

The Minimum Oracle Circuit Size Problem

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Abstract

We consider variants of the Minimum Circuit Size Problem MCSP, where the goal is to minimize the size of *oracle* circuits computing a given function. When the oracle is QBF, the resulting problem $MCSP^{QBF}$ is known to be complete for PSPACE under ZPP reductions. We show that it is *not* complete under logspace reductions, and indeed it is not even hard for TC^0 under uniform AC^0 reductions. We obtain a variety of consequences that follow if oracle versions of MCSP are hard for various complexity classes under different types of reductions. We also prove analogous results for the problem of determining the resource-bounded Kolmogorov complexity of strings, for certain types of Kolmogorov complexity measures.

1 Introduction

The Minimum Circuit Size Problem (MCSP) asks to decide, for a given truth table f of a Boolean function and a parameter s, whether f is computable by a Boolean circuit of size at most s. MCSP is a well-known example of a problem in NP that is widely believed to be intractable, although it is not known to be NP-complete. MCSP is known to be hard for the complexity class SZK under BPP-Turing reductions [AD14], which provides strong evidence for intractability. On the other hand, Kabanets and Cai showed [KC00] that if MCSP is NP-complete under the "usual" sort of polynomial-time reductions, then EXP $\not\subseteq$ P/poly. This can not be interpreted as strong evidence that it may be difficult to provide an NP-completeness proof.

However, there are other ways to define what the "usual" sort of reductions are: e.g., logspace, (uniform) TC^0 , AC^0 , or NC^0 . The overwhelming majority of problems that are known to be NP-complete are, in fact, NP-complete under very restricted kinds of reductions. Can we rule out NP-hardness of MCSP under such reductions?

Very recently, Murray and Williams [MW14] have shown that MCSP is not even P-hard under uniform NC^0 reductions. Can MCSP be NP-hard under slightly stronger reductions, e.g., uniform AC^0 reductions? We suspect that the answer is 'No', but so far we (like Murray and Williams) can

only show that P-hardness of MCSP under uniform AC^0 , TC^0 , or logspace reductions would imply new (likely) complexity lower bounds (in the spirit of [KC00]).

The main focus of the present paper is an *oracle version* of MCSP, denoted MCSP^A for a language A, which asks to decide for a given truth table f and a parameter s if f is computable by an A-oracle circuit of size at most s. We prove a number of implications of hardness of MCSP^A for various choices of the oracle A, and various reductions. In particular, we prove for a PSPACE-complete A that MCSP^A is *not* P-hard under uniform AC⁰ reductions.

The results presented here (along with the results recently reported by Murray and Williams [MW14]) are the first results giving unlikely consequences that would follow if variants of MCSP or the various oracle circuit minimization problems are hard under a natural notion of reducibility. We also show that analogous results hold in the Kolmogorov complexity setting due to the correspondence between circuit size and Kolmogorov complexity, using the minimum-KT complexity problem defined in this paper.

Below we provide a summary of our main results.

1.1 Our results

Most of our results follow the template:

If $MCSP^A$ is hard for a complexity class C under reductions of type \mathcal{R} , then complexity statement S is true.

Table 1 below states our results for different instantiations of A, C, \mathcal{R} , and \mathcal{S} ; note that $\mathcal{S} = \bot$ means that the assumption is false, i.e., MCSP^A is *not* C-hard under \mathcal{R} -reductions. Throughout, we assume that the reader is familiar with complexity classes such as NP, PP, PSPACE, NEXP, etc. We denote the polynomial hierarchy by PH, and its linear-time version (linear-time hierarchy) by LTH. The Counting Hierarchy, denoted CH, is the union of the classes PP, PP^{PP}, etc.

Table 1: Summary of main results: If $MCSP^A$ is	s C -hard under \mathcal{R} , then \mathcal{S} . The last column shows
the theorem where the result is stated in the pap	per.

oracle A	class \mathcal{C}	reductions \mathcal{R}	statement \mathcal{S}	Theorem
PH-hard	TC^0	uniform AC^0	\perp	Theorem 3.9
any	TC^0	uniform AC^0	$LTH \not\subseteq io-SIZE^A[2^{\Omega(n)}]$	Lemma 3.10
any	TC^0	uniform AC^0	$NP^A \not\subseteq SIZE^A[poly]$	Corollary 3.13
any in CH	Р	uniform TC^0	$P \neq PP$	Corollary 3.2
Ø	Р	logspace	$P \neq PSPACE$	Corollary 3.3
QBF	Р	logspace	EXP = PSPACE	Corollary 3.7
QBF	NP	logspace	NEXP = PSPACE	Theorem 3.6
QBF	PSPACE	logspace	\perp	Corollary 3.8
EXP-complete	NP	polytime	NEXP = EXP	Theorem 3.4

