# Partial Fairness in Secure Two-Party Computation 

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#### Abstract

Complete fairness is impossible to achieve, in general, in secure two-party computation. In light of this, various techniques for obtaining partial fairness in this setting have been suggested. We explore the possibility of achieving partial fairness with respect to a strong, simulation-based definition of security within the standard real/ideal world paradigm. We show feasibility with respect to this definition for randomized functionalities where each player may possibly receive a different output, as long as at least one of the domains or ranges of the functionality are polynomial in size. When one of the domains is polynomial size, our protocol is also secure-with-abort. In contrast to much of the earlier work on partial fairness, we rely on standard assumptions only (namely, enhanced trapdoor permutations).

We also provide evidence that our results are, in general, optimal. Specifically, we show a boolean function defined on a domain of super-polynomial size for which it is impossible to achieve both partial fairness and security with abort, and provide evidence that partial fairness is impossible altogether for functions whose domains and ranges all have super-polynomial size.


## 1 Introduction

In the setting of secure two-parry computation, two parties wish to run some protocol that will enable each of them to learn a (possibly different) function of their inputs while preserving, to the extent possible, security properties such as privacy, correctness, input independence, etc. These requirements, and more, are typically formalized by comparing a real-world execution of the protocol to an ideal world where there is a trusted entity who performs the computation on behalf of the parties. Informally, a given protocol is said to be "secure" if for any real-world adversary $\mathcal{A}$ there exists a corresponding ideal-world adversary $\mathcal{S}$ (corrupting the same party as $\mathcal{A}$ ) such that the result of executing the protocol in the real world with $\mathcal{A}$ is computationally indistinguishable from the result of computing the function in the ideal world with $\mathcal{S}$.

One desirable security property is fairness which, intuitively, means that either both parties learn the output, or else neither party does. In a "true" ideal world - and this is the ideal world used for proving security in the multi-party setting when a majority of parties are honest - fairness is ensured since the trusted party would, in fact, provide output to both parties. Unfortunately, results of Cleve [9] show that complete fairness is impossible to achieve, in general, in the two-party setting. For this reason, the usual treatment of secure two-party computation within the real/ideal world paradigm (see [17]) weakens the ideal-world model to one in which fairness is not guaranteed at all. A protocol is then defined to be "secure-with-abort" if it can be simulated (as described above) with respect to this, less-satisfying ideal-world model.

[^0]Various methods for achieving partial fairness have been suggested; we provide an extensive discussion in Section 1.2. With the exception of [16], however, all previous work has departed from the real/ideal paradigm in defining partial fairness; this deficiency is explicitly noted, for example, by Goldreich [17, Section 7.7.1.1]. Our aim is to define, and achieve, a meaningful notion of partial fairness while staying within the traditional real/ideal paradigm. Furthermore, many previously-suggested protocols only apply in certain settings (e.g., fair exchange of signatures) or under certain assumptions on the parties' inputs (e.g., that inputs are chosen uniformly at random) but do not give a "general-purpose" solution that can be used for arbitrary functions computed on arbitrary inputs. In contrast, protocols analyzed within the real/ideal paradigm do not suffer from these drawbacks. Finally, we note that much previous work on partial fairness requires strong cryptographic assumptions (e.g., regarding the precise amount of time needed to perform some computation, even using parallelism); we would prefer a solution based on standard assumptions.

As we have remarked already, the most desirable (but, in general, unachievable) definition of security requires computational indistinguishability between the real world and a "true" ideal world where both parties receive output. The standard relaxation of security-with-abort [17] leaves unchanged the requirement of computational indistinguishability, but weakens the ideal world to one in which fairness is no longer guaranteed at all. Katz [21], in a slightly different context, suggested an alternate relaxation: keep the ideal world unchanged, but relax the notion of simulation and require instead that the real and ideal worlds be distinguishable with probability at most $\frac{1}{p}+$ negl, for $p$ some specified polynomial ${ }^{1}$ (see Definition 1). We refer to a protocol satisfying this definition as being " $\frac{1}{p}$-secure". Cleve [9] and Moran et al. [24] show $\frac{1}{p}$-secure protocols for two-party coin tossing with $\mathcal{O}\left(p^{2}\right)$ and $\mathcal{O}(p)$ rounds, respectively. ${ }^{2}$ We are not aware of any other results that satisfy our definition and, in particular, none of the previous approaches for achieving partial fairness appear to yield protocols that are $\frac{1}{p}$-secure. See Section 1.2 for further discussion.

### 1.1 Our Results

We show general feasibility results for partial fairness in the two-party setting, with respect to the definition of $\frac{1}{p}$-security. Specifically, let $f_{n}: X_{n} \times Y_{n} \rightarrow Z_{n}^{1} \times Z_{N}^{2}$ be a (randomized) functionality where player 1 (resp., player 2) provides input $x \in X_{n}$ (resp., $y \in Y_{n}$ ) and receives output $z^{1} \in Z_{n}^{1}$ (resp., $z^{2} \in Z_{n}^{2}$ ). (Throughout this paper, $n$ denotes the security parameter.) Assuming the existence of enhanced trapdoor permutations we prove, for arbitrary polynomial $p$, the existence of a $\frac{1}{p}$-secure protocol for computing $f_{n}$ as long as at least one of $X_{n}, Y_{n}, Z_{n}^{1}, Z_{n}^{2}$ is polynomial size (in $n$ ). When either $X_{n}$ or $Y_{n}$ is polynomial-size, our protocol is also secure-with-abort.

As far as general feasibility results go, we also show that our results are essentially the best possible. First, we show an example of a deterministic, boolean function $f_{n}: X_{n} \times Y_{n} \rightarrow\{0,1\}$, where $\left|X_{n}\right|$ and $\left|Y_{n}\right|$ are both super-polynomial, for which no protocol computing $f_{n}$ can achieve both security-with-abort and $\frac{1}{p}$-security (for $p>4$ ) simultaneously. We also show that if exponentiallystrong one-way functions exist, then there exists a deterministic function $f_{n}: X_{n} \times Y_{n} \rightarrow Z_{n}$, with each of $X_{n}, Y_{n}, Z_{n}$ super-polynomial in size, such that $f$ cannot be $\frac{1}{p}$-securely computed for $p>2$. Taken together, our work thus settles the main open questions in this direction.

[^1]
### 1.2 Prior Work

There is an extensive literature devoted to the problem of achieving partial fairness when an honest majority is not present, both for the case of specific functionalities like coin tossing [9, 10, $24]$ and contract signing/exchanging secrets [ $5,22,13,4,11]$, as well as for the case of general functionalities $[26,15,3,19,14,6,25,16]$. Prior work (with the exception of [16]), however, does not consider a simulation-based security definition of the sort we do here. Moreover, to the best of our knowledge none of the previous approaches (with the exception of [9, 24], that deal only with coin tossing) can be proven $\frac{1}{p}$-secure.

Garay et al. [16] give a simulation-based formalization of "gradual release" [15, 3, 11, 6, 25] within the universal composability framework [8]. The guarantee their protocol provides, informally, is that at any point in the protocol both the adversary and the honest party can obtain their entire output by investing a "similar" amount of work. Somewhat unsatisfying is that the decision of whether an honest party should invest the necessary work and recover the output is not mandated by the protocol, but is somehow supposed to be decided "externally". A gradual release approach also seems problematic in defending against an adversary who runs in polynomial time, but has more computational power than honest parties are able to invest. Finally, we add that the proof of security in [16] relies on a strong, non-standard assumption regarding the precise time required to solve a specific computational problem.

Gordon et al. [20] recently showed that, in contrast to the accepted folklore, complete fairness is possible in the two-party setting for certain non-trivial functions. Work continuing that direction should be viewed as complementary to our work here: while we do not yet have a complete characterization of what can be computed with complete fairness, we know that there certainly do exist some functions that cannot be computed with complete fairness [9] and so some relaxation must be considered (at least for some functions). Note further that currently the only (non-trivial) feasibility results for complete fairness in the two-party setting [20] pertain to single-output, deterministic, boolean functions over polynomial-size domains. Our feasibility results here apply to substantially-larger classes of functions.

### 1.3 Overview of our Approach

We now give an informal description of our feasibility result. Let $x$ denote the input of $P_{1}$, let $y$ denote the input of $P_{2}$, and let $f: X \times Y \rightarrow Z$ denote the function they are trying to compute. (For simplicity, here we focus on the case when each party receives the same output; the more general case is handled when we formally describe our protocols.) As in [21, 20, 24], our protocols will be composed of two stages, where the first stage can be viewed as a "pre-processing" step and the second stage takes place in a sequence of $r=r(n)$ iterations. The stages have the following form:

First stage The first stage includes the following steps:

1. First, a value $i^{*} \in\{1, \ldots, r\}$ is chosen according to some distribution.
2. Values $a_{1}, \ldots, a_{r}$ and $b_{1}, \ldots, b_{r}$ are generated. For $i<i^{*}$, the value $a_{i}$ (resp., $b_{i}$ ) is chosen according to some distribution that is independent of $y$ (resp., $x$ ). For $i \geq i^{*}$, however, it holds that $a_{i}=b_{i}=f(x, y)$.
3. Each $a_{i}$ is randomly shared as $a_{i}^{(1)}, a_{i}^{(2)}$ with $a_{i}^{(1)} \oplus a_{i}^{(2)}=a_{i}$ (and similarly for each $b_{i}$ ). The stage concludes with $P_{1}$ being given $a_{1}^{(1)}, b_{1}^{(1)}, \ldots, a_{r}^{(1)}, b_{r}^{(1)}$, and $P_{2}$ being given
$a_{1}^{(2)}, b_{1}^{(2)}, \ldots, a_{r}^{(2)}, b_{r}^{(2)}$. (These shares are also authenticated using an information-theoretic MAC, but we omit this in the current description of the protocol.)

