

# NP-hardness of the Minimum Circuit Size Problem from Well-Studied Assumptions

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**Abstract**—Whether the Minimum Circuit Size Problem (MCSP) is NP-hard or not is a long-standing open question. Indeed, Levin delayed the publication of his fundamental work on the theory of NP-completeness because he hoped to prove NP-completeness of MCSP.

In this paper, we present the first plausible assumptions under which MCSP is NP-hard. Specifically, we prove that MCSP is NP-hard under deterministic quasi-polynomial-time nonadaptive reductions, assuming:

- subexponentially-secure non-interactive witness indistinguishable proof systems for SAT exist,
- coNP requires subexponential-size non-deterministic circuits, and
- $P^{NP}/poly$  requires circuits of size  $\Omega(2^n/n)$ .

This is arguably the first evidence that MCSP is not in coNP, which indicates that there is no short proof that witnesses the hardness of a function.

**Index Terms**—Minimum Circuit Size Problem, NP-hardness, Cryptography, Meta-complexity

## I. INTRODUCTION

The Cook–Levin theorem is a cornerstone of computational complexity theory, introducing the notion of NP-completeness and establishing that the Satisfiability problem (SAT) is NP-complete. This landmark result was independently proved by Cook [1] and Levin [2] during the Cold War. The paper of Levin was published in 1973, two years after the publication of Cook. An intriguing anecdote<sup>1</sup> is that Levin intentionally delayed his publication because he could not prove NP-completeness of “popular” problems, such as graph isomorphism, integer factorization, and “circuit minimization” — nowadays called the *Minimum Circuit Size Problem* (MCSP) [3]. The graph isomorphism problem is known to be in  $NP \cap coAM$  [4], and, similarly, (the decision version of) the integer factorization problem is in  $NP \cap coNP$  [5]. This implies that these problems cannot be NP-complete (even under randomized polynomial-time adaptive reductions) unless  $NP \subseteq coAM$ , providing strong complexity-theoretic evidence against their NP-completeness. However, no evidence *for* or *against* NP-completeness of MCSP is known. One of the

oldest open questions, dating back to the beginning of the theory of NP-completeness, is the following:

**Question.** *Is MCSP NP-complete? Are there plausible assumptions under which this is true or false?*

MCSP is formally defined as follows. For a function  $f: \{0, 1\}^n \rightarrow \{0, 1\}$ , the *truth table* of  $f$  is defined as the string of length  $N = 2^n$  obtained by concatenating  $f(x)$  for all  $x \in \{0, 1\}^n$  in the lexicographical order. The *Minimum Circuit Size Problem* (MCSP) is the following decision problem.

Input: the truth table of a function  $f: \{0, 1\}^n \rightarrow \{0, 1\}$  and a size parameter  $s \in \mathbb{N}$ .

Task: decide whether there is a Boolean circuit of size  $s$  that computes  $f$ .

It is easy to see that MCSP is in  $E = DTIME(2^{O(N)})$ : Since any  $n$ -bit function can be computed by a circuit of size  $(1 + o(1)) \cdot 2^n/n$  and such circuits can be described by  $(1 + o(1)) \cdot 2^n$ -bit strings [6], the minimum circuit size can be calculated by exhaustively searching all the  $2^{(1+o(1)) \cdot 2^n}$  circuits in time  $2^{O(N)} = 2^{O(2^n)}$ .<sup>2</sup> Similarly, it is easy to see that MCSP is in NP. One of the best hardness results was given by Allender and Das [9], who proved that MCSP is hard for SZK (Statistical Zero Knowledge). Since SZK contains cryptographic problems believed to be intractable, such as quadratic residuosity [10] and integer factorization [11], the SZK-hardness of MCSP provides evidence for the computational intractability of MCSP.

**Meta-Complexity:** MCSP is one of the central problems in the paradigm of *meta-complexity*, i.e., “complexity of complexity.” The complexity of MCSP is referred to as meta-complexity because the problem itself asks for computing the *circuit complexity* of a given function. Another important example of meta-complexity problem is the problem of computing the *t-time-bounded Kolmogorov complexity*  $K^t(x)$  of a given string  $x$ , i.e., the length of a shortest program that prints  $x$  in time  $t$ . Meta-complexity has deep connections to many areas of theoretical computer science, such as computational learning theory [12], cryptography [13, 14], circuit lower bounds [15, 3, 16], derandomization [17, 18], and average-case complexity [19]. For example, Hirahara [19] proved that NP-hardness of an approximation version of MCSP implies the

Shuichi Hirahara was supported by JST FOREST Program, Grant Number JPMJFR226Y. Rahul Ilango was supported by NSF CCF-2420092 and an NSF graduate research fellowship.

<sup>1</sup>See Levin’s webpage (<https://www.cs.bu.edu/fac/ln/research/hard.htm>) for details.

<sup>2</sup>In fact, MCSP can be solved by a circuit of size  $2^{(0.8+o(1)) \cdot N}$  [7, 8].

equivalence between the worst- and average-case complexities of NP — one of the central problems in the theory of average-case complexity, known as the exclusion of Heuristica from Impagliazzo’s five possible worlds [20]. Meta-complexity has been actively investigated, with the hope of ruling out Heuristica [19, 21, 17, 22, 23, 24, 25, 26, 27, 28, 29, 30], Pessiland [14, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 13, 41, 42], both of Heuristica and Pessiland [43, 40, 44, 45], and Minicrypt [46, 47].

#### *Circumstantial Evidence for NP-hardness of MCSP:*

Many variants of MCSP have been shown to be NP-hard, providing circumstantial evidence for NP-hardness of MCSP. The original paper of Levin [2] presented six NP-complete problems, one of which is DNF-MCSP<sup>\*</sup> — the problem of deciding whether a given *partial* function  $f: \{0, 1\}^n \rightarrow \{0, 1, \star\}$  (represented as its truth table) has a DNF formula of size  $s$  for a given parameter  $s$ , where  $f(x) = \star$  indicates that  $f$  is undefined on input  $x$ . Masek [48] proved NP-completeness of DNF-MCSP, i.e., MCSP for DNF formulas (see also [49]). Hirahara et al. [50] extended this to DNF  $\circ$  XOR formulas. Ilango [51] proved NP-hardness of an oracle circuit version of MCSP. Ilango [52] proved NP-hardness of MCSP for constant depth formulas (under randomized quasi-polynomial-time reductions). Ilango et al. [53] proved NP-hardness of MCSP for multi-output functions. Ilango [54] proved ETH-hardness of the formula version of MCSP. Hirahara [55] proved NP-hardness of MCSP<sup>\*</sup>, i.e., the partial function variant of MCSP under randomized polynomial-time reductions. Huang et al. [56] gave improved inapproximability of the oracle circuit version of MCSP and first suggested the idea of using cryptographic assumptions for proving the hardness of MCSP. Ilango [57] proved that SAT reduces to the  $\mathcal{O}$ -oracle circuit variant of MCSP, denoted MCSP <sup>$\mathcal{O}$</sup> , with probability 1, where  $\mathcal{O}$  is a random oracle. This result provides a concrete candidate reduction for showing NP-hardness of MCSP, as the random oracle could be replaced with practical hash functions. Still, it is an outstanding question whether MCSP is NP-hard under plausible assumptions.

#### *Circumstantial Evidence against NP-hardness of MCSP:*

One difficulty is that restricted classes of reductions are not capable of establishing NP-hardness of MCSP. One of the strongest results in this direction was recently presented by Mazon and Pass [58]. They showed that a 2.01-factor approximation of MCSP is not NP-hard under randomized Levin reductions, assuming plausible cryptographic assumptions, such as the existence of an indistinguishable obfuscation. Informally, a *Levin reduction* from  $L \in \text{NP}$  to  $L' \in \text{NP}$  is a reduction that efficiently maps a witness of  $L$  to  $L'$  and vice versa. The reduction of Ilango [57] is a Levin reduction under a random oracle, which poses a fundamental barrier to extend it to NP-hardness of MCSP.

In addition to the barrier of [58], any NP-hardness proof of MCSP must avoid all the other barriers presented in the literature. Murray and Williams [59] showed that MCSP is not NP-hard under  $n^{49}$ -time reductions. An analogous result for time-bounded Kolmogorov complexity was proved by

Saks and Santhanam [60]. In fact, almost all NP-complete problems, such as SAT, Vertex Cover, and Hamiltonian Path, are also complete under  $\text{poly}(\log n)$ -time reductions (see [59] and references therein). Thus, these barriers suggest that any NP-hardness proof of MCSP must look quite different from standard NP-hardness proofs. Hirahara and Watanabe [61] proved that there is no NP-hardness reduction to the  $A$ -oracle version MCSP <sup>$A$</sup>  for some oracle  $A$  unless  $\text{NP} \not\subseteq \text{coAM}$ . Ko [62] (resp. Ren and Santhanam [63]) presented oracles under which approximating  $t$ -time-bounded Kolmogorov complexity is not NP-hard (resp. MCSP does not admit a search-to-decision reduction, and, in particular, is not NP-hard). A related line of research [3, 59, 64, 61, 65] presented the difficulty of establishing NP-hardness of MCSP unconditionally under *deterministic* reductions, by deriving complexity class separations, such as  $\text{EXP} \neq \text{ZPP}$ , from deterministic NP-hardness reductions. We also mention that an approximation version of MCSP is provably NP-intermediate if the approximation error is sufficiently large and a one-way function exists [66].

In summary, no conclusive results for NP-hardness of MCSP have been obtained so far. The result of Ilango [57] comes close to NP-hardness of MCSP, but it is subject to the barrier of Mazon and Pass [58], suggesting that a new idea would be necessary.

#### *A. Our Results*

In this paper, we present the first plausible assumptions under which MCSP is NP-hard, circumventing all the barriers in the literature [58, 60, 59, 61, 62, 63] and answering the oldest open question from the beginning of the theory of NP-completeness.

**Theorem 1.1.** *(A constant factor approximation of) the Minimum Circuit Size Problem is NP-hard under deterministic quasipolynomial-time non-adaptive reductions assuming all of the following:*

- 1) *subexponentially-secure non-interactive witness indistinguishable proof systems for SAT exist,*
- 2) *coNP requires subexponential-size non-deterministic circuits almost everywhere, and*
- 3)  *$\text{P}^{\text{NP}}/\text{poly}$  requires circuits of size  $\delta 2^n/n$  almost everywhere for some constant  $\delta > 0$ .*

All the assumptions are plausible and involve only well-studied concepts. We describe the details of each assumption below.

- 1) The first assumption states that a *non-interactive witness indistinguishable proof systems* (NIWI) [67] for SAT exist. This is a proof system in which a prover can convince a verifier that a given formula is satisfiable without revealing which satisfying assignment the prover used to generate the proof. There are several different constructions of NIWIs from various widely believed assumptions, such as the existence of a trapdoor one-way permutation and the non-deterministic exponential circuit lower bound for E [67]. See Section III-A for the

precise definition of NIWIs and the list of constructions of NIWIs.

- 2) The second assumption states that  $\text{coNP} \not\subseteq \text{i.o.NSIZE}(2^{n^\epsilon})$  for some constant  $\epsilon > 0$ . Here,  $\text{i.o.}$  stands for infinitely often, and  $\text{i.o.NSIZE}(2^{n^\epsilon})$  is the class of problems that can be solved by a non-deterministic circuit of size  $2^{n^\epsilon}$  on inputs of length  $n$  for infinitely many  $n \in \mathbb{N}$ . The assumption is a sub-exponential and almost-everywhere-hardness version of the widely believed assumption that  $\text{coNP} \not\subseteq \text{NP/poly}$ , which holds unless the Polynomial Hierarchy collapses [68].
- 3) The third assumption states that  $\text{P}^{\text{NP}}/\text{poly} \not\subseteq \text{i.o.SIZE}(\delta 2^n/n)$  for some constant  $\delta > 0$ , where  $\text{SIZE}(\delta 2^n/n)$  is the class of problems that can be solved by circuits of size  $\delta 2^n/n$  on inputs of length  $n$ . Chen et al. [69], Li [70] proved that an exponential-time version  $S_2^E$  of  $S_2^P \subseteq \text{P}^{\text{NP}}/\text{poly}$  [71] requires circuits of nearly maximum size, i.e.,  $S_2^E \not\subseteq \text{i.o.SIZE}(2^n/n)$ . Our assumption is that a similar circuit lower bound holds for  $\text{P}^{\text{NP}}/\text{poly}$ . It can be shown that this holds with probability 1 under a random oracle; see Appendix A.