For the most restricted reductions, uniform AC^0 , we get that $MCSP^A$ is not TC^0 -hard for any oracle A such that $PH \subseteq SIZE^A[poly]$ (Theorem 3.9), e.g., for $A = \oplus P$ (Corollary 3.12). For any oracle A, we conclude new circuit lower bounds for the linear-time hierarchy and for NP^A

(Lemma 3.10 and Corollary 3.13^1).

If MCSP is P-hard under uniform TC^0 or logspace reductions, then P is different from PP or from PSPACE (Corollaries 3.2 and 3.3).

One of the more interesting oracle circuit minimization problems is $MCSP^{QBF}$. It was shown in $[ABK^+06]$ that $MCSP^{QBF}$ is complete for PSPACE under ZPP-Turing reductions, but the question of whether it is complete for PSPACE under more restrictive reductions was left open. For most natural complexity classes C above PSPACE, there is a corresponding oracle circuit minimization problem (which we will sometimes denote $MCSP^{C}$) that is known to be complete under P/poly reductions, but is not known to be complete under more restrictive reductions [ABK+06]. For the particular case of C = PSPACE, we denote this as $MCSP^{QBF}$. We show that $MCSP^{QBF}$ is not PSPACE-complete under logspace reductions (Corollary 3.8). Furthermore, it is not even TC^{0} -hard under uniform AC^{0} reductions (Theorem 3.9).

Finally, for even more powerful oracles A, we can handle even general polynomial-time reductions. We show that if $SAT \leq_m^p MCSP^{EXP}$, then EXP = NEXP (Theorem 3.4).

We believe that MCSP is not TC^0 -hard under even *nonuniform* AC^0 reductions. While we are unable to prove this, we can rule out restricted AC^0 reductions for a certain gap version of MCSP. Define gap-MCSP as follows: Given a truth table f and a parameter s, output 'Yes' if f requires circuit size s, and output 'No' if f can be computed by a circuit of size at most s/2. Call a mapping from *n*-bit strings to *m*-bit strings $\alpha(n)$ -stretching if $m \leq n \cdot \alpha(n)$, for some function $\alpha : \mathbb{N} \to \mathbb{R}^{\geq 0}$.

We prove that gap-MCSP is not TC^0 -hard under nonuniform AC^0 reductions that are $n^{1/31}$ -stretching (Theorem 3.17).

1.2 Related work

The most closely related is the recent paper by Murray and Williams [MW14], which also considers the question whether MCSP is NP-complete under weak reductions, and proves a number of conditional and unconditional results. The main unconditional result is that MCSP is *not* TC⁰-hard under uniform NC⁰ reductions (or more generally, under $O(n^{1/2-\epsilon})$ -time projections, for every $\epsilon > 0$); we give an alternative proof of this result (Theorem 3.15). For conditional results, [MW14] shows that if MCSP is NP-hard under uniform AC⁰ reductions, then NP $\not\subset$ P/poly and E $\not\subset$ io-SIZE[2^{$\Omega(n)$}] (also implied by our Corollary 3.13 and Lemma 3.10), and that NP-hardness of MCSP under general polynomial-time reductions implies EXP \neq ZPP.

MCSP, MCSP^{QBF} and other oracle circuit minimization problems are closely related to notions of resource-bounded Kolmogorov complexity. Briefly, a small (oracle) circuit is a short description of the string that represents the truth-table of the function computed by the circuit. Notions of resource-bounded Kolmogorov complexity were presented and investigated in [ABK⁺06] that are roughly equivalent to (oracle) circuit size.

In particular, there is a space-bounded notion of Kolmogorov complexity, KS, such that the set of KS-random strings (denoted R_{KS}) is complete for PSPACE under ZPP reductions. It is shown in [ABK⁺06] that R_{KS} is not even hard for TC⁰ under AC⁰ reductions, and R_{KS} is not hard for PSPACE under logspace-Turing reductions. The proof of this non-hardness result also carries over to show that a set such as $\{f : f \text{ is the truth table of a function on } n \text{ variables that has QBF}$ circuits of size at most $2^{n/2}\}$ is also not hard for TC⁰ under AC⁰ reductions, and is not hard for PSPACE under logspace-Turing reductions. However it does *not* immediately carry over to MCSP^{QBF}, which is defined as $\{(f, i) : f \text{ is the truth table of a function on } n \text{ variables that has}$

¹Prior to our work, Murray and Williams have shown that if $SAT \leq_m^{AC^0} MCSP$, then NP $\not\subseteq$ P/poly [MW14]. Their result is similar to (and is implied by) our Corollary 3.13 for the case of $A = \emptyset$.