Note that, at the end of this stage, each party only has a set of random shares that reveal nothing about the other party's input. This stage can therefore be carried out by any twoparty protocol that is secure-with-abort.

Second stage In each iteration $i$, for $i=1, \ldots, r$, the parties do the following: First, $P_{2}$ sends $a_{i}^{(2)}$ to $P_{1}$ who reconstructs $a_{i}$; then $P_{1}$ sends $b_{i}^{(1)}$ to $P_{1}$ who reconstructs $b_{i}$. (In the real protocol, parties must also verify validity of the shares but we omit this step here.) If a party (say, $P_{1}$ ) aborts or otherwise fails to send a valid message in some iteration $i$, then the other party (here, $P_{2}$ ) outputs the value reconstructed in the previous iteration (i.e., $b_{i-1}$ ). Otherwise, parties reach the end of the protocol and output $a_{r}$ and $b_{r}$, respectively.
The above defines a generic template, but to fully specify the protocol we must specify the distribution of $i^{*}$ as well as the distribution of the $a_{i}, b_{i}$ for $i<i^{*}$. As in [21, 24], we choose $i^{*}$ uniformly from $\{1, \ldots, r\}$. (In [20] a geometric distribution was used. That would work in our context as well, but would result in worse round complexity.) For the case when $X$ and $Y$ (the domains of $f$ ) are polynomial size, we follow [20] and set $a_{i}=f(x, \hat{y})$ for $\hat{y}$ chosen uniformly from $Y$, and set $b_{i}=f(\hat{x}, y)$ for $\hat{x}$ chosen uniformly (and independently) from $X$. Note that $a_{i}$ (resp., $b_{i}$ ) is independent of $y$ (resp. $x$ ), as desired.

Intuitively, this is partially fair for the following reasons: fairness is only violated if $P_{1}$ aborts exactly in iteration $i^{*}$ (indeed, if it aborts before iteration $i^{*}$ then neither party learns the "correct" value of the function, while if it aborts subsequently then both parties learn the correct value). But even if $P_{1}$ knows the value of $z=f(x, y)$, it cannot determine when iteration $i^{*}$ occurs with certainty because $a_{i}=z$ with some noticeable probability $\alpha$ even when $i<i^{*}$. In [24] (which deals with coin tossing), the distribution of $a_{i}$ for $i<i^{*}$ is identical to the distribution of $a_{i^{*}}$, and so it is relatively straightforward to see that $P_{1}$ cannot abort in iteration $i^{*}$ except with probability at most $1 / r$, where $r$ is the number of iterations. In our case things are complicated by the fact that the distribution of $a_{i}$ for $i<i^{*}$ is different from the distribution of $a_{i^{*}}$, and furthermore the adversary may know (partial information about) the correct result $a_{i^{*}}=f(x, y)$. Nevertheless, we use a combinatorial argument (see Lemma 1) to prove that the adversary cannot abort in iteration $i^{*}$ except with probability at most $1 / \alpha r$. Taking $\alpha^{*}=\min _{z \in f(x, Y)}\left\{\operatorname{Pr}\left[a_{i}=z \mid i<i^{*}\right]\right\}$, we conclude that setting $r=p / \alpha^{*}$ suffices to achieve $\frac{1}{p}$-security. As long as $Y$ is polynomial-size, $\alpha^{*} \geq 1 /|Y|$ is noticeable and we get a protocol with polynomially-many rounds.

The above approach does not work without additional modifications in the case when $Y$ has super-polynomial size. Essentially, this is because it could be the case that $f(x, y)=z$ for some value $z$ for which the probability that $f(x, \hat{y})=z$ when $\hat{y}$ is chosen uniformly from $Y$ is negligible. In this case, conditioned on $a_{i}=z$ is it overwhelmingly likely that $i=i^{*}$, and this gives a strategy for the adversary to abort exactly in iteration $i^{*}$ with overwhelming probability. (Namely, abort in the first iteration when $a_{i}=z$.) To fix this, we must ensure that every possible output in $Z$ (the range of $f$ ) occurs with noticeable probability. We do this by changing the distribution of $a_{i}$ (for $i<i^{*}$ ) as follows: with probability $1-1 / q$ choose $a_{i}$ as before, but with probability $1 / q$ choose $a_{i}$ uniformly from $Z$. (We will set $q$ in a moment, but it will be polynomial in $n$.) Defining $\alpha^{*}=\min _{z \in Z}\left\{\operatorname{Pr}\left[a_{i}=z \mid i<i^{*}\right]\right\}$ we now have $\alpha^{*} \geq 1 / q|Z|$, which is noticeable when $|Z|$ is polynomial. Furthermore, as before, setting $r=p / \alpha^{*} \leq p q|Z|$ suffices to ensure that the adversary cannot abort in iteration $i^{*}$ except with probability at most $1 / p$.

By changing the distribution of $a_{i}$, however, we introduce a new problem: if a malicious $P_{2}$ aborts in some iteration prior to $i^{*}$, then the output of the honest $P_{1}$ in the real world cannot necessarily be simulated in the ideal world. We show, however, that it can be simulated to within statistical difference $O(1 / q)$. Taking $q=p$ (along with $r=p q|Z|$ ) thus gives a protocol with polynomially-many rounds that is $\frac{1}{p}$-secure.

We refer the reader to Section 3 for further details regarding our protocols.

### 1.4 Organization of the Paper

We present our definitions in Section 2. These are largely standard, except for our definition of $\frac{1}{p}-$ security (Definition 1). The reader may also want to quickly review our notation for functionalities. In Section 3 we describe protocols showing feasibility of $\frac{1}{p}$-secure computation for a large class of functionalities. Our impossibility results are given in Section 4.

## 2 Definitions

We denote the security parameter by $n$.

### 2.1 Preliminaries

A function $\mu(\cdot)$ is negligible if for every positive polynomial $p(\cdot)$ and all sufficiently large $n$ it holds that $\mu(n)<1 / p(n)$. A distribution ensemble $X=\{X(a, n)\}_{a \in \mathcal{D}_{n}, n \in \mathbb{N}}$ is an infinite sequence of random variables indexed by $a \in \mathcal{D}_{n}$ and $n \in \mathbb{N}$, where $\mathcal{D}_{n}$ is a set that may depend on $n$. For a fixed function $p$, two distribution ensembles $X=\{X(a, n)\}_{a \in \mathcal{D}_{n}, n \in \mathbb{N}}$ and $Y=\{Y(a, n)\}_{a \in \mathcal{D}_{n}, n \in \mathbb{N}}$ are computationally $\frac{1}{p}$-indistinguishable, denoted $X \stackrel{1 / p}{\approx} Y$, if for every non-uniform polynomial-time algorithm $D$ there exists a negligible function $\mu(\cdot)$ such that for every $n$ and every $a \in \mathcal{D}_{n}$

$$
|\operatorname{Pr}[D(X(a, n))=1]-\operatorname{Pr}[D(Y(a, n))=1]| \leq \frac{1}{p(n)}+\mu(n) .
$$

Two distribution ensembles are computationally indistinguishable, denoted $X \xlongequal{\equiv} Y$, if for every $c \in \mathbb{N}$ they are computationally $\frac{1}{n^{c}}$-indistinguishable.

The statistical difference between two distributions $X(a, n)$ and $Y(a, n)$ is defined as

$$
\mathrm{SD}(X(a, n), Y(a, n))=\frac{1}{2} \cdot \sum_{s}|\operatorname{Pr}[X(a, n)=s]-\operatorname{Pr}[Y(a, n)=s]|,
$$

where the sum ranges over $s$ in the support of either $X(a, n)$ or $Y(a, n)$. Two distribution ensembles $X=\{X(a, n)\}_{a \in \mathcal{D}_{n}, n \in \mathbb{N}}$ and $Y=\{Y(a, n)\}_{a \in \mathcal{D}_{n}, n \in \mathbb{N}}$ are statistically close, denoted $X \stackrel{\text { s }}{\equiv} Y$, if there is a negligible function $\mu(\cdot)$ such that for every $n$ and every $a \in \mathcal{D}_{n}$, it holds that $\mathrm{SD}(X(a, n), Y(a, n)) \leq \mu(n)$.

### 2.2 Two-Party Computation

Functionalities. In the two-party setting, a functionality $\mathcal{F}=\left\{f_{n}\right\}_{n \in \mathbb{N}}$ is a sequence of randomized processes for which $f_{n}$ can be computed in time $\operatorname{poly}(n)$. We view $f_{n}$ as a (randomized)
mapping $f_{n}: X_{n} \times Y_{n} \rightarrow Z_{n}^{1} \times Z_{n}^{2}$, where $X_{n}$ (resp., $Y_{n}$ ) denotes the valid inputs of the first (resp., second) party and we assume that elements of $X_{n}, Y_{n}, Z_{n}^{1}$, and $Z_{n}^{2}$ can be described by strings of length $\operatorname{poly}(n)$. We write $f_{n}=\left(f_{n}^{1}, f_{n}^{2}\right)$ if we wish to emphasize the two outputs of $f_{n}$, but stress that if $f_{n}^{1}$ and $f_{n}^{2}$ are randomized then the outputs of $f_{n}^{1}$ and $f_{n}^{2}$ are correlated random variables. If $\operatorname{Pr}\left[f_{n}^{1}(x, y)=f_{n}^{2}(x, y)\right]=1$ for all $x, y$, then we call $f_{n}$ a single-output functionality and write it as $f_{n}: X_{n} \times Y_{n} \rightarrow Z_{n}$.

