A Round-Collapse Theorem for Computationally-Sound Protocols; or, TFNP is Hard (on Average) in Pessiland

Rafael Pass∗ Muthuramakrishnan Venkitasubramaniam†

Abstract

Consider the following two fundamental open problems in complexity theory:

• Does a hard-on-average language in NP imply the existence of one-way functions?
• Does a hard-on-average language in NP imply a hard problem in TFNP (i.e., the class of total NP search problem)?

We show that the answer to (at least) one of these questions is yes. In other words, in Impagliazzo’s Pessiland (where NP is hard-on-average, but one-way functions do not exist), TFNP is unconditionally hard (on average).

This result follows from a more general theory of interactive average-case complexity, and in particular, a novel round-collapse theorem for computationally-sound protocols, analogous to Babai-Moran’s celebrated round-collapse theorem for information-theoretically sound protocols. As another consequence of this treatment, we show that the existence of $O(1)$-round public-coin non-trivial arguments (i.e., argument systems that are not proofs) imply the existence of a hard-on-average problem in NP/poly.

∗Cornell Tech
†University of Rochester
1 Introduction

We initiate a complexity-theoretic study of interactive computational puzzles: 2-player interactive games between a polynomial-time public-coin challenger C and an attacker satisfying the following two properties:

Computational Soundness: There does not exist a probabilistic polynomial-time (PPT) attacker $A^*$ and polynomial $p$ such that $A^*(1^n)$ succeeds in making $C(1^n)$ output 1 with probability $\frac{1}{p(n)}$ for all sufficiently large $n \in N$.

Completeness/Non-triviality: There exists a negligible function $\mu$ and an inefficient attacker $A$ that on input $1^n$ succeeds in making $C(1^n)$ output 1 with probability $1 - \mu(n)$ for all $n \in N$.

In other words, (a) no polynomial-time attacker, $A^*$, can make $C$ output 1 with inverse polynomial probability, yet (b) there exists a computationally unbounded attacker $A$ that makes $C$ output 1 with overwhelming probability. We refer to $C$ as a $k(\cdot)$-round computational puzzle (or simply a $k(\cdot)$-round puzzle) if $C$ satisfies the above completeness and computational soundness conditions, while restricting $C(1^n)$ to communicate with $A$ in $k(n)$ rounds.

As an example of a 2-round puzzle, let $f$ be a one-way permutation and consider a game where $C(1^n)$ samples a random $y \in \{0, 1\}^n$ and requires the adversary to output a preimage $x$ such that $f(x) = y$. Since $f$ is a permutation, this puzzle has “perfect” completeness—an unbounded attacker $A$ can always find a preimage $x$. By the one-wayness of $f$ (and the permutatation property of $f$), we also have that no PPT adversary $A^*$ can find such an $x$ (with inverse polynomial probability), and thus soundness holds.

In fact, the existence of 2-round puzzles is “essentially” equivalent to the existence of an average-case hard problem in NP: any 2-round puzzle trivially implies a hard-on-average search problem (w.r.t. the uniform distribution) in NP and thus by [IL90] also a hard-on-average decision problem in NP. Furthermore, “almost-everywhere” hard-on-average languages in NP also imply the existence of a 2-round puzzle (by simply sampling many random instances $x$ and asking the attacker to provide a witness for at least, say, 1/3 of the instances).\(^2\)

Proposition 1.1 (informally stated). The existence of an (almost-everywhere) hard-on-average language in NP implies the existence of a 2-round puzzle. Furthermore, the existence of a 2-round puzzle implies the existence of a hard-on-average language in NP.

Thus, 2-round puzzles are “morally” (up to the infinitely-often/almost-everywhere issue) equivalent to the existence of a hard-on-average language in NP. As such, $k(\cdot)$-round puzzles are a natural way to generalize average-case hardness in NP.

While the game-based modeling in the notion of a puzzle is common in the cryptographic literature—most notably, it is commonly used to model cryptographic assumptions [Nao03, Pas11, GW11], complexity-theoretic consequences or properties of puzzles have remained largely unexplored. In this work, we initiate such a treatment. Furthermore, we show that such an interactive treatment of average-case complexity leads to a new tool set also for answering “classic” questions regarding average-case hardness in NP. Most notably, relying on this interactive treatment of puzzles, we demonstrate the following result:

If NP is (almost everywhere) hard-on-average, then either (1) one-way functions exist, or 2) TFNP (i.e., the class of total search problems) is hard-on-average.

\(^1\)That is, a language in NP such that for every $\delta > 0$, no PPT attacker $A$ can decide random instances with probability greater than $\frac{1}{2} + \delta$ for infinitely many (as opposed to all) $n \in N$. Such an “almost-everywhere” notion is more commonly used in the cryptographic literature.

\(^2\)The reason we need the language to be almost-everywhere hard-on-average is to guarantee that YES instances exists for every sufficiently large input length, or else completeness would not hold.
1.1 The Round-Complexity of Puzzles

Perhaps the most basic question regarding the existence of interactive puzzles is whether the existence of a $k$-round puzzle is actually a weaker assumption than the existence of a $(k-1)$ round puzzle. In particular, do interactive puzzles actually generalize beyond just average-case hardness in $\text{NP}$:

Does the existence of a $k$-round puzzle imply the existence of $(k-1)$-round puzzle?

At first sight, one would hope the classic “round-reduction” theorem due to Babai-Moran (BM) [BM88] can be applied to collapse any $O(1)$-round puzzle into a 2-round puzzle (i.e., a hard-on-average $\text{NP}$ problem). Unfortunately, while BM’s round reduction technique indeed works for all information-theoretically sound protocols, Wee [Wee06] demonstrated that BM’s round reduction fails for computationally sound protocols. In particular, Wee shows that black-box proofs of security cannot be used to prove that BM’s transformation preserves soundness even when applied to just 3-round protocols, and demonstrates (under computational assumptions) a concrete 4-round protocol for which BM’s round-reduction results in an unsound protocol.

As BM’s round reduction is the only known round-reduction technique (which does not rely on any assumptions), it was generally conjectured that the existence of a $k$-round puzzle is a strictly stronger assumption than the existence of a $(k+1)$-round puzzle—in particular, this would imply the existence of infinitely many worlds between Impagliazzo’s Pessiland and Heuristica [Imp95] (i.e., infinitely many worlds where $\text{NP} \neq \text{P}$ yet average-case $\text{NP}$ hardness does not exist). Further evidence in this direction comes from a work by Gertner et al. [GKM+00] which shows a black-box separation between $k$-round puzzles and $(k+1)$-round puzzles for a particular cryptographic task (namely that of a key-agreement scheme).

In contrast to the above negative results, our main technical result provides an affirmative answer to the above question—we demonstrate a round-reduction theorem for puzzles.

**Theorem 1.1** (informally stated). For every constant $c$, the existence of a $k(\cdot)$-round puzzle is equivalent to the existence of a $(k(\cdot)-c)$-round puzzle.

In particular, as corollary of this result, we get that the assumption that a $O(1)$-round puzzle exists is not weaker than the assumption that average-case hardness in $\text{NP}$ exists:

**Corollary 1.2** (informally stated). The existence of an $O(1)$-round puzzle implies the existence of a hard-on-average problem in $\text{NP}$.

Perhaps paradoxically, we strongly rely on BM’s round reduction technique, yet we rely on a non-black-box security analysis. Our main technical lemma shows that if infinitely-often one-way functions do not exist (i.e., if we can invert any function for all sufficiently large input lengths), then BM’s round reduction actually works:

**Lemma 1.2** (informally stated). Either infinitely-often one-way functions exist, or BM’s round-reduction transformation turns a $k(\cdot)$-round puzzle into a $(k(\cdot)-1)$-round puzzle.

We provide a proof outline of Lemma 1.2 in Section 1.5. The proof of Theorem 1.1 now easily follows by considering two cases:

**Case 1: (Infinitely-often) one-way functions exists.** In such a world, we can rely on Rompel’s construction of a universal one-way hashfunction [NY89, Rom90] to get a 2-round puzzle.

---

3The example from [GKM+00] isn’t quite captured by our notion of a computational puzzle as their challenger is not public coin.

4Recall that a one-way function $f$ is a function that is efficiently computable, yet there does not exist a PPT attacker $A$ and polynomial $p(\cdot)$ such that $A$ inverts $f$ with probability $\frac{1}{p(n)}$ for infinitely many inputs lengths $n \in N$. A function $f$ is infinitely often one-way if the same conditions hold except that we only require that no PPT attacker $A$ succeeds in inverting $f$ with probability $\frac{1}{p(n)}$ for all sufficiently large $n \in N$—i.e., it is hard for invert $f$ “infinitely often”
**Case 2: (Infinitely-often) one-way functions does not exist.** In such a world, by Lemma 1.2, BM’s round reduction preserves soundness of the underlying protocol and thus we have gotten a puzzle with one round less. We can next iterate BM’s round reduction any constant number of times.

A natural question is whether we can collapse more than a constant number of rounds. Our next result—which characterizes the existence of poly(n)-round puzzles—shows that this is unlikely.

**Theorem 1.3** (informally stated). For every $\epsilon > 0$, there exists an $n^\epsilon$-round puzzle if and only if $\text{PSPACE} \nsubseteq \text{BPP}$.

In particular, if $n^\epsilon$-round puzzles imply $O(1)$-round puzzles, then by combining Theorem 1.1 and Theorem 1.3, we have that $\text{PSPACE} \nsubseteq \text{BPP}$ implies the existence of a hard-on-average problem in NP, which seems unlikely. Theorem 1.3 also shows that the notion of an interactive puzzle (with a super constant-number of rounds) indeed is a non-trivial generalization of average-case hardness in NP.

Theorem 1.3 follows using standard techniques: Any puzzle $C$ can be broken using a PSPACE oracle (as the optimal strategy can be found using a PSPACE oracle), so if $\text{PSPACE} \subseteq \text{BPP}$, it can also be broken by a probabilistic polynomial-time algorithm. For the other direction, recall that worst-case to average-case reductions are known for PSPACE [FF93, BFNW93]. In other words, there exists a language $L \in \text{PSPACE}$ that is hard-on-average assuming $\text{PSPACE} \nsubseteq \text{BPP}$. Additionally, recall that PSPACE is closed under complement. We then construct a puzzle where $C$ first samples a hard instance for $L$ and then asks $A$ to determine whether $x \in L$ and next provide an interactive proof—using [Sha92, LFKN92] which is public-coin—for containment or non containment in $L$. This puzzle clearly satisfies the completeness condition. Computational soundness, on the other hand, follows directly from the hard-on-average property of $L$ (and the unconditional soundness of the interactive proof of [Sha92]).

We next present some complexity-theoretic consequences of our treatment of interactive computational puzzles.

### 1.2 Perfect Completeness and TFNP Hardness

We consider two fundamental open problems in complexity theory:

- Does the existence of a hard-on-average language in NP imply the existence of one-way function?
- Does the existence of a hard-on-average language in NP imply the existence of a hard problem in TFNP (i.e., the class of total search problems)?

Roughly speaking, our main corollary demonstrates that one of the above open problems has a positive answer. Let us elaborate.

**One-way functions from Average-case Hardness** Perhaps the most important open problem in the foundation of Cryptography is whether the existence of a hard-on-average problem in NP implies the existence of one-way functions. One-way functions are both necessary [IL89] and sufficient for many of the central cryptographic tasks (e.g., pseudorandom generators [HILL99], pseudorandom functions [GGM84], private-key encryption [GM84, BM84]). In more complexity-theoretic terms, one-way functions are equivalent to the existence of an efficient method for sampling hard instances in NP together with their witnesses. Impagliazzo refers to a world where hard-on-average problems in NP exist, but one-way functions do not, as Pessiland [Imp95].

---

5To see why the existence of such a sampling method implies the existence of one-way functions, consider the function $f$ which takes the random coins used by the sampling method and outputs the instance generated by it.
As far as we know, there are only two approaches towards demonstrating the existence of one-way functions from average-case NP hardness: (1) Ostrovsky and Wigderson [OW93a] demonstrate such an implication assuming that NP has zero-knowledge proofs [GMW91], (2) Komargodski et al. [KMN+14] demonstrate the implication (in fact, an even stronger implication, showing worst-case hardness of NP implies one-way functions) assuming the existence of indistinguishability obfuscators [BGI01]. Both of these additional assumptions are not known to imply one-way functions on their own (in fact, they unconditionally exist if NP ⊆ BPP).

TFNP Hardness Another central problem in complexity theory concerns the hardness of total search problems [MP91]: the class TFNP (total function NP) is the search analog of NP with the additional guarantee that any instance has a solution. In other words, TFNP is class of search problems in NP ∩ coNP (i.e., $F(\text{NP} \cap \text{coNP})$). In recent years, TFNP has attracted extensive attention due to its natural syntactic subclasses that capture the computational complexity of important search problems from algorithmic game theory, combinatorial optimization and computational topology—perhaps most notable among those are the classes PPAD [Pap94, GP16], which characterizes the hardness of computing Nash equilibrium [DGP09, CDT09, DP11], and PLS [JPY85], which characterizes the hardness of local search. A central open problem is whether (average-case) NP hardness implies (average-case) TFNP hardness. A recent elegant result by Hubacek, Naor, and Yogev [HNY17] shows that under “derandomization” assumptions [NW94, IW97, MV05, BOV07]—the existence of certain Nisan-Wigderson (NW) [NW94] type pseudorandom generators that fool non-deterministic bounded size circuits— (almost everywhere) average-case hardness of NP implies average-case hardness of TFNP. As a consequence, they showed that in Pessiland, TFNP is average-case hard under NW-type derandomization assumptions.

But it remains open whether just average-case hardness of NP suffices. Hubacek et al. also present another condition under which TFNP is average-case hard: assuming the existence of one-way functions and non-interactive witness indistinguishable proofs (NIWI) [FS90, DN00, BOV07] for NP.

TFNP Hardness in Pessiland By using our interactive average-case hardness treatment, we are able to unconditionally show that in Pessiland (where NP is hard-on-average but one-way functions do not exist), TFNP is hard (on average).

More precisely, we show that one of the above open problems must has a positive resolution: The existence of an (almost everywhere) hard-on-average problem in NP unconditionally implies either (1) one-way functions, or (2) average-case hardness of TFNP.