QBF circuits of size at most i}; similarly it does not carry over to the set $\{(x, i) : \mathsf{KS}(x) \leq i\}$. Also, the techniques presented in [ABK⁺06] have not seemed to provide any tools to derive consequences assuming completeness results for oracle circuit minimization problems for oracles less powerful than PSPACE. We should point out, however, that [ABK⁺06] proves a result similar to (and weaker than) our Lemma 3.10 in the context of time-bounded Kolmogorov complexity: if R_{KT} is TC^0 -hard under AC^0 many-one reductions, then $\mathsf{PH} \not\subseteq \mathsf{SIZE} \left[2^{n^{o(1)}}\right]$.

1.3 Our techniques

To illustrate our proof techniques, let us sketch a proof of one of our results: If MCSP is P-hard under uniform logspace reductions, then $P \neq PSPACE$ (Corollary 3.3).

The proof is by contradiction. Suppose that P = PSPACE. Our logspace reduction maps *n*-bit instances of QBF to n^c -bit instances (f, s) of MCSP so that each bit of f is computable in $O(\log n)$ space.

- Imagine that our reduction is given as input a succinct version of QBF, where some poly(log n)-size circuit D on each log n-bit input 1 ≤ i ≤ n computes the ith bit of the QBF instance. It is not hard to see that our reduction, given the circuit D, can compute each bit of f in poly(log n) space. Thus the Boolean function with the truth table f is computable by a PSPACE = P algorithm (which also has the circuit D as an input). It follows that this function f is computable by some polynomial-size Boolean circuit.
- 2. Next, since we know that f has at most polynomial circuit complexity, to decide the MCSP instance (f, s), we only need to consider the case where s < poly (since for big values of s, the answer is 'Yes'). But deciding such MCSP instances (which we call *succinct* MCSP) is possible in Σ_2^p : guess a circuit of size at most s, and verify that it agrees with the given polynomial-size circuit for f on all inputs.
- 3. Finally, since $\Sigma_2^p \subseteq \mathsf{PSPACE} = \mathsf{P}$, we get that our succinct MCSP instances can be decided in P . The reduction from succinct QBF to succinct MCSP is also in $\mathsf{PSPACE} = \mathsf{P}$. Hence, succinct QBF is in P . But, succinct QBF is $\mathsf{EXPSPACE}$ -complete, and so we get the collapse $\mathsf{EXPSPACE} = \mathsf{P}$, contradicting the hierarchy theorems.

In step (1) of the sketched proof, the uniformity of an assumed reduction to MCSP is used to argue that the truth table f produced by the reduction is in fact "easy" to compute uniformly. The uniform complexity of computing the function f is roughly the "exponential" analogue of the uniform complexity of the reduction. For circuit classes such as AC⁰ and TC⁰, we use the well-known connection between the "exponential" analog of uniform AC⁰ and PH, and between the "exponential" analog of uniform TC⁰ and CH.

We use the uniform easiness of the function f to conclude that f has small circuit complexity (and hence our reduction actually outputs instances of *succinct* MCSP). To get that conclusion, we need to assume (or derive) the collapse to P/poly of the uniform complexity class that contains f; in our example above, we got it from the assumption that PSPACE = P.

Step (2) exploits the fact that succinct MCSP does *not* become "exponentially harder" (unlike the usual succinct versions of hard problems), but is actually computable in Σ_2^p .

In Step (3), we combine the algorithm for our reduction and the algorithm for succinct MCSP to get an "efficient" algorithm for the succinct version of the input problem (succinct QBF in our example). Since the succinct version of the input problem *does* become exponentially harder

than its non-succinct counterpart, we get some impossible collapse (which can be disproved by diagonalization).

We use this style of proof for all our results involving reductions computable by uniform TC^0 and above. However, for the case of uniform AC^0 (and below), we get stronger results by replacing the diagonalization argument of Step (3) with the nonuniform AC^0 circuit lower bound for PARITY [Hås86].