In the rest of the paper, we frequently drop the explicit dependence on $n$. It is important to keep in mind, however, that the domain and range of $f$ depend $n$, since the complexity of our protocols will depend on their sizes.

Two-party computation. A two-party protocol for computing a functionality $\mathcal{F}=\left\{\left(f^{1}, f^{2}\right)\right\}$ is a protocol running in polynomial time and satisfying the following functional requirement: if party $P_{1}$ begins by holding $1^{n}$ and input $x \in X$, and party $P_{2}$ holds $1^{n}$ and input $y \in Y$, then the joint distribution of the outputs of the parties is statistically close to $\left(f^{1}(x, y), f^{2}(x, y)\right)$.

In what follows, we define what we mean by a secure protocol. Our definition uses the standard real/ideal paradigm of [17] (based on $[23,2,7]$ ), except that we will sometimes require only $\frac{1}{p}$-indistinguishability rather than indistinguishability. We consider active adversaries, who may deviate from the protocol in an arbitrary manner, and static corruptions.

Security of protocols (informal). The security of a protocol is analyzed by comparing what an adversary can do in a real protocol execution to what it can do in an ideal scenario that is secure by definition. This is formalized by considering an ideal computation involving an incorruptible trusted party to whom the parties send their inputs. The trusted party computes the functionality on the inputs and returns to each party its respective output. Loosely speaking, a protocol is secure if any adversary interacting in the real protocol (where no trusted party exists) can do no more harm than if it was involved in the above-described ideal computation.
Execution in the ideal model. The parties are $P_{1}$ and $P_{2}$, and there is an adversary $\mathcal{A}$ who has corrupted one of them. An ideal execution for the computation of $\mathcal{F}=\left\{f_{n}\right\}$ proceeds as follows:

Inputs: $P_{1}$ and $P_{2}$ hold $1^{n}$ and inputs $x \in X_{n}$ and $y \in Y_{n}$, respectively; the adversary $\mathcal{A}$ receives an auxiliary input aux.

Send inputs to trusted party: The honest party sends its input to the trusted party. The corrupted party controlled by $\mathcal{A}$ may send any value of its choice. Denote the pair of inputs sent to the trusted party by $\left(x^{\prime}, y^{\prime}\right)$.

Trusted party sends outputs: If $x^{\prime} \notin X_{n}$ the trusted party sets $x^{\prime}$ to some default element $x_{0} \in X$ (and likewise if $y^{\prime} \notin Y_{n}$ ). Then, the trusted party chooses $r$ uniformly at random and sends $f_{n}^{1}\left(x^{\prime}, y^{\prime} ; r\right)$ to $P_{1}$ and $f_{n}^{2}\left(x^{\prime}, y^{\prime} ; r\right)$ to party $P_{2}$.

Outputs: The honest party outputs whatever it was sent by the trusted party, the corrupted party outputs nothing, and $\mathcal{A}$ outputs any arbitrary (probabilistic polynomial-time computable) function of its view.

We let $\operatorname{IDEAL}_{\mathcal{F}, \mathcal{A}(\operatorname{aux})}(x, y, n)$ be the random variable consisting of the output of the adversary and the output of the honest party following an execution in the ideal model as described above.

Execution in the real model. We next consider the real model in which a two-party protocol $\pi$ is executed by $P_{1}$ and $P_{2}$ (and there is no trusted party). In this case, the adversary $\mathcal{A}$ gets the
inputs of the corrupted party and sends all messages on behalf of this party, using an arbitrary polynomial-time strategy. The honest party follows the instructions of $\pi$.

Let $\pi$ be a two-party protocol computing $\mathcal{F}$. Let $\mathcal{A}$ be a non-uniform probabilistic polynomialtime machine with auxiliary input aux. We let $\operatorname{REAL}_{\pi, \mathcal{A}(\text { aux })}(x, y, n)$ be the random variable consisting of the view of the adversary and the output of the honest party, following an execution of $\pi$ where $P_{1}$ begins by holding $1^{n}$ and input $x$, and $P_{2}$ begins by holding $1^{n}$ and input $y$.
Security as emulation of an ideal execution in the real model. Having defined the ideal and real models, we can now define security of a protocol. Loosely speaking, the definition asserts that a secure protocol (in the real model) emulates the ideal model (in which a trusted party exists). This is formulated as follows:

Definition 1 Let $\mathcal{F}$, $\pi$ be as above, and fix a function $p$. Protocol $\pi$ is said to $\frac{1}{p}$-securely compute $\mathcal{F}$ if for every non-uniform probabilistic polynomial-time adversary $\mathcal{A}$ in the real model, there exists a non-uniform probabilistic polynomial-time adversary $\mathcal{S}$ in the ideal model such that

$$
\left\{\operatorname{IDEAL}_{\mathcal{F}, \mathcal{S}(\mathrm{aux})}(x, y, n)\right\}_{(x, y) \in X \times Y, \text { aux } \in\{0,1\}^{*}} \stackrel{1 / p}{\approx}\left\{\operatorname{REAL}_{\pi, \mathcal{A}(\mathrm{aux})}(x, y, n)\right\}_{(x, y) \in X \times Y, \text { aux } \in\{0,1\}^{*}}
$$

We stress that, in the above definition, we compare the real-world execution of the protocol to an ideal world in which fairness is guaranteed. $\frac{1}{p}$-security thus guarantees, in particular, that fairness is guaranteed in the real world except with probability at most $\frac{1}{p}$.

We also define security-with-abort in the standard way [17]. We remark that the notions of $\frac{1}{p}$-security and security-with-abort are incomparable.

### 2.3 Information-Theoretic MACs

We briefly review the standard definition for information-theoretically secure message authentication codes (MACs). A message authentication code consists of three polynomial-time algorithms (Gen, Mac, Vrfy). The key-generation algorithm Gen takes as input the security parameter $1^{n}$ in unary and outputs a key $k$. The message authentication algorithm Mac takes as input a key $k$ and a message $M \in\{0,1\} \leq n$, and outputs a tag $t$; we write this as $t=\operatorname{Mac}_{k}(M)$. The verification algorithm Vrfy takes as input a key $k$, a message $M \in\{0,1\} \leq n$, and a tag $t$, and outputs a bit $b$; we write this as $b=\operatorname{Vrfy}_{k}(M, t)$. We regard $b=1$ as acceptance and $b=0$ as rejection, and require that for all $n$, all $k$ output by $\operatorname{Gen}\left(1^{n}\right)$, all $M \in\{0,1\}^{\leq n}$, it holds that $\operatorname{Vrfy}_{k}\left(M, \operatorname{Mac}_{k}(M)\right)=1$.

We say (Gen, Mac, Vrfy) is a secure $m$-time MAC, where $m$ may be a function of $n$, if no computationally-unbounded adversary can output a valid tag on a new message after seeing valid tags on $m$ other messages. For our purposes, we do not require security against an adversary who adaptively chooses its $m$ messages for which to obtain a valid tag; it suffices to consider a nonadaptive definition where the $m$ messages are fixed in advance. (Nevertheless, known constructions satisfy the stronger requirement.) Formally:

Definition 2 Message authentication code (Gen, Mac, Vrfy) is an information-theoretically secure $m$-time MAC if for any sequence of messages $M_{1}, \ldots, M_{m}$ and any adversary $\mathcal{A}$, the following is negligible in the security parameter $n$ :

$$
\operatorname{Pr}\left[\begin{array}{c}
k \leftarrow \operatorname{Gen}\left(1^{n}\right) ; \forall i: t_{i}=\operatorname{Mac}_{k}\left(M_{i}\right) ; \\
\left(M^{\prime}, t^{\prime}\right) \leftarrow \mathcal{A}\left(M_{1}, t_{1}, \ldots, M_{m}, t_{m}\right)
\end{array}: \operatorname{Vrfy}_{k}\left(M^{\prime}, t^{\prime}\right)=1 \bigwedge M^{\prime} \notin\left\{M_{1}, \ldots, M_{m}\right\}\right] .
$$

## $3 \quad \frac{1}{\mathrm{p}}$-Secure Computation of General Functionalities

We begin in Section 3.1 by stating and proving a combinatorial lemma that will form an essential piece of our analysis in the two sections that follow. The reader who is willing to accept the results of that lemma on faith is welcome to skip to Section 3.2 where we demonstrate a $\frac{1}{p}$-secure protocol that works for functionalities defined on polynomial-size domains. A slight modification of this protocol is also secure-with-abort. To keep the exposition as simple as possible, we restrict our attention there to single-output functionalities, though the techniques extend easily to the case where each party receives a different function of the inputs. In Section 3.3 we show how to adapt the protocol for functionalities defined over domains of super-polynomial size (but polynomial range), and also generalize to the case of functionalities generating different outputs for each party.