**Theorem 1.4.** If NP contains an (almost-everywhere) hard-on-average problem, then either (1) one-way functions exists, or 2) TFNP is hard-on-average.

By combining Theorem 1.4 with the result of [HNY17] (that NIWI for NP and one-way functions imply hardness of TFNP), we get that it suffices to assume NIWI for NP and an (almost-everywhere) hard-on-average problem in NP to conclude hardness of TFNP. As far as we know, this constitutes the first result where witness indistinguishability [FS90] can be non-trivially used without assuming the existence of one-way functions.

Theorem 1.4 is proven in the following steps: (1) As mentioned above, an (almost-everywhere) hard-on-average problem in NP yields a 2-round puzzle; (2) We can next use a standard technique from the literature on interactive proofs (namely the result of [FGM+89]) to turn this puzzle into a 3-round puzzle with perfect completeness. (3) We next observe that the BM transformation preserves perfect completeness of the protocol.

---

6Such pseudorandom generators are known to exist based on the assumption that $\text{E} = \text{DTIME}(2^{O(n)})$ has a function of non-deterministic circuit complexity $2^{\Omega(n)}$ [MV05].

7They also show that average-case hardness of NP implies an average-case hard problem in TFNP/poly (i.e., TFNP with a non-uniform verifier). In essence, this follows since non-uniformity enables unconditional derandomization.
Thus, by Lemma 1.2, either infinitely-often one-way functions exist, or we can get a 2-round puzzle with perfect completeness. (4) We finally observe that the existence of a 2-round puzzle with perfect completeness is (syntactically) equivalent to the existence of a hard-on-average problem in TFNP (with respect to the uniform distribution on instances).

The above proof approach actually only concludes a slightly weaker form of Theorem 1.4—we only show that either TFNP is hard or infinitely-often one-way functions exist. But we can get the proof also of the stronger conclusion (i.e., conclude the existence of standard (i.e., “almost-everywhere”) one-way functions), by noting that an almost-everywhere hard-on-average language in NP actually implies an 2-round puzzle satisfying a “almost-everywhere” notion of soundness, and for such “almost-everywhere puzzles”, Lemma 1.2 can be strengthened to show that either one-way functions exist, or BM’s round-reduction works.\footnote{More precisely, the variant of Lemma 1.2 says that either one-way functions exist, or the existence of a $k$-round almost-everywhere puzzle yields the existence of a $k-1$-round puzzle (with the standard, infinitely-often, notion of soundness).}

1.3 The Complexity of Non-trivial Public-coin Arguments

Soon after the introduction of interactive proof by Goldwasser, Micali and Rackoff \cite{GMR89} and Babai and Moran \cite{BM88}, Brassard, Chaum and Crepeau \cite{BCC88} introduced the notion of an interactive argument. Interactive arguments are defined identically to interactive proofs, but we relax the soundness condition to only hold with respect to non-uniform PPT algorithms (i.e., no non-uniform PPT algorithm can produce proofs of false statements, except with negligible probability).

Interactive arguments have proven extremely useful in the cryptographic literature, most notably due to the feasibility (assuming the existence of collision-resistant hashfunctions) of succinct public-coin argument systems for NP—namely, argument systems with sublinear, or even polylogarithmic communication complexity \cite{Kil92, Mic00}. Under widely believed complexity assumptions (i.e., NP not being solvable in subexponential time), interactive proofs cannot be succinct \cite{GH98}.

A fundamental problem regarding interactive arguments involves characterizing the complexity of non-trivial argument systems—namely interactive arguments that are not interactive proofs (in other words, the soundness condition is inherently computational). While we do not have an explicit reference for the discussion of non-trivial arguments, this notion and the problem of understanding whether the existence of non-trivial arguments implies some notion of one-wayness or average-case hardness in NP has been discussed in community for at least fifteen years.

We focus our attention on public-coin arguments (similar to our treatment of puzzles). Using our interactive-average-case hardness treatment, we are able to establish an “almost-tight” characterization of constant-round public-coin non-trivial arguments.

**Theorem 1.5** (informally stated). The existence of a $O(1)$-round public-coin non-trivial argument for any language $L$ implies a hard-on-average language in $NP/\text{poly}$. Conversely, the existence of a hard-on-average language in $NP$ implies an (efficient-prover) 2-round public-coin non-trivial argument for NP.

The first part of the theorem is shown by observing that any public-coin non-trivial argument can be turned into a non-uniform puzzle (where the challenger is a non-uniform PPT algorithm), and next observing that our round-collapse theorem also applies to non-uniform puzzles. The second part follows from the observation that we can take any NP proof for some language $L$ and extending it into a 2-round argument where the verifier samples a random statement $x'$ from a hard-on-average language $L$ and next requiring the prover to provide a witness $w$ that either $x \in L$ or $x' \in L'$. Completeness follows trivially, and computational soundness follows directly if $L'$ is sufficiently hard-on-average (in the sense that it is hard to find witnesses to true statements with inverse polynomial probability). This argument system is not a proof, though, since by \footnote{Wee \cite{Wee05} considers a notion of a non-trivial argument, but his notion refers to what today is called a succinct argument.}
the hard-on-average property of \(L'\), there must exist infinitely many input lengths for which random instances are contained in \(L'\) with inverse polynomial probability.

We finally observe that the existence of \(n^\epsilon\)-round non-trivial public-coin arguments is equivalent to \(\text{PSPACE} \not\subseteq \text{P}/\text{poly}\).

**Theorem 1.6** (informally stated). For every \(\epsilon > 0\), there exists an (efficient-prover) \(n^\epsilon\)-round non-trivial public-coin argument (for \(\text{NP}\)) if and only if \(\text{PSPACE} \not\subseteq \text{P}/\text{poly}\).

The “only-if” direction follows just as the only-if direction of Theorem 1.3. The “if” direction follows by combining a standard \(\text{NP}\) proof with the puzzle from Theorem 1.3 (which becomes sound w.r.t. nu PPT attacker assuming \(\text{PSPACE} \not\subseteq \text{P}/\text{poly}\)), and requiring the prover to either provide the \(\text{NP}\) witness, or to provide a solution to the puzzle.

### 1.4 Perspective

Our results reveal that an interactive notion of average-case hardness (i.e., interactive computational puzzles) is intriguing not only in its own right, but also leads to insights into classic questions in complexity theory and cryptography; most notably, we are able to unconditionally establish that \(\text{TFNP}\) is (average-case) hard in Pessiland.

Theorem 1.3 demonstrates that there may exist a natural hierarchy of average-case hard problems between \(\text{NP}\) and \(\text{PSPACE}\), characterized by the round complexity of the puzzles. By Theorem 1.1, gaps in the hierarchy need to come from super constant increases in the number rounds. We leave open the intriguing question of characterizing the complexity of \(k(\cdot)\)-round puzzles for \(k(\cdot) \in \omega(1) \cup o(n)\); most notably, the question of understanding the complexity-theoretic implications of \(O(\log n)\)-round puzzles is open.

### 1.5 Proof Overview for Lemma 1.2

We here provide a proof overview of our main technical lemma. As mentioned, we shall show that if one-way functions do not exist, then Babai-Moran’s round reduction method actually works. Towards this we will rely on two tools:

- **Pre-image sampling.** By the result of Impagliazzo and Levin [IL90], the existence of so-called “distributional one-way functions” (function for which it is hard to sample a uniform pre-image) imply the existence of one-way function. So if one-way functions do not exist, we have that for every efficient function \(f\), given a sample \(f(x)\) for a random input \(x\), we can efficiently sample a (close to random) pre-image \(x'\).

- **Raz’s sampling lemma** (from the literature on parallel repetition for 2-prover games and interactive arguments [Raz98, HPWP10, CP15]). This lemma states that if we sample \(\ell\) uniform \(n\)-bit random variables \(R_1, R_2, \ldots, R_\ell\) conditioned on some event \(W\) that happens with sufficiently large probability \(\epsilon\), then the conditional distribution \(R_i\) of a randomly selected index \(i\) will be close to uniform. More precisely, the statistical distance will be \(\sqrt{\frac{\log(\frac{1}{\epsilon})}{\ell}}\), so even if \(\epsilon\) is tiny, as long as we have sufficiently many repetitions \(\ell\), the distance will be small.\(^{10}\)

To see how we will use these tools, let us first recall the BM transformation (and its proof for the case of information-theoretically sound protocols). To simplify our discussion, we here focus on showing how to collapse a 3-round public-coin protocol between a prover \(P\) and a public-coin verifier \(V\) into a 2-round protocol. We denote a transcript of the 3-round protocol \((p_1, r_1, p_2)\) where \(p_1\) and \(p_2\) are the prover messages and \(r_1\)

\(^{10}\)Earlier works [HPWP10, CP15] always used Raz’ lemma when \(\epsilon\) was non-negligible. In contrast, we will here use it also when \(\epsilon\) is actually negligible.
is the randomness of the verifier. Let \( n = |p_1| \) be the length of the prover message. The BM transformation collapses this protocol into a 2-round protocol in the following two steps:

**Step 1: Reducing soundness error:** First, use a form of parallel repetition to make the soundness error \( 2^{-n^2} \) (i.e., extremely small). More precisely, consider a 3-round protocol where \( P \) first still send just \( p_1 \), next the verifier picks \( \ell = n^2 \) random strings \( \vec{r} = (r_1^1, \ldots, r_1^\ell) \), and finally \( P \) needs to provide accepting answers \( p_2^1 = (p_2^1, \ldots, p_2^\ell) \) to all of the queries \( \vec{r} \) (so that for every \( i \in [\ell] \), \( (p_1, r_1^i, p_2^i) \) is accepting transcript).

**Step 2: Swap order of messages:** Once the soundness error is small, yet the length of the first message is short, we can simply allow the prover to pick it first message \( p_1 \) after having \( \vec{r} \). In other words, we now have a 2-round protocol where \( V \) first picks \( \vec{r} \), then the prover responds by sending \( p_1, p_2 \). This swapping preserves soundness by a simple union bound: since (by soundness) for every string \( p_1 \), the probability over \( \vec{r} \) that there exists some accepting response \( \vec{r} \) is \( 2^{-n^2} \), it follows that with probability at most \( 2^n \times 2^{-n^2} = 2^{-n} \) over \( \vec{r} \), there exists some \( p_1 \) that has an accepting \( p_2 \) (as the number of possible first messages \( p_1 \) is \( 2^n \)). Thus soundness still holds (with a \( 2^n \) degradation) if we allow \( P \) to choose \( p_1 \) after seeing \( \vec{r} \).

For the case of computationally sound protocols, the “logic” behind both steps fail: (1) it is not known how to use parallel repetition to reduce soundness error beyond being negligible, (2) the union bound cannot be applied since, for computationally sound protocols, it is not the case that responses \( p_2 \) do not exist, rather, they are just hard to find. Yet, as we shall see, using the above tools, we present a different proof strategy. More precisely, to capture computational hardness, we show a reduction from any polynomial-time attacker \( A \) that breaks soundness of the collapsed protocol with some inverse polynomial probability \( \epsilon \), to a polynomial-time attacker \( B \) that breaks soundness of the original 3-round protocol.

\( B \) starts by sampling a random string \( \vec{r} \) and computes \( A \)’s response given this challenge \( (p_1^1, p_2^1) \leftarrow A(\vec{r}) \). If the response is not an accepting transcript, simply abort; otherwise, take \( p_1^1 \) and forward externally as \( B \)’s first message. (Since \( A \) is successful in breaking soundness, we have that \( B \) won’t abort with probability \( \epsilon \).) Next, \( B \) gets a verifier challenge \( r \) from the external verifier and needs to figure out how to provide an answer to it. If \( B \) is lucky and \( r \) is one of the challenges \( r^i \) in \( \vec{r} \), then \( B \) could provide the appropriate \( p_2 \) message, but this unfortunately will only happen with negligible probability. Rather, \( B \) will try to get \( A \) to produce another accepting transcript \( (p_1^2, r^i, p_2^i) \) that (1) still contains \( p_1^1 \) as the prover’s first message (i.e., \( p_1^1 = p_1^i \)), and (2) contains \( r \) in some coordinate \( i \) of \( r^i \). To do this, \( B \) will consider the function \( f(\vec{r}, z, i) \)---which runs \( (p_1, p_2) \leftarrow A(\vec{r}; z) \) (i.e., \( A \) has its randomness fixed to \( z \)) and outputs \( (p_1, r^i) \) if \( (p_1, p_2) \) is accepting and \( \bot \) otherwise---and runs the pre-image sampler for this function \( f \) on \( (p_1^i, r^i) \) to recover some new verifier challenge, randomness, index tuple \( (r^j, z, i) \) which leads \( A(r^j; z) \) to produce a transcript \( (p_1^i, r^j, p_2^i) \) of the desired form, and \( B \) can subsequently forward externally the \( i \)’th coordinate of \( p_2^i \) as its response and convince the external verifier.

So, as long as the pre-image sampler indeed succeeds with high enough probability, we have managed to break soundness of the original 3-round protocol. The problem is that the pre-image sampler is only required to work given outputs that are correctly distributed over the range of the function \( f \), and the input \( (p_1, r) \) that we now feed it may not be so—for instance, perhaps \( A(\vec{r}) \) chooses the string \( p_1 \) as a function of \( \vec{r} \). So, whereas the marginal distribution of both \( p_1 \) and \( r \) are correct, the joint distribution is not. In particular, the distribution of \( r \) conditioned on \( p_1 \) may be off. We, however, show how to use Raz’s lemma to argue that if the number of repetitions \( \ell \) is sufficiently bigger than the length of \( p_1 \), the conditional distribution of \( r \) cannot be too far off from being uniform (and thus the pre-image sampler will work). On a high-level, we proceed as follows:

* Note that in the one-way function experiment, we can think of the output distribution \( (p_1, r) \) of \( f \) on a random input, as having been produced by first sampling \( p_1 \) and next, if \( p_1 \neq \bot \), sampling \( r \) conditioned
on the event $W_{p_1}$ that $A$ generates a successful transcript with first-round prover message $p_1$, and finally sampling a random index $i$ and outputting $p_1$ and $r_i$ (and otherwise output ⊥).

- Note that by an averaging argument, we have that with probability at least $\frac{\epsilon}{2}$ over the choice of $p_1$, $\Pr[W_{p_1}] \geq \frac{\epsilon}{2n+\ell}$ (otherwise, the probability that $A$ succeeds would need to be smaller than $\frac{\epsilon}{2} + 2^n \times 2^{n+\ell} = \epsilon$, which is a contradiction).