Remainder of the paper. We state the necessary definitions and auxiliary results in Section 2. Our main results are proved in Section 3, and some generalizations are given in Section 4. We give concluding remarks in Section 5.

2 Definitions

Definition 2.1. The minimum circuit size problem MCSP, as defined in [KC00], is defined as $\{(f,s)|f$ has circuits of size $s\}$, where f is a string of length 2^m encoding the entire truth-table of some *m*-variate Boolean function. (Versions of this problem have been studied long prior to [KC00]. See [AD14, Tra84] for a discussion of this history.) We will also consider the analogous problem for circuits with oracles, the Minimum A-Circuit Size problem MCSP^A, defined analogously, where instead of ordinary circuits, we use circuits that also have oracle gates that query the oracle A. When A is a standard complete problem for some complexity class C, we may refer to this as MCSP^C.

We will not need to be very specific about the precise definition of the "size" of a circuit. Our results hold if the "size" of a circuit is the number of gates (including oracle gates), or the number of "wires", or the number of bits used to describe a circuit in some standard encoding. It is perhaps worth mentioning that the different versions of MCSP that one obtains using these different notions of "size" are not known to be efficiently reducible to each other.

Circuit size relative to oracle A is polynomially-related to a version of time-bounded Kolmogorov complexity, denoted KT^A , which was defined and studied in [ABK⁺06].

Definition 2.2. $\operatorname{KT}^{A}(x) = \min\{|d| + t : \forall b \in \{0, 1, *\} \forall i \leq |x| + 1 \ U^{A}(d, i, b) \text{ accepts in } t \text{ steps iff } x_{i} = b\}$. Here, U is some fixed universal Turing machine, which has random access to the oracle A and to the input string (or "description") d; x_{i} denotes the *i*-th symbol of x, where $x_{|x|+1} = *$.

By analogy to $MCSP^A$, we define the "minimum KT problem":

Definition 2.3. $\mathsf{MKTP}^A = \{(x, i) | \mathsf{KT}^A(x) \le i\}.$

All of our results that deal with $MCSP^A$ also apply to $MKTP^A$.

We wish to warn the reader that one's intuition can be a poor guide, when judging how $MCSP^A$ and $MCSP^B$ compare to each other, for given oracles A and B. For instance, it is known that $MCSP^{SAT}$ ZPP-Turing reduces to $MCSP^{QBF}$ [ABK+06], but no deterministic reduction is known. Similarly, no efficient reduction of any sort is known between MCSP and $MCSP^{SAT}$. Some of our theorems derive consequences from the assumption that $MCSP^{SAT}$ is hard for some complexity class under AC^0 reductions. Although one might suspect that this is a weaker hypothesis than assuming that $MCSP^{SAT}$ is hard for the same complexity class under AC^0 reductions – certainly the best upper bound for $MCSP^{SAT}$ is worse than the best known upper bound for $MCSP^{-}$ nonetheless we are not able to derive the same consequences assuming only that MCSP is hard. For essentially

all time- and space-bounded complexity classes C that contain PSPACE, MCSP^C is complete for C/poly under P/poly reductions [ABK⁺06, AKRR10], but uniform reductions are known only for two cases [ABK⁺06]: when C = PSPACE (MCSP^{QBF} is complete for PSPACE under ZPP reductions) and when C = EXP (MCSP^{EXP} is complete for EXP under NP-Turing reductions).

2.1 Succinct Problems

The study of succinct encodings of computational problems was introduced by [GW83, PY86], and has been studied since then by [Wag86, BLT92], among others. Succinct encodings play an important role in the proofs of our main results.

Definition 2.4. Given a language L, we define the succinct version of L (denoted succ. L) to be the language $\{C|tt(C) \in L\}$ where C is a Boolean Circuit and tt(C) is the truth-table for C.

It will be necessary for us to consider "succinctly-presented" problems, where the circuit that constitutes the succinct description is itself an *oracle* circuit:

Definition 2.5. Given a language L and an oracle A, we define the A-succinct version of L (denoted A-succ.L) to be the language $\{C|tt(C) \in L\}$ where C is a Boolean Circuit with oracle gates, and tt(C) is the truth-table for C, when it is evaluated with oracle A. If $A = \emptyset$, we denote this language as succ.L.

The typical situation that arises is that the succinct version of a problem A has exponentially greater complexity than A. In particular, this happens when A is complete for a complexity class under "logtime reductions".

Definition 2.6. We say that a function f can be computed in logarithmic time if there exists a random-access Turing machine that, given (x, i), computes the *i*th bit f(x) in time $O(\log |x|)$.

