^{q, j}\right\}_{j \in[d], q \in[k]}\right)$.

## RobRec:

1. For each $q$ in 1 to $\lambda$ do in parallel:
(a) Run $\operatorname{BasRobRec}_{1}\left(\left\{\operatorname{Sh}_{i}, r_{i}^{q}, S^{q, i},\left\{k_{i \rightarrow v_{j}^{q, i}}^{q, j}, \sigma_{i \rightarrow v_{j}^{q, i}}^{q, j}\right\}_{j \in[d]}\right\}_{i \in[n]}\right)$ to obtain the set $\Gamma_{q}$.
(b) Set $m_{q}:=\operatorname{Rec}\left(\left\{\operatorname{Sh}_{i}\right\}_{i \in \Gamma_{q}}\right)$.
2. If there is a majority value $m$ in the sequence $\left(m_{1}, \ldots, m_{\lambda}\right)$ then output $m$. Else, output $\perp$.

Fig. 2. Improved Construction of Robust Secret Sharing

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[^0]:    ${ }^{3}$ Note that if the adversary is allowed to modify any more shares, then correct reconstruction would be impossible.

[^1]:    ${ }^{4}$ Throughout this work, we will be concerned only with threshold secret sharing, and thus we leave out this specification hereafter.

[^2]:    ${ }^{5}$ Throughout the introduction, we use $\tilde{O}$ to hide polylogarithmic factors in $\lambda, n$, and $m$.

[^3]:    ${ }^{6}$ We note that our graph allows multi-edges and self-loops.

[^4]:    ${ }^{7}$ Of course, the conflict graph depends on the labeling. However, we choose not to have $L$ in the notation to avoid cumbersomeness.
    ${ }^{8}$ Equivalently, the solution must be a vertex of the polytope defined by the constraints.

[^5]:    ${ }^{9}$ The constant 0.15 here can be replaced with any constant less than $1 / 6$. We only use 0.15 to avoid introducing an additional parameter.

[^6]:    ${ }^{10}$ Recall the definition of multiplicity of a multiset and $\|$ from Section 2.

[^7]:    ${ }^{11}$ From Remark 1, we also satisfy the stronger notion of adaptive privacy.

[^8]:    ${ }^{12} 0$ denoting some universally fixed element in $\mathcal{M}$