- Thus, whenever we pick such a “good” $p_1$ (i.e., a $p_1$ such that $\Pr[W_{p_1}] \geq \frac{\epsilon}{2n+\ell}$), by Raz’ lemma the distribution of $r_i$ for a random $i$ can be made $\frac{1}{p(n)}$ close to uniform for any polynomial $p$ by choosing $\ell$ to be sufficiently large (yet polynomial). Note that even though the lower bound on $\Pr[W_{p_1}]$ is negligible, the key point is that it is independent of $\ell$ and as such we can still rely on Raz lemma by choosing a sufficiently large $\ell$. (As we pointed out above, this usage of Raz’ lemma even on very “rare” events—with negligible probability mass—is different from how it was previously applied to argue soundness for computationally sound protocols [HPWP10, CP15].)

- It follows that conditioned on picking such a “good” $p_1$, the pre-image sampler will also successfully generate correctly distributed preimages if we feed him $p_1, r$ where $r$ is randomly sampled. But this is exactly the distribution that $B$ feeds to the pre-image sampler, so we conclude that with probability $\frac{\epsilon}{2}$ over the choice of $p_1$, $B$ will manage to convince the outside verifier with probability close to 1.

This concludes the proof overview for 3-round protocols. When the protocol has more than 3 rounds, we can apply a similar method to collapse the last rounds of the protocol. The analysis now needs to be appropriately modified to condition also on the prefix of the partial execution up until the last rounds.

2 Preliminaries

We assume familiarity with basic concepts such as Turing machines, interactive Turing machine, polynomial-time algorithms, probabilistic polynomial-time algorithms (PPT), non-uniform polynomial-time and non-uniform PPT algorithms. A function $\mu$ is said to be negligible if for every polynomial $p(\cdot)$ there exists some $n_0$ such that for all $n > n_0$, $\mu(n) \leq \frac{1}{p(n)}$. For any two random variables $X$ and $Y$, we let $\text{SD}(X,Y) = \max_{T \subseteq U} |\Pr[X \in T] - \Pr[Y \in T]|$ denote the statistical distance between $X$ and $Y$.

Basic Complexity Classes Recall that $\mathsf{P}$ is the class of languages $L$ decidable in polynomial time (i.e., there exists a polynomial-time algorithm $M$ such that for every $x \in \{0,1\}^*$, $M(x) = L(x)$), $\mathsf{P}/\text{poly}$ is the class of languages decidable in non-uniform polynomial time, and $\mathsf{BPP}$ is the class of languages decidable in probabilistic polynomial time with probability $2/3$ (i.e., there exists a PPT $M$ such that for every $x \in \{0,1\}^*$, $\Pr[M(x) = L(x)] \geq 2/3$ where we abuse of notation and define $L(x) = 1$ if $x \in L$ and $L(x) = 0$ otherwise.)

We refer to a relation $\mathcal{R}$ over pairs $(x,y)$ as being polynomially bounded if there exists a polynomial $p(\cdot)$ such that for every $(x,y) \in \mathcal{R}$, $|y| \leq p(|x|)$. We denote by $L_{\mathcal{R}}$ the language characterized by the “witness relation” $\mathcal{R}$—i.e., $x \in L$ iff there exists some $y$ such that $(x,y) \in \mathcal{R}$. We say that a relation $\mathcal{R}$ is polynomial-time (resp. non-uniform polynomial-time) if $\mathcal{R}$ is polynomially-bounded and the languages consisting of pairs $(x,y) \in \mathcal{R}$ is in $\mathsf{P}$ (resp. $\mathsf{P}/\text{poly}$). $\mathsf{NP}$ (resp $\mathsf{NP}/\text{poly}$) is the class of languages $L$ for which there exists a polynomial-time (resp. non-uniform polynomial-time) relation $\mathcal{R}$ such that $x \in L$ iff there exists some $y$ such that $(x,y) \in \mathcal{R}$.

Search Problems A search problem $\mathcal{R}$ is simply a polynomially-bounded relation. We say that the search problem is solvable in polynomial-time (resp. non-uniform polynomial time) if there exists a polynomial-time (resp. non-uniform polynomial-time) algorithm $M$ that for every $x \in L_{\mathcal{R}}$ outputs a “witness” $y$ such that
Let \((x, y) \in \mathcal{R}\). Analogously, \(\mathcal{R}\) is solvable in \(\text{PPT}\) if there exists some \(\text{PPT}\) \(M\) that for every \(x \in L_\mathcal{R}\) outputs a “witness” \(y\) such that \((x, y) \in \mathcal{R}\) with probability \(2/3\).

An \(\text{NP}\) search problem \(\mathcal{R}\) is total if for every \(x \in \{0, 1\}^*\) there exists some \(y\) such that \((x, y) \in \mathcal{R}\) (i.e., every instance has a witness). We refer to \(\text{FNP}\) (function \(\text{NP}\)) as the class of \(\text{NP}\) search problems and \(\text{TFNP}\) (total-function \(\text{NP}\)) as the class of total \(\text{NP}\) search problems.

### 2.1 One-way functions

We recall the definition of one-way functions (see e.g., [Gol01]). Roughly speaking, a function \(f\) is one-way if it is polynomial-time computable, but hard to invert for \(\text{PPT}\) attackers. The standard (cryptographic) definition of a one-way function requires every \(\text{PPT}\) attacker to fail (with high probability) on all sufficiently large input lengths. We will also consider a weaker notion of an infinitely-often one-way function [OW93a] which only requires the \(\text{PPT}\) attacker to fail for infinitely many inputs length (in other words, there is no \(\text{PPT}\) attacker that succeeds on all sufficiently large input lengths, analogously to complexity-theoretic notions of hardness).

**Definition 2.1.** Let \(f : \{0, 1\}^* \rightarrow \{0, 1\}^*\) be a polynomial-time computable function. \(f\) is said to be a one-way function (\(\text{OWF}\)) if for every \(\text{PPT}\) algorithm \(A\), there exists a negligible function \(\mu\) such that for all \(n \in \mathbb{N}\),

\[
\Pr[x \leftarrow \{0, 1\}^n; y = f(x) : A(1^n, y) \in f^{-1}(f(x))] \leq \mu(n)
\]

\(f\) is said to be an infinitely-often one-way function (\(\text{ioOWF}\)) if the above condition holds for infinitely many \(n \in \mathbb{N}\) (as opposed to all).

We may also consider a notion of a non-uniform (a.k.a. “auxiliary-input”) one way function, which is identically defined except that (a) we allow \(f\) to be computable by a non-uniform \(\text{PPT}\), and (b) the attacker \(A\) is also allowed to be a non-uniform \(\text{PPT}\).

### 2.2 Interactive Proofs and Arguments

We recall basic definitions of interactive proofs [GMR89, BM88] and arguments [BCC88]. An interactive protocol \((P, V)\) is a pair of interactive Turing machine; we denote by \((P_1, P_2)(x)\) the output of \(P_2\) in an interaction between \(P_1\) and \(P_2\) on common input \(x\).

**Definition 2.2.** An interactive protocol \((P, V)\) is an interactive proof system for a language \(L \subseteq \{0, 1\}^*\) if \(V\) is \(\text{PPT}\) and the following conditions hold:

**Completeness:** There exists a negligible function \(\mu(\cdot)\) such that for every \(x \in L\),

\[
\Pr[(P, V)(x) = 1] \geq 1 - \mu(|x|)
\]

**Soundness:** For every \(\text{Turing machine}\) \(P^*\), there exists a negligible function \(\mu(\cdot)\) such that for every \(x \notin L\),

\[
\Pr[(P^*, V)(x) = 1] \leq \mu(|x|)
\]

If the soundness condition is relaxed to only hold for all non-uniform \(\text{PPT}\) \(P^*\), we refer to \((P, V)\) as an interactive argument for \(L\). We refer to \((P, V)\) as a public-coin proof/argument system if \(V\) simply sends the outcomes of its coin tosses to the prover (and only performs computation to determine its final verdict).

Whenever \(L \in \text{NP}\), we say that \((P, V)\) has an efficient prover if there exists some witness relation \(\mathcal{R}\) that characterizes \(L\) (i.e., \(L_\mathcal{R} = L\)) and a \(\text{PPT}\) \(\tilde{P}\) such that \(P(x) = \tilde{P}(x, w)\) satisfies the completeness condition for every \((x, w) \in \mathcal{R}\).
2.3 Average-Case Complexity

We recall some basic notions from average-case complexity. A distributional problem is a pair $(L, D)$ where $L \subseteq \{0, 1\}^*$ and $D$ is a PPT; we say that $(L, D)$ is an NP (resp. NP/poly) distributional problem if $L \in$ NP (resp. $L \in$ NP/poly). Roughly speaking, a distributional problem $(L, D)$ is hard-on-average if there does not exist some PPT algorithm that can decide instances drawn from $D$ with probability significantly better than 1/2.

**Definition 2.3** ($\delta$-hard-on-the-average). We say that a distributional problem $(L, D)$ is $\delta$-hard-on-the-average ($\delta$-HOA) if there does not exist some PPT $A$ such that for every sufficiently large $n \in \mathbb{N}$,

$$\Pr[x \leftarrow D(1^n) : A(1^n, x) = L(x)] > 1 - \delta$$

We say that a distributional problem $(L, D)$ is simply hard-on-the-average (HOA) if it is $\delta$-HOA for some $\delta > 0$.

We also define an notion of HOA w.r.t. non-uniform PPT algorithm (nuHAO) in exactly the same way but where we allow $A$ to be a non-uniform PPT (as opposed to just a PPT).

The above notion average-case hardness (traditionally used in the complexity-theory literature) is defined analogously to the notion of an infinitely-often one-way function: we simply require every PPT “decider” to fail for infinitely many $n \in \mathbb{N}$. For our purposes, we will also rely on an “almost-everywhere” notion of average-case hardness (similar to standard definitions in the cryptography, and analogously to the definition of a one-way function), where we require that every decider fails on all (sufficiently large) input lengths.

**Definition 2.4** (almost-everywhere hard-on-the-average (aeHOA)). We say that a distributional problem $(L, D)$ is almost-everywhere $\delta$ hard-on-the-average ($\delta$-aeHOA) if there does not exist some PPT $A$ such that for infinitely many $n \in \mathbb{N}$,

$$\Pr[x \leftarrow D(1^n) : A(1^n, x) = L(x)] > 1 - \delta$$

We say $(L, D)$ is almost-everywhere hard-on-the-average (aeHOA) if $(L, D)$ is $\delta$-aeHOA for some $\delta > 0$.

We move on to defining hard-on-the-average search problems. A distributional search problem is a pair $(\mathcal{R}, D)$ where $\mathcal{R}$ is a search problem and $D$ is a PPT. If $\mathcal{R}$ is an NP search problem (resp. NP/poly search problem), we refer to $(\mathcal{R}, D)$ as a distributional NP (resp. NP/poly) search problem.

**Definition 2.5** (hard-on-the-average search (SearchHOA)). We say that a distributional search problem $(\mathcal{R}, D)$ is $\delta$-hard-on-the-average ($\delta$-SearchHOA) if there does not exist some PPT $A$ such that for every sufficiently large $n \in \mathbb{N}$,

$$\Pr[x \leftarrow D(1^n); (b, x) \leftarrow A(1^n, x) : (b = L_{\mathcal{R}}(x)) \land ((b = 1) \Rightarrow (x, w) \in \mathcal{R})] > 1 - \delta$$

$(\mathcal{R}, D)$ is simply SearchHOA if there exists $\delta > 0$ such that $(\mathcal{R}, D)$ is $\delta$-SearchHOA.

We can analogously define an almost-everywhere notion, aeSearchHOA, of SearchHOA (by replacing “for every sufficiently large $n \in \mathbb{N}$” with “for infinitely many $n \in \mathbb{N}$”) as well as a non-uniform notion, nuSearchHOA, (by replacing PPT with non-uniform PPT).

The following lemmas which essentially directly follow from the result of [IL90, BCGL92, Tre05] will be useful to us. (These results were originally only stated for the standard notion of HOA, whereas we will require it also for the almost-everywhere notion; as we explain in more detail in Appendix A, these results however directly apply also for the almost-everywhere notion of HOA.) The first results from [IL90] (combined with [Tre05]) shows that without loss of generality, we can restrict our attention to the uniform distribution over statements $x$; we denote by $U_p$ a PPT such that $U_p(1^n)$ simply samples a random string in $\{0, 1\}^{p(n)}$. 

12
Lemma 2.1 (Private to public distributions). Suppose there exists a distributional NP problem \((L, D)\) that is HOA (resp., aeHOA or nuHOA). Then, there exists a polynomial \(p\) and an NP-language \(L'\) such that \((L', U_p)\) is HAO (resp. aeHOA or nuHOA).

The next result from \([Tre05]\) shows that when the distribution over instances is uniform, we can amplify the hardness.

Lemma 2.2 (Hardness amplification). Let \(p\) be a polynomial and suppose there exists a distributional NP-problem \((L, U_p)\) that is HOA (resp., aeHOA or nuHOA). Then, for every \(\delta < \frac{1}{2}\), there exists some polynomial \(p'\) and NP-language \(L'\) such that \((L', U_{p'})\) is \(\delta\)-HOA (resp., \(\delta\)-aeHOA or \(\delta\)-nuHOA).

Finally, by \([BCGL92]\) (combined with \([Tre05]\), \([IL90]\)) we have a decision-to-search reduction.

Lemma 2.3 (Search to decision). Suppose there exists a distributional NP (resp. NP/poly) search problem \((R, D)\) that is SearchHOA (resp., nuSearchHOA). Then, there a polynomial \(p\) and an NP (resp. NP/poly) language \(L\) such that \((L, U_p)\) is HOA (resp., nuHOA).

3 Interactive Computational Puzzles

Roughly speaking, an interactive computational puzzle is described by an interactive polynomial-time public-coin challenger \(C\) having the property that (a) there exists an inefficient \(A\) that succeeds in convincing \(C(1^n)\) with probability negligibly close to 1, yet (b) no PPT attacker \(A^*\) can make \(C(1^n)\) output 1 with inverse polynomial probability for sufficiently large \(n\).