### 3.1 A Useful Lemma

In this section, we analyze an abstract game $\Gamma$ between a challenger and an (unbounded) adversary $\mathcal{A}$. The game is parameterized by a value $\alpha \in(0,1]$ and an integer $r \geq 1$. Fix (arbitrary) distributions $D_{1}, D_{2}$ such that for every $z$ it holds that

$$
\begin{equation*}
\operatorname{Pr}_{a \leftarrow D_{1}}[a=z] \geq \alpha \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a=z] . \tag{1}
\end{equation*}
$$

The game $\Gamma(\alpha, r)$ proceeds as follows:

1. The challenger chooses $i^{*}$ uniformly from $\{1, \ldots, r\}$, and then prepares values $a_{1}, \ldots, a_{r}$ as follows:

- For $i<i^{*}$, it chooses $a_{i} \leftarrow D_{1}$.
- For $i \geq i^{*}$, it sets $a_{i} \leftarrow D_{2}$.

2. The challenger and the adversary then interact in a sequence of at most $r$ iterations. In iteration $i$ :

- The challenger gives $a_{i}$ to the adversary.
- The adversary can either abort or continue. In the former case, the game stops. In the latter case, the game continues to the next iteration.

3. $\mathcal{A}$ wins if it aborts the game in iteration $i^{*}$. (Since $\mathcal{A}$ can no longer win once iteration $i^{*}$ has passed, we may simply assume the game stops if that ever occurs.)

Let $\operatorname{Win}(\alpha, r)$ denote the maximum probability with which $\mathcal{A}$ can win the above game.
Lemma 1 For any $D_{1}, D_{2}$ satisfying Equation (1), $\operatorname{Win}(\alpha, r) \leq 1 / \alpha r$.
Proof: Fix $D_{1}, D_{2}$ satisfying Equation (1). We prove the lemma by induction on $r$. When $r=1$ the lemma is trivially true; for completeness, we also directly analyze the case $r=2$. Since $\mathcal{A}$ is unbounded we may assume it is deterministic. Then, without loss of generality, we may assume
the adversary's strategy is determined by a set $S$ in the support of $D_{2}$ such that $\mathcal{A}$ aborts in the first iteration iff $a_{1} \in S$, and otherwise aborts in the second iteration (no matter what). We have

$$
\begin{aligned}
\operatorname{Pr}[\mathcal{A} \text { wins }] & =\operatorname{Pr}\left[\mathcal{A} \text { wins and } i^{*}=1\right]+\operatorname{Pr}\left[\mathcal{A} \text { wins and } i^{*}=2\right] \\
& =\frac{1}{2} \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]+\frac{1}{2} \cdot\left(1-\operatorname{Pr}_{a \leftarrow D_{1}}[a \in S]\right) \\
& \leq \frac{1}{2} \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]+\frac{1}{2} \cdot\left(1-\alpha \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]\right) \\
& =\frac{1}{2}+\frac{1}{2} \cdot\left((1-\alpha) \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]\right) \leq 1-\alpha / 2,
\end{aligned}
$$

where the first inequality is due to Equation (1). One can easily verify that $1-\alpha / 2 \leq 1 / 2 \alpha$ when $\alpha>0$. We have thus proved $\operatorname{Win}(\alpha, 2) \leq 1 / 2 \alpha$.

Assume $\operatorname{Win}(\alpha, r) \leq 1 / \alpha r$, and we now bound $\operatorname{Win}(\alpha, r+1)$. As above, let $S$ denote a set in the support of $D_{2}$ such that $\mathcal{A}$ aborts in the first iteration iff $a_{1} \in S$. If $\mathcal{A}$ does not abort in the first iteration, and the game does not end, then the conditional distribution of $i^{*}$ is uniform in $\{2, \ldots, r+1\}$ and the game $\Gamma(\alpha, r+1)$ from this point forward is exactly equivalent to the game $\Gamma(\alpha, r)$. In particular, conditioned on the game $\Gamma(\alpha, r+1)$ not ending after the first iteration, the best strategy for $\mathcal{A}$ is to play whatever is the best strategy in game $\Gamma(\alpha, r)$. We thus have

$$
\begin{aligned}
\operatorname{Pr}[\mathcal{A} \text { wins }] & =\operatorname{Pr}\left[\mathcal{A} \text { wins and } i^{*}=1\right]+\operatorname{Pr}\left[\mathcal{A} \text { wins and } i^{*}>1\right] \\
& =\frac{1}{r+1} \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]+\frac{r}{r+1} \cdot\left(1-\operatorname{Pr}_{a \leftarrow D_{1}}[a \in S]\right) \cdot \operatorname{Win}(\alpha, r) \\
& \leq \frac{1}{r+1} \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]+\frac{1}{\alpha(r+1)} \cdot\left(1-\alpha \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a \in S]\right) . \\
& =\frac{1}{\alpha(r+1)} .
\end{aligned}
$$

This completes the proof.

### 3.2 A $\frac{1}{\mathrm{p}}$-Secure Protocol for Functions over Polynomial-Size Domains

In this section, we describe an approach that works for functions where at least one of the domains is polynomial-size. Although the protocol we describe would work even when the parties receive different outputs, for simplicity we assume here that the parties compute a single-output function. (We will return to the more general setting in the following section.) We prove the following:

Theorem 3 Let $\mathcal{F}=\left\{f_{n}: X_{n} \times Y_{n} \rightarrow Z_{n}\right\}$ be a sequence of (randomized) functions where $\left|Y_{n}\right|=$ poly $(n)$. Then, assuming the existence of enhanced trapdoor permutations, for any polynomial $p$ there exists an $\mathcal{O}\left(p \cdot\left|Y_{n}\right|\right)$-round protocol that $\frac{1}{p}$-securely computes $\mathcal{F}$.

Proof: As described in Section 1.3, our protocol $\Pi$ consists of two stages. Let $p$ be an arbitrary polynomial, and set $r=p \cdot\left|Y_{n}\right|$. We will implement the first stage using a sub-protocol for computing a randomized functionality ShareGen $_{r}$ defined in Figure 1. (ShareGen ${ }_{r}$ is parameterized by a polynomial $r$.) This functionality returns shares to each party, authenticated using an informationtheoretically secure $r$-time MAC, and in the second stage of the protocol the parties exchange these shares in a sequence of $r$ iterations as described in Figure 2.

## ShareGen $_{r}$

Inputs: The security parameter is $n$. Let the inputs to ShareGen $_{r}$ be $x \in X_{n}$ and $y \in Y_{n}$. (If one of the received inputs is not in the correct domain, a default input is substituted.)

## Computation:

1. Define values $a_{1}, \ldots, a_{r}$ and $b_{1}, \ldots, b_{r}$ in the following way:

- Choose $i^{*}$ uniformly at random from $\{1, \ldots, r\}$.
- For $i=1$ to $i^{*}-1$ do:
- Choose $\hat{y} \leftarrow Y_{n}$ and set $a_{i}=f(x, \hat{y})$.
- Choose $\hat{x} \leftarrow X_{n}$ and set $b_{i}=f(\hat{x}, y)$.
- For $i=i^{*}$ to $r$, set $a_{i}=b_{i}=f(x, y)$.

2. For $1 \leq i \leq r$, choose $\left(a_{i}^{(1)}, a_{i}^{(2)}\right)$ and $\left(b_{i}^{(1)}, b_{i}^{(2)}\right)$ as random secret sharings of $a_{i}$ and $b_{i}$, respectively. (E.g., $a_{i}^{(1)}$ is random and $a_{i}^{(1)} \oplus a_{i}^{(2)}=a_{i}$.)
3. Compute $k_{a}, k_{b} \leftarrow \operatorname{Gen}\left(1^{n}\right)$. For $1 \leq i \leq r$, let $t_{i}^{a}=\operatorname{Mac}_{k_{a}}\left(i \| a_{i}^{(2)}\right)$ and $t_{i}^{b}=\operatorname{Mac}_{k_{b}}\left(i \| b_{i}^{(1)}\right)$.

## Output:

1. Send to $P_{1}$ the values $a_{1}^{(1)}, \ldots, a_{r}^{(1)}$ and $\left(b_{1}^{(1)}, t_{1}^{b}\right), \ldots,\left(b_{r}^{(1)}, t_{r}^{b}\right)$, and the MAC-key $k_{a}$.
2. Send to $P_{2}$ the values $\left(a_{1}^{(2)}, t_{1}^{a}\right), \ldots,\left(a_{r}^{(2)}, t_{r}^{a}\right)$ and $b_{1}^{(2)}, \ldots, b_{r}^{(2)}$, and the MAC-key $k_{b}$.

Figure 1: Functionality ShareGen $_{r}$.

We analyze our protocol in a hybrid model where there is a trusted party computing ShareGen ${ }_{r}$. (In this ideal world, if $P_{1}$ is adversarial it can abort the trusted party computing ShareGen ${ }_{r}$ before the trusted party sends output to the honest party.) We will prove $\frac{1}{p}$-security of $\Pi$ in this hybrid model (in fact, we will prove that an execution in the hybrid world has statistical difference at most $\frac{1}{p}+\operatorname{negl}(n)$ from an ideal-world computation of $\left.\mathcal{F}\right)$; it follows as in [7] that if we use a sub-protocol for computing ShareGen ${ }_{r}$ that is secure-with-abort, then $\Pi$ is $\frac{1}{p}$-secure.