Building on prior work of [PY86, GW83, Wag86], Balcázar, Lozano, and Torán presented a large list of complexity classes (C_1 , C_2), where C_1 is defined in terms of some resource bound B(n)and C_2 is defined in the same way, with resource bound $B(2^n)$, such that if a set A is complete for C_1 under logtime reductions, then succ. A is complete for C_2 under polynomial-time many-one reductions [BLT92].

Somewhat surprisingly, the complexity of succ.MCSP appears *not* to be exponentially greater than that of MCSP. (Related observations were made earlier by Williams [Wil12].)

Theorem 2.7. succ. $MCSP \in \Sigma_2^p$

Proof. We present an algorithm in Σ_2^p that decides succ.MCSP. Given an instance of succinct MCSP C, note that $C \in \text{succ.MCSP}$ iff z is a string of the form $(f, s) \in \text{MCSP}$, where z = tt(C). By definition, |z| must be a power of 2, say $|z| = 2^r$, and |f| must also be a power of 2, say $|f| = 2^m$ for some m < r. Note also that if $s > |f| = 2^m$, then (f, s) should obviously be accepted, since every m-variate Boolean function has a circuit of size 2^m . To be precise, we will choose one particular convention for encoding the pair (f, s); other reasonable conventions will also yield a Σ_2^p upper bound. Let us encode (f, s) as a string of length 2^{m+1} , where the first 2^m bits give the truth table for f, and the second 2^m bits give s in binary. Note that this means that C has m + 1 input variables, and hardwiring the high-order input bit of C to 0 results in a circuit C' for f (of size at most |C|).

Using this encoding, the "interesting" instances (f, s) are of the form where the second half of the string is all zeros, except possibly for the low-order m bits (encoding a number $s \leq 2^m = |f|$.

The low-order m bits can be computed deterministically in polynomial time, given C, by evaluating C on inputs $1^{m+1-\log m}0^{\log m}, 1^{m+1-\log m}0^{-1+\log m}1, \ldots, 1^{m+1}$. Let the number encoded by the low-order m bits be s'. Then C (an encoding of (f, s)) is in succ.MCSP iff

- there is some bit position j corresponding to one of the high-order $2^m m$ bits of s such that C(j) = 1, or
- there exists a circuit D of size at most s' such that, for all i, D(i) = C'(i) and for all bit positions j corresponding to one of the high-order $2^m m$ bits of s, C(j) = 0 (and thus s = s').

It is easily seen that this can be checked in Σ_2^p .

Because this proof relativizes, we obtain:

Corollary 2.8. Let A and B be oracles such that $B \leq_T^p A$. Then B-succ.MCSP^A is in $(\Sigma_2^p)^A$.

Proof. We use the same encoding as in Theorem 2.7. Thus, an oracle circuit C encoding an instance (f, s) (where f is an m-ary function) has m+1 input variables, and hardwiring the high-order input bit of C to 0 results in an oracle circuit C' (with oracle B) for f (of size at most |C|). But if $B \leq_T^p A$, then this also gives us an oracle circuit C'' (with oracle A) for f (of size at most $|C|^k$ for some k), where we can obtain C'' from C in polynomial time.

Then C (an encoding of (f, s)) is in B-succ.MCSP^A iff

- there is some bit position j corresponding to one of the high-order $2^m m$ bits of s such that $C^B(j) = 1$, or
- there exists a circuit D of size at most s' such that, for all $i, D^A(i) = C''^A(i)$ and for all bit positions j corresponding to one of the high-order $2^m m$ bits of $s, C^B(j) = 0$ (and thus s = s').

It is easily seen that this can be checked in $(\Sigma_2^p)^A$.

An analogous result also holds for MKTP^A .

Theorem 2.9. Let A and B be oracles such that $B \leq_T^p A$. Then B-succ.MKTP^A is in $(\Sigma_2^p)^A$.