Definition 3.1 (interactive puzzle). An interactive algorithm \(C\) is referred to as a \(k(\cdot)\)-round (computational) puzzle if the following conditions hold:

\(k(\cdot)\)-round public-coin: \(C\) is an (interactive) PPT that on input \(1^n\) (a) only communicates in \(k(n)\) communication rounds, (b) simply sends the outcomes of its coin tosses in each communication round, and (c) only performs some deterministic computation to determine its final verdict (after having received the message in round \(k(n)\)).

Completeness/Non-triviality: There exists a (possibly unbounded) Turing machine \(A\) and a negligible function \(\mu_C(\cdot)\) such that for all \(n \in \mathbb{N}\),

\[
\Pr[(A, C)(1^n) = 1] = 1 - \mu(n)
\]

Computational Soundness: There does not exist a PPT machine \(A^*\) and polynomial \(p(\cdot)\) such that for all sufficiently large \(n \in \mathbb{N}\),

\[
\Pr[(A^*, C)(1^n) = 1] \geq \frac{1}{p(n)}
\]

In other words, a \(k(\cdot)\)-round puzzle, \(C\), gives rise to an \(k(\cdot)\)-round public-coin interactive proof \((P, V)\) (where \(P = A, V = C\)) for the “trivial” language \(L = \{0, 1\}^*\) with the property that there does not exist a PPT prover that succeeds in convincing the verifier with inverse polynomial probability for all sufficiently large \(n\).

We may also define an almost-everywhere notion of a puzzle by replacing “for all sufficiently large \(n \in \mathbb{N}\)” in the soundness condition with “for infinitely many \(n \in \mathbb{N}\)”, and a non-uniform notion of a puzzle \(C\) which allows both \(C\) and \(A^*\) to be non-uniform PPT (as opposed to just PPT).

A puzzle \(C\) is said to have perfect completeness if the “completeness error”, \(\mu_C(n)\), is 0—in other words, the completeness condition holds with probability 1.
Remark 3.1. One can consider a more relaxed notion of a \((c(\cdot), s(\cdot))\)-puzzle for \(c(n) = s(n) + 1/\text{poly}(n)\), where the completeness condition is required to hold with probability \(c(\cdot)\) for every sufficiently large \(n \in \mathbb{N}\), and the soundness condition holds with probability \(s(\cdot)\) for every sufficiently large \(n \in \mathbb{N}\). But, by “Chernoff-type” parallel-repetition theorems for computationally-sound public-coin protocols [PV12, HPWP10, CL10, CP15], the existence of such a \(k(\cdot)\)-round \((c(\cdot), s(\cdot))\)-puzzle implies the existence of a \(k(\cdot)\)-round puzzle. The same holds for almost-everywhere (resp. non-uniform) puzzles.

3.1 Characterizing 2-round Puzzles

In this section we make some basic observations regarding 2-round puzzle; these results mostly follow using standard results in the literature. We begin by observing that the existence of ioOWF imply the existence of 2-round puzzles.

Proposition 3.2. Assume the existence of ioOWFs (resp. non-uniform ioOWF). Then, there exists a 2-round puzzle (resp. non-uniform puzzle).

Proof: By the result of Rompel [Rom90] (see also [KK05, HHR+10]), we have that ioOWFs imply the existence of infinitely-often “second-preimage” resistant hash-function families that compress \(n\) bits to \(n/2\) bits. This, in turn, directly yields a simple 2-round puzzle where the challenger \(C(1^n)\) uniformly samples a hashfunction \(h\) and input \(x \in \{0,1\}^n\) and sends \((h, x)\) to the adversary; \(C\) accepts a response \(x'\) if \(|x'| = n\), \(x' \neq x\) and \(h(x') = h(x)\). Since the hash function is compressing, we have that there exists a negligible function \(\mu\) such that with probability \(1 - \mu(n)\), a random \(x \in \{0,1\}^n\) will have a “collision” \(x'\) and thus an unbounded \(\mathcal{A}\) can easily find a collision and thus completeness follows. Computational soundness, on the other hand, directly from the (infinitely-often) second-preimage resistance property. The same result holds also if we start with non-uniform ioOWFs, except that we now get a non-uniform puzzle.

We turn to showing that any aeHOA distributional NP problem implies a 2-round puzzle. (In fact, it even implies an almost-everywhere puzzle.)

Lemma 3.3. Suppose there exists a distributional NP problem \((L, D)\) that is aeHOA. Then there exist an (almost-everywhere) 2-round puzzle.

Proof: Assume there exists a distributional problem \((L, D)\) such that \(L \in \text{NP}\) and \((L, D)\) is aeHOA. From Lemma 2.2 and Lemma 2.1, we can conclude that there exists a polynomial \(p\) and a distributional NP problem \((L', \mathcal{U}_p)\) that is \(\delta\)-aeHOA for \(\delta = 3/8\). Let \(\mathcal{R}'\) be some NP relation corresponding to \(L'\). Consider a puzzle \(C\) where \(C(1^n)\) samples a random \(x \in \{0,1\}^p(n)\) and accepts a response \(y\) if \((x, y) \in \mathcal{R}'\). We will show that \(C\) is a \((\frac{3}{8}, \frac{1}{4})\)-puzzle which by Remark 3.1 implies the existence of a 2-round almost-everywhere puzzle. To show completeness, consider an inefficient algorithm \(\mathcal{A}\) that on input \((1^n, x)\) tries to find a witness \(y\) (using brute-force) such that \((x, y) \in \mathcal{R}'\) and if it is successful sends it to \(C\) (and otherwise simply aborts). Observe that for all sufficiently large \(n \in \mathbb{N}\), for a random \(x \leftarrow \{0,1\}^p(n)\) we have that \(\text{Pr}[x \in L'] > \frac{3}{4}\); otherwise, \((L', \mathcal{U}_p)\) can be decided with probability \(1 - \frac{3}{8}\) for infinitely many \(n \in \mathbb{N}\) contradicting its \(\frac{3}{8}\)-aeHOA property. It follows that for all sufficiently large \(n \in \mathbb{N}\), \(\mathcal{A}\) convinces \(C\) with probability \(\frac{3}{4}\) and thus completeness of \(C\) follows.

To prove soundness, assume for contradiction that there exists a PPT algorithm \(\mathcal{A}^*\) such that \(\text{Pr}[\langle \mathcal{A}^*, C \rangle(1^n) = 1] > \frac{1}{4}\) for infinitely many \(n\). Consider the machine \(M(x)\) that runs \(\mathcal{A}^*(1|x, x)\) and outputs 1 if \(\mathcal{A}^*\) outputs

---

11Roughly speaking, a family of public coin hashfunctions \(H\) having the property that for a random \(h \in H\) and random input \(x\), it is hard for any PPT to find a different \(x'\) of the same length that collides with \(x\) under \(h\) (that is, \(h(x) = h(x')\), \(|x| = |x'|\) yet \(x \neq x'\). Rompel’s theorem was only stated for standard OWFs (as opposed to ioOWFs, but the construction and proof directly also works for the infinitely-often variant as well.

12Note that this is where were are crucially relying on the almost-everywhere hardness of the distributional problem.
a valid witness for \( x \) and otherwise outputs a random bit. By definition, \( M \) solves the distributional problem \((L', U_d)\) with probability \( \frac{1}{2} + \frac{1}{2} (1 - \frac{1}{2}) = \frac{5}{8} = 1 - \frac{3}{8} \) for infinitely many \( n \), which contradicts the \( \frac{3}{8} \)-aeHAO property of \((L', U_p)\). \( \square \)

We now turn to showing that 2-round puzzles imply a HOA distributional NP problem. It will be useful for the sequel to note that the same result also holds in the non-uniform setting.

**Lemma 3.4.** Suppose there exists a 2-round puzzle (resp. a non-uniform puzzle). Then, there exists a distributional NP problem (resp. distributional NP/poly problem) that is HOA (resp. nuHOA).

**Proof:** Let \( C \) be a 2-round puzzle (resp. 2-round non-uniform puzzle). Let \( \ell(\cdot) \) be an upper bound on the amount of randomness used by \( C \). Consider the NP-relation (resp. NP/poly-relation) \( R \) that includes all tuples \((\text{pad}, x), y)\) such that \( C(1^{\text{pad}}) \) given randomness \( x \in \ell(|\text{pad}|) \) accepts upon receiving \( y \), and the sampler \( D(1^n) \) that picks a random \( x \in \{0, 1\}^{\ell(n)} \) and outputs \( (1^n, x) \). We argue next that \((R, D)\) is \( \frac{1}{8} \)-SearchHOA (resp. \( \frac{3}{8} \)-nuSearchHOA), which concludes the proof by applying Lemma 2.3. Assume for contradiction that there exists a PPT (resp. non-uniform PPT) machine \( M \) that solves \((R, D)\) with probability \( > 1 - \frac{1}{8} = \frac{7}{8} \) for all \( n > n_0 \). By the completeness of \( C \), there exists some \( A, n_1 \) such that such that for every \( n > n_1 \), \( \Pr[(A, C)(1^n) = 1] > \frac{7}{8} \). This implies that for all \( n > n_1 \), at most an \( \frac{1}{8} \) fraction of \( \ell(n) \)-bit strings \( x, (1^n, x) \notin L_R \). In particular, for every \( n > \max(n_0, n_1) \), for a random \( x \in \{0, 1\}^{\ell(n)} \), \( M(1^n, x) \) must output a valid witness \( y \) for \( x \) with probability \( > \frac{7}{8} - \frac{1}{8} = \frac{1}{2} \), and can thus be used to break the soundness of the puzzle with probability \( > \frac{1}{2} \) for all sufficiently large \( n \) which is a contradiction. \( \square \)

If the 2-round puzzle has perfect completeness, essentially the same proof gives a SearchHOA problem in TFNP as the relation \( R \) constructed in the proof of Lemma 3.4 is total if the puzzle has perfect completeness.

**Lemma 3.5.** Suppose there exists a 2-round puzzle (resp. almost-everywhere puzzle) with perfect completeness. Then, there exists some search problem \( R \in \text{TFNP} \) and some PPT \( D \) such that the distributional search problem \((R, D)\) that is SearchHAO (aeSearchHOA).

### 4 The Round-Collapse Theorem

In this section, we prove our main technical lemma—a round-collapse theorem for \( O(1) \)-round puzzles.

#### 4.1 An Efficient Babai-Moran Theorem

Our main lemma shows that if ioOWF do not exist, the the Babai-Moran transformation preserves computational soundness.

**Lemma 4.1.** Assume there exists a \( k(\cdot) \)-round puzzle such that \( k(n) \geq 3 \). Then, either there exists an ioOWF, or there exists a \( (k(\cdot) - 1) \)-round puzzle. Moreover, if the \( k(\cdot) \)-round puzzle has perfect completeness, then either there exists an ioOWF, or a \( (k(\cdot) - 1) \)-round puzzle with perfect-completeness.

**Proof:** Consider some \( k(\cdot) \)-round puzzle \( C \) and assume for contradiction that ioOWF do not exist. We will show that Babai-Moran’s (BM) [BM88] round reduction works in this setting and thus we can obtain a \( (k(\cdot) - 1) \)-round puzzle.

Note that if ioOWF do not exist, every polynomial-time computable function is “invertible” with inverse polynomial probability for all sufficiently long input lengths \( n \). In fact, since by [IL90], the existence of distributional one-way functions implies the existence of one-way functions (and this results also works in the infinitely-often setting), we can conclude that if ioOWF do not exist, for any polynomial \( p(\cdot) \), and any polynomial-time computable function \( f \), there exists a PPT algorithm \( \text{Inv} \) such that, for sufficiently large \( n \), the following distributions are \( \frac{1}{p(n)} \)-statistically close.
• \( \{ x \leftarrow \{0,1\}^n : (x, f(x)) \} \)
• \( \{ x \leftarrow \{0,1\}^n ; y = f(x) : (\text{Inv}(y), y) \} \)

In this case, we will say that \( \text{Inv} \) inverts \( f \) with \( \frac{1}{p(n)} \)-statistical closeness. We now proceed to show how to use such an inverter to prove that BM’s round-collapse transformation works on \( C \). To simplify notation, we will make the following assumptions that are without loss of generality:

• \( C \) has at least 4 communication rounds and \( C \) sends the first message; we can always add an initial dummy message to achieve this, while only increasing the number of round by 1. We will then construct a new puzzle that has \( k(\cdot) - 2 \) rounds (which concludes the theorem). Since, in any puzzle, \( A \) sends the final message, this implies we can assume \( k(\cdot) \) is even. To make our notations easier to read, we show how to reduce a \( 2k(\cdot) \)-round protocol to a \( 2k(\cdot) - 2 \) rounds.

• There exists polynomials \( \ell_c, \ell_a \) such that all messages from \( C(1^n) \) are of (the same) length \( \ell_c(n) \) and all the messages from \( A(1^n) \) need to be of length \( \ell_a(n) \) (or else \( C \) rejects). Furthermore, \( \ell_a(\cdot) \) and \( \ell_c(\cdot) \) are polynomial-time computable, and strictly increasing.

We denote by \( C(1^n, p_1, p_2, \ldots, p_i; r_c) \) the \( (i + 1)^{st} \)-message (i.e., the message to be sent in round \( 2i + 1 \) round) from \( C \) where \( r_c \) is \( C \)’s randomness randomness and \( p_1, p_2, \ldots, p_i \) are bit strings (representing the messages received from \( A \) in the first \( 2i \) rounds). Let \( m(n) \) be \( (\ell_a(n) + 4)(\log(n))^2 \) rounded upwards to the next power of two.\(^\text{13}\) We will show that the BM transformation works (if ioOWF do not exist), when using \( m(\cdot) \) repetitions. More precisely, consider the following \( (2k(\cdot) - 2) \)-round puzzle challenger \( \tilde{C} \) that on input \((1^n, p_1, \ldots, p_i; r_c)\) proceeds as follows:

1. If \( i < k(n) - 2 \), output \( r_{i+1} \) (i.e., proceed just like \( C \) before round \( 2k(n) - 3 \));

2. If \( i = k(n) - 2 \), output \( (r_{k(n)-1}, r_{k(n)}^1, \ldots, r_{k(n)}^{m(n)}) \) (i.e., in round \( 2k(n) - 3 \), send the original challenge for round \( 2k(n) - 3 \) as well as a “\( m(n) \)-wise parallel-repetition” challenge for the original round \( 2k(n) - 1 \));

3. If \( i = k(n) - 1 \) (i.e., after receiving the message in the last round), output 1 if and only if

\[
C(1^n, p_1, p_2, \ldots, p_{k(n)-1}, p_{k(n)}^1; r_1, r_2, \ldots, r_{k(n)-1}, r_{k(n)}^i) = 1
\]

for every \( i \in [m(n)] \) (i.e., all the parallel instances are accepting),

where \( r_c^i \) is interpreted as \( (r_1, r_2, \ldots, r_{k(n)-1}, r_{k(n)}^1, \ldots, r_{k(n)}^{m(n)}) \)

We will show that \( \tilde{C} \) is a \((99/100, 1/2)\)-puzzle, and thus by Remark 3.1 this implies a puzzle with the same number of rounds.