Claim 1 For every non-uniform, polynomial-time adversary $\mathcal{A}$ corrupting $P_{1}$ and running $\Pi$ in a hybrid model with access to an ideal functionality computing ShareGen ${ }_{r}$ (with abort), there exists a non-uniform, polynomial-time adversary $\mathcal{S}$ corrupting $P_{1}$ and running in the ideal world with access to an ideal functionality computing $f_{n}$ (with complete fairness), such that

Proof: We construct a simulator $\mathcal{S}$ that is given black-box access to $\mathcal{A}$. For readability in what follows, we ignore the presence of the MAC-tags and keys. That is, we do not mention the fact that $\mathcal{S}$ computes MAC-tags for messages it gives to $\mathcal{A}$, nor do we mention the fact that $\mathcal{S}$ must verify the MAC-tags on the messages sent by $\mathcal{A}$. When we say that $\mathcal{A}$ "aborts", we include in this the event that $\mathcal{A}$ sends an invalid message, or a message whose tag does not pass verification. We also drop the subscript $n$ from our notation.

1. $\mathcal{S}$ invokes $\mathcal{A}$ on the input ${ }^{3} x^{\prime}$, the auxiliary input, and the security parameter $n$. The simulator also chooses $\hat{x} \in X$ uniformly at random (it will send $\hat{x}$ to the trusted party, if needed).
[^2]
## Protocol 1

Inputs: Party $P_{1}$ has input $x$ and party $P_{2}$ has input $y$. The security parameter is $n$. Let $r=p \cdot\left|Y_{n}\right|$. The protocol:

## 1. Preliminary phase:

(a) $P_{1}$ chooses $\hat{y} \in Y_{n}$ uniformly at random, and sets $a_{0}=f(x, \hat{y})$. Similarly, $P_{2}$ chooses $\hat{x} \in X_{n}$ uniformly at random, and sets $b_{0}=f(\hat{x}, y)$.
(b) Parties $P_{1}$ and $P_{2}$ compute ShareGen ${ }_{r}$, using their inputs $x$ and $y$.
(c) If $P_{2}$ receives $\perp$ from the above computation, it outputs $b_{0}$ and halts. Otherwise, the parties proceed to the next step.
(d) Denote the output of $P_{1}$ from $\pi$ by $a_{1}^{(1)}, \ldots, a_{r}^{(1)},\left(b_{1}^{(1)}, t_{1}^{b}\right), \ldots,\left(b_{r}^{(1)}, t_{r}^{b}\right)$, and $k_{a}$.
(e) Denote the output of $P_{2}$ from $\pi$ by $\left(a_{1}^{(2)}, t_{1}^{a}\right), \ldots,\left(a_{r}^{(2)}, t_{r}^{a}\right), b_{1}^{(2)}, \ldots, b_{r}^{(2)}$, and $k_{b}$.
2. For $i=1, \ldots, r$ do:
$P_{2}$ sends the next share to $P_{1}$ :
(a) $P_{2}$ sends $\left(a_{i}^{(2)}, t_{i}^{a}\right)$ to $P_{1}$.
(b) $P_{1}$ receives $\left(a_{i}^{(2)}, t_{i}^{a}\right.$ ) from $P_{2}$. If $\mathrm{Vrfy}_{k_{a}}\left(i \| a_{i}^{(2)}, t_{i}^{a}\right)=0$ (or if $P_{1}$ received an invalid message, or no message), then $P_{1}$ outputs $a_{i-1}$ and halts.
(c) If $\mathrm{Vrfy}_{k_{a}}\left(i \| a_{i}^{(2)}, t_{i}^{a}\right)=1$, then $P_{1}$ sets $a_{i}=a_{i}^{(1)} \oplus a_{i}^{(2)}$ (and continues running the protocol).
$\boldsymbol{P}_{\mathbf{1}}$ sends the next share to $\boldsymbol{P}_{\mathbf{2}}$ :
(a) $P_{1}$ sends $\left(b_{i}^{(1)}, t_{i}^{b}\right)$ to $P_{2}$.
(b) $P_{2}$ receives $\left(b_{i}^{(1)}, t_{i}^{b}\right)$ from $P_{1}$. If $\mathrm{Vrfy}_{k_{b}}\left(i \| b_{i}^{(1)}, t_{i}^{b}\right)=0$ (or if $P_{2}$ received an invalid message, or no message), then $P_{2}$ outputs $b_{i-1}$ and halts.
(c) If $\mathrm{Vrfy}_{k_{b}}\left(i \| b_{i}^{(1)}, t_{i}^{b}\right)=1$, then $P_{2}$ sets $b_{i}=b_{i}^{(1)} \oplus b_{i}^{(2)}$ (and continues running the protocol).
3. If all $r$ iterations have been run, party $P_{1}$ outputs $a_{r}$ and party $P_{2}$ outputs $b_{r}$.

Figure 2: Generic protocol for computing a function $f_{n}$.
2. $\mathcal{S}$ receives the input $x$ of $\mathcal{A}$ to the computation of the functionality ShareGen $_{r}$. (If $x \notin X$ a default input is substituted.)
3. $\mathcal{S}$ sets $r=p \cdot|Y|$, and chooses uniformly-distributed shares $a_{1}^{(1)}, \ldots, a_{r}^{(1)}$ and $b_{1}^{(1)}, \ldots, b_{r}^{(1)}$. Then, $\mathcal{S}$ gives these shares to $\mathcal{A}$ as its output from the computation of ShareGen .
4. If $\mathcal{A}$ sends abort to the trusted party computing ShareGen $_{r}$, then $\mathcal{S}$ sends $\hat{x}$ to the trusted party computing $f$, outputs whatever $\mathcal{A}$ outputs, and halts. Otherwise (i.e., if $\mathcal{A}$ sends continue), $\mathcal{S}$ proceeds as below.
5. Choose $i^{*}$ uniformly from $\{1, \ldots, r\}$
6. For $i=1$ to $i^{*}-1$ :
(a) $\mathcal{S}$ chooses $\hat{y} \in Y$ uniformly at random, computes $a_{i}=f(x, \hat{y})$, and sets $a_{i}^{(2)}=a_{i}^{(1)} \oplus a_{i}$. It gives $a_{i}^{(2)}$ to $\mathcal{A}$. (Note that a fresh $\hat{y}$ is chosen in every iteration.)
(b) If $\mathcal{A}$ aborts, then $\mathcal{S}$ sends $\hat{x}$ to the trusted party, outputs whatever $\mathcal{A}$ outputs, and halts.
7. For $i=i^{*}$ to $r$ :
(a) If $i=i^{*}$ then $\mathcal{S}$ sends $x$ to the trusted party computing $f$ and receives $z=f(x, y)$.
(b) $\mathcal{S}$ sets $a_{i}^{(2)}=a_{i}^{(1)} \oplus z$ and gives $a_{i}^{(2)}$ to $\mathcal{A}$.
(c) If $\mathcal{A}$ aborts, then $\mathcal{S}$ then outputs whatever $\mathcal{A}$ outputs, and halts. If $\mathcal{A}$ does not abort, then $\mathcal{S}$ proceeds.
8. If $\mathcal{A}$ has never aborted (and all $r$ iterations are done), then $\mathcal{S}$ outputs whatever $\mathcal{A}$ outputs and halts.

Ignoring the possibility of a MAC forgery, we claim that the statistical difference between an execution of $\mathcal{A}$, running $\Pi$ in a hybrid world with access to an ideal functionality computing ShareGen ${ }_{r}$, and an execution of $\mathcal{S}$, running in an ideal world with access to an ideal functionality computing $f$, is at most $1 / p$. (Thus, taking into account the possibility of a MAC forgery makes the statistical difference at most $1 / p+\mu(n)$ for some negligible function $\mu$.) To see this, let $y$ denote the input of the honest $P_{2}$ and consider three cases depending on when the adversary aborts:

1. $\mathcal{A}$ aborts in round $i<i^{*}$. Conditioned on this event, the view of $\mathcal{A}$ is identically distributed in the two worlds (and is independent of $y$ ), and the output of the honest party is $f(\hat{x}, y)$ for $\hat{x}$ chosen uniformly in $X$.
2. $\mathcal{A}$ aborts in round $i>i^{*}$, or never aborts. Conditioned on this event, the view of $\mathcal{A}$ is again distributed identically in the two worlds, and in both worlds the output of the honest party is $f(x, y)$.
3. $\mathcal{A}$ aborts in round $i=i^{*}$ : here the distributions are not identical in the two worlds. Although the view of $\mathcal{A}$ is identical in both worlds, the output of the honest party is not: in the hybrid world the honest party will output $f(\hat{x}, y)$, for $\hat{x}$ chosen uniformly in $X$, while in the ideal world the honest party will output $f(x, y)$.
However, we use Lemma 1 to claim that this event occurs with probability at most $1 / p$. To see this, let $D_{1}$ denote the distribution of $a_{i}$ for $i<i^{*}$, and let $D_{2}$ denote the distribution of $a_{i^{*}}$. By construction of the protocol, we have

$$
\begin{aligned}
\operatorname{Pr}_{a \leftarrow D_{1}}[a=z] & \stackrel{\text { def }}{=} \operatorname{Pr}_{\hat{y} \leftarrow Y}[f(x, \hat{y})=z] \\
& \geq \frac{1}{|Y|} \cdot \operatorname{Pr}[f(x, y)=z]=\frac{1}{|Y|} \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a=z]
\end{aligned}
$$

Taking $\alpha=1 /|Y|$ and applying Lemma 1 , we see that $\mathcal{A}$ aborts in iteration $i^{*}$ with probability at most $1 / \alpha r=|Y| /|Y| p=1 / p$.