Proof. Given an instance of *B*-succ.MKTP^A *C*, note that $C \in B$ -succ.MKTP^A only if *z* is a string of the form (x, i), where z = tt(C). Let us settle on a suitable encoding for pairs; the number *i* should be at most 2|x| (a generous overestimate of how large $KT^A(x)$ could be), and thus should consist of at $O(\log |x|)$ bits. In order to mark the location of the "comma" separating *x* and *i*, we use the familiar convention of doubling each bit of *i*, and using the symbols 10 to mark the position of the "comma". Thus, given a circuit *C* with *n* variables, a $(\Sigma_2^p)^B$ machine can compute the length of the encoded string *x* as follows:

- 1. Using nondeterminism, guess a position ℓ and verify that $C^B(\ell) = 1$ and $C^B(\ell+1) = 0$.
- 2. Using co-nondeterminism, verify that for all $\ell' > \ell$ it is not the case that $C^B(\ell') = 1$ and $C^B(\ell'+1) = 0$. (If this test passes, then the tt(C) is of the form (x,i) for some x and i, although it allows the possibility that absurdly large numbers i are provided.)
- 3. Reject if the number of bits used in the encoding of i is more than 4n (which is greater than $4\log |x|$).

This $(\Sigma_2^p)^B$ computation can be simulated by a $(\Sigma_2^p)^A$, by our assumption that $B \leq_T^p A$.

The rest of the algorithm follows closely the algorithm that we presented for MCSP. Given a circuit C, guess the number ℓ such that tt(C) = (x, i) for some string x of length ℓ . Without loss of generality, we can assume that the universal Turing machine used to define KT takes a description of a program and the input to the program and runs the program on that input. The universal oracle machine U, given a description d of length $|C| + |\ell| + O(1)$, along with (i, b) can output * if $i > \ell$ and otherwise can use oracle A to simulate $C^B(i)$ and accept iff the answer is b. The running time will be at most $(|C| + |\ell|)^k$ for some k, which gives us an upper bound on $\mathrm{KT}^A(x)$.

The $(\Sigma_2^p)^A$ algorithm for *B*-succ.MKTP^A is thus:

- 1. Guess and verify ℓ as above, and in parallel:
- 2. Evaluate $C^B(j)$ for the 4n largest positions $j < 2^n$ (using oracle A), and thus obtain the encoding of *i*.
- 3. Accept if $i \ge (|C| + |\ell|)^k$.
- 4. Guess a description d' of length at most i. Reject if $U^A(d, \ell+1, *)$ does not accept.
- 5. Using co-nondeterminism, verify that for all $j \leq \ell$ and all $b \in \{0,1\}$, $U^A(d,j,b)$ accepts iff $C^B(j) = b$.

2.2 Constant-Depth Reductions

Proposition 2.10. Suppose that f is a uniform AC^0 reduction from a problem A to a problem B. Let C be an instance of succ.A. Then, the language $\{(C, i)|$ the *i*th bit of f(tt(C)) is 1 $\}$ is in LTH (the linear-time hierarchy).

Proof. Consider the unary version of the above language: $\{1^{(C,i)}|$ the *i*th bit of f(tt(C)) is 1}; we claim that this language is in uniform AC^0 . To see this, note that after computing the length of the input (in binary), and thus obtaining a description of C (of length $\log n$), an AC^0 algorithm can compute each bit of tt(C). For instance, the *i*th bit of tt(C) can be computed by guessing a bit vector of length $\log n$ recording the value of each gate of C on input *i*, and then verifying that all of the guessed values are consistent. Once the bits of tt(C) are available, then the AC^0 algorithm computes f(tt(C)).

The result is now immediate, from [AG93, Proposition 5], which shows that the rudimentary languages (that is, the languages in the linear-time version LTH of the polynomial-time hierarchy PH) are precisely the sets whose unary encodings are in Dlogtime-uniform AC^0 .

By an entirely analogous argument, we obtain:

Proposition 2.11. Suppose that f is a uniform TC^0 reduction from a problem A to a problem B. Let C be an instance of succ. A. Then, the language $\{(C, i) | \text{ the ith bit of } f(tt(C)) \text{ is } 1\}$ is in CH.

3 Main Results

3.1 Conditional collapses and separations of complexity classes

Our first theorem shows that significant conclusions follow if MCSP is hard for P under AC^0 reductions.

Theorem 3.1. If there is any set A in the polynomial hierarchy such that $MCSP^A$ (or $MKTP^A$) is hard for P under AC^0 reductions, then $P \neq NP$.

Proof. We present only the proof for $MCSP^A$; the proof for $MKTP^A$ is identical. Suppose that P = NP and $MCSP^A$ is hard for P under AC^0 reductions. Thus, there is a family $\{C_n\}$ of AC^0 circuits reducing SAT to $MCSP^A$, such that $C_n(\phi) = f(\phi)$, where f is the reduction function and ϕ is an instance of SAT.