How our proof circumvents the barriers of [58, 60, 59, 61, 62, 63] will be discussed in Section II-C.

As an immediate corollary, we obtain the first plausible and well-studied assumptions under which MCSP is not in  $\text{coNP}$ .

**Corollary I.2.** *MCSP  $\not\subseteq \text{coNP}$  under the three assumptions of Theorem I.1.*

This indicates the “impossibility of witnessing hardness” — that there is no short proof for the statements for circuit lower bounds, i.e., the statements that a certain function  $f$  cannot be computed by circuits of size  $s$ . The non-existence of an efficient proof system for this  $\text{coNP}$  statement (which corresponds to YES instances of the complement of MCSP) is equivalent to  $\text{MCSP} \not\subseteq \text{coNP}$ . It is widely recognized that proving circuit lower bounds for explicit functions is extremely difficult, as the current best circuit lower bound for any problem in NP is only  $(3.1 - o(1)) \cdot n$  [72]. One explanation for this difficulty of proving circuit lower bounds is the lack of short proofs for the circuit lower bound statements. Our results indicate that every efficient proof system fails to prove that a function  $f$  cannot be computed by small circuits for some  $f$ ; in other words, no worst-case variant of NP-constructive natural proofs exists in the sense of Razborov and Rudich [15].

Previously, Rudich [73] conjectured some average-case variant of  $\text{MCSP} \not\subseteq \text{coNP}$ , based on the non-deterministic hardness of a specific pseudorandom generator [74]. Our result comes somewhat close to basing his conjecture on the well-studied assumptions because of the equivalence between the worst-case complexity of the approximation version of MCSP and the average-case complexity of MCSP [19].

It is natural to ask whether our assumptions are necessary. Because our reduction is deterministic, the results in the line of research [3, 59, 64, 61, 65] indicate that some assumptions

would be necessary. (For example, Murray and Williams [59] showed that NP-hardness of MCSP under deterministic reductions implies  $\text{EXP} \neq \text{ZPP}$ .) In fact, using non-uniform reductions, we can remove the third assumption of Theorem I.1, at the cost of worst-case hardness. Below, being NP-hard on average roughly means that solving any NP problem on average with respect to the uniform distribution reduces to MCSP.<sup>3</sup>

**Corollary I.3** (informal; see Corollary V.9). *Assume that*

- 1) *subexponentially-secure NIWIs for SAT exist, and*
- 2)  *$\text{coNP}$  requires subexponential-size non-deterministic circuits almost everywhere.*

*Then, MCSP is NP-hard on average under infinitely-often non-uniform subexponential-time non-adaptive reductions.*

## II. TECHNICAL OVERVIEW

Our starting point is (a simplified version of) Hirahara’s proof that it is NP-hard to compute the circuit complexity of a partial function  $f: \{0, 1\}^n \rightarrow \{0, 1, \star\}$  [55]. We begin by reviewing this in Section II-A. We present the ideas of our reduction in Section II-B. Lastly, we explain how our proof circumvents all the barriers in the literature in Section II-C.

### A. An Exposition of Hirahara’s NP-hardness of MCSP\*

a) *Starting Problem: Vertex Cover:* The reduction is from the NP-hard problem Vertex Cover.<sup>4</sup> Recall that a vertex cover is a subset of vertices that intersect every edge. It is NP-hard to calculate the minimum size OPT of a vertex cover in a given graph.

b) *Tool: Extractable Information Theoretic Encryption:* It will be useful to consider the following information-theoretic symmetric encryption scheme, closely related to Shannon’s one-time pad and the standard construction of an encryption scheme from a pseudorandom function.

- *Key Generation:* sample a uniformly random function  $H: [\lambda] \rightarrow \{0, 1\}$  where  $\lambda$  is the security parameter.
- *Encryption:* To encrypt a bit  $b$ , one samples  $r \leftarrow [\lambda]$  uniformly at random and outputs  $\text{ct} = (r, b \oplus H(r))$ . We denote this algorithm by  $\text{Enc}(H, b)$ .
- *Decryption:* To decrypt  $\text{ct} = (r, c)$ , one outputs  $c \oplus H(r)$ . We denote this algorithm by  $\text{Dec}(H, \text{ct})$ .

An useful property of this scheme is *extractability*. If an adversary  $A$  can distinguish an encryption of 1 from an encryption of 0, then one can extract out the secret key (in a certain quantitative sense). We delay describing this until we have to use it, but we mention that it will follow immediately from the distinguisher-to-predictor lemma [75].

<sup>3</sup>The uniform distribution can be replaced with any polynomial-time samplable distribution because Impagliazzo and Levin [13] reduced solving NP on any polynomial-time samplable distribution to solving NP on the uniform distribution.

<sup>4</sup>In our actual proof, we will need to switch to hypergraph vertex cover to get better hardness of approximation. We ignore this in this high-level overview.

Given an  $n$ -vertex graph with edge set  $E$ :

- 1) Sample secret keys  $H_v: [\lambda] \rightarrow \{0, 1\}$  for all  $v \in [n]$
- 2) Define the partial function  $f: E \times ([\lambda] \times \{0, 1\}) \times ([\lambda] \times \{0, 1\}) \rightarrow \{0, 1, \star\}$  as follows:

$$f(e = (v, w), ct_v, ct_w) = \begin{cases} \text{Dec}(H_v, ct_v) & \text{if } \text{Dec}(H_v, ct_v) = \text{Dec}(H_w, ct_w), \\ \star & \text{otherwise.} \end{cases}$$

- 3) Output that OPT is “roughly” the circuit complexity of  $f$  divided by  $\lambda$ .

Fig. 1. The reduction from Vertex Cover to MCSP\*.

*c) Reduction:* We can now describe the reduction. Given a graph with edge set  $E$ , one first samples independent secret keys for each vertex. Then one considers the function  $f$  that takes as input an edge  $e = (v, w) \in E$  and two “purported” encryptions  $ct_v$  and  $ct_w$ . The function is undefined in the case where  $ct_v$  and  $ct_w$  encrypt different bits (according to the secret keys of the respective vertices). But if they encrypt the same bit  $b$ , the function outputs  $b$ . We have written the reduction formally in Figure 1 (parameterized by a choice of security parameter  $\lambda$ ).

The intuition is that in order to compute  $f$  one ought to memorize the secret key of at least one vertex in each edge (so that you can decrypt at least one of the ciphertexts). Hence, one should need to memorize at least OPT (the optimal vertex cover size) many  $\lambda$ -length secret keys. So the complexity of  $f$  should be roughly  $\text{OPT} \cdot \lambda$ .

For the upper bound, one can show that indeed  $f$  can be computed by a circuit of size roughly  $\text{OPT} \cdot \lambda$  (as long as we set  $\lambda$  to be a sufficiently large polynomial in  $n$ ).

*d) Lower Bound Proof:* The proof of the lower bound works as follows. Fix a circuit  $C$  computing  $f$ . Fix an arbitrary  $e$  in  $E$ . Now we will use the encryption scheme’s extractability property. By the definition of  $f$ , we have that

$$\Pr_{\substack{b \leftarrow \{0,1\}, \\ ct_v \leftarrow \text{Enc}(H_v, b) \\ ct_w \leftarrow \text{Enc}(H_w, b)}} [C(e, ct_v, ct_w) = b] = 1.$$

On the other hand, if we instead sampled  $ct_v$  and  $ct_w$  as truly random strings, the chance  $C(e, ct_v, ct_w) = b$  is exactly one half (because the inputs to  $C$  are independent of  $b$ ). Hence, by a hybrid argument and the distinguisher-to-predictor lemma (Lemma III.3), one gets that for at least one element of  $e$  (for simplicity, assume  $v$ ) we have

$$K_{1/2+\Omega(1)}^C(H_v) \leq O(n + \log \lambda),$$

where we use the notation  $K_p^C(z)$  to denote the length of the shortest program that, given oracle access to  $C$ , outputs a string that agrees with  $z$  on a  $p$ -fraction of coordinates.<sup>5</sup> By incorporating error-correction into the encryption scheme, one can improve this to instead be

$$K^C(H_v) \leq O(n + \log \lambda).$$

Now since the argument above held for an arbitrary edge, we in fact get that in every edge there is a vertex  $v$  with

$$K^C(H_v) \leq O(n + \log \lambda).$$

Thus, the above bound holds for OPT many vertices. Label these vertices  $v_1, \dots, v_{\text{OPT}}$ . One can then show that

$$K^C(H_{v_1} \dots H_{v_{\text{OPT}}}) \leq O((n^2 + \log \lambda) \cdot \text{OPT})$$

so

$$K(H_{v_1} \dots H_{v_{\text{OPT}}}) \leq O((n^2 + \log \lambda) \cdot \text{OPT}) + |C|.$$

Because we choose the  $\lambda$ -length secret keys uniformly at random, we expect  $K(H_{v_1} \dots H_{v_{\text{OPT}}}) \approx \text{OPT} \cdot \lambda$ . Hence, setting  $\lambda = \text{poly}(n)$  sufficient large we get that

$$|C| \approx \text{OPT} \cdot \lambda,$$

as desired.

## B. Our Reduction

The reason why Hirahara’s reduction requires partial functions is to handle the case of two ciphertexts that encrypt different bits. We will fix this by modifying  $f$  to also include a proof  $\pi$  that the two ciphertexts are equivalent. However, this approach faces two immediate problems.

*a) Less Serious Problem: The Statement Lacks Short Proofs:* Recall, that we have chosen the secret keys  $H_v$  and  $H_w$  uniformly at random. It does not seem possible in this case to give a proof that ciphertexts  $ct_v$  and  $ct_w$  decrypt to the same bit (under their corresponding secret keys), unless we assume the verifier reading the proof knows a significant amount of information about our random choice of  $H_v$  and  $H_w$ . This seems impossible in our setting because we actually want the verifier to *not* know one of these secret keys.

To fix this, we do not choose our secret keys at random. Imagine instead that we choose the  $\lambda$ -length secret keys  $H_v$  and  $H_w$  such that any bit in the secret key corresponds to a  $O(\log \lambda)$ -length instance of an  $\text{NP} \cap \text{coNP}$  language (in fact, by being more clever one can even consider a  $\text{P}^{\text{NP}}$  language, but for now consider  $\text{NP} \cap \text{coNP}$ ).

Then to prove that two ciphertexts decrypt to the same value we can do the following. Recall that a ciphertext is of the form  $(r_v, c_v)$  where  $c_v = b \oplus H_v(r_v)$  and  $b$  is the bit being encrypted. Also, recall  $H_v(r_v)$  is a bit that corresponds to an instance of a  $\text{NP} \cap \text{coNP}$  problem. Hence, to prove that

<sup>5</sup>See Definition III.2 for a formal definition.

Given an  $n$ -vertex graph with edge set  $E$ :

- 1) Let  $\{H_v : \{\lambda\} \rightarrow \{0, 1\}\}_{v \in [n]}$  be “very hard”  $P^{NP}$  functions.
- 2) For a formula  $\psi$ , define the function  $f_\psi$  as follows:

$$f_\psi(e = (v, w) \in E, ct_v, ct_w, \pi) = \begin{cases} \text{Dec}(H_v, ct_v) & \text{if } \pi \text{ is a NIWI-proof that} \\ & \text{“either } \text{Dec}(H_v, ct_v) = \text{Dec}(H_w, ct_w) \text{ or } \psi \text{ is satisfiable”,} \\ 0 & \text{otherwise.} \end{cases}$$

- 3) Output that OPT is “roughly” the maximum (over all “small”<sup>a</sup> unsatisfiable  $\psi$ ) of the circuit complexity of  $f_\psi$  divided by  $\lambda$ .