This completes the proof of the claim.
Claim 2 For every non-uniform, polynomial-time adversary $\mathcal{A}$ corrupting $P_{2}$ and running $\Pi$ in a hybrid model with access to an ideal functionality computing ShareGen $_{r}$ (with abort), there exists a non-uniform, polynomial-time adversary $\mathcal{S}$ corrupting $P_{2}$ and running in the ideal world with access to an ideal functionality computing $f$ (with complete fairness), such that

$$
\left\{\operatorname{IDEAL}_{f, \mathcal{S}}(x, y, n)\right\}_{x \in X_{n}, y \in Y_{n}, n \in \mathbb{N}} \stackrel{c}{\equiv}\left\{\operatorname{HYBRID}_{\Pi, \mathcal{A}}^{\operatorname{ShareGen}_{r}}(x, y, n)\right\}_{x \in X_{n}, y \in Y_{n}, n \in \mathbb{N}}
$$

## ShareGen ${ }_{p, r}^{\prime}$

Inputs: The security parameter is $n$. Let the inputs to ShareGen ${ }_{p, r}^{\prime}$ be $x \in X_{n}$ and $y \in Y_{n}$. (If one of the received inputs is not in the correct domain, a default input is substituted.)

## Computation:

1. Define values $a_{1}, \ldots, a_{r}$ and $b_{1}, \ldots, b_{r}$ in the following way:

- Choose $i^{*}$ uniformly at random from $\{1, \ldots, r\}$
- For $i=1$ to $i^{*}-1$ do:
- Choose $\hat{x} \leftarrow X$ and set $b_{i}=f^{2}(\hat{x}, y)$.
- With probability $\frac{1}{p}$, choose $z \leftarrow Z_{n}^{1}$ and set $a_{i}=z$
- With the remaining probability $1-\frac{1}{p}$, choose $\hat{y} \leftarrow Y$ and set $a_{i}=f^{1}(x, \hat{y})$.
- For $i=i^{*}$ to $r$, set $a_{i}=f^{1}(x, y)$ and $b_{i}=f^{2}(x, y)$.

2. For $1 \leq i \leq r$, choose $\left(a_{i}^{(1)}, a_{i}^{(2)}\right)$ and $\left(b_{i}^{(1)}, b_{i}^{(2)}\right)$ as random secret sharings of $a_{i}$ and $b_{i}$, respectively. (E.g., $a_{i}^{(1)}$ is random and $a_{i}^{(1)} \oplus a_{i}^{(2)}=a_{i}$.)
3. Compute $k_{a}, k_{b} \leftarrow \operatorname{Gen}\left(1^{n}\right)$. For $1 \leq i \leq r$, let $t_{i}^{a}=\operatorname{Mac}_{k_{a}}\left(i \| a_{i}^{(2)}\right)$ and $t_{i}^{b}=\operatorname{Mac}_{k_{b}}\left(i \| b_{i}^{(1)}\right)$.

Output:

1. Send to $P_{1}$ the values $a_{1}^{(1)}, \ldots, a_{r}^{(1)}$ and $\left(b_{1}^{(1)}, t_{1}^{b}\right), \ldots,\left(b_{r}^{(1)}, t_{r}^{b}\right)$, and the MAC-key $k_{a}$.
2. Send to $P_{2}$ the values $\left(a_{1}^{(2)}, t_{1}^{a}\right), \ldots,\left(a_{r}^{(2)}, t_{r}^{a}\right)$ and $b_{1}^{(2)}, \ldots, b_{r}^{(2)}$, and the MAC-key $k_{b}$.

Figure 3: Functionality ShareGen ${ }_{p, r}^{\prime}$.
The simulation is identical to the one in Claim 1, though the proof is much simpler (and we can prove a stronger notion of security) since $P_{1}$ always "gets the output first" in every iteration.

Achieving security-with-abort. As written, the protocol is not secure-with-abort. However, it is easy to modify it so that it is (without affecting $\frac{1}{p}$-security): simply have ShareGen ${ }_{r}$ set $b_{i^{*}-1}=\perp$, where $\perp$ is some distinguished value outside the range of $f$. Although this allows a malicious $P_{2}$ to identify exactly when iteration $i^{*}$ occurs, this does not affect security (as reflected by Claim 2).

### 3.3 A $\frac{1}{\mathrm{p}}$-Secure Protocol for Functions over Arbitrary Domains

The protocol from the previous section does not directly apply to functions on arbitrary (i.e., not necessarily polynomial-size) domains, since the round complexity of the protocol is polynomial in the smaller domain. In this section we demonstrate how to extend the protocol so as to handle arbitrary domains, as long as the range of the function is polynomial size (for at least one of the parties). For completeness, we now also explicitly take into account the case when parties obtain different outputs. Intuition for the changes we introduce is given in Section 1.3.

Theorem 4 Let $\mathcal{F}=\left\{f_{n}: X_{n} \times Y_{n} \rightarrow Z_{n}^{1} \times Z_{n}^{2}\right\}$ be a sequence of (randomized) functions, with $\left|Z_{n}^{1}\right|=\operatorname{poly}(n)$. Then, assuming the existence of enhanced trapdoor permutations, for any polynomial $p$ there exists an $\mathcal{O}\left(p^{2} \cdot\left|Z_{n}^{1}\right|\right)$-round protocol that $\frac{1}{p}$-securely computes $\mathcal{F}$.

Proof: Our protocol $\Pi$ is, once again, composed of two stages. The second stage will be identical to the second stage of the previous protocol (see Figure 2), except that the number of iterations $r$
will now be set to $r=p^{2} \cdot\left|Z_{n}^{1}\right|$. The first stage will generate shares using a new sub-routine ShareGen ${ }_{p, r}^{\prime}$, parameterized by both $p$ and $r$, as described in Figure 3.

We will again analyze our protocol in a hybrid model, but where there is now a trusted party computing ShareGen ${ }_{p, r}^{\prime}$. (Once again, $P_{1}$ can abort the computation of ShareGen ${ }_{p, r}^{\prime}$ even in the ideal world.) We will prove $\frac{1}{p}$-security in this hybrid model, which implies that if the parties use a secure-with-abort protocol for computing ShareGen ${ }_{p, r}^{\prime}$ then the entire protocol $\Pi$ is $\frac{1}{p}$-secure.

Claim 3 For every non-uniform, polynomial-time adversary $\mathcal{A}$ corrupting $P_{1}$ and running $\Pi$ in a hybrid model with access to an ideal functionality computing ShareGen ${ }_{p, r}^{\prime}$ (with abort), there exists a non-uniform, polynomial-time adversary $\mathcal{S}$ corrupting $P_{1}$ and running in the ideal world with access to an ideal functionality computing $\mathcal{F}$ (with complete fairness), such that

$$
\left\{\operatorname{IDEAL}_{\mathcal{F}, \mathcal{S}(\mathrm{aux})}(x, y, n)\right\}_{x \in X_{n}, y \in Y_{n}, \mathrm{aux} \in\{0,1\}^{*}} \stackrel{1 / p}{\approx}\left\{\operatorname{HYBRID}_{\Pi, \mathcal{A}(\mathrm{aux})}^{\text {ShareGen }_{p, r}^{\prime}}(x, y, n)\right\}_{x \in X_{n}, y \in Y_{n}, \mathrm{aux} \in\{0,1\}^{*}} .
$$

Proof: The simulator is essentially the same as the simulator used in the proof of Claim 1, except that in step 6(a) the distribution on $a_{i}\left(\right.$ for $\left.i<i^{*}\right)$ is changed to the one used by ShareGen ${ }_{p, r}^{\prime}$. The analysis is similar, too, except for bounding the probability that $\mathcal{A}$ aborts in iteration $i^{*}$. To bound this probability we will again rely on Lemma 1, but now distribution $D_{1}$ (i.e., the distribution of $a_{i}$ for $\left.i<i^{*}\right)$ is different. Let $y$ denote the input of $P_{2}$. Note that, by construction of ShareGen ${ }_{p, r}^{\prime}$, for any $z \in Z_{n}^{1}$ we have $\operatorname{Pr}_{a \leftarrow D_{1}}[a=z] \geq \frac{1}{p} \cdot \frac{1}{\mid Z_{n}^{n}}$. Regardless of $f^{1}$ and $y$, it therefore holds for all $z \in Z_{n}^{1}$ that

$$
\operatorname{Pr}_{a \leftarrow D_{1}}[a=z] \geq \frac{1}{p \cdot\left|Z_{n}^{1}\right|} \cdot \operatorname{Pr}_{a \leftarrow D_{2}}[a=z] .
$$

Setting $\alpha=1 / p \cdot\left|Z_{n}^{1}\right|$ and applying Lemma 1 , we see that $\mathcal{A}$ aborts in iteration $i^{*}$ with probability at most

$$
\frac{1}{\alpha r}=\frac{p \cdot\left|Z_{n}^{1}\right|}{p^{2} \cdot\left|Z_{n}^{1}\right|}=\frac{1}{p} .
$$

This completes the proof.
The following claim considers the case of a malicious $P_{2}$. We stress that, in contrast to Claim 2, here we claim only $\frac{1}{p}$-indistinguishability.