We first define some notation:

• Given a transcript \( T = (r_1, p_1, \ldots, p_{k(n)-2}, r_{k(n)-1}, r_{k(n)}^1, \ldots, r_{k(n)}^{m(n)}; p_{k(n)-1}, p_{k(n)}^1, \ldots, p_{k(n)}^{m(n)}) \) of an interaction between \( \tilde{C} \) and an adversary, we let \( T_{\leq k-1} = (r_1, p_1, \ldots, p_{k(n)-2}, r_{k(n)-1}) \) denote the transcript up to and including the round where \( \tilde{C} \) (in the emulation done by \( \tilde{C} \)) sends it \((k(n)-1)\)’st message.

• We say that \( T \) is accepting if

\[
\tilde{C}(p_1, \ldots, p_{k(n)-2}, p_{k(n)-1}, p_{k(n)}^1, \ldots, p_{k(n)}^{m(n)}; r_1, \ldots, r_{k(n)-1}, r_{k(n)}^1, \ldots, r_{k(n)}^{m(n)}) = 1
\]

(i.e., if \( \tilde{C} \) is accepting in the transcript).

\(^{13}\)We round to the next power of 2 to make it easy to sample a random number in \([m(n)]\); this is just to simplify presentation/analysis
Completeness: Completeness (in fact with all but negligible probability) follows directly from original proof by Babai-Moran [BM88].

Soundness: Assume for contradiction that there exists a PPT algorithm $A^*$ that convinces $\widetilde{C}$ on common input $1^n$ with probability $\epsilon(n)$ such that $\epsilon(n) > \frac{1}{2}$ for all sufficiently large $n$. Let $h(\cdot)$ be a polynomial such that $A^*$ runs in time at most $h(n)$ when its first input is $1^n$. We assume without loss of generality that $A^*$ only sends a real last message if $\widetilde{C}$ will be accepting it (note that since $\widetilde{C}$ is public coin, $A^*$ can verify this, so it is without loss of generality), and otherwise sends ⊥ as its last message.

On a high-level, using $A^*$ and the fact that polynomial-time computable functions are “invertible”, we will construct a PPT $B$ such that $\Pr[B(C)(1^n) = 1] \geq \frac{1}{64}$ for sufficiently large $n$, which contradicts the soundness of the original $2k(n)$-round puzzle $C$. Towards constructing $B$, we first define a polynomial-time algorithm $M$ on which we will apply the inverter $Inv$. As described in the introduction, we will consider an algorithm $M$ that operates on inputs of the form $(1^n, i, r_M)$ where $i$ is an index of one of the $m(n)$ parallel sessions and $r_M$ contains the randomness of $A$ and $\widetilde{C}$. To correctly parse such inputs, let $\ell_M(n) = n + \log(m(n)) + h(n) + \epsilon(n) \cdot (k(n) - 1 + m(n))$ and note that by our assumption on $\ell_c(n)$ and $\ell_a(n)$, this is a strictly increasing and polynomial-time computable function. In the rest of the proof, whenever the security parameter $n$ is clear from context, we omit it and let $k = k(n), \ell_c = \ell_c(n), \ell_a = \ell_a(n), \epsilon = \epsilon(n), m = m(n)$ and $h = h(n)$. Now, consider the machine $M$ that on input $u$ internally incorporates the code of $A^*$ and proceeds as follows:

1. $M$ finds an $n$ such that $\ell_M(n) = |u|$ (simply by enumerating different $n$ from $1$ up to $|u|$). If no such $n$ exists, then $M$ outputs ⊥ and halts. Otherwise, $M$ interprets $u$ as $(\text{pad}, i, r_M)$ such that $|\text{pad}| = n, |i| = \log(m(n)), r_M = (z, r_1, r_2, \ldots, r_{k-1}, r_k^1, \ldots, r_k^m), z \in \{0, 1\}^h, \text{ and all the strings } r_1, r_2, \ldots, r_{k-1}, r_k^1, \ldots, r_k^m \text{ are in } \{0, 1\}^{\ell_c}$.

2. It internally emulates an execution between $A^*$ and $\widetilde{C}$ on common input $1^n$ and respectively using randomness $z$ and $(r_1, r_2, \ldots, r_{k-1}, r_k^1, \ldots, r_k^m)$. Let

$$T = (r_1, p_1, \ldots, p_{k-2}, r_{k-1}, r_k^1, \ldots, r_k^m, p_{k-1}, p_k^1, \ldots, p_k^m)$$

denote the transcript of the interaction.

3. If $T$ is accepting, then $M$ outputs $(1^{||\text{pad}||}, T_{\leq k-1}, p_{k-1}, r_k^i)$,

and otherwise ⊥.

Let $Inv$ be an inverter for $M$ with $\frac{1}{n}$ statistical-closeness for all sufficiently large $n$—such an inverter exists due to our assumption on the non-existence of ioOWFs.

We are now ready to describe our adversary $B$ for the $2k$-round puzzle. $B$ on input $(1^n, r_1, r_2, \ldots, r_i; r_B)$ proceeds as follows:

1. $B$ interprets $r_B$ as $(z, \text{pad}, s_{k-1}^1, \ldots, s_{k-1}^m)$ such that $z \in \{0, 1\}^h, \text{pad} \in \{0, 1\}^n$ and all the strings $s_{k-1}^1, \ldots, s_{k-1}^m$ are in $\{0, 1\}^{\ell_c}$.

2. If $i < k - 1$, $B$ outputs $p_i = A^*(1^n, r_1, r_2, \ldots, r_i; z)$ (i.e., $B$ proceeds just as $A^*$ in the first $2k - 4$ rounds).

3. If $i = k - 1$ (i.e., in round $2k - 2$), $B$ lets $(p_{k-1}, p_k^1, \ldots, p_k^m) = A^*(1^n, r_1, r_2, \ldots, r_{k-1}, s_k^1, \ldots, s_k^m; z)$ and outputs $p_{k-1}$.

4. If $i = k$ (i.e., in round $2k$), then:
We now proceed to analyze the success probability of $B$ against $\mathcal{C}$. In particular, we shall show that for all sufficiently large $n$, $\Pr[(B, \mathcal{C})(1^n) = 1] > \frac{1}{4}$ which will conclude the proof of Lemma 4.1. We denote by $\text{View}_{A^*}(\langle A^*, \mathcal{C}\rangle(1^n))$ the random variable that represents the view of the adversary $A^*$ in an interaction with $\mathcal{C}$ on common input $1^n$—for convenience, we describe this view $v = (z, T)$ by $A^*$’s random coin tosses $z$, as well as the transcript $T$ of the interaction between $A^*$ and $\mathcal{C}$. Towards analyzing $B$, we consider a sequence of hybrid experiments $\text{Expt}_0, \text{Expt}_1, \text{Expt}_2, \text{Expt}_3$—formally, an experiment defines a probability space and a probability density function over it. All experiments will be defined over the same probability space so we can consider the same random variables over all of them. To simplify notation, we abuse of notation and let $\text{Expt}_0(n)$ also denote a random variable describing the output of the experiment $\text{Expt}_1(n)$.

$\text{Expt}_0(n)$ will simply consider an execution between $B^{\text{Inv}}$ and $\mathcal{C}$ on common input $1^n$ and will output 1 if $\mathcal{C}$ is accepting and 0 otherwise; see Figure 1 for a formalization. To simplify the transition to later experiments, we formalize $\text{Expt}_0(n)$ as first sampling a full transcript $T$ of an execution between $A^*$ and $\mathcal{C}$, keeping only the prefix $T_{\leq k-1}$ (this gives exactly the same distribution as an interaction between $B$ and $\mathcal{C}$ up to round $2k - 2$), next sampling an “external” random message $r$ (just as $\mathcal{C}$ would in round $2k - 1$), and finally producing the last message just as $B$ does. (We additionally sample a random index $i \in [m]$ is not used in the current experiment, but will be useful in later experiments.) We thus directly have:

**Claim 1.** $\Pr[(B, \mathcal{C})(1^n) = 1] = \Pr[\text{Expt}_0(n) = 1]$.

We now slowly transform the experiment into one that becomes easy to analyze. See Figure 1 for a formal description of the experiments.

1. We first define an “good” event $G = W \cap G'$, where $W$ is the event that the originally sampled transcript is accepting and $G'$ is the event that the “prefix” $(T_{\leq k-1}, p_{k-1})$ is “good” in a well defined sense (roughly speaking, that continuations conditioned $T_{\leq k-1}$ are successful with high probability, and that that in such successful continuations $p_{k-1}$ is used with not “too low” probability). $\text{Expt}_1$ will next proceed just like $\text{Expt}_0$ except that we additionally fail if the event $G$ does not happen. We thus have that the probability of $\text{Expt}_0(n)$ outputting 1 is at least as high as the probability of $\text{Expt}_1(n)$ outputting 1.

**Claim 2.** $\Pr[\text{Expt}_0(n) = 1] \geq \Pr[\text{Expt}_1(n) = 1]$.

Additionally, as we shall show (using an averaging argument), the event $G$ happens with non-negligible probability, not just in $\text{Expt}_1$ but also in all the other experiments $\text{Expt}_j$ for $j \in \{1, 2, 3\}$ (as Step 1 of the experiment remains unchanged in all of them).

**Claim 3.** For $j \in \{1, 2, 3\}$, $\Pr[G] \geq \frac{\epsilon^2}{8}$, where the probability is over the randomness in experiment $\text{Expt}_j(n)$.

2. We next transition to an experiment $\text{Expt}_2$ where instead of choosing the message $r_k$ at random (as it was in $\text{Expt}_1$), we select it as the message in the $i$'th repetition of $\mathcal{C}$’s $k - 1$st message in the initially

---

14This is a bit redundant—as the messages sent by $A^*$ can of course be recomputed given just the randomness of $A^*$ and the messages from $\mathcal{C}$, but will simplify notation.
Experiment Expt$_0(n)$.

1. Sample $(z, T) \leftarrow \text{View}_{A^*}(\langle A^*, \hat{C} \rangle(1^n)); \text{pad} \leftarrow \{0, 1\}^n; i \leftarrow [m]; r \leftarrow \{0, 1\}^r$. Interpret $T$ as $(r_1, p_1, \ldots, p_{k-2}, r_{k-1}, s_1^k, \ldots, s_m^k, p_{k-1}^1, \ldots, p_{k}^m)$.

2. Let $r_k = r$ and let $y = (\text{pad}, T_{\leq k-1}, p_{k-1}, r_k)$ if $T$ is accepting and $y = \bot$ otherwise.

3. Let $u \leftarrow \text{Inv}(y)$; interpret $u$ as $(p, j, r_M)$ where $|\text{pad}| = n$, $|j| = \log_2(m)$, and $r_M = (z', t_1, t_2, \ldots, t_{k-1}, t_k^1, \ldots, t_k^m)$ just as $B$ does and let $(q_{k-1}, q_k^1, \ldots, q_k^m) = A^*(1^n, r_1, r_2, \ldots, r_{k-1}, t_k^1, \ldots, t_k^m; z')$.

4. Output 1 iff $T' = (T_{\leq k-1}, p_{k-1}, r_k, q_k^1)$ is accepting, and 0 otherwise.

Experiment Expt$_1(n)$.

1. Sample $(z, T) \leftarrow \text{View}_{A^*}(\langle A^*, \hat{C} \rangle(1^n)); \text{pad} \leftarrow \{0, 1\}^n; i \leftarrow [m]; r \leftarrow \{0, 1\}^r$. Interpret $T$ as $(r_1, p_1, \ldots, p_{k-2}, r_{k-1}, s_1^k, \ldots, s_m^k, p_{k-1}^1, \ldots, p_{k}^m)$.

2. Let $r_k = r$ and let $y = (\text{pad}, T_{\leq k-1}, p_{k-1}, r_k)$ if $T$ is accepting and $y = \bot$ otherwise.

3. Let $u \leftarrow \text{Inv}(y)$; interpret $u$ as $(p, j, r_M)$ where $|\text{pad}| = n$, $|j| = \log_2(m)$, and $r_M = (z', t_1, t_2, \ldots, t_{k-1}, t_k^1, \ldots, t_k^m)$ just as $B$ does and let $(q_{k-1}, q_k^1, \ldots, q_k^m) = A^*(1^n, r_1, r_2, \ldots, r_{k-1}, t_k^1, \ldots, t_k^m; z')$.

4. Output 1 iff $T' = (T_{\leq k-1}, p_{k-1}, r_k, q_k^1)$ is accepting and $G$ holds, and 0 otherwise.

Distribution Expt$_2^i(n)$.

1. Sample $(z, T) \leftarrow \text{View}_{A^*}(\langle A^*, \hat{C} \rangle(1^n)); \text{pad} \leftarrow \{0, 1\}^n; i \leftarrow [m]; r \leftarrow \{0, 1\}^r$. Interpret $T$ as $(r_1, p_1, \ldots, p_{k-2}, r_{k-1}, s_1^k, \ldots, s_m^k, p_{k-1}^1, \ldots, p_{k}^m)$.

2. Let $r_k = s_k^i$ and let $y = (\text{pad}, T_{\leq k-1}, p_{k-1}, r_k)$ if $T$ is accepting and $y = \bot$ otherwise.

3. Let $u \leftarrow \text{Inv}(y)$; interpret $u$ as $(p, j, r_M)$ where $|\text{pad}| = n$, $|j| = \log_2(m)$, and $r_M = (z', t_1, t_2, \ldots, t_{k-1}, t_k^1, \ldots, t_k^m)$ just as $B$ does and let $(q_{k-1}, q_k^1, \ldots, q_k^m) = A^*(1^n, r_1, r_2, \ldots, r_{k-1}, t_k^1, \ldots, t_k^m; z')$.