<sup>a</sup>For example, one setting of “small” we use is  $\text{poly log}(n)$ .

Fig. 2. The informal description of our reduction from Vertex Cover to MCSP.

two ciphertexts  $(r_v, c_v)$  and  $(r_w, c_w)$  encrypt the same bit, we can provide the  $NP \cap \text{coNP}$  witnesses that reveal the value of  $H_v(r_v)$  and  $H_w(r_w)$  respectively. Given this, we can easily check if the two ciphertexts encrypt the same bit.

Now at least a short proof of equivalence exists, but the proof we described above is actually not good for us. In the process of proving that the ciphertexts are equivalent it also reveals the bit they encrypt. As a result, the function  $f$  actually becomes easy to compute, regardless of the minimum vertex cover size.

*b) More Serious Problem: How Do You Keep  $f$  Hard:*

Indeed, this brings us to our main technical challenge. How can we prove that the ciphertexts are equivalent while still maintaining the hardness of  $f$ ? A natural idea is to use a zero knowledge proof [10], which guarantees the verifier “learns nothing except the validity of the statement.”

Unfortunately, this idea runs into two significant issues. First, in our setting the proof is *completely non-interactive*: the circuit is just given a proof  $\pi$  and needs to decide whether it is valid or not. Second, we need the proof to have *perfect soundness*. The verifier can never be convinced that two ciphertexts are equivalent when they are not. No known zero knowledge proofs have either of these two properties, and indeed, Goldreich and Oren [76] show that achieving either of them is impossible (under a standard complexity assumption).

Luckily, we do not actually need the full strength of zero knowledge. We only need a very specific security property to hold: that computing  $f$  is still hard. A proof system with such a property was constructed by Kuykendall and Zhandry [77]. For every NP-search problem  $R$  that is hard on average, they constructed a non-uniform *witness-hiding* proof system, in which a prover can convince a verifier that an input  $x$  has some witness in  $R$  without revealing the witness: no efficient adversary can solve  $R$  on average even given a proof from the prover.

*c) Using Non-interactive Witness Indistinguishable Proofs:* The proof system of Kuykendall and Zhandry [77] is based on non-interactive witness indistinguishable proofs (NIWIs). This is one relaxation of zero knowledge that is

achievable with no interaction and perfect soundness. A NIWI proof that a circuit  $C$  is satisfiable has the following “witness indistinguishability” security property: one cannot tell which satisfying assignment to  $C$  was used to generate the proof (except if  $C$  is uniquely satisfiable).

Building on Kuykendall and Zhandry [77], we will do the following. Let  $\psi$  be a formula that we choose later. We modify  $f$  to include a NIWI proof  $\pi$  that “either these two ciphertexts encrypt the same bit or  $\psi$  is satisfiable.”

Now if  $\psi$  is satisfiable, then this proof says nothing about whether the two ciphertexts encrypt the same bit. However, precisely because of this, one can show (via the witness indistinguishability property) that providing such a proof is useless and that  $f$  is still hard. In fact,  $f$  is even hard “on average,” which will be useful for us later.

On the other hand if  $\psi$  is unsatisfiable, then a proof that “either these two ciphertexts encrypt the same bit or  $\psi$  is satisfiable” does indicate that the ciphertexts indeed encrypt the same bit. However, it is not clear whether  $f$  is still hard.

Luckily, using the ideas of Kuykendall and Zhandry, we get that there should be *some* unsatisfiable  $\psi$  for which  $f$  is still hard. The reason is that  $f$  being easy is a short certificate that  $\psi$  is unsatisfiable. This is because (as mentioned earlier)  $f$  is indeed hard, even on average, when  $\psi$  is satisfiable. So if you are given a circuit that computes  $f$  on random inputs, it must be that  $\psi$  is unsatisfiable. (We note that this argument is subtle and breaks down for super-linear-time-bounded Kolmogorov complexity.)

But if one believes  $NP \neq \text{coNP}$  (and its extensions), then it is impossible to certify every unsatisfiable formula is indeed unsatisfiable. Hence,  $f$  should be hard for some unsatisfiable  $\psi$ . In our reduction, we will brute force over all small unsatisfiable  $\psi$  in order to find one with this property.

*d) The Reduction:* We give a very informal description our reduction in Figure 2.

### C. How the Barriers Are Circumvented

Here, we explain how our proof circumvents all the barriers in the literature.

a) *The Barrier of Mazon and Pass*: Mazon and Pass [58] show that, under cryptographic assumptions, approximating MCSP to a factor of 2.01 is not NP-hard under Levin reductions. Because our results give hardness of approximation to any constant factor, our reduction must overcome this barrier. Indeed, we overcome it in a rather remarkable way: our reduction is not Levin, but it is in some sense *indistinguishable* from a Levin reduction.

In more detail, let  $R$  be a reduction that maps instances  $x$  of a language  $L \in \text{NP}$  to an instance  $R(x)$  of another language  $L' \in \text{NP}$ . For this to be a valid reduction, it must be sound, meaning that if  $R(x) \in L'$ , then  $x \in L$ . A *Levin* reduction has the property that its soundness proof is algorithmic: there is a polynomial-time algorithm that maps a witness that  $R(x) \in L$  to a witness that  $x \in L$ .<sup>6</sup> For the vast majority of NP-hardness reductions, such an algorithm usually follows immediately by a reduction's proof of soundness, which are almost always constructive. Consequently, this barrier seems formidable: what non-constructive techniques could be useful in a soundness proof?

The key idea for us is as follows. In our reduction, we iterate through all unsatisfiable formulas  $\psi$ . But in our analysis, we say that — because of non-deterministic hardness of UNSAT — this should actually be indistinguishable from the case where we iterate through at least one satisfiable  $\psi$ . For a satisfiable  $\psi$ , however, we can actually prove soundness algorithmically. Specifically, we can use the encryption scheme's extractability property to extract from any small circuit a satisfying assignment.

In this sense, our reduction is not Levin (because we only iterate through unsatisfiable  $\psi$ ), but almost Levin (because in our analysis we pretend  $\psi$  is satisfiable).

b) *The Barrier of Saks and Santhanam, Murray and Williams*: Saks and Santhanam [59] showed that any randomized nonadaptive reduction that shows NP-hardness of approximating  $t$ -time-bounded Kolmogorov complexity must spend at least  $t^{\Omega(1)}$  time under plausible assumptions. Similarly,<sup>7</sup> Murray and Williams [59] showed that there is no deterministic  $n^{49}$ -time reduction from NP to MCSP.

A high-level intuition behind these barriers is that any  $t$ -time (deterministic) reduction cannot create a string with high  $t$ -time-bounded Kolmogorov complexity. Our proof overcomes the barriers, by spending (at least)  $\text{poly}(t)$  time to construct some string with nearly maximum  $t$ -time-bounded Kolmogorov complexity, which can be obtained from the hard function in  $\text{P}^{\text{NP}}/\text{poly}$  that requires the nearly maximum circuit complexity (Lemma V.6).

c) *The Barrier of Hirahara and Watanabe*: Hirahara and Watanabe [61] showed that there is no randomized NP-hardness reduction to  $\text{MCSP}^A$  for some oracle  $A$  unless  $\text{NP} \not\subseteq \text{coAM}$ . We bypass this barrier because our result does not generalize to  $\text{MCSP}^f$ , where  $f$  is the hard function in  $\text{P}^{\text{NP}}/\text{poly}$  with nearly maximum circuit complexity. This is in

contrast to the NP-hardness of  $\text{MCSP}^*$  [55], which can be generalized to  $\text{MCSP}^{*,A}$  for every oracle  $A$ .

d) *The Relativization Barriers of Ko [62], Ren and Santhanam [63]*: Our proof does not relativize, just as in the case of NP-hardness of  $\text{MCSP}^*$  [55]. Hirahara also proved NP-hardness of a problem called MINLT [62], which is provably non-relativizing. The primary non-relativizing component in our reduction (and [55]) is the use of the NP-hardness of approximating (hypergraph) vertex cover [79], which relies on the PCP theorem. The PCP theorem is non-relativizing [80].

## Organization

The remainder of this paper is organized as follows. After presenting preliminaries in Section III, we introduce the notion of somewhat  $\text{NP} \cap \text{coNP}$  computable functions, which enables us to reduce the computation of  $\text{P}^{\text{NP}}$  to some form of non-uniform  $\text{NP} \cap \text{coNP}$  computations in Section IV. The proof of Theorem I.1 is given in Section V. In Appendix A, a strong lower bound for NP is proved under a random oracle.

## III. PRELIMINARIES

a) *Notation*:  $[n]$  denotes  $\{1, \dots, n\}$ . The *circuit complexity* of a function  $f: \{0, 1\}^n \rightarrow \{0, 1\}$ , denoted by  $\text{CC}(f)$ , is the minimum size of any circuit that computes  $f$ . We use the same notation  $\text{CC}(t)$  for the truth table  $t \in \{0, 1\}^{2^n}$  of  $f$ . For a string  $x \in \{0, 1\}^*$ , let  $|x|$  denote the length of  $x$ .

### A. Non-Interactive Witness Indistinguishable Proofs

We recall the definition of a non-interactive witness indistinguishable proof system (NIWI) [81, 82, 67] for SAT. (In this paper, whenever we refer to NIWIs, we refer to NIWIs for SAT.)

**Definition III.1** (Non-Interactive Witness Indistinguishable Proof (NIWI)). A *non-interactive witness indistinguishable proof system* is a tuple of algorithms  $(\text{NIWI.Prove}, \text{NIWI.Verify})$  satisfying all of the following properties:

- $\text{NIWI.Prove}(\text{formula } \varphi, \text{satisfying assignment } w, \text{security parameter } 1^\lambda)$  is a uniform randomized polynomial time algorithm that outputs a binary string (usually denoted  $\pi$ ).
- $\text{NIWI.Verify}(\text{formula } \varphi, \text{purported proof } \pi)$  is a uniform polynomial-time algorithm that outputs either 0 or 1.
- **Completeness**: For all formulas  $\varphi$  with  $\varphi(w) = 1$  and all  $\lambda$ ,

$$\Pr[\text{NIWI.Verify}(\varphi, \text{NIWI.Prove}(\varphi, w, 1^\lambda), 1^\lambda) = 1] = 1,$$

where the probability is taken over the internal randomness of  $\text{NIWI.Prove}$ .

- **Soundness**: For all unsatisfiable formulas  $\varphi$ , all strings  $\pi$ , and all  $\lambda$ ,

$$\text{NIWI.Verify}(\varphi, \pi, 1^\lambda) = 0.$$

- **Security (Witness Indistinguishability)**: There is a negligible function  $\epsilon$  such that for all formulas  $\varphi$  and all

<sup>6</sup>It also has the vice-versa property, but that will be less important to us.

<sup>7</sup>MCSP can be regarded as the problem of computing sublinear-time-bounded Kolmogorov complexity [78].

assignments  $w$  and  $w'$  with  $\varphi(w) = \varphi(w') = 1$ , for all adversary circuits  $A$  of size at most  $1/\epsilon(\lambda)$ , we have that

$$\left| \Pr[A(\text{NIWI.Prove}(\varphi, w, 1^\lambda)) = 1] - \Pr[A(\text{NIWI.Prove}(\varphi, w', 1^\lambda)) = 1] \right| < \epsilon(\lambda), \quad (1)$$

where the probability is taken over the internal randomness of  $\text{NIWI.Prove}$ .