Claim 4 For every non-uniform, polynomial-time adversary $\mathcal{A}$ corrupting $P_{2}$ and running $\Pi$ in a hybrid model with access to an ideal functionality computing ShareGen ${ }_{p, r}^{\prime}$ (with abort), there exists a non-uniform, polynomial-time adversary $\mathcal{S}$ corrupting $P_{2}$ and running in the ideal world with access to an ideal functionality computing $\mathcal{F}$ (with complete fairness), such that

$$
\left\{\operatorname{IDEAL}_{\mathcal{F}, \mathcal{S}(\mathrm{aux})}(x, y, n)\right\}_{x \in X_{n}, y \in Y_{n}, \mathrm{aux} \in\{0,1\}^{*}} \stackrel{1 / p}{\approx}\left\{\operatorname{HYBRID}_{\Pi, \mathcal{A}(\mathrm{aux})} \operatorname{Sharen}_{p}^{\prime}, r(x, y, n)\right\}_{x \in X_{n}, y \in Y_{n}, \mathrm{aux} \in\{0,1\}^{*}} .
$$

Proof: The simulator $\mathcal{S}$ in this case is fairly obvious, but we include it for completeness. Once again, for readability we ignore the presence of the MAC-tags and keys.

1. $\mathcal{S}$ invokes $\mathcal{A}$ on the input $y^{\prime}$, the auxiliary input, and the security parameter $n$. The simulator also chooses $\hat{y} \in Y$ uniformly at random (it will send $\hat{y}$ to the trusted party, if needed).
2. $\mathcal{S}$ receives the input $y$ of $\mathcal{A}$ to the computation of the functionality ShareGen ${ }_{p, r}^{\prime}$. (If $y \notin Y$ a default input is substituted.)
3. $\mathcal{S}$ sets $r=p^{2} \cdot\left|Z^{1}\right|$, and chooses uniformly-distributed shares $a_{1}^{(1)}, \ldots, a_{r}^{(1)}$ and $b_{1}^{(1)}, \ldots, b_{r}^{(1)}$. Then, $\mathcal{S}$ gives these shares to $\mathcal{A}$ as its output from the computation of ShareGen ${ }_{p, r}^{\prime}$.
4. Choose $i^{*}$ uniformly from $\{1, \ldots, r\}$
5. For $i=1$ to $i^{*}-1$ :
(a) $\mathcal{S}$ chooses $\hat{x} \in X$ uniformly at random, computes $b_{i}=f^{2}(\hat{x}, y)$, and sets $b_{i}^{(1)}=b_{i}^{(2)} \oplus b_{i}$. It gives $b_{i}^{(1)}$ to $\mathcal{A}$. (Note that a fresh $\hat{x}$ is chosen in every iteration.)
(b) If $\mathcal{A}$ aborts, then $\mathcal{S}$ sends $\hat{y}$ to the trusted party, outputs whatever $\mathcal{A}$ outputs, and halts.
6. For $i=i^{*}$ to $r$ :
(a) If $i=i^{*}$ then $\mathcal{S}$ sends $y$ to the trusted party computing $f$ and receives $z=f^{2}(x, y)$.
(b) $\mathcal{S}$ sets $b_{i}^{(1)}=b_{i}^{(2)} \oplus z$ and gives $b_{i}^{(1)}$ to $\mathcal{A}$.
(c) If $\mathcal{A}$ aborts, then $\mathcal{S}$ then outputs whatever $\mathcal{A}$ outputs, and halts. If $\mathcal{A}$ does not abort, then $\mathcal{S}$ proceeds.
7. If $\mathcal{A}$ has never aborted (and all $r$ iterations are done), then $\mathcal{S}$ outputs whatever $\mathcal{A}$ outputs and halts.

Ignoring the possibility of a MAC forgery, we claim that the statistical difference between an execution of $\mathcal{A}$, running $\Pi$ in a hybrid world with access to an ideal functionality computing ShareGen ${ }_{p, r}^{\prime}$, and an execution of $\mathcal{S}$, running in an ideal world with access to an ideal functionality computing $\mathcal{F}$, is at most $1 / p$. (Thus, taking into account the possibility of a MAC forgery makes the statistical difference at most $1 / p+\mu(n)$ for some negligible function $\mu$.) In fact, the view of $\mathcal{A}$ is identical in the two worlds; the only issue is the output of the honest $P_{1}$ holding input $x$. Specifically, if $\mathcal{A}$ aborts in any iteration prior to $i^{*}$ then, in the ideal world interaction with $\mathcal{S}$, party $P_{1}$ outputs $f^{1}(x, \hat{y})$ for a uniformly-chosen $\hat{y} \in Y$. In the hybrid world, however, the output of $P_{1}$ is given by the distribution of $a_{i}\left(\right.$ for $\left.i<i^{*}\right)$ as determined by ShareGen ${ }_{p, r}^{\prime}$. However, these two distributions are within statistical difference (at most) $1 / p$. The claim follows.

## 4 Optimality of Our Results

In this section, we prove two impossibility results showing that the results of the previous section are optimal as far as generic feasibility results go. The first result is unconditional, while the second relies on cryptographic assumptions.

### 4.1 Impossibility of $\frac{1}{p}$-Security and Security-with-Abort for Functions over Arbitrary Domains

As mentioned earlier, the notions of security-with-abort and $\frac{1}{p}$-security are incomparable. Although we are able to achieve $\frac{1}{p}$-security and security-with-abort simultaneously for the case of functionalities where at least one of the domains is polynomial-size (cf. Section 3.2), we were not able to
do so for the case of arbitrary domains. We show, in fact, that it is impossible to achieve both of these criteria (in general) in this case.

Consider the equality function EQ defined over domains of super-polynomial size. (Formally, consider EQ : $X_{n} \times Y_{n} \rightarrow\{0,1\}$ where $X_{n}=Y_{n}=\{0,1\}^{\ell(n)}$ for some $\ell(n)=\omega(\log n)$.) Let $\Pi$ be some protocol computing EQ, where we assume without loss of generality that $P_{2}$ goes first and $P_{1}$ goes last. Say $\Pi$ has $r=r(n)$ iterations for some polynomial $r$. (An iteration consists of a message sent by $P_{2}$ followed by a message from $P_{1}$.) We show that $\Pi$ cannot be simultaneously secure-with-abort and $\frac{1}{p}$-secure for any $p>4+\frac{1}{\operatorname{poly}(n)}$.

Let $a_{0}$ denote the value that $P_{1}$ outputs if $P_{2}$ sends nothing, and let $a_{i}$, for $1 \leq i \leq r$, denote the value that $P_{1}$ outputs if $P_{2}$ aborts after sending its iteration- $i$ message. Similarly, let $b_{0}$ denote the value that $P_{2}$ outputs if $P_{1}$ sends nothing, and let $b_{i}$, for $1 \leq i \leq r$, denote the value that $P_{2}$ outputs if $P_{1}$ aborts after sending its iteration- $i$ message. We may assume without loss of generality that, for all $i, a_{i} \in\{0,1\}$ and $b_{i} \in\{0,1, \perp\}$.

We will consider two experiments involving an execution of $\Pi$. In the first, $x$ and $y$ are chosen uniformly and independently from $\{0,1\}^{\ell(n)}$; the parties are given inputs $x$ and $y$, respectively; and the parties then run protocol $\Pi$ honestly. We denote the probability of events in this probability space by $\operatorname{Pr}_{\text {rand }}[\cdot]$. In the second experiment, $x$ is chosen uniformly from $\{0,1\}^{\ell(n)}$ and $y$ is set equal to $x$; these inputs are given to the parties and they run the protocol honestly as before. We denote the probability of events in this probability space by $\operatorname{Pr}_{\text {eq }}[\mathrm{l}]$.

Claim 5 If $\Pi$ is secure-with-abort, then $\operatorname{Pr}_{\text {rand }}\left[a_{0}=1 \vee \cdots \vee a_{r}=1\right]$ and $\operatorname{Pr}_{\text {rand }}\left[b_{0}=1 \vee \cdots \vee b_{r}=1\right]$ are negligible.

Proof: If, say, $\operatorname{Pr}_{\text {rand }}\left[a_{0}=1 \vee \cdots \vee a_{r}=1\right]$ were not negligible, then we could consider an adversarial $P_{2}$ that runs the protocol honestly but aborts at a random round. This would result in the honest $P_{1}$ outputting 1 with non-negligible probability in the real world, whereas this occurs with only negligible probability in the ideal world (when the parties are given independent, random inputs). A similar argument applies to $\operatorname{Pr}_{\text {rand }}\left[b_{0}=1 \vee \cdots \vee b_{r}=1\right]$.

Assume for simplicity that $\Pi$ has perfect correctness, i.e., that $a_{r}=b_{r}=\mathrm{EQ}(x, y)$ when the two parties run the protocol honestly holding initial inputs $x$ and $y$. (This assumption is not necessary, but allows us to avoid introducing burdensome notation.) Then $\operatorname{Pr}_{\text {eq }}\left[a_{0}=1 \vee \cdots \vee a_{r}=1\right]=$ $\operatorname{Pr}_{\text {eq }}\left[b_{0}=1 \vee \cdots \vee b_{r}=1\right]=1$. Let $i^{*}$ denote the lowest index, if any, for which $a_{i^{*}}=1$. Similarly, let $j^{*}$ denote the lowest index, if any, for which $b_{j^{*}}=1$. Since

$$
\operatorname{Pr}_{\mathrm{eq}}\left[i^{*} \leq j^{*}\right]+\operatorname{Pr}_{\mathrm{eq}}\left[i^{*}>j^{*}\right]=\operatorname{Pr}_{\mathrm{eq}}\left[i^{*} \text { and } j^{*} \text { are defined }\right]=1,
$$

at least one of the two terms on the left-hand side is at least $1 / 2$. We assume $\operatorname{Pr}_{\text {eq }}\left[i^{*} \leq j^{*}\right] \geq 1 / 2$ in what follows, but the exact same argument (swapping the roles of the parties) applies in case $\operatorname{Pr}_{\mathrm{eq}}\left[i^{*}>j^{*}\right] \geq 1 / 2$.