4. Output 1 iff $T' = (T_{\leq k-1}, p_{k-1}, r_k, q_k^1)$ is accepting and $G$ holds, and 0 otherwise.

Distribution Expt$_3^i(n)$.

1. Sample $(z, T) \leftarrow \text{View}_{A^*}(\langle A^*, \hat{C} \rangle(1^n)); \text{pad} \leftarrow \{0, 1\}^n; i \leftarrow [m]; r \leftarrow \{0, 1\}^r$. Interpret $T$ as $(r_1, p_1, \ldots, p_{k-2}, r_{k-1}, s_1^k, \ldots, s_m^k, p_{k-1}^1, \ldots, p_{k}^m)$.

2. Let $r_k = s_k^i$ and let $y = (\text{pad}, T_{\leq k-1}, p_{k-1}, r_k)$ if $T$ is accepting and $y = \bot$ otherwise.

3. Let $u \leftarrow \text{PInv}(y)$; interpret $u$ as $(p, j, r_M)$ where $|\text{pad}| = n$, $|j| = \log_2(m)$, and $r_M = (z', t_1, t_2, \ldots, t_{k-1}, t_k^1, \ldots, t_k^m)$ just as $B$ does and let $(q_{k-1}, q_k^1, \ldots, q_k^m) = A^*(1^n, r_1, r_2, \ldots, r_{k-1}, t_k^1, \ldots, t_k^m; z')$.

4. Output 1 iff $T' = (T_{\leq k-1}, p_{k-1}, r_k, q_k^1)$ is accepting and $G$ holds, and 0 otherwise.

Figure 1: Description of intermediate experiments.
sampled transcript $T$, where $i$ is a randomly sampled index $i \in [m]$. The reason for defining this experiment is that, in it, we are applying the one-way function inverter on the "right" distribution (just as in the definition of $M$). The central claim to show is that this change does not change the success probability by too much. As discussed in the introduction, we shall prove it using Raz’s sampling lemma.

**Claim 4.** $\Pr[\text{Expt}_1(n) = 1] \geq \Pr[\text{Expt}_2(n) = 1] - \frac{1}{\log(n)}$.

3. Finally, we transition to an experiment $\text{Expt}_2$ where we employ a perfect inverter $\text{PInv}$—that always samples uniform preimages to $M$, instead of the (imperfect) inverter $\text{Inv}$. It directly follows from the fact that $\text{Inv}$ is an inverter with statistical closeness $\frac{1}{n}$ and that the inverter is applied to an element that is sampled as a uniform image of $M$ that the statistical distance between $\text{Expt}_2(n)$ and $\text{Expt}_3(n)$ is bounded by $\frac{1}{n}$ for sufficiently large $n$. In particular,

**Claim 5.** For all sufficiently large $n$,

$$\Pr[\text{Expt}_2(n) = 1] \geq \Pr[\text{Expt}_3(n) = 1] - \frac{1}{n}.$$

4. We finally note that in $\text{Expt}_3$, there are only two reason the experiment can output 0: (1) The originally sampled transcript $T$ is not accepting (i.e., the event $W$ does not hold); if is is accepting, the perfect inverter will make sure that $T'$ is also accepting, or (2) the event $G$ does not hold. Additionally note, since $G$ is defined as $W \cap G'$, we have that whenever $G$ holds, $W$ holds as well and thus the experiment must output 1. Thus, by Claim 3, we have:

**Claim 6.**

$$\Pr[\text{Expt}_3(n) = 1] \geq \frac{\epsilon^2}{8}$$

5. By combining claims 1, 2, 4, 6, we have that for all sufficiently large $n$,

$$\Pr[(B, C) (1^n) = 1] = \Pr[\text{Expt}_0(n) = 1] \geq \Pr[\text{Expt}_1(n) = 1] \geq \Pr[\text{Expt}_2(n) = 1] - \frac{1}{\log(n)} \geq \Pr[\text{Expt}_3(n) = 1] - \frac{2}{\log(n)} \geq \frac{\epsilon^2}{8} - \frac{2}{\log(n)} > \frac{1}{64}$$

which is a contradiction.

To conclude the proof of Lemma 4.1, it just remains to formalize the event $G = W \cap G'$ and proving Claim 3 and Claim 5.

### 4.2 The Good Event $G$ and the Proof of Claim 3

We begin by defining some random variables over the probability space over which $\text{Expt}_1$ is defined. Note that the probability space is the same for $\text{Expt}_0, \text{Expt}_1, \text{Expt}_2$ and as such random variables and events over $\text{Expt}_1$ are also defined over all the other experiments. We use boldface to denote random variables describing the outcome of variables in the experiments—for instance, we let $T$ denote a random variable describing the value of $T$ as sampled in the experiments.

Let $W$ denote the event that $T$ is accepting (i.e., the transcript sampled in Step 1 is accepting) and let $\Theta$ be the set of partial transcripts $\theta$ such that

$$\Pr[W \mid T_{\leq (k-1)} = \theta] \geq \frac{\epsilon}{2}.$$  

\[15\] Note that we here rely on the fact that $y = \perp$ when $T$ is not accepting.
where the probability is over \( \text{Expt}_1(n) \). That is, \( \Theta \) is the set of “good” partial transcripts conditioned on which \( \mathcal{A}^* \) has a reasonable probability of succeeding. Note that by a standard averaging argument, we have that such transcripts occur often:

\[
\Pr[T_{\leq(k-1)} \in \Theta] \geq \frac{\epsilon}{2}. \tag{1}
\]

Now, consider the event \( W_p \) that \( W \) holds and \( p_{k-1} = p \); let \( P(\theta) \) be the set of messages \( p \in \{0,1\}^{T_\theta} \) for which

\[
\Pr[W_p \mid T_{\leq(k-1)} = \theta] \geq \frac{\epsilon}{2^{T_{\theta}+2}}.
\]

In other words, \( P(\theta) \) is the set of “good” (adversary) messages \( p \) such that conditioned on the partial transcript \( \theta \), the probability that \( \mathcal{A}^* \) succeeds while using \( p \) as its \( k-1 \)'st message is greater than \( \frac{\epsilon}{2^{T_{\theta}+2}} \). As we shall now show using another (standard) averaging argument, for every \( \theta \in \Theta \), we have

\[
\Pr[p_{k-1} \in P(\theta) \mid T_{\leq(k-1)} = \theta] \geq \frac{\epsilon}{4}. \tag{2}
\]

Suppose for contradiction that for some \( \theta \in \Theta \), Equation 2 does not hold. Then, we have

\[
\Pr[W \mid T_{\leq(k-1)} = \theta] = \sum_{p \in \{0,1\}^{T_{\theta}}} \Pr[W_p \mid T_{\leq(k-1)} = \theta] \\
= \sum_{p \in P(\theta)} \Pr[W_p \mid T_{\leq(k-1)} = \theta] + \sum_{p \in \{0,1\}^{T_{\theta}} - P(\theta)} \Pr[W_p \mid T_{\leq(k-1)} = \theta] \\
\leq \Pr[p_{k-1} \in P(\theta) \mid T_{\leq(k-1)} = \theta] + \sum_{p \in \{0,1\}^{T_{\theta}} - P(\theta)} \Pr[W_p \mid T_{\leq(k-1)} = \theta] \\
< \frac{\epsilon}{4} + \sum_{p \in \{0,1\}^{T_{\theta}} - P(\theta)} \Pr[W_p \mid T_{\leq(k-1)} = \theta] \\
\leq \frac{\epsilon}{4} + \sum_{p \in \{0,1\}^{T_{\theta}} - P(\theta)} \frac{\epsilon}{2^{T_{\theta}+2}} \\
\leq \frac{\epsilon}{4} + 2^{T_{\theta}} \cdot \frac{\epsilon}{2^{T_{\theta}+2}} = \frac{\epsilon}{2}
\]

which contradicts that \( \theta \in \Theta \).

Next, define \( G' \) to be the event that \( T_{\leq(k-1)} \in \Theta \) and \( p_{k-1} \in P(T_{\leq(k-1)}) \), and define \( G \) as holding when \( W \) and \( G' \) both hold (i.e., the originally sampled transcript is accepting and \( G' \) holds). Note that \( G' \) in fact implies that \( W \) holds (since \( p_{k-1} \in P(T_{\leq(k-1)}) \) implies that \( p_{k-1} \neq \perp \) which by our assumption on \( A^* \) means that \( T \) must be accepting), thus in fact \( G' = G \). By combining Equations 1 and 2, we have:

\[
\Pr[G] = \Pr[G'] = \Pr[T_{\leq(k-1)} \in \Theta \land p_{k-1} \in P(T_{\leq(k-1)})] \\
= \Pr[T_{\leq(k-1)} \in \Theta] \times \Pr[p_{k-1} \in P(T_{\leq(k-1)}) \mid T_{\leq(k-1)} \in \Theta] \\
\geq \frac{\epsilon}{2} \times \frac{\epsilon}{4} = \frac{\epsilon^2}{8}
\]

where the probability is taken over \( \text{Expt}_1(n) \). Finally, note that since Step 1 (whose outcome determines whether \( G \) happens) remains unchanged in all the experiments, we can conclude that \( \Pr[G] \geq \frac{\epsilon^2}{8} \) where the probability is taken over \( \text{Expt}_j(n) \) for every \( j \in \{1,2,3\} \), which concludes the proof of Claim 3.

### 4.3 Proof of Claim 4

Recall that we need to show that \( \Pr[\text{Expt}_1(n) = 1] \geq \Pr[\text{Expt}_2(n) = 1] - \frac{1}{\log(n)} \). Observe that the only difference between experiments \( \text{Expt}_1 \) and \( \text{Expt}_2 \) is that, in \( \text{Expt}_1 \), we set \( r_k = r \) and in \( \text{Expt}_2 \), we set
In other words, we need to upper bound, statistical distance between $r$ between $T_{\leq k-1} = \theta$ and $W_p$. So we can ignore everything that happens after this. Additionally, note that the only variables that are relevant after this point are $\text{pad}, T_{\leq k-1}, p_{k-1}, i$ and $r$. Note that $\text{pad}, i$ are both independent of the events $T_{\leq k-1} = \theta, W_p$ and thus still independently and uniformly sampled in both experiments. $T_{\leq k-1}$ and $p_{k-1}$, on the other hand are fixed (constant) conditioned on $T_{\leq k-1} = \theta, W_p$. Thus, to bound the statistical difference between $\{\text{Expt}_1(n) \mid T_{\leq k-1} = \theta, W_p\}$ and $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta, W_p\}$, it suffices to bound the statistical distance between $r_k$ in $\{\text{Expt}_1(n) \mid T_{\leq k-1} = \theta, W_p\}$ and $r_k$ in $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta, W_p\}$. In other words, we need to upper bound,

$$\Delta = \text{SD}(r, s_k^i | W_p) = \text{SD}(s_k^i, s_k^j | W_p) \leq \sum_{j \in [m]} \frac{1}{m} \text{SD}(s_k^i, s_k^j | W_p)$$

over $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta\}$ since for each $j$, $s_k^i$ is sampled uniformly at random, independent of $T_{\leq k-1}$ (and independent of $s_k^{j'}$ for $j' \neq j$). Towards bounding this quantity, we will rely on Raz’s sampling lemma.

**Lemma 4.2** ([Raz98]). Let $X_1, \ldots, X_m$ be independent random variables on a finite domain $U$. Let $E$ be an event over $\bar{X} = (X_1, \ldots, X_m)$. Then,

$$\frac{1}{m} \cdot \sum_{i=1}^{m} \text{SD}(X_i, X_i | E) \leq \sqrt{\frac{1}{m} \cdot \log \frac{1}{\Pr[E]}}$$

By applying Raz’s lemma, we directly get that

$$\Delta \leq \sqrt{\frac{1}{m} \cdot \log \frac{1}{\Pr[W_p]}}$$

where the probability is over $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta\}$. Since by our assumption $p \in P(\theta)$, we have that the probability of $W_p$ conditioned on $T_{\leq k-1} = \theta$ is at least $\frac{\epsilon}{2^{a+\epsilon}}$, thus

$$\Delta \leq \sqrt{\frac{1}{m} \cdot (\ell_a + 2 - \log(\epsilon)) \leq \frac{1}{\log(m)}$$

since $\epsilon > \frac{1}{2}$ and $m = (\ell_a + 4)(\log(n))^2 > (\ell_a + 2 - \log(\epsilon))(\log(n))^2$. 

---

$r_k = s_k^i$. Furthermore, both the experiments sample $((z, T), \text{pad}, i, r)$ from the same distributions and output 0 whenever $G$ does not hold (which is a function only of $T$). It follows that the statistical distance between $\text{Expt}_1(n)$ and $\text{Expt}_2(n)$ is bounded by the statistical distance of $\text{Expt}_1(n)$ and $\text{Expt}_2(n)$ conditioned on the event $G$. Note that we can rephrase the event $G$ as

$$G = \bigcup_{\theta \in \Theta, p \in P(\theta)} W_p \cap (T_{\leq k-1} = \theta)$$

Below, we shall show that for every $\theta \in \Theta, p \in P(\theta)$, it holds that the statistical distance between $\{\text{Expt}_1(n) \mid T_{\leq k-1} = \theta, W_p\}$ and $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta, W_p\}$ is bounded by $\frac{1}{\log(n)}$, which concludes the proof of Claim 5.

Consider some $\theta \in \Theta, p \in P(\theta)$ and consider the experiments $\{\text{Expt}_1(n) \mid T_{\leq k-1} = \theta, W_p\}$ and $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta, W_p\}$. Note both experiments proceed exactly the same after $r_k$ is defined in Step 2, so we can ignore everything that happens after this. Additionally, note that the only variables that are relevant after this point are $\text{pad}, T_{\leq k-1}, p_{k-1}, i$ and $r$. Note that $\text{pad}, i$ are both independent of the events $T_{\leq k-1} = \theta, W_p$ and thus still independently and uniformly sampled in both experiments. $T_{\leq k-1}$ and $p_{k-1}$, on the other hand are fixed (constant) conditioned on $T_{\leq k-1} = \theta, W_p$. Thus, to bound the statistical difference between $\{\text{Expt}_1(n) \mid T_{\leq k-1} = \theta, W_p\}$ and $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta, W_p\}$, it suffices to bound the statistical distance between $r_k$ in $\{\text{Expt}_1(n) \mid T_{\leq k-1} = \theta, W_p\}$ and $r_k$ in $\{\text{Expt}_2(n) \mid T_{\leq k-1} = \theta, W_p\}$. In other words, we need to upper bound,
4.4 Variations

Using essentially the same proofs, we can directly get the following vacations of 4.1. The first variant simply states that the same result holds for almost-everywhere puzzles.