There are several different constructions of NIWIs from various widely-believed cryptographic assumptions. For example, NIWIs exist either

- if indistinguishability obfuscation and one-way permutations exist [83],
- if a trapdoor one-way permutation exists and  $E$  requires  $2^{\Omega(n)}$ -size non-deterministic circuits [67],
- if indistinguishability obfuscation and one-way functions exist and  $E$  requires  $2^{\Omega(n)}$ -size non-deterministic circuits [67, 83], or
- assuming certain hardness [84] for bilinear groups [85].

### B. Time-Bounded Kolmogorov Complexity

Fix an efficient universal Turing machine  $U$ . We write  $U^t(d)$  to denote the output of running  $U$  on  $d$  for  $t$  steps.

**Definition III.2.** The  $p$ -average  $t$ -time  $\mathcal{O}$ -oracle Kolmogorov complexity of a string  $x$  is  $K_p^{t, \mathcal{O}}(x) = \min\{|d| : U^t(d, i) = x_i \text{ for a } p\text{-fraction of } i \in [|x|]\}$ . We omit writing  $p$  when  $p = 1$ . We omit writing  $\mathcal{O}$  when  $\mathcal{O} = \emptyset$ .

The following lemma will be useful to us.

**Lemma III.3** (Distinguisher to Predictor [75]). *Let  $f: \{0, 1\}^n \rightarrow \{0, 1\}$ . Let  $\mathcal{D}$  be a distribution on  $\{0, 1\}^q$ . Let  $D$  be a circuit that outputs a bit and satisfies*

$$\left| \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D} \\ b = f(r)}} [D(z, r, b) = 1] - \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D} \\ b \leftarrow \{0, 1\}}} [D(z, r, b) = 1] \right| > \epsilon \quad (2)$$

Then  $K_{1/2+\epsilon}^{\text{poly}(n, |z|), D}(f) \leq |z| + O(1)$ .

*Proof.* Without loss of generality, assume (2) holds without the absolute value signs (if not, then the argument is similar, except one negates the output of  $D$ ). Consider the randomized algorithm  $P(r)$  that samples  $b \leftarrow \{0, 1\}$ ,  $z \leftarrow \mathcal{D}$  and outputs  $b \oplus 1 \oplus D(z, r, b)$ . Then we have that

$$\begin{aligned} & \Pr_{r \leftarrow \{0, 1\}^n} [P(r) = f(r)] \\ &= \frac{1}{2} \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D}}} [D(z, r, f(r)) = 1] + \frac{1}{2} \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D}}} [D(z, r, 1 \oplus f(r)) = 0] \\ &= \frac{1}{2} \left( \mathbb{E}_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D}}} [D(z, r, f(r)) + 1 - D(z, r, 1 \oplus f(r))] \right) \\ &= \frac{1}{2} + \frac{1}{2} \left( \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D}}} [D(z, r, f(r)) = 1] - \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D}}} [D(z, r, 1 \oplus f(r)) = 1] \right) \\ &= \frac{1}{2} + \frac{1}{2} \cdot 2 \left( \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D} \\ b = f(r)}} [D(z, r, f(r)) = 1] - \Pr_{\substack{r \leftarrow \{0, 1\}^n \\ z \leftarrow \mathcal{D} \\ b \leftarrow \{0, 1\}}} [D(z, r, 1 \oplus f(r)) = 1] \right) \\ &\geq \frac{1}{2} + \epsilon. \end{aligned}$$

Then by an averaging argument we can fix  $b \in \{0, 1\}$  and  $z \in \{0, 1\}^q$ , so that  $P_{b, z}$  is a deterministic algorithm satisfying  $\Pr_r [P(r) = f(r)] \geq 1/2 + \epsilon$ . Thus, we get that  $K^{\text{poly}(n, q), D}(f) \leq q + O(1)$ .  $\square$

We will also use the following bound on the complexity of biased strings.

**Theorem III.4** (Time-Bounded Kolmogorov Complexity of Biased Strings [86]). *Let  $x \in \{0, 1\}^n$  with  $k \leq n/2$  ones. Then for all  $t$ ,*

$$K^{O(t^2)}(x) \leq \log \binom{n}{k} + O\left(\frac{n}{t}\right) \leq nH\left(\frac{k}{n}\right) + O\left(\frac{n}{t}\right)$$

where  $H$  is the binary entropy function which satisfies  $H(x) \leq 3\sqrt{x}$ .

### C. Hardness Amplification

Yao's XOR lemma implies that one can amplify average-case hardness in the following way.

**Lemma III.5.** *There is a deterministic algorithm that takes as input  $\epsilon > 0$  such that  $\epsilon^{-1} \in \mathbb{N}$  and the truth table of a Boolean function  $f: \{0, 1\}^n \rightarrow \{0, 1\}$  and outputs in time  $\text{poly}(2^{n'})$  the truth table of a function  $\text{Amp}(f, \epsilon): \{0, 1\}^{n'} \rightarrow \{0, 1\}$  with the following properties:*

- $n' \leq O(n/\epsilon^2)$ .
- $\text{Amp}(f, \epsilon)$  is computable by an oracle circuit of size  $\text{poly}(\frac{n}{\epsilon^2})$  that makes  $O(1/\epsilon^2)$  non-adaptive oracle queries to  $f$ .
- The following holds for all  $t$ :

$$K_{1-\epsilon}^{\text{poly}(tn/\epsilon)}(f) \leq K_{1/2+\epsilon}^t(\text{Amp}(f, \epsilon)) + \text{poly}\left(\frac{n}{\epsilon}\right).$$

*Proof Sketch.* We define  $\text{Amp}(f, \epsilon)$  as the  $k$ -wise XOR of  $f$  such that

$$\text{Amp}(f, \epsilon)(x_1, \dots, x_k) := f(x_1) \oplus \dots \oplus f(x_k)$$

for some parameter  $k$  chosen later. Impagliazzo et al. [87] proved a uniform version of Yao's XOR lemma: There exists a randomized algorithm  $R$  that, given oracle access to a function  $g$  that agrees with  $\text{Amp}(f, \epsilon)$  on a  $(1/2 + \epsilon)$  fraction of inputs, outputs a list of  $O(1/\epsilon^2)$  many oracle circuits  $C^g$ , one of which computes  $f$  on a  $(1 - \delta)$  fraction of inputs with probability  $\Omega(\epsilon)$  in time  $\text{poly}(n, k, 1/\epsilon)$ , where  $\delta = O((\log(1/\epsilon))/k)$ . We choose  $k := O(1/\epsilon^2)$  so that  $\delta \leq \epsilon$ . Then we have  $n' = nk = O(n/\epsilon^2)$ .

To see the last property, let  $A$  be a  $t$ -time machine that computes  $\text{Amp}(f, \epsilon)$  on a  $(1/2 + \epsilon)$  fraction of inputs. By running  $R$  with oracle access to  $A$ , with a positive probability,  $R^A$  outputs a list of machines, one of which computes  $f$  on a  $(1 - \delta)$  fraction of inputs. Such a machine can be specified by the internal randomness of  $R^A$  and the index of the machine in the list, which cost  $\text{poly}(n/\epsilon)$  bits and  $O(\log 1/\epsilon)$  bits, respectively.  $\square$

#### D. Hypergraph Vertex Cover

A  $\tau$ -uniform hypergraph is specified by the vertex set  $[n]$  and an edge set  $E \subseteq [n]^\tau$ . The minimum vertex cover size for such a hypergraph is defined as

$$\min\{|V| : |V \cap e| \geq 1 \text{ for all } e \in E\}.$$

The following NP-hardness is known for this problem.

**Theorem III.6** ([79]). *There is a universal constant  $\gamma \geq 1$  such that the following promise problem is NP-hard for all  $\gamma < \tau \leq \log^{1/\gamma} n$  under  $n^{O(\log \tau)}$ -time reductions*

- **Given:** A  $\tau$ -uniform hypergraph on vertex set  $[n]$  and edge set  $E \subseteq [n]^\tau$
- **Output YES:** if the minimum vertex cover size is at most  $\gamma \frac{n}{\tau}$
- **Output NO:** if the minimum vertex cover size is at least  $n(1 - \gamma/\tau)$

We refer to this problem as Gap  $\tau$ -Uniform Vertex Cover.

Throughout this paper,  $\gamma$  refers to the universal constant in Theorem III.6.

#### E. Uhlig's Theorem

Uhlig [88, 89] shows an efficient way of computing a function on multiple inputs, at least when the function is very hard.

**Theorem III.7** (Uhlig's theorem [88, 89]; see also [90]). *Let  $f : \{0, 1\}^n \rightarrow \{0, 1\}$ . Let  $f^k : (\{0, 1\}^n)^k \rightarrow \{0, 1\}^k$  be the  $k$ -wise direct sum of  $f$ . If  $k \leq 2^{n/\log^2 n}$  and  $n$  is sufficiently large, then there is a circuit for  $f^k$  of size at most  $2 \cdot \frac{2^n}{n}$*

#### IV. SOMEWHAT NP $\cap$ coNP COMPUTABLE FUNCTIONS

We define a certain class of functions that we call *somewhat NP  $\cap$  coNP computable*.

**Definition IV.1.** We say a function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  is  $(S, s)$ -somewhat NP  $\cap$  coNP-computable if there exists a  $S$ -size circuit  $\text{Gen}$  that on every input  $x \in \{0, 1\}^n$  outputs a circuit  $D_x$  of size at most  $s$  such that all the following hold:

- The range of  $D_x$  is  $\{0, 1, \perp\}$ ,
- if  $D_x(w) \neq \perp$ , then  $D_x(w) = f(x)$ , and
- $D_x(w) = f(x)$  for some  $w$ .

Any  $\text{P}^{\text{NP}}$ /poly computation can be simulated by somewhat NP  $\cap$  coNP computation.

**Proposition IV.2.** *Let  $m \in [n]$ . Let  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  be computed by a SAT-oracle circuit  $C^{\text{SAT}}$ . Then  $f$  is  $(S, s)$ -somewhat NP  $\cap$  coNP-computable, where*

$$S \leq n^2 2^{n-m} |C| + \text{poly}(|C|, 2^m)$$

and

$$s \leq \text{poly}(|C|, 2^m).$$

Moreover, the  $S$ -size generator for  $f$  can be computed in exponential time given as input the description of the oracle circuit  $C$ .

*Proof.* The idea is that we can divide the set of all inputs into buckets. In each bucket  $B$ , we reveal the value  $v_{i,B}$  of how many inputs in that bucket have their  $i$ 'th oracle query equal to YES. Then to certify the value of  $f(x)$  you reveal witnesses to every YES oracle query for every input in the same bucket as  $x$ .

In more detail, let  $\mathcal{B}$  be a partition of  $\{0, 1\}^n$  that we choose later. For a bucket  $B \in \mathcal{B}$ , let  $v_{i,B}$  be given by  $v_{i,B} = |\{x \in B : \text{the answer to the } i\text{'th oracle query in } C^{\text{SAT}}(x) \text{ is YES}\}|$ .

For an input  $x$  in the bucket  $B$ , we construct the verifier circuit  $D_x$  as follows.

##### Circuit $D_x$

Given a collection of witnesses  $\mathcal{W}_i$ :

- 1) Simulate running  $C^{\text{SAT}}(y)$  on every  $y \in B$ . Answer an oracle query  $\varphi$  as follows: evaluate  $\varphi$  on every element of  $\mathcal{W}$ . Reply YES if it ever outputs one and reply NO otherwise.
- 2) Let  $\tilde{v}_i$  be the number of times we reply YES on the  $i$ 'th query in the above simulation (i.e., summed over all  $y \in B$ ).
- 3) If  $\tilde{v}_i \neq v_{i,B}$  for some  $i$ , then output  $\perp$ . (To do this, we hardcode in the values of  $v_{i,B}$  for all  $i$ .)
- 4) Otherwise, output the simulated output of  $C^{\text{SAT}}(x)$

It is easy to see that we can construct a  $D_x$  of size at most  $\text{poly}(|C|, |B|)$ . Furthermore, it is easy to see that we have that both

$$D_x(\mathcal{W}) \neq \perp \implies f(x) = D_x(\mathcal{W})$$

and

$$D_x(\mathcal{W}) \neq \perp \text{ for some } \mathcal{W}.$$



It remains to construct Gen. Let  $m \in [n]$  be a parameter we choose later. We choose buckets to be sets of the form

$$B[z] = \{x \in \{0, 1\}^n : \text{the first } (n - m) \text{ bits of } x \text{ are equal to } z\},$$

where  $z \in \{0, 1\}^{n-m}$ .