Consider now a third experiment that is a mixture of the previous two. Specifically, in this experiment a random bit $b$ is chosen; if $b=0$ then the parties are given inputs $x$ and $y$ as in the first experiment (i.e., chosen uniformly and independently at random), while if $b=1$ then the parties are given (random) $x=y$ as in the second experiment. The parties then run protocol $\Pi$ honestly. We denote the probability of events in this probability space by $\operatorname{Pr}_{3}^{\text {real }}[\cdot]$. We use the superscript real to distinguish this from an ideal-world version of this experiment where the bit $b$ is chosen uniformly and the parties are given $x$ and $y$ generated accordingly, but now the parties
interact with an ideal party computing EQ without abort. We denote the probability of events in this probability space by $\operatorname{Pr}_{3}^{\text {ideal }}[\cdot]$.

Consider an execution of the third experiment (in either the real or ideal worlds), in the case when $P_{1}$ is malicious. Let guess denote the event that $P_{1}$ correctly guesses the value of the bit $b$, and let out ${ }_{2}$ denote the output of $P_{2}$. A simple calculation shows that

$$
\begin{equation*}
\operatorname{Pr}_{3}^{\text {ideal }}\left[\text { guess } \wedge \text { out }_{2} \neq 1\right] \leq \frac{1}{2}+\operatorname{negl}(n) \tag{2}
\end{equation*}
$$

Now take the following real-world adversary $\mathcal{A}$ corrupting $P_{1}$ : upon receiving input $x$, run $\Pi$ honestly but compute $a_{i}$ after receiving each iteration- $i$ message from $P_{2}$. There are two cases:

- If, at some point, some $a_{i}=1$ then abort the protocol (before sending the iteration- $i$ message on behalf of $P_{1}$ ) and output the guess " $b=1$ ".
- If $a_{i}=0$ for all $i$, then simply run the protocol to the end (including the final message of the protocol) and output the guess " $b=0$ ".

We have:

$$
\begin{align*}
\operatorname{Pr}_{3}^{\text {real }}\left[\text { guess } \wedge \text { out }_{2} \neq 1\right] & =\frac{1}{2} \cdot \operatorname{Pr}_{\text {rand }}\left[\text { guess } \wedge \text { out }_{2} \neq 1\right]+\frac{1}{2} \cdot \operatorname{Pr}_{\mathrm{eq}}\left[\text { guess } \wedge \text { out }_{2} \neq 1\right] \\
& \geq \frac{1}{2} \cdot \operatorname{Pr}_{\text {rand }}\left[a_{1}=0 \wedge \cdots \wedge a_{r}=0 \wedge b_{r}=0\right]+\frac{1}{2} \cdot \operatorname{Pr}_{\mathrm{eq}}\left[i^{*} \leq j^{*}\right] \\
& \geq \frac{1}{2} \cdot(1-\operatorname{negl}(n))+\frac{1}{4}=\frac{3}{4}-\operatorname{negl}(n) \tag{3}
\end{align*}
$$

Taken together, Claim 5 and Equations (2) and (3) show that if $\Pi$ is secure with abort, then it cannot also be $\frac{1}{p}$-secure for any $p>4+\frac{1}{\operatorname{poly}(n)}$.

### 4.2 Impossibility of $\frac{1}{p}$-Security for General Functions

Our results in Section 3.2 and 3.3 imply that $\frac{1}{p}$-security is achievable for any function $f: X \times Y \rightarrow$ $Z^{1} \times Z^{2}$ as long as at least one of $X, Y, Z^{1}, Z^{2}$ are polynomial size. Here, we give evidence that this limitation is inherent by showing, under a reasonable cryptographic assumption, that there is a deterministic, single-output function $f: X \times Y \rightarrow Z$ with $|X|,|Y|,|Z|=\omega(\operatorname{poly}(n))$ that cannot be $\frac{1}{p}$-securely computed for any $p>2+\frac{1}{\operatorname{poly}(n)}$. The proof is not difficult, and we would not be surprised if similar proofs have appeared previously in the literature; however, we were unable to track down any such references.

The assumption we will need is that exponentially-strong one-way functions exist. (We note, however, that our proof rules out partial fairness for functions over domains of size $2^{n^{\epsilon}}$ if standard one-way functions exist.) This is a family of functions $\left\{f_{n}:\{0,1\}^{n} \rightarrow\{0,1\}^{n}\right\}$ such that, for some $\delta$, we have

$$
\operatorname{Pr}\left[\mathcal{A}(f(x)) \in f^{-1}(f(x))\right] \leq 2^{-\delta n}
$$

for all $\mathcal{A}$ running in time $2^{\delta n}$. This implies that $f$ is one-way (in the standard sense) when its input is chosen from $\{0,1\}^{\omega(\log n)}$.

Given such an $f$, we define the function

$$
\text { Swap : }\left(\{0,1\}^{\omega(\log n)}\right)^{2} \times\left(\{0,1\}^{\omega(\log n)}\right)^{2} \rightarrow\left(\{0,1\}^{\omega(\log n)}\right)^{2}
$$

as follows: $\operatorname{Swap}\left(\left(x_{1}, y_{2}\right),\left(x_{2}, y_{1}\right)\right)$ outputs $\left(x_{1}, x_{2}\right)$ iff $f\left(x_{1}\right)=y_{1}$ and $f\left(x_{2}\right)=y_{2}$, and outputs $\perp$ otherwise.

Consider an ideal-world computation of Swap, where $x_{1}, x_{2}$ are chosen uniformly at random, $y_{1}=f\left(x_{1}\right)$ and $y_{2}=f\left(x_{2}\right)$, and $P_{1}$ is given $\left(x_{1}, y_{2}\right)$ while $P_{2}$ is given $\left(x_{2}, y_{1}\right)$. We can easily observe that, e.g., an adversarial $P_{1}$ cannot simultaneously output an inverse of $y_{2}$ while causing $P_{2}$ to fail to output a valid inverse of $y_{1}$, except with negligible probability. In what follows, we refer to this event as a win for $P_{1}$.

In any real-world computation of Swap, however, there must be one party who "gets its output first" with probability at least $1 / 2$. More formally, say we have an $r$-iteration protocol computing Swap where $P_{2}$ sends the first message and $P_{1}$ sends the last message, and let $a_{i}$, for $i=0, \ldots, a_{r}$, denote the second component of the value $P_{1}$ would output if $P_{2}$ aborts the protocol after sending its iteration- $i$ message. Similarly, let $b_{i}$ denote the first component of the value that $P_{2}$ would output if $P_{1}$ aborts the protocol after sending its iteration- $i$ message. Each value $a_{i}$ and $b_{i}$ can be computed in polynomial time after receiving the other party's iteration- $i$ message. We can therefore define an adversary $P_{1}^{*}$ that acts as follows:

Run the protocol honestly until the first round in which $f\left(a_{i}\right)=y_{2}$; then abort.
An adversary $P_{2}^{*}$ can be defined analogously. Let $i_{1}$ be a random variable denoting the first round in which $f\left(a_{i}\right)=y_{2}$, and let $i_{2}$ denote the first round in which $f\left(b_{i}\right)=y_{1}$. Since

$$
\operatorname{Pr}\left[i_{1} \leq i_{2}\right]+\operatorname{Pr}\left[i_{1}>i_{2}\right]=1,
$$

while $\operatorname{Pr}\left[P_{1}^{*}\right.$ wins $]=\operatorname{Pr}\left[i_{1} \leq i_{2}\right]$ and $\operatorname{Pr}\left[P_{2}^{*}\right.$ wins $]=\operatorname{Pr}\left[i_{1}>i_{2}\right]$, it is clear that either $P_{1}^{*}$ or $P_{2}^{*}$ wins with probability at least $1 / 2$. Since an adversary wins in the ideal world with negligible probability, this rules out $\frac{1}{p}$-security if $\frac{1}{2}-\frac{1}{p}$ is noticeable.

The above does not contradict the result of [11], or any other work on partial fairness that aims to solve exactly this problem. The reason is that in previous work on partial fairness the running time of the honest parties is not bounded by a fixed polynomial, whereas in our setting we require this to be the case. We add further that in previous work on partial fairness (this is made explicit in [16]), the decision of how much "effort" honest parties should invest is not determined by the protocol, but instead is supposed to be determined is some other (unspecified) manner.

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[^1]:    ${ }^{1}$ This definition is similar in spirit to (but weaker than) the notion of $\epsilon$-zero knowledge [12] and is analogous to the definition used in [18] in the context of password-based key exchange, although there the value of $p$ is fixed by the size of the password dictionary. A similar idea, formalized differently and in a different context, is also used in [1].
    ${ }^{2}$ Actually, they prove something weaker but one can show that their protocols satisfy our stronger definition.

[^2]:    ${ }^{3}$ To simplify notation, we reserve $x$ for the value input by $\mathcal{A}$ to the computation of ShareGen ${ }_{r}$.