**Lemma 4.3** (Almost-everywhere variant 1). Assume there exists a \( k(\cdot) \)-round almost-everywhere puzzle such that \( k(n) \geq 3 \). Then, either there exists an ioOWF, or there exists a \((k(\cdot) - 1)\)-round almost-everywhere puzzle. Moreover, if the \( k(\cdot) \)-round puzzle has perfect completeness, then either there exists an ioOWF, or a \((k(\cdot) - 1)\)-round almost-everywhere puzzle with perfect-completeness.

The next variant shows that if we start off with an almost-everywhere puzzle, we can either get a (standard) one-way function or a puzzle with one less round (but this new puzzle no longer satisfies almost-everywhere security). This follows from the fact that if the attacker \( A^* \) succeeds on all sufficiently large input lengths, then it suffices for \( Inv \) to work on infinitely many input lengths, to conclude that \( B^{Inv} \) works on infinitely many inputs length (thus violating almost-everywhere security of the original puzzle).

**Lemma 4.4** (Almost-everywhere variant 2). Assume there exists a \( k(\cdot) \)-round almost-everywhere puzzle such that \( k(n) \geq 3 \). Then, either there exists an OWF, or there exists a \((k(\cdot) - 1)\)-round puzzle. Moreover, if the \( k(\cdot) \)-round puzzle has perfect completeness, then either there exists an OWF, or a \((k(\cdot) - 1)\)-round puzzle with perfect-completeness.

We additionally consider a variant for non-uniform puzzles. As the challenger now may be a non-uniform PPT, the function \( M \) that we are required to invert is also a non-uniform PPT and thus we can only conclude the existence of non-uniform OWFs.

**Lemma 4.5** (Non-uniform variant). Assume there exists a \( k(\cdot) \)-round non-uniform puzzle such that \( k(n) \geq 3 \). Then, either there exists a non-uniform ioOWF, or there exists a \((k(\cdot) - 1)\)-round non-uniform puzzle.\(^{16}\)

4.5 Characterizing O(1)-Round Puzzles

We next apply our round-collapse theorem (and its variants) to get a characterization of \( O(1) \)-round puzzles. This characterization applies to both standard puzzles and non-uniform puzzles.

**Corollary 4.1.** Assume the existence of a \( O(1) \)-round (resp. \( O(1) \)-round non-uniform puzzle). Then there exists a 2-round puzzle (resp. 2-round non-uniform puzzle) and thus a distributional NP problem (resp. distributional NP/poly problem) that is HOA (resp. nuHOA).

**Proof:** If (non-uniform) ioOWF exists, then by applying Proposition 3.2 we have that 2-round (non-uniform) puzzles exist. If (non-uniform) ioOWF do not exist, we can apply Lemma 4.1 (Lemma 4.5) iteratively to collapse any constant-round protocol to a 2-round protocol. (Note that we can only apply Lemma 4.1 a constant number of times, as the communication complexity of the resulting protocol grows polynomially with each application.). Thus in either case, we conclude that the existence of a \( O(1) \)-round (non-uniform) puzzle implies a 2-round (non-uniform) puzzle. The corollary is concluded by applying Lemma 3.4. \( \square \)

We remark that the reason we cannot get an (unconditional) characterization of almost-everywhere puzzles is that ioOWFs are not known to imply 2-round almost-everywhere puzzles.

5 Characterizing Polynomial-round Puzzles

We observe that the existence of a poly-round puzzle is equivalent to the statement that \( \text{PSPACE} \not\subseteq \text{BPP} \). A consequence of this result (combined with Lemma 3.4) is that any round-collapse theorem that (unconditionally) can transform a polynomial-round puzzle into a \( O(1) \)-round puzzle, must show the existence of

\(^{16}\)The transformation still preserves perfect completeness, but this will not be of relevance for us.
a HAO distributional NP problem based on the assumption that PSPACE \( \not\subseteq \) BPP (which would be highly unexpected).

**Theorem 5.1.** For every \( \epsilon > 0 \), there exists an \( n^\epsilon \)-round puzzle (resp. a non-uniform puzzle) if and only if PSPACE \( \not\subseteq \) BPP (resp. PSPACE \( \not\subseteq \) P/poly).

**Proof:** For the “only-if” direction, note that using the same proof as (the easy direction) in IP = PSPACE \[Sha92, LFKN92\], we can use a PSPACE oracle to implement the optimal adversary strategy in every puzzle and thus (due to the completeness condition of the puzzle) break the soundness of every puzzle using a PSPACE oracle. So, if PSPACE \( \subseteq \) BPP, soundness of every puzzle can be broken in PPT and thus puzzles cannot exist.

For the “if” direction, recall that by the classic result of \[BFNW93\] (see also \[TV07\]), if PSPACE \( \not\subseteq \) BPP, then there is PSPACE language \( L' \), constant \( c \in \mathbb{N} \), and a polynomial \( p(\cdot) \) such that \( (L', U_p) \) is \( \frac{1}{n^c} \)-HOA. We will now use this HAO language \( L' \) together with the fact that by \[Sha92, LFKN92\] all of PSPACE has a public-coin interactive proof (and the fact that PSPACE is closed under complement, to get a puzzle. The puzzle challenger \( C(1^n) \) simply samples a random statement \( x \in \{0, 1\}^{p(n)} \) and sends it to the adversary. The adversary next announces a bit \( b \) (determining whether \( x \in L' \) or not) and next if \( b = 1 \), \( C \) runs the IP verifier for \( x \in L' \) and if \( b = 0 \) instead runs the IP verifier for \( x \notin L' \). Due to \[Sha92, LFKN92\], we may assume without loss of generality that the IP has completeness 1 and soundness error \( 2^{-n} \). As we shall now argue \( C \) is a \((1, 1 - \frac{2}{n^c})\)-puzzle which by remark Remark 3.1 implies a puzzle. Completeness follows directly from the completeness of the IP. For soundness, consider a PPT machine \( A^* \) that convinces \( C \) with probability better than \( 1 - \frac{2}{n^c} \). We construct a machine \( B \) that breaks the \( \frac{1}{n^c} \)-HOA property of \( L' \). \( B(1^n, x) \) simply emulates an interaction between \( C(1^n) \) and \( A^* \) while fixing \( C \)’s first message to \( x \) and accepts \( x \) if \( C \) is accepting, and rejects otherwise. Since \( B \) is feeding \( A^* \) messages according to the same distribution as in the real execution (with \( C \)), we have that \( A^* \) convinces \( C \) in the emulation by \( B \) with probability at least \( 1 - \frac{2}{n^c} \). By the soundness of the IP, we have that except with probability \( 2^{-n} \), whenever the proof is accepting, the bit \( b \) must correctly decide \( x \). We conclude (by a union bound) that \( B \) correctly decides \( x \) with probability \( 1 - \frac{2}{n^c} - 2^{-n} > 1 - \frac{1}{n^c} \) for all sufficiently large \( n \in \mathbb{N} \).

The non-uniform version of the theorem follows using exactly the same proof. 

### 6 Perfect Completeness and TFNP hardness

We show that any 2-round puzzle can be transformed into a 3-round puzzle with perfect completeness; next, we shall use this result together with our round-reduction theorem to conclude our results on the hardness of TFNP.

#### 6.1 From Imperfect to Perfect Completeness (by Adding a Round)

Furer et al. \[FGM+89\] showed how to transform any 2-round public-coin proof system into a 3-round public-coin proof system with perfect completeness. We will rely on the same protocol transformation to transform any 2-round puzzle into a 3-round puzzle with perfect completeness. The perfect completeness condition will follow directly from their proof; we simply must argue that the transformation also preserves computational soundness (as they only showed that it preserves information-theoretic soundness).

**Theorem 6.1.** Suppose there exists 2-round puzzle. Then there exists a 3-round puzzle with perfect completeness.

**Proof:** Let \( C \) be a 2-round puzzle. Let \( \ell_c, \ell_a \) be polynomials such that the message from \( C(1^n) \) is of length \( \ell_c(n) \) and the message from \( A(1^n) \) is of length \( \ell_a(n) \); we assume without loss of generality that \( \ell_c(n) > 2 \). When the security parameter \( n \) is clear from the context we will omit it and let \( \ell_c(n) = \ell_c \) and \( \ell_a(n) = \ell_a \).
We now apply the Furer et al. [FGM+89] transformation to this puzzle to create a 3-round puzzle $\tilde{C}$. The puzzle will proceed by first having the adversary sending $\ell_c$ “pads” $z_1, \ldots, z_{\ell_c}$ to $\tilde{C}$; $\tilde{C}$ next sends back a random message $r_{\tilde{C}} \in \{0,1\}^{\ell_c}$, and the adversary is next supposed to find a response $i, p$ such that $(r \oplus z_i, p)$ is a valid transcript for the original puzzle (i.e., the adversary wins in one of the parallel “padded” instances of the original puzzle). More formally, $\tilde{C}(1^n, (z_1, \ldots, z_{\ell_c}), (i, p); r_{\tilde{C}}) = 1$ if and only if $C(1^n, p; r_{\tilde{C}} \oplus z_i)$ outputs 1. Perfect completeness of $\tilde{C}$ follows directly from the original proof by [FGM+89].

For completeness, we recall their proof. From the (imperfect) completeness of $C$, we have that there exists some adversary $A$ such that $\Pr[\langle A, C \rangle(1^n) = 1] \geq 1 - \frac{1}{n}$ for all sufficiently large $n$; without loss of $A$ is deterministic. Fix some $n > 2$ for which this holds. Let $S \subseteq \{0,1\}^{\ell_c}$ be the set of challenges for which $A$ provides an accepting response; the probability that a random challenge $z \in \{0,1\}^{\ell_c}$ is inside $S$ is thus at least $1 - \frac{1}{n}$. We will show that there exists “pads” $z_1, \ldots, z_{\ell_c}$ such that for every $r \in \{0,1\}^{\ell_c}$, there exists some $i$ such that $r \oplus z_i \in S$, which concludes that an unbounded attacker $\tilde{A}$ can succeed with probability 1 (by selecting those pads and next providing an accepting response). Note that for every fixed $r$, for a randomly chosen pad $z_i$, the probability that $r \oplus z_i \notin S$ is at most $\frac{1}{n}$; and thus the probability over randomly chosen pads $z_1, \ldots, z_{\ell_c}$ that $r \oplus z_i \notin S$ for all $i$ is at most $\frac{1}{n^{\ell_c}}$. We conclude, by a union bound, that the probability over randomly chosen pads $z_1, \ldots, z_{\ell_c}$ that there exists some $r \in \{0,1\}^{\ell_c}$ such that $r \oplus z_i \notin S$ for all $i$ is at most $\frac{2^{\ell_c}}{n^{\ell_c}} < 1$. Thus, there exists pads $z_1, \ldots, z_{\ell_c}$ such that for every $r \in \{0,1\}^{\ell_c}$ there exists some $i$ such that $r \oplus z_i \notin S$, which concludes perfect completeness.

We now turn to proving computational soundness. Consider some adversary $\tilde{A}^*$ that succeeds in convincing $\tilde{C}$ with probability $\epsilon(n)$ for all $n \in \mathbb{N}$. We construct an adversary $A^*$ that convinces $C$ with probability $\frac{\epsilon(n)}{\ell_c}$, which is a contradiction. $A^*(1^n)$ picks a random tape $r_{\tilde{A}^*}$ for $\tilde{A}^*$, lets $(z_1, \ldots, z_{\ell_c}) = \tilde{A}^*(1^n; r_{\tilde{A}^*})$, picks a random index $i \in [\ell_c]$ and outputs $z_i$. Upon receiving a “challenge” $r$, it lets $(j, p) = \tilde{A}^*(1^n, r \oplus z_i; r_{\tilde{A}^*})$ outputs $p$ if $i = j$ and $\bot$ otherwise. First, note that in the emulation by $A^*$, $A^*$ feeds $\tilde{A}^*$ the same distribution of messages as $\tilde{A}^*$ would see in a “real” interaction with $\tilde{C}$; thus, we have that the $(j, p)$ is an accepting message (w.r.t., the challenge $r \oplus z_i$) with probability $\epsilon$. Additionally, since $r \oplus z_i$ information-theoretically hides $i$ (as $r$ is completely random), we have that the probability that $i = j$ is $\frac{1}{\ell_c}$ and furthermore, the event that this happens is independent of whether the message $(j, p)$ is accepting. We conclude that $A^*$ convinces $C$ with probability $\frac{\epsilon(n)}{\ell_c}$, which concludes the soundness proof. ■

### 6.2 TFNP Hardness in Pessiland

We now conclude that TFNP is average-case hard in Pessiland.

**Theorem 6.2.** Suppose there exists a distributional NP problem $(L, D)$ that is aeHOA. Then, either of the following holds:

- There exists a OWF;
- There exists some $R \in \text{TFNP}$ and some PPT $D$ such that $\langle R, D \rangle$ is SearchHAO.

**Proof:** The theorem follows by simply applying our earlier proved results:

- From Lemma 3.3, we have that an aeHOA distributional NP problem implies a 2-round almost-everywhere puzzle.
- By Theorem 6.1 (perfect completeness through adding a round), this implies a 3-round almost-everywhere puzzle with perfect completeness.
- Applying Lemma 4.4 (round-collapse, variant 2), we conclude that either one-way functions exists, or there exists a 2-round puzzle with perfect completeness.

25
Finally, by applying Lemma 3.5, the 2-round puzzle with perfect completeness implies the existence of some $\mathcal{R} \in \text{TFNP}$ and some PPT $\mathcal{D}$ such that $(\mathcal{R}, \mathcal{D})$ is SearchHAO.

By replacing the use of Lemma 4.4 with Lemma 4.3 (round-collapse, variant 1), we instead get the following variant.