Then given  $z, x \in B[z]$  and the values of  $v_{i, B[z]}$  for all  $i$ , one can construct  $D_x$  in time  $\text{poly}(|C|, |B[z]|) = \text{poly}(|C|, 2^m)$ . Furthermore, each  $v_{i, B[z]}$  is an integer between 0 and  $2^m$ , so it requires  $m$  bits to specify. Hence, using the trivial DNF construction, there is a circuit of size  $mn2^{n-m}|C| \leq n^22^{n-m}|C|$  that takes as input  $z$  and outputs  $v_{i, B[z]}$  for all  $i$ . Thus, Gen has size at most

$$n^22^{n-m}|C| + \text{poly}(|C|, 2^m).$$

□

We will also use the fact that being somewhat NP  $\cap$  coNP computable is closed under non-adaptive oracle calls.

**Proposition IV.3.** *Let  $f$  be  $(S, s)$ -somewhat NP  $\cap$  coNP-computable. If  $f'$  is computable by a  $t$ -size  $f$ -oracle circuit  $C$  making  $q$  non-adaptive queries, then  $f'$  is  $(S', s')$ -somewhat NP  $\cap$  coNP-computable where*

$$S' = t + q \cdot S + \text{poly}(q)$$

and

$$s' = t + q \cdot s + \text{poly}(q).$$

Moreover, given the corresponding  $S$ -size Gen for  $f$  and the circuit  $C$ , one can generate the corresponding  $S'$ -size Gen' for  $f'$  in uniform polynomial-time.

*Proof.* Let Gen be the circuit witnessing that  $f$  is  $(S, s)$ -somewhat NP  $\cap$  coNP-computable. The proposition follows from the following construction of Gen' for  $f'$ .

On input  $x$ :

- 1) Run  $C(x)$  to generate the  $q$  oracle queries  $y_1, \dots, y_q$ .
- 2) Let  $D_i = \text{Gen}(y_i)$  for all  $i \in [q]$ .
- 3) Output the circuit  $D'_x$  that takes inputs to each  $D_i$  and does the following. Output  $\perp$  if  $D_i$  outputs  $\perp$  for some  $i \in [q]$ . Otherwise simulate running  $C(x)$  using the output of  $D_i$  to answer the  $i$ 'th query and output the simulated answer.

□

## V. PROOF OF NP-HARDNESS OF MCSP

### A. Reduction

We present the formal description of our reduction from Gap  $\tau$ -Uniform Vertex Cover to MCSP below:

#### • Parameterized by:

- a formula size value  $|\psi| \leq n$ ,
- an integer “security parameter”  $1 \leq \lambda \leq 2^n$ ,
- the hypergraph uniformity  $\tau \leq \log n$ ,

- a function  $f: [n] \times [\lambda] \rightarrow \{0, 1\}$  that is  $(Q, q)$ -somewhat NP  $\cap$  coNP-computable by a  $Q$ -sized circuit  $\text{Gen}_f$  and  $q \leq n$ ,
- an  $\epsilon$ -secure non-interactive witness indistinguishable proof system NIWI with its security parameter set to  $|\psi|$ , and
- a size threshold  $S$ .

#### • Notation:

- Let  $H: [n] \times [\lambda'] \rightarrow \{0, 1\}$  be the function given by

$$H(v, i) = \text{Amp}\left(f(v, \cdot), \frac{1}{5\tau}\right)(i),$$

where  $\text{Amp}(\cdot)$  is the hardness amplification procedure of Lemma III.5. Note that  $\lambda' \leq \lambda^{O(\tau^2)}$ . By Proposition IV.3, note that  $H$  is  $(Q', q')$ -somewhat NP  $\cap$  coNP-computable by some  $Q'$ -sized circuit Gen where  $Q' = O(Q\tau^2)$  and  $q' = O(q\tau^2)$ . Moreover, Gen can be produced in uniform  $\text{poly}(Q, \tau)$ -time given  $\text{Gen}_f$  and  $\tau$ .

- For a  $\tau$ -uniform hypergraph on  $n$  vertices with edge set  $E$  and a formula  $\psi$ , define the function

$$f_{\psi, E}\left(x = (e \in [n]^\tau, \{r_v \in [\lambda']\}_{v \in e}, \{c_v \in \{0, 1\}\}_{v \in e}), \pi \in \{0, 1\}^{\text{poly}(|\psi|, q\tau)}\right) =$$

$$\begin{cases} c_w \oplus H(w, r_w) & \text{if NIWI.Verify}(\psi \vee \varphi_x, \pi) = 1 \\ & \text{and } e \in E \text{ where } \varphi_x \text{ is} \\ & \text{defined below and } w \text{ is the} \\ & \text{first entry of } e \\ 0 & \text{otherwise.} \end{cases}$$

To define  $\varphi_x$ , let  $D_v = \text{Gen}(v, r_v)$  for all  $v \in e$ . We define  $\varphi_x$  as the  $\text{poly}(q, \tau)$ -sized formula that is satisfiable if and only if there exist  $b \in \{0, 1\}$  and  $\{w_v\}_{v \in e}$  such that for all  $v \in e$  we have  $D_v(w_v) = b \oplus c_v$ .

#### • Reduction:

Given a  $\tau$ -uniform hypergraph on  $n$  vertices with edge set  $E$ :

- 1) Accept if and only if for every unsatisfiable formula  $\psi$  of size at most  $|\psi|$  (we iterate through such formulas by brute force)  $f_{\psi, E}$  has complexity at most  $S$ .

### B. Hardness

By construction, we have the following bound on the running time of the reduction (including the  $n^{O(\log \tau)}$  time to run the reduction from SAT to Gap  $\tau$ -Uniform Vertex Cover).

**Proposition V.1.** *The reduction can be implemented in uniform time  $(n\lambda)^{O(\tau^2)}2^{\text{poly}(|\psi|q\tau)}$ , given  $\text{Gen}_f$ .*

In Section V-C, we prove the following upper bound of the complexity of  $f_{\psi, E}$  when  $E$  is a YES instance.

**Lemma V.2** (Upper Bound on YES Instances). *If  $\psi$  is unsatisfiable and  $E$  is the edge set of a YES instance of Gap  $\tau$ -Uniform Vertex Cover, then*

$$\text{CC}(f_{\psi,E}) \leq O\left(\frac{\gamma \cdot n \cdot \lambda}{\tau \log \lambda} + \tau^3 Q\right) + \text{poly}(n^\tau).$$

In Section V-D, we prove the following lower bound of the complexity of  $f_{\psi,E}$  when  $E$  is a NO instance.

**Lemma V.3** (Lower Bound on NO Instances). *There is a universal constant  $c$  such that the following holds. Assume that*

- UNSAT on formulas of size  $|\psi|$  requires non-deterministic circuits of size greater than  $(n\lambda't)^c$ ,
- $\epsilon(|\psi|) < \frac{1}{(n\lambda't)^2}$ , and
- $n$  is sufficiently large.

*Then, for every edge set  $E$  of a NO instance of Gap  $\tau$ -Uniform Vertex Cover, there exists an unsatisfiable  $|\psi|$ -size formula  $\psi$  such that*

$$\text{K}^t(f_{\psi,E}) \geq \text{K}^{\text{poly}(t \log(n\lambda'))}(f) - 4\sqrt{\frac{2\gamma}{\tau}} n\lambda - \text{poly}(n\tau).$$

The proof of the main results will be presented in Section V-E.

### C. YES instance analysis

We now prove Lemma V.2.

**Lemma V.2** (Upper Bound on YES Instances). *If  $\psi$  is unsatisfiable and  $E$  is the edge set of a YES instance of Gap  $\tau$ -Uniform Vertex Cover, then*

$$\text{CC}(f_{\psi,E}) \leq O\left(\frac{\gamma \cdot n \cdot \lambda}{\tau \log \lambda} + \tau^3 Q\right) + \text{poly}(n^\tau).$$

*Proof.* Fix an optimal vertex cover  $V \subseteq [n]$  of size  $\text{OPT} \leq \gamma \frac{n}{\tau}$ . We can compute  $f_{\psi,E}$  as follows.

On input  $x = (e, \{r_v\}_{v \in e}, \{c_v\}_{v \in e})$  and  $\pi$ :

- 1) Check if  $e \in E$ .
- 2) Compute  $D_v = \text{Gen}(v, r_v)$  for all  $v \in e$ .
- 3) Construct  $\varphi_x$ .
- 4) Reject if  $\text{NIWI.Verify}(\psi \vee \varphi_x, \pi) \neq 1$ .
- 5) Fix the lexicographically first vertex  $v' \in e$  in the vertex cover  $V$ .
- 6) Compute  $H(v', r_{v'})$ . (We will do this step carefully to make sure it is efficient.)
- 7) Output  $c_{v'} \oplus H(v', r_{v'})$ .

It is easy to see that this procedure does indeed compute  $f_{\psi,E}$  as long as we have that

$$c_{v'} \oplus H(v', r_{v'}) = c_w \oplus H(w, r_w)$$

where  $w$  (as in the reduction) is the first entry of  $e$ . This is true because of the following argument. Since  $\text{NIWI.Verify}(\psi \vee \varphi_x, \pi) = 1$ , we get (by the perfect soundness of the NIWI) that “either  $\psi$  or  $\varphi_x$  is satisfiable.” Then because  $\psi$  is unsatisfiable

(by the assumption in this lemma), we get that  $\varphi_x$  is satisfiable. Then by construction of  $\varphi_x$ , we get that

$$c_{v'} \oplus H(v', r_{v'}) = c_v \oplus H(v, r_v)$$

for all  $v \in e$ , in particular including  $w$ . This completes our argument that this procedure computes  $f_{\psi,E}$ .

Thus, it just remains to bound the complexity of each step in the procedure.

- 1) This step can be done by a DNF of size  $\text{poly}(n^\tau)$ .
- 2) This step can be done by a circuit of size  $O(\tau Q') = O(\tau^3 Q)$  essentially by construction.
- 3) By construction of  $\varphi_x$ , this can be done in size  $\text{poly}(q, \tau)$ .
- 4) By the efficiency of the NIWI, this can be done in size  $\text{poly}(|\pi| + |\psi| + \tau q) = \text{poly}(|\psi|, \tau q) = \text{poly}(n)$ .
- 5) This step can be done in size  $\text{poly}(n)$ .
- 6) Let  $H|_V: V \times [\lambda'] \rightarrow \{0, 1\}$  be the restriction of  $H$  to its first input being in  $V$ . The circuit complexity of this step in the procedure is exactly the complexity of computing  $H|_V$ .