Now we claim that succ.SAT \leq_m^p succ.MCSP^A. To see this, consider an instance D of succ.SAT (that is, a circuit D on n variables that, when given input i, outputs the ith bit of a SAT instance of size 2^n). This problem has been shown to be complete for NEXP[Pap03]. By Proposition 2.10, we have that the language $\{(D, i) | \text{ the } i\text{th } \text{bit of } f(tt(D)) \text{ is } 1\}$ is in PH. By our assumption that P = NP, we have that this language is in P. Let E_m be a family of circuits deciding this language. The function that takes input D and outputs $E_{|(D,n)|}$ (with D hardwired in) is a polynomial-time reduction from succ.SAT to succ.MCSP^A, which is in $(\Sigma_2^p)^A$, by Corollary 2.8. Since $A \in P$ (by our assumption that P = NP), we have that NEXP $\subseteq P$, which is a contradiction.

Corollary 3.2. If there is any set $A \in CH$ such that $MCSP^A$ (or $MKTP^A$) is hard for P under TC^0 reductions, then $P \neq PP$.

Proof. The proof is similar to that of the preceding theorem. If P = PP, and there is a TC^0 reduction f from SAT to $MCSP^A$, then the language $\{(D, i) | \text{ the } i\text{ th bit of } f(tt(D)) \text{ is } 1\}$ is in CH (by Proposition 2.11), and hence is in P.

Now, just as above, we use the circuit family recognizing this language, in order to construct a polynomial-time reduction from succ.SAT to succ.MCSP^A, leading to the contradiction that NEXP = P.

Corollary 3.3. Suppose that MCSP (or MKTP) is hard for P under logspace many-one reductions. Then $P \neq PSPACE$.

Proof. The proof proceeds along similar lines. Assume P = PSPACE. Consider an instance D of succ.SAT, where there is a reduction f computable in logspace reducing SAT to MCSP. Then the language $\{(D, i)|$ the *i*th bit of f(tt(D)) is 1} is in PSPACE, since polynomial space suffices in order to compute f on an exponentially-large input. (We don't need to store the string tt(D), the bits of tt(D) can re-computed when they are needed.) By our assumption that P = PSPACE, this language is in P, and hence is recognized by a uniform circuit family (E_m) .

Now, as above, the function that maps D to $E_{|(D,n)|}$ (with D hardwired in) is a polynomial-time reduction from succ.SAT to succ.MCSP, which allows us to conclude that NEXP = P.

Theorem 3.4. Suppose that $MCSP^{EXP}$ is hard for NP under polynomial-time reductions. Then NEXP = EXP.

Proof. Let f be the reduction taking an instance of SAT to an instance of MCSP^{EXP}. We construct a reduction from succ.SAT to B-succ.MCSP^{EXP} for some $B \in EXP$.

Consider the language $L = \{(C, i) | \text{ the } i\text{ th } \text{ bit of } f(\phi_C) \text{ is } 1\}$, where ϕ_C is the formula described by the circuit C, viewed as an instance of succ.SAT with n input variables. We can decide L in exponential time because we can write down ϕ_C in exponential time, and then we can compute $f(\phi_C)$ in exponential time because f is a poly-time reduction on an exponentially large instance. Let $\{D_m\}$ be a family of oracle circuits for L, using an oracle for an EXP-complete language B. Thus the mapping $C \mapsto D_{|C|+n}$ is a polynomial-time reduction from succ.SAT to B-succ.MCSP^{EXP}, which is in $(\Sigma_2^p)^{\mathsf{EXP}} = \mathsf{EXP}$ (see, e.g., [AKRR10, Theorem 24]), and thus $\mathsf{EXP} = \mathsf{NEXP}$. **Corollary 3.5.** Consider Levin's time-bounded Kolmogorov complexity measure Kt [Lev84]. Suppose that $\{(x, i) : Kt(x) \le i\}$ is hard for NP under polynomial-time reductions. Then NEXP = EXP.

Proof. As discussed in [ABK⁺06], there is essentially no difference between Kt(x) and $KT^{\mathsf{EXP}}(x)$. Thus the proof is immediate, given the proof of Theorem 3.4.

Theorem 3.6. If $MCSP^{QBF}$ or $MKTP^{QBF}$ is hard for NP under logspace reductions, then NEXP = PSPACE.

Proof. Let f be the reduction taking an instance of SAT to an instance of MCSP^{QBF}. We construct a reduction from succ.SAT to QBF-succ.MCSP^{QBF}.