**Theorem 6.3.** Suppose there exists a distributional $\text{NP}$ problem $(L, \mathcal{D})$ that is aeHOA. Then, either of the following holds:

- There exists an ioOWF;
- There exists some $\mathcal{R} \in \text{TFNP}$ and some PPT $\mathcal{D}$ such that $(\mathcal{R}, \mathcal{D})$ is aeSearchHAO.

### 7 Characterizing Non-trivial Public-coin Arguments

We finally apply our round-collapse theorem to arguments systems.

**Non-trivial arguments** We first define the notion of a non-trivial argument.\(^{17}\) We simply say that an argument system is non-trivial if it is not a proof system—i.e., the computation aspect of the soundness condition is “real”.

**Definition 7.1** (non-trivial arguments). An argument system $(P, V)$ for a language $L$ is called non-trivial if $(P, V)$ is not an interactive proof system for $L$.

We focus our attention on public-coin arguments. We show that the existence of any $O(1)$-round public-coin non-trivial argument implies the existence of distributional $\text{NP/poly}$ problem that is nuHAO.

**Theorem 7.2.** Assume there exists a $O(1)$-round public-coin non-trivial argument for some language $L$. Then, there exists a distributional $\text{NP/poly}$ problem that is nuHAO.

**Proof:** Consider some $k$-round non-trivial public-coin argument system $(P, V)$. We show that this implies the existence of a $k$-round non-uniform puzzle. The theorem next follows by applying Corollary 4.1.

Since $(P, V)$ is not a proof system, there must exist some polynomial $p(\cdot)$, an unbounded prover $B$, and sequences $I = \{n_1, n_2, \ldots\}$ and $\{x_{n_i}\}_{i \in \mathbb{N}}$ such that for all $i \in \mathbb{N}$, $|x_{n_i}| = n_i$, $x_{n_i} \notin L$ yet $B$ convinces $V$ on common input $x_{n_i}$ with probability $\frac{1}{p(n_i)}$.

Now consider the $k$-round non-uniform puzzle $C$ that for each $n$ receives $(1, x_n)$ as non-uniform advice if $n \in I$ and otherwise $(0, \perp)$. Given non-uniform advice $(b, x)$, $C(1^n)$ simply accepts if $b = 0$ and otherwise runs the verifier $V(x_n)$. We shall argue that $C$ is a $(\frac{1}{p(n)}, \frac{1}{2p(n)})$ puzzle which by Remark 3.1 implies a puzzle. Completeness follows directly from the existence of $B$ (when $b = 0$, we have completeness 1 and otherwise, we have completeness $\frac{1}{p(n)}$ by construction). To show soundness, notice that any non-uniform PPT adversary $A^*$ that breaks soundness of the puzzle with probability $\frac{1}{2p(n)}$ for all sufficiently large $n$, must in particular break it for infinitely many $n \in I$, and as such breaks the soundness of $(P, V)$ for infinitely many $x \in \{0, 1\}^* - L$ with probability $\frac{1}{2p(|x|)}$, which contradicts the soundness of $(P, V)$.

We next remark that the implication is almost tight. The existence of a nuHOA problem in $\text{NP}$ (as opposed to $\text{NP/poly}$) implies a 2-round non-trivial public-coin argument for $\text{NP}$.

\(^{17}\)Whereas such a notion of a non-trivial argument has been discussed in the community for at least 15 years, we are not aware of an explicit formalization in the literature.
Lemma 7.1. Suppose there exists a distributional NP problem \((L, D)\) that is nuHOA. Then, for every language \(L' \in \text{NP}\), there exists a non-trivial 2-round public-coin argument for \(L'\) with an efficient prover.

Proof: We first observe that by the same proof as for Lemma 3.3, a nuHOA NP problem implies a 2-round puzzle satisfying a “weak” completeness property, where completeness only holds for infinitely many \(n \in \mathbb{N}\), but where soundness holds also against non-uniform PPT algorithms. (Recall that in the proof of Lemma 3.3, we only relied on the almost-everywhere HOA property of the NP problem to ensure that completeness held for all sufficiently large input lengths.) We next simply combine this “weakly-complete” puzzle with a standard NP proof for \(L\) to get a non-trivial 2-round argument for \(L\). More precisely, the verifier \(V(x)\) samples the first message of the puzzle and sends it to the prover; next the verifier accepts the prover’s response if it is either a witness for \(x \in L\) (for some witness relation for \(L\)), or if the response is a valid response to the puzzle. The honest prover \(P\) simply sends a valid witness for \(x\). Completeness of \((P, V)\) trivially holds. Soundness holds due to the soundness of the puzzle (w.r.t. nu PPT). By the weak completeness property of the puzzle, we additionally have that \((P, V)\) is not an interactive proof (since there are infinitely many input lengths on which an unbounded prover can find a puzzle solution and thus break soundness). ■

We finally observe that the existence of \(n^\epsilon\)-round non-trivial public-coin arguments is equivalent to \(\text{PSPACE} \not\subseteq \text{P/poly}\).

Theorem 7.3 (informally stated). For every \(\epsilon > 0\), there exists an (efficient-prover) \(n^\epsilon\)-round non-trivial public-coin argument (for NP) if and only if \(\text{PSPACE} \not\subseteq \text{P/poly}\).

Proof: The “only-if” direction follows just as the only-if direction of Theorem 5.1. The “if” direction follows by combining a standard NP proof with the puzzle from Theorem 5.1 and requiring the prover to either provide the NP witness, or to provide a solution to puzzle. ■

Round Collapse for Succinct Arguments We proceed to remark that the proof of our round-collapse theorem also has consequences for succinct [Kil92] and universal [Mic00, BG02] argument systems.

Theorem 7.4. Assume there exists a \(k\)-round public-coin (efficient-prover) argument system for \(L\) with communication complexity \(\ell(\cdot)\), where \(k\) is a constant. Then, either non-uniform ioOWFs exists, or there exists a \(2\)-round public-coin (efficient-prover) argument for \(L\) with communication complexity \(O(\ell(n)\text{polylog}(n))^{k(n)−1}\).

Proof: We apply the BM round-collapse transformation to the \(k\)-round argument system \(k−1\) times, and between each application repeat the protocol in parallel \(\log^2 n\) times (where \(n\) is the length of the statement to be proven). Completeness (also w.r.t. efficient provers) follows directly from the classic proof of the BM round collapse [BM88]. To show soundness, as before, we consider a single application of the round-collapse transformation. Consider an adversary that breaks the soundness of the \(k−1\)-round argument system with probability \(\epsilon(n)\) for infinitely many \(\{x^n\}_n\); by [PV12, HPWP10] such an adversary can be turned into an adversary that break the soundness of a single of the \(\log^2 n\) repetitions of the protocol obtained after the BM transformation with probability \(\epsilon'(n) > \frac{1}{2}\) for infinitely many \(\{x^n\}_n\). If non-uniform ioOWFs does not exist, then we can rely on the same construction as in the proof of Lemma 4.1 to construct an adversary \(B^*\) that breaks the \(k\)-round argument system for the same statements \(\{x^n\}_n\) with probability \(\frac{1}{64}\), which contradicts the soundness of the \(k\)-round argument.

Note that each step of the round-collapse transformation has a multiplicative overhead of \(O(\ell_{a}(n)\text{polylog}(n))\) where \(\ell_{a}(n)\) bounds the length of the prover messages. Therefore, iterating the round collapse transformation \(i\) times will result in a multiplicative overhead of \(O((\ell(n)\text{polylog}(n))^i)\). ■

Theorem 7.4 thus shows that the existence of a \(O(1)\)-round succinct (i.e., with sublinear or polylogarithmic communication complexity) public-coin argument systems can either be collapsed into a 2-round public-coin succinct argument for the same language (and while preserving communication complexity up to polylogarithmic factors, as well as prover efficiency), or non-uniform ioOWF exist.
It is worthwhile to also note that if the underlying $O(1)$-round protocol satisfies some notion of resettable privacy for the prover (e.g., resettable witness indistinguishability (WI) or witness hiding (WH) \cite{CGGM00,FS90}), then so will the resulting 2-round protocol. (The reason we do not consider resettable zero-knowledge is that due to \cite{OW93b} even just plain zero-knowledge protocols for non-trivial languages imply the existence of a non-uniform ioOWF; thus for resettable zero-knowledge, the result would hold vacuously assuming $NP \subseteq BPP$. However, it is not known whether (resettable) WI or WH arguments for non-trivial languages imply non-uniform ioOWFs.)

8 Acknowledgements

We are grateful to Johan Håstad and Salil Vadhan for discussions about non-trivial arguments back in 2005.

References

\begin{itemize}
\item \cite{BFNW93} László Babai, Lance Fortnow, Noam Nisan, and Avi Wigderson. BPP has subexponential time simulations unless EXPTIME has publishable proofs. \textit{Computational Complexity}, 3:307–318, 1993.
\item \cite{BG02} Boaz Barak and Oded Goldreich. Universal arguments and their applications. In \textit{IEEE Conference on Computational Complexity}, pages 194–203, 2002.
\item \cite{CGGM00} Ran Canetti, Oded Goldreich, Shafi Goldwasser, and Silvio Micali. Resettable zero-knowledge (extended abstract). In \textit{STOC ’00}, pages 235–244, 2000.
\end{itemize}


A Some Theorems from Average Case Complexity

In this section, we provide formal justifications for Lemmas 2.1, 2.2 and 2.3. We recall some previous results on average-case complexity relevant to our work.

**Theorem A.1 ([Tre05]).** Suppose that there exists a NP language $L$ and polynomials $\ell(\cdot)$ and $p(\cdot)$ such that $(L, U_\ell) \leq \frac{1}{p(n)}$-HOA (resp., $\frac{1}{p(n)}$-aeHOA and $\frac{1}{p(n)}$-nuHOA). Then there exists a NP language $L'$ and polynomial $\ell'(\cdot)$ such that $(L', U_{\ell'}) \leq \frac{1}{(\log n)^{\alpha}}$-HOA (resp., $\frac{1}{(\log n)^{\alpha}}$-aeHOA and $\frac{1}{(\log n)^{\alpha}}$-nuHOA). The value $\alpha > 0$ is an absolute constant.
Theorem A.2 ([IL90]). Suppose there exists a distributional NP search problem $(\mathcal{R}, \mathcal{D})$ that is $\frac{1}{p(n)}$-SearchHOA (resp., $\frac{1}{p(n)}$-aeSearchHOA and $\frac{1}{p(n)}$-nuSearchHOA) for some polynomial $p(\cdot)$. Then there exists a search problem $\mathcal{R}'$ and polynomials $\ell(\cdot)$ and $q(\cdot)$ such that $(\mathcal{R}', \mathcal{U}_\ell)$ is $\frac{1}{q(n)}$-SearchHOA (resp., $\frac{1}{q(n)}$-aeSearchHOA and $\frac{1}{q(n)}$-nuSearchHOA).

Theorem A.3 ([BCGL92]). Suppose that there exists a distributional NP search problem $(\mathcal{R}, \mathcal{U}_\ell)$ that is $\frac{1}{p(n)}$-SearchHOA (resp., $\frac{1}{p(n)}$-aeSearchHOA and $\frac{1}{p(n)}$-nuSearchHOA) for some polynomials $p(\cdot)$ and $\ell(\cdot)$. Then there is a NP-language $L'$ and polynomials $\ell'(\cdot)$ and $q'(\cdot)$ such that $(L', \mathcal{U}_{\ell'})$ is $\frac{1}{q(n)}$-HOA (resp., $\frac{1}{q(n)}$-aeHOA and $\frac{1}{q(n)}$-nuHOA). If we start with a distributional NP/poly search problem $(\mathcal{R}, \mathcal{U}_\ell)$ that is $\frac{1}{p(n)}$-nuSearchHOA, then we obtain $L' \in$ NP/poly such that $(L', \mathcal{U}_{\ell'})$ is $\frac{1}{q(n)}$-nuHOA.

Theorems A.1, A.2 and A.3 are stated in a slightly different form in [Tre05, IL90, BCGL92]. Namely, we highlight that the reduction is in fact “length-regular” in that solving instances of size $\ell(n)$ in the target language helps solving instances of size $n$ in the source language. We will require this stronger property for the reductions to hold in the case of almost-everywhere hardness.

Proof of Lemma 2.2. This follows immediately from Theorem A.1.

Proof of Lemma 2.3. Suppose there exists a distributional NP-search problem $(\mathcal{R}, \mathcal{D})$ that is SearchHOA (resp., aeSearchHOA and nuSearchHOA). By Theorem A.2, there exists a search problem $\mathcal{R}'$ and polynomials $\ell(\cdot), q(\cdot)$ such that $(\mathcal{R}', \mathcal{U}_\ell)$ is $\frac{1}{q(n)}$-SearchHOA (resp., aeSearchHOA and nuSearchHOA). Next, by Theorem A.3, there is a NP-language $L'$ and polynomials $\ell'(\cdot)$ and $q'(\cdot)$ such that $(L', \mathcal{U}_{\ell'})$ is $\frac{1}{q(n)}$-HOA (resp., aeHOA and nuHAO), which when combined with Theorem A.1 yields a NP language $L''$ and a polynomial $\ell''$ such that $(L'', \mathcal{U}_{\ell''})$ is $\frac{1}{(\log n)^{\alpha}}$-HOA (resp., aeHOA and nuHAO). This implies that $(L'', \mathcal{U}_{\ell''})$ is HOA (resp., aeHOA and nuHOA).

Proof of Lemma 2.1. Suppose $(L, \mathcal{D})$ is a distributional NP problem that is HOA (resp. a distributional NP/poly problem that is nuHOA), then $(\mathcal{R}, \mathcal{D})$ is SearchHOA (resp., nuSearchHOA) where $\mathcal{R}$ is the witness relation corresponding to $L$. By Lemma 2.3, we can obtain a NP (resp. NP/poly) language $L'$ and polynomial $\ell'$ such that $(L', \mathcal{U}_{\ell'})$ is HOA (resp., nuHOA).

---

A length-regular function $f : \{0,1\}^* \rightarrow \{0,1\}^*$ satisfies the properties that: (1) $|x| = |y| \Leftrightarrow |f(x)| = |f(y)|$, and (2) $|x| < |y| \Leftrightarrow |f(x)| < |f(y)|$ for any two strings $x,y$. We require the “length-regular” property for Turing (Cook) reductions where solving an instance $x$ on the target language requires queries the oracle only on instances of size $\ell(|x|)$ on the source language where $\ell$ is a non-decreasing function.