To compute  $H|_V$ , it will be helpful to compute another function. Set  $k = \lfloor \log^3 n \rfloor$ . Let  $f^k|_V: (V \times [\lambda])^k \rightarrow \{0, 1\}^k$  be given by

$$f^k|_V(v_1, i_1, \dots, v_k, i_k) = f(v_1, i_1) \cdots f(v_k, i_k).$$

Since  $\lambda \geq n$ , for sufficiently large  $n$  we have that

$$\begin{aligned} k = \lfloor \log^3 n \rfloor &\leq n^{1/\log^2 \log(2n)} \\ &\leq (\text{OPT} \cdot \lambda)^{1/\log^2 \log(2\text{OPT} \cdot \lambda)}. \end{aligned}$$

Hence, we can apply Uhlig’s theorem (Theorem III.7) to say there is a circuit for  $f^k|_V$  of size at most

$$4 \frac{\text{OPT} \cdot \lambda}{\log(\text{OPT} \cdot \lambda)}.$$

Now recall  $H(v, \cdot) = \text{Amp}(f(v, \cdot), \frac{1}{5\tau})$ . By the properties of Amp this means that there a circuit of size  $\text{poly}(n\tau)$  computes  $H(v, \cdot)$  given  $O(\tau^2) = O(\log^2 n)$  non-adaptive oracle queries to  $f(v, \cdot)$ . Thus, for sufficiently large  $n$ , we can answer the non-adaptive  $f(v, \cdot)$  queries using our circuit for  $f^k|_V$  whenever  $v \in V$ . Thus, there is a circuit  $H_V$  of size at most

$$\text{poly}(n\tau) + \frac{\text{OPT} \cdot \lambda}{\log(\text{OPT} \cdot \lambda)}.$$

- 7) We can do this with  $O(1)$  gates.

Thus, in total we obtain the bound

$$\text{CC}(f_{\psi,E}) \leq \text{poly}(n^\tau) + O\left(\tau^3 Q + \frac{\text{OPT} \cdot \lambda}{\log(\text{OPT} \cdot \lambda)}\right),$$

which proves the lemma plugging in  $\text{OPT} \leq \gamma \frac{n}{\tau}$ .  $\square$

#### D. NO instance analysis

We begin by introducing some notation.

- For  $\varphi_x$  as in the code of the reduction, let  $w_{\varphi_x}$  be the lexicographically first witness to  $\varphi_x$  (which will always exist in the cases we refer to it).
- For  $0 \leq p \leq 1$ , we say a Turing machine  $A$  is  $p$ -successful on  $f_{\psi,E}$  if for all  $e \in E$  we have that

$$\Pr_{\substack{b \leftarrow \{0,1\}, \\ r_v \leftarrow [\lambda'] \text{ for all } v \in e \\ c_v = b \oplus H(v, r_v) \text{ for all } v \in e \\ \pi \leftarrow \text{NIWI.Prove}(\varphi_x \vee \psi, w_{\varphi_x})}} [A(e, \{r_v\}_{v \in e}, \{c_v\}_{v \in e}, \pi) = b]$$

is at least  $p$  where  $x$  is defined as  $(e, \{r_v \in [\lambda']\}_{v \in e}, \{c_v \in \{0,1\}\}_{v \in e})$ .

- We let  $|A|$  be short hand for  $|d|$  where our universal Turing machine  $U(d, i)$  computes  $A(i)$ .

By definition of  $f_{\psi,E}$ , we get the following.

**Proposition V.4.** *For all  $0 \leq p \leq 1$ , every  $A$  that computes  $f_{\psi,E}$  is also  $p$ -successful on  $f_{\psi,E}$ .*

Now we prove a lower bound on the size of any successful Turing machine if one chooses a satisfiable  $\psi$ .

**Lemma V.5.** *Let  $\psi$  be a satisfiable formula. Let  $E$  be the edge set of a NO instance of Gap  $\tau$ -Uniform Vertex Cover. Assume  $\epsilon(|\psi|) < \frac{1}{(n\lambda t)^2}$ . If  $A$  is .9-successful on  $f_{\psi,E}$  and runs in  $t$ -time, then*

$$|A| \geq K^{\text{poly}(t \log(n\lambda'))}(f) - 4\sqrt{\frac{2\gamma}{\tau}}n\lambda - \text{poly}(n\tau).$$

*Proof.* For contradiction suppose the bound does not hold. Fix an arbitrary edge  $e \in E$ . We now use a hybrid argument. Let  $w_{\psi}$  be the lexicographically first witness to  $\psi$ . First, for sufficiently large  $n$  we have

$$\epsilon(|\psi|) \leq \frac{1}{(n\lambda t)^2} \leq \frac{1}{t^2 \cdot (K^{t+\text{poly}(n\lambda')}(f))^2} \leq \frac{1}{t^2 \cdot |A|^2},$$

so the security of the NIWI says we can replace  $\pi$  to come from the witness for  $\psi$  and get

$$\Pr_{\substack{b \leftarrow \{0,1\}, \\ r_v \leftarrow [\lambda'] \text{ for all } v \in e \\ c_v = b \oplus H(v, r_v) \text{ for all } v \in e \\ \pi \leftarrow \text{NIWI.Prove}(\varphi_x \vee \psi, w_{\psi})}} [A(e, \{r_v\}_{v \in e}, \{c_v\}_{v \in e}, \pi) = b] \geq .9 - \epsilon(|\psi|) \geq .89,$$

for sufficiently large  $n$ .

On the other hand if we replace all the  $c_v$  with uniformly random bits, then the chance of outputting  $b$  must be one half (because  $b$  is independent of all the inputs to  $A$ ). In other words,

$$\Pr_{\substack{b \leftarrow \{0,1\}, \\ r_v \leftarrow [\lambda'] \text{ for all } v \in e \\ c_v \leftarrow \{0,1\} \text{ for all } v \in e \\ \pi \leftarrow \text{NIWI.Prove}(\varphi_x \vee \psi, w_{\psi})}} [A(e, \{r_v\}_{v \in e}, \{c_v\}_{v \in e}, \pi) = b] = 1/2,$$

Thus, by a hybrid argument and the distinguisher-to-predictor lemma (Lemma III.3), for some  $v \in e$  we must have that

$$K_{1/2+1/(3\tau)}^{\text{poly}(\log(n\lambda'))}(H_v) \leq O(\tau \log \lambda') + |\pi| \leq \text{poly}(n\tau),$$

where  $H_v$  denotes the truth table of  $H(v, \cdot)$ . Hence by the construction of  $H$  and the properties of Amp (Lemma III.5), we get that

$$K_{1-1/\tau}^{\text{poly}(\log(n\lambda'))}(f_v) \leq \text{poly}(n\tau), \quad (3)$$

where  $f_v$  denote the truth table of  $f(v, \cdot)$ , and we used that  $\tau \leq \log n$  and  $\lambda \leq 2^n$

Now, recall we picked an arbitrary  $e \in E$ . Thus, for every  $e \in E$ , there exists a vertex  $v \in e$  such that (3) holds. Since the smallest vertex cover is of size at least  $n(1 - \frac{\gamma}{\tau})$ , we then get that (3) holds for all but a  $\frac{\gamma}{\tau}$  fraction of vertices. This immediately implies that

$$K_{1-2\gamma/\tau}^{\text{poly}(\log(n\lambda'))}(f) \leq \text{poly}(n\tau).$$

By hardcoding in corrections to the small number of errors (Theorem III.4) we get (noting  $\tau \leq \log n$ ) that

$$K^{\text{poly}(\log(n\lambda'))}(f) \leq 4\sqrt{\frac{2\gamma}{\tau}}n\lambda + \text{poly}(n\tau).$$

Hence, using that  $A$  runs in time  $t$ , we have that

$$K^{\text{poly}(t \log(n\lambda'))}(f) \leq |A| + 4\sqrt{\frac{2\gamma}{\tau}}n\lambda + \text{poly}(n\tau),$$

which proves the lemma.  $\square$

We can then exploit the non-deterministic hardness of solving UNSAT to prove the following lemma.

**Lemma V.3** (Lower Bound on NO Instances). *There is a universal constant  $c$  such that the following holds. Assume that*

- UNSAT on formulas of size  $|\psi|$  requires non-deterministic circuits of size greater than  $(n\lambda't)^c$ ,
- $\epsilon(|\psi|) < \frac{1}{(n\lambda t)^2}$ , and
- $n$  is sufficiently large.

*Then, for every edge set  $E$  of a NO instance of Gap  $\tau$ -Uniform Vertex Cover, there exists an unsatisfiable  $|\psi|$ -size formula  $\psi$  such that*

$$K^t(f_{\psi,E}) \geq K^{\text{poly}(t \log(n\lambda'))}(f) - 4\sqrt{\frac{2\gamma}{\tau}}n\lambda - \text{poly}(n\tau).$$

*Proof.* Set

$$S = K^{\text{poly}(t \log(n\lambda'))}(f) - 4\sqrt{\frac{2\gamma}{\tau}}n\lambda - \text{poly}(n\tau).$$

so that Lemma V.5 holds. Note that we always have that  $S \leq O(n\lambda)$ .

We prove the contrapositive. Assume there exists an edge set  $E$  that is a NO instance of  $n$ -vertex Gap  $\tau$ -Uniform Vertex Cover such that for every  $|\psi|$ -size unsatisfiable  $\psi$ ,

$$K^t(f_{\psi,E}) < S. \quad (4)$$

On the other hand, Lemma V.5 says that for every satisfiable  $\psi$  every  $t$ -time machine  $A$  that is .9-successful on  $f_{\psi,E}$  has

$$|A| \geq S. \quad (5)$$

Hence, the following yields a randomized non-deterministic circuit for checking whether  $\psi$  is unsatisfiable:

Given  $\psi$ :

- 1) Non-deterministically guess an  $S$ -size Turing machine  $A$
- 2) Using randomness, empirically estimate to additive error .01 (allowing failure probability at most 1/3) the probability

$$\Pr[A(e, \{r_v\}_{v \in e}, \{c_v\}_{v \in e}, \pi) = b \text{ in } t\text{-time}]$$

where  $b \leftarrow \{0, 1\}$ ,  $r_v \leftarrow [\lambda']$  for all  $v \in e$ ,  $c_v = b \oplus H(v, r_v)$  for all  $v \in e$ , and  $\pi \leftarrow \text{NIWI.Prove}(\varphi_x \vee \psi, w_{\varphi_x})$ . Reject if this probability is less than .95. In order to do this empirical estimate efficiently, we hardcode in the truth table of  $H$  and also hardcode in the lexicographically first witnesses to each circuit output by  $\text{Gen}(v, r_v)$  for all  $v \in [n]$  and  $r_v \in [\lambda']$ . From the latter information, we can compute  $w_{\varphi_x}$  easily for any  $\varphi_x$ .

- 3) Accept (i.e., output that  $\psi$  is unsatisfiable.)

From (4) and the fact (Proposition V.4) that any  $A$  computing  $f_{\psi, E}$  is 1-successful, we get that this circuit accepts every unsatisfiable  $\psi$ . From (5), the circuit rejects all satisfiable  $\psi$ .

Thus, we just need to argue for efficiency. Step 1 requires size  $O(S) = \text{poly}(n\lambda)$ . Step 2 can be done in size  $\text{poly}(qn\lambda't) = \text{poly}(n\lambda't)$ . Thus, this circuit has size at most  $\text{poly}(n\lambda't)$ .

Finally, using Adleman's trick we can remove the randomness from this non-deterministic circuit to obtain a (standard) non-deterministic circuit of size at most  $\text{poly}(n\lambda't|\psi|) = \text{poly}(n\lambda't)$ .  $\square$

#### E. Proof of the Main Results

We need one additional lemma, which essentially states that, given a function with nearly maximum circuit complexity, one can construct a string with high time-bounded Kolmogorov complexity.

**Lemma V.6.** *For every oracle  $A$ , the following are equivalent.*

- 1)  $P^A/\text{poly} \not\subseteq \text{i.o.E}/\delta 2^n$  for some constant  $\delta > 0$ .
- 2)  $P^A/\text{poly} \not\subseteq \text{i.o.SIZE}(\delta 2^n/n)$  for some constant  $\delta > 0$ .

*Proof.* The high-level idea is this. (1) implies (2) because Turing machines can simulate circuits. (2) implies (1) because of the following argument. If one breaks the truth table of a high circuit complexity function  $f$  into small pieces, then at least one small piece must have high time-bounded Kolmogorov complexity (relative to its length). If this were not the case, then one could build a small circuit for  $f$ .