Consider the language $L = \{(C, i) | \text{ the } i\text{th bit of } f(\phi_C) \text{ is } 1\}$, where ϕ_C is the formula described by the circuit C, viewed as an instance of succ.QBF with n input variables. We can decide L in PSPACE, because we can compute $f(\phi_C)$ by building the bits of ϕ_C as they are needed. Let $\{D_m\}$ be a family of oracle circuits for L, using an oracle for QBF. Thus the mapping $C \mapsto D_{|C|+n}$ is a polynomial-time reduction from succ.QBF to QBF-succ.MCSP^{QBF}, which is in $(\Sigma_2^p)^{\text{QBF}} =$ PSPACE, implying NEXP = PSPACE.

Corollary 3.7. If $MCSP^{QBF}$ (or $MKTP^{QBF}$) is hard for P under logspace reductions, then EXP = PSPACE.

Proof. The proof is identical to the proof of the preceding theorem, with NP replaced by P, and with NEXP replaced by EXP. \Box

If we carry out a similar argument, replacing NP with PSPACE, we obtain the contradiction EXPSPACE = PSPACE, yielding the following.

Corollary 3.8. Neither MCSP^{QBF} nor MKTP^{QBF} is hard for PSPACE under logspace reductions.

3.2 Impossibility of uniform AC⁰ reductions

Theorem 3.9. For any language A that is hard for PH under P/poly reductions, $MCSP^A$ is not hard for TC^0 under uniform AC^0 reductions.

The theorem will follow from the next lemma. Recall that LTH (linear-time hierarchy) stands for the linear-time version of the polynomial-time hierarchy PH.

Lemma 3.10. Suppose that, for some language A, MCSP^A is TC^0 -hard under uniform AC^0 reductions. Then $\mathsf{LTH} \not\subseteq \mathsf{io}\mathsf{-SIZE}^A[2^{\Omega(n)}]$.

Proof. It is shown in [Agr11, Theorems 5.1 and 6.2] that if a set is hard for any class C that is closed under TC^0 reductions under uniform AC^0 reductions, then it is hard under length-increasing (uniform AC^0)-uniform NC^0 reductions. (Although Theorems 5.1 and 6.2 in [Agr11] are stated only for sets that are *complete* for C, they do hold also assuming only hardness [Agr14], using exactly the same proofs.) Here, the notion " AC^0 -uniform NC^0 " refers to NC^0 circuits with the property that direct connection language $DCL = \{(n, t, i, j) | \text{ gate } i \text{ of } F_n \text{ has type } t \text{ and has an edge leading from gate } j\}$ with n in unary is in Dlogtime-uniform AC^0 .

Hence, if $MCSP^A$ is hard for TC^0 under uniform AC^0 reductions, then we get that PARITY is reducible to $MCSP^A$ under a length-increasing (uniform AC^0)-uniform NC^0 reduction. Such a reduction R maps PARITY instances $x \in \{0, 1\}^n$ to $MCSP^A$ instances (f, s), where f is the truth table of a Boolean function, $f \in \{0, 1\}^m$, for some m such that $n \le m \le n^{O(1)}$, and $0 \le s \le m$ is the size parameter in binary, and hence $|s| \le O(\log n)$.

Being the output of an NC⁰ reduction, the binary string s depends on at most $O(\log n)$ bits in the input string x. Imagine fixing these bits in x to achieve the minimum value of the parameter s. Denote this minimum value of s by v. (We do not need for v to be efficiently computable in any sense.) We get a *nonuniform* NC⁰ reduction from PARITY on $n - O(\log n) \ge n/2$ bit strings to MCSP^A with the size parameter fixed to the value v.

Claim 3.11. For any language A and any $0 \le v \le m$, MCSP^A on inputs $f \in \{0,1\}^m$, with the size parameter fixed to v, is solved by a DNF formula of size $O(m \cdot 2^{v^2 \log v})$.

Proof of Claim 3.11. Each A-oracle circuit of size v on $\log m$ inputs can be described by a binary string of length at most $O(v^2 \log v)$, since each of v gates has at most v inputs. Thus, there are at most $2^{O(v^2 \log v)}$ Boolean functions on $\log m$ inputs that are computable by A-oracle circuits of size at most v. Checking if any one of these truth tables equals to the input truth table f can be done by a DNF, where we take an OR over all easy functions, and for each easy function we use an AND gate to check equality to the input f.

We conclude that PARITY on n/2-bit strings is solvable by AC^0 circuits of depth 3 and size $O(m \cdot 2^{v^2 \log v})$. Indeed, each bit of the truth table f is computable by an NC^0 circuit