Now we give the formal proof. Since any  $n$ -input circuit of size  $s$  can be encoded as a binary string of  $O(s \log(s+n))$ , it holds that  $\text{SIZE}(\delta 2^n/n) \subseteq \text{E}/\alpha(n)$ , where

$$\alpha(n) = O\left(\frac{\delta 2^n}{n} \cdot \log \frac{\delta 2^n}{n}\right) = O(\delta 2^n).$$

Thus, the first item implies the second item.

To see the converse, let  $f = \{f_n : \{0, 1\}^n \rightarrow \{0, 1\}\}$  be a function in  $P^A/\text{poly} \setminus \text{i.o.SIZE}(\delta 2^n/n)$ . This assumption is

equivalent to saying that for all large  $n \in \mathbb{N}$ , the function  $f_n$  can be computed by some  $A$ -oracle circuit of size  $\text{poly}(n)$  but cannot be computed by any circuit of size  $\delta 2^n/n$ . Fix an arbitrary constant  $c$ . Let  $\epsilon, \delta' > 0$  be sufficiently small constants chosen later. (We will construct some function outside  $\text{DTIME}(2^{cn})/\delta' 2^n$ .) Fix sufficiently large  $n \in \mathbb{N}$ . For every  $z \in \{0, 1\}^{(1-\epsilon)n}$ , define  $f_{n,z} : \{0, 1\}^{\epsilon n} \rightarrow \{0, 1\}$  as the function such that  $f_{n,z}(x) = f_n(z, x)$ .

We claim that there exists some  $z$  such that  $K^t(f_{n,z}) \geq \delta' 2^{\epsilon n}$  for  $t := 2^{cn} \leq 2^{n/2}$  (we choose  $\epsilon \leq c/2$ ). This is sufficient, as  $f_{n,z}$  can be computed by an  $A$ -oracle circuit of size  $\text{poly}(n)$ . To see the claim, assume for contradiction that for every  $z$ , it holds that  $K^t(f_{n,z}) < \delta' 2^n$ . That is, for every  $z$ , there exists  $d_z \in \{0, 1\}^{\delta' 2^n}$  such that  $U^t(d_z, x) = f_{n,z}(x)$ . For every  $i \in [\delta' 2^{\epsilon n}]$ , let  $\alpha_i$  be the function that takes  $z \in \{0, 1\}^{(1-\epsilon)n}$  and outputs the  $i$ 'th bit of  $d_z$ . Then,  $\alpha_i$  can be implemented by a circuit of size

$$(1 + o(1)) \cdot \frac{2^{(1-\epsilon)n}}{\log 2^{(1-\epsilon)n}} = (1 + \epsilon + o(1)) \cdot \frac{2^{(1-\epsilon)n}}{n}.$$

Thus, given  $z$  as input, the advice string  $d_z$  can be computed by a circuit of size

$$\delta' 2^{\epsilon n} \cdot O\left(\frac{2^{(1-\epsilon)n}}{n}\right) = O\left(\frac{\delta' 2^n}{n}\right).$$

Now we construct a small circuit that computes  $f_n$  as follows: Given input  $w \in \{0, 1\}^n$ , let  $z$  and  $x$  be the first  $(1-\epsilon)n$  bits and the last  $\epsilon n$  bits of  $w$ , respectively. Compute  $d_z$  by the circuit of size  $O(\delta' 2^n/n)$ , and simulate and output  $U^t(d_z, x)$ . This algorithm can be implemented by a circuit of size

$$O\left(\delta' \frac{2^n}{n}\right) + \tilde{O}(t) \leq \delta 2^n,$$

where the last inequality holds by choosing a sufficiently small  $\delta' > 0$ . This is a contradiction.  $\square$

Now we are ready to prove our main theorem.

**Theorem V.7** (a restatement of Theorem I.1). *Assume all of the following:*

- 1) *subexponentially-secure NIWIs exist,*
- 2)  $\text{coNP} \not\subseteq \text{i.o.NSIZE}(2^{n^\epsilon})$  for some constant  $\epsilon > 0$ , and
- 3)  $P^{\text{NP}}/\text{poly} \not\subseteq \text{i.o.SIZE}(\delta 2^n/n)$  for some constant  $\delta > 0$ .

*Then, for any constant  $g \geq 1$  and any polynomial  $t$ , it is NP-hard under deterministic quasipolynomial-time non-adaptive reductions to solve the following promise problem  $\Pi$ :*

*Input: the truth table of  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  and a size parameter  $S \leq 2^n/n$ .*

*Yes:  $\text{CC}(f) \leq S$  and  $K^{O(S \log S)}(f) \leq S \log S$ .*

*No:  $\text{CC}(f) > S \cdot g$  and  $K^{t(S)}(f) > S \log S \cdot g$ .*

*In particular, both MCSP and  $K^t$  are NP-hard to approximate to within a constant factor.*

*Proof.* By Lemma V.6, the third assumption implies that for any polynomial  $t$ , there exists a function  $f = \{f_m : \{0, 1\}^m \rightarrow \{0, 1\}\}_{m \in \mathbb{N}}$  such that  $f_m$  is computable by a SAT-oracle circuit  $C^{\text{SAT}}$  of size  $\text{poly}(m)$  and  $K^{t(2^m)}(f_m) \geq$

$\delta 2^m$ . It follows from Proposition IV.2 that  $f_m$  is  $(Q, q)$ -somewhat NP  $\cap$  coNP-computable by some  $Q$ -size circuit  $\text{Gen}_{\text{SAT}}$ , where  $Q = 2^m/m^5$  and  $q = \text{poly}(m)$ . We may regard  $f_m: [n] \times [\lambda] \rightarrow \{0, 1\}$ , where  $2^m = n \cdot \lambda$  for parameters  $n$  and  $\lambda$  chosen later.<sup>8</sup>

Our final reduction builds on the reduction presented in Section V-A. Let  $([n], E)$  be an instance of Gap  $\tau$ -Uniform Vertex Cover. We enumerate all the SAT-oracle circuits  $C^{\text{SAT}}$  of size  $\text{poly}(m)$ , and output “Yes” if and only if for any SAT-oracle  $m$ -input circuit  $C^{\text{SAT}}$  of size  $\text{poly}(m)$ , all the queries  $(f_{\psi, E}, S)$  of the reduction of Section V-A given the generator  $\text{Gen}_{\text{SAT}}$  are answered with “Yes” by the oracle that solves  $\Pi$ , where  $f_{\psi, E}$  is the function defined in Section V-A. Note that this reduction can be implemented in time  $2^{\text{poly}(m)}$ , where  $m \leq \text{poly}(\log n)$ . We will choose the parameters during the analysis of this reduction.

To see the completeness, let  $E$  be the edge set of a YES instance of Gap  $\tau$ -Uniform Vertex Cover. By Lemma V.2,

$$\text{CC}(f_{\psi, E}) \leq O\left(\frac{\gamma \cdot n \cdot \lambda}{\tau \log \lambda} + \tau^3 Q\right) + \text{poly}(n^\tau)$$

Since  $\tau \leq \log n \leq m$ , we have  $\tau^3 Q \leq 2^m/m^2 \leq n \cdot \lambda/(\log \lambda)^2$ . Let  $c$  be a sufficiently large constant. We set  $\lambda = n^{c\tau}$ . Then we have  $\text{poly}(n^\tau) \leq \lambda/(\log \lambda)^2$  by choosing a sufficiently large  $c$ . Since  $\tau \leq \log \lambda$ , it follows that

$$\begin{aligned} \text{CC}(f_{\psi, E}) &\leq O\left(\frac{\gamma \cdot n \cdot \lambda}{\tau \log \lambda} + \tau^3 Q\right) + \text{poly}(n^\tau) \\ &\leq O\left(\frac{n\lambda}{\tau \log \lambda}\right) = S, \end{aligned}$$

where we define  $S$  so that this equality holds. This further implies that

$$\text{K}^{O(n\lambda)}(f_{\psi, E}) \leq O\left(\frac{n\lambda}{\tau \log n} \log\left(\frac{n\lambda}{\tau \log n}\right)\right) \leq O\left(\frac{n\lambda}{\tau}\right).$$

Now we argue for soundness. Let  $E$  be the edge set of a NO instance of Gap  $\tau$ -Uniform Vertex Cover. Let  $t = t(2^m)$ . Let  $C^{\text{SAT}}$  be the SAT-oracle circuit of size  $\text{poly}(m)$  that computes  $f_m$  such that  $\text{K}^{\text{poly}(t)}(f_m) \geq \delta 2^m = \delta n\lambda$ , where  $\text{poly}$  is some polynomial chosen later. It suffices to claim that the reduction of Section V-A given  $\text{Gen}_{\text{SAT}}$  queries the function  $f_{\psi, E}$  that is answered with “No” for some unsatisfiable formula  $\psi$ . Let  $c_0$  be the universal constant of Lemma V.3. To satisfy the first assumption of Lemma V.3, we choose  $|\psi|$  so that  $2^{|\psi|^\epsilon} \geq (n\lambda^{O(\tau^2)}t)^{c_0}$ . For all sufficiently large  $n$ , Lemma V.3 implies that

$$\begin{aligned} \text{K}^t(f_{\psi, E}) &\geq \text{K}^{\text{poly}(t \log(n\lambda'))}(f) - 4\sqrt{\frac{2\gamma}{\tau}}n\lambda - \text{poly}(n\tau) \\ &\geq \delta n\lambda - \frac{\delta}{2} \cdot n\lambda \\ &= \frac{\delta}{2} \cdot n\lambda, \end{aligned}$$

<sup>8</sup>Strictly speaking, we choose  $2^{m-1} \leq n\lambda \leq 2^m$ , and the truth table of length  $n\lambda$  is constructed from the first  $n\lambda$  bits of the truth table of  $f_m$ .

where the second inequality holds by choosing a sufficiently large  $\tau \geq O(\delta^{-2})$ . This further implies that

$$\text{CC}(f_{\psi, E}) \geq \Omega\left(\delta \cdot \frac{n\lambda}{\log(n\lambda)}\right) \geq \Omega(\tau\delta \cdot S) > g \cdot S,$$

where the last inequality holds by choosing a sufficiently large constant  $\tau$ .  $\square$

**Remark V.8.** Choosing  $\tau := (\log n)^\epsilon$  for a small constant  $\epsilon > 0$ , the inapproximability factor of Theorem V.7 can actually be improved to  $(\log |f|)^{\Omega(1)}$ . Moreover, the third assumption can be weakened to  $\text{P}^{\text{NP}}/\text{poly} \not\subseteq \text{i.o.SIZE}(2^n/n^{1+\Omega(1)})$ .

We may remove the third assumption using a case analysis.

**Corollary V.9** (The formal version of Corollary I.3). *Assume that*

- 1) *subexponentially-secure NIWIs for SAT exist, and*
- 2) *coNP requires subexponential-size non-deterministic circuits almost everywhere.*

*Then, for every function  $f = \{f_n: \{0, 1\}^n \rightarrow \{0, 1\}\}_{n \in \mathbb{N}}$  in NP and every polynomial  $p$ , there exist infinitely many  $n \in \mathbb{N}$  and an  $n$ -input subexponential-size non-adaptive oracle circuit  $C$  such that*

$$\Pr_{x \leftarrow \{0, 1\}^n} [f_n(x) = C^{\text{MCSP}}(x)] \geq 1 - \frac{1}{p(n)}.$$

*Proof.* A high-level idea is to consider two cases: whether  $\text{P}^{\text{NP}}/\text{poly} \not\subseteq \text{i.o.SIZE}(2^n/2n)$  or not. If it is, we conclude from Theorem I.1 that MCSP is NP-hard (in the worst case). Otherwise, we will show that an MCSP oracle can be used to break the security of a pseudorandom generator based on a candidate hard function  $f \in \text{NP}$ , from which we obtain an average-case MCSP-oracle circuit to compute  $f$ . Details follow.

Let  $\epsilon > 0$  be an arbitrary small constant. Fix  $n \in \mathbb{N}$ . Define  $m = m(n) := n^\epsilon$ . As in Lemma III.5, we define  $g := f^{\oplus k}$ , i.e., the  $k$ -wise XOR function of  $f$ , where  $k = O(m(n)/p(n))$ . Let  $\text{NW}^g: \{0, 1\}^{\text{poly}(n)} \rightarrow \{0, 1\}^{2^m}$  be the Nisan–Wigderson generator [91] instantiated with a candidate hard function  $g$  and the combinatorial design  $\{S_i\}_{i \in [2^m]}$  based on polynomials. For each seed  $z$ , we regard  $\text{NW}^g(z)$  as the truth table of a function  $h_z: \{0, 1\}^m \rightarrow \{0, 1\}$  on  $m$ -bit inputs. As observed in [12],  $h_z$  can be computed by a  $g$ -oracle circuit of size  $\text{poly}(n)$  because the  $i$ 'th set  $S_i$  in the combinatorial design can be computed by a polynomial-size circuit given  $i$  as input. In particular,  $h_z$  can be computed by a polynomial-size SAT-oracle circuit.

Consider the case where for all sufficiently large  $n \in \mathbb{N}$ , it holds that  $\text{CC}(\text{NW}^g(z)) \geq \frac{2^m}{2m}$  for some  $z$ . In this case, the function  $h_z$  defined by its truth table  $\text{NW}^g(z)$  has circuit complexity at least  $2^m/2m$ , and moreover  $h_z$  can be computed by a polynomial-size SAT-oracle circuit. Thus, we obtain  $\text{P}^{\text{NP}}/\text{poly} \not\subseteq \text{i.o.SIZE}(2^n/2n)$ . By Theorem I.1, MCSP is NP-hard under quasi-polynomial-time non-adaptive reductions, and we are done.

Now consider the other case: Assume that for infinitely many  $n \in \mathbb{N}$ , it holds that  $\text{CC}(\text{NW}^g(z)) < 2^m/2m$  for all  $z$ . Since  $\text{CC}(w) \geq 2^m/2m$  with high probability over a uniformly random  $w \in \{0,1\}^{2^m}$ , this implies that MCSP is a statistical test that distinguishes the output distribution of  $\text{NW}^g(\cdot)$  from the uniform distribution. By the reconstruction property of the Nisan–Wigderson pseudorandom generator, there exists an oracle circuit  $C_0$  of size  $2^{O(m)}$  such that

$$\Pr_{y \leftarrow \{0,1\}^{\text{poly}(n)}} [g(y) = C_0^{\text{MCSP}}(y)] \geq \frac{1}{2} + 2^{-m}.$$

By the property of hardness amplification (Yao’s XOR lemma; see the proof of Lemma III.5), we obtain an oracle circuit  $C$  of size  $2^{O(m)} = 2^{O(n^\epsilon)}$  such that

$$\Pr_{x \leftarrow \{0,1\}^n} [f(x) = C^{\text{MCSP}}(x)] \geq 1 - \frac{1}{p(n)},$$

as desired.  $\square$

## APPENDIX

In this section, we justify one of our assumptions under the random oracle model.

**Theorem A.1.** *Let  $\mathcal{O}$  be a random oracle. Then,*

$$\Pr_{\mathcal{O}} [\text{NP}^{\mathcal{O}} \text{ is not in } \text{DTIME}[2^{n^2}]^{\mathcal{O}} / (\delta 2^n) \text{ infinitely often}] = 1$$

for some constant<sup>9</sup>  $\delta > 0$ .

**Lemma A.2** (Borel–Cantelli). *Let  $E_n$  be a sequence of events. If  $\sum_n \Pr[E_n]$  converges, then*

$$\Pr[E_n \text{ occurs for at most finitely many } n] = 1$$

*Proof of Theorem A.1.* Without loss of generality, we can view the oracle  $\mathcal{O}$  as taking as input pairs  $(x, y)$  where  $|y| = |x|^3$  and outputting a string of length  $|x|^3 + 1$ . Our hard language will be the following:

$$L^{\mathcal{O}} = \{x : \mathcal{O}(x, y) \text{ outputs all zeroes for some } y\}.$$

It is easy to see that  $L^{\mathcal{O}} \in \text{NP}^{\mathcal{O}}$ .

It remains to show the lower bound on  $L^{\mathcal{O}}$ . We make the following claim.

**Claim A.3.** *There is a constant  $n_0$  such that the following holds. Let  $A$  be an oracle circuit that makes  $q = 2^{2n^2}$  queries. If  $n \geq n_0$ , then*

$$\Pr_{\mathcal{O}} [A^{\mathcal{O}} \text{ computes } L^{\mathcal{O}} \text{ on all inputs of length } n] \leq (.9)^{2^n}.$$

Assuming Claim A.3 holds, we prove the desired lower bound. Let  $A_1, A_2, A_3, \dots$  be an enumeration of all (uniform) algorithms that run in time also take an advice string as input. Let  $E_n$  be the event that there exists an  $i \in [n]$  and an advice string  $\alpha \in \{0,1\}^{\delta 2^n}$  such that all the following holds:

- $A^{\mathcal{O}}(x, \alpha) = L^{\mathcal{O}}(x)$  for all  $x \in \{0,1\}^n$ , and
- $A^{\mathcal{O}}(x, \alpha)$  makes at most  $2^{2n^2}$  oracle queries for all  $x \in \{0,1\}^n$ .

<sup>9</sup>We have not attempted to calculate or optimize for this value.

By Claim A.3 and a union bound, we have that

$$\Pr_{\mathcal{O}} [E_n] \leq n 2^{\delta 2^n} (.9)^{2^n} \leq O\left(\frac{1}{n^2}\right),$$

by setting  $\delta$  to be a sufficiently small constant. Since  $\sum_n \Pr[E_n] < \infty$ , the Borel–Cantelli lemma (Lemma A.2) implies that

$$\Pr_{\mathcal{O}} [\text{infinitely many } E_n \text{ occurs}] = 0.$$

By the definition of  $E_n$ , this implies that

$$\Pr_{\mathcal{O}} [L^{\mathcal{O}} \text{ is in } \text{DTIME}[2^{n^2}]^{\mathcal{O}} / (\delta 2^n) \text{ infinitely often}] = 0.$$

Thus, it just remains to prove Claim A.3. To do so, we introduce some notation. Let  $\tilde{\mathcal{O}}$  denote a partial function with the same input/output behavior as  $\mathcal{O}$  (the difference is that  $\tilde{\mathcal{O}}$  can output  $\star$  for “undefined” values). We say that  $\mathcal{O}$  *agrees* with  $\tilde{\mathcal{O}}$  if they agree on all values where they are defined. We say that  $\tilde{\mathcal{O}}$  is *good* for  $x$  if both of the following hold:

- $\tilde{\mathcal{O}}(x, y)$  is not all zeroes (i.e.,  $\tilde{\mathcal{O}}(x, y) \notin \{0\}^*$ ) for any  $y \in \{0,1\}^{|x|^3}$ , and
- $\tilde{\mathcal{O}}(x, y) \neq \star$  for at most  $q 2^n$  many  $y \in \{0,1\}^{|x|^3}$ .

We make an intermediary claim.

**Claim A.4.** *Let  $b \in \{0,1\}$ . If  $\tilde{\mathcal{O}}$  is good for  $x \in \{0,1\}^n$  and  $n \geq 10$  and  $q \leq 2^{2n^2}$ , then*

$$\Pr_{\mathcal{O}} [L^{\mathcal{O}}(x) = b \mid \mathcal{O} \text{ agrees with } \tilde{\mathcal{O}}] \leq 3/4.$$

*Proof.* Since  $\tilde{\mathcal{O}}$  is good for  $x$ , a union bound implies

$$\Pr_{\mathcal{O}} [L^{\mathcal{O}}(x) = 1 \mid \mathcal{O} \text{ agrees with } \tilde{\mathcal{O}}] \leq 2^{-n^3-1} 2^{n^3} = 1/2$$

On the other hand, we also have

$$\begin{aligned} \Pr_{\mathcal{O}} [L^{\mathcal{O}}(x) = 0 \mid \mathcal{O} \text{ agrees with } \tilde{\mathcal{O}}] \\ \leq (1 - 2^{-n^3-1})^{2^{n^3} - q 2^n} \\ \leq e^{-1/2 + 2^{3n^2} - n^3 - 1} \leq 3/4 \end{aligned}$$

when  $n \geq 10$ .  $\square$

Inductively applying Claim A.4 yields the following intermediary claim.

**Claim A.5.** *Let  $f: \{0,1\}^n \rightarrow \{0,1\}$ . Let  $V \subseteq \{0,1\}^n$ . If  $\tilde{\mathcal{O}}$  is good for every element of  $V$  and  $n \geq 10$  and  $q \leq 2^{2n^2}$ , then*

$$\Pr_{\mathcal{O}} [L^{\mathcal{O}}(x) \text{ computes } f(x) \mid \mathcal{O} \text{ agrees with } \tilde{\mathcal{O}}] \leq (3/4)^{|V|}.$$

*Proof.* We use the lazy sampling technique. Imagine slowly revealing  $\mathcal{O}$  as follows. We start the original  $\tilde{\mathcal{O}}$ . Then we reveal  $\mathcal{O}$  on all inputs of the form  $(x, y)$  where  $x$  is the lexicographically first element of  $V$ . By Claim A.4, the probability that  $f(x) = L^{\mathcal{O}}(x)$  is at most  $3/4$ .

Now remove  $x$  from  $V$  and consider the new partial function  $\tilde{\mathcal{O}}$  that now includes the values we have revealed. Observe that we have preserved the invariant that (the updated)  $\tilde{\mathcal{O}}$  is good

for every element of (the updated)  $V$ . This is because we only revealed values that began with the  $x$  we removed, and because the definition of good for  $x'$  only cares about inputs that start with  $x'$ . Repeating this argument until  $V$  is empty yields the bound.  $\square$

Finally, we prove Claim A.3

*Proof of Claim A.3.* We will again use the lazy sampling technique. Imagine we go over every  $x \in \{0, 1\}^n$  and reveal all the values of  $\mathcal{O}$  that  $A^\mathcal{O}(x)$  queries and set  $f: \{0, 1\}^n \rightarrow \{0, 1\}$  to the function that  $A^\mathcal{O}$  computes. In total, this reveals at most  $2^n q$  values of  $\mathcal{O}$ . Let  $\tilde{\mathcal{O}}$  be the corresponding partial function. Since so little values are defined, the only way for  $\tilde{\mathcal{O}}$  not to be good for an  $x \in \{0, 1\}^n$  is if we revealed an  $\mathcal{O}(x, y)$  equal to all zeroes.

The number of  $x$  for which this occurs is at most the number  $B$  of all zero outputs we have revealed. Observe that  $B$  is the sum of  $q2^n \leq 2^{3n^2}$  independent Bernoulli random variables each with expectation  $2^{-n^3-1}$ . So  $B$  has expectation at most  $2^{-n^3+3n^2-1}$ . Hence, by a Chernoff Bound, the probability that  $|B| \geq 2^n/2$  is at most

$$\exp(-2^n/3)$$

for sufficiently large  $n$ .

On the other hand if  $|B| \leq 2^n/2$ , then by Claim A.5, the probability that  $f$  computes  $L^\mathcal{O}$  on inputs of length  $n$  is at most

$$(3/4)^{2^n/2}.$$

Thus, we have that the probability  $f$  computes  $L^\mathcal{O}$  on inputs of length  $n$  is at most

$$\max\{\exp(-2^n/3), (3/4)^{2^n/2}\} \leq (.9)^{2^n}$$

for sufficiently large  $n$ .  $\square$

#### ACKNOWLEDGMENT

We deeply thank Marshall Ball and Yael Tauman Kalai for helpful discussion.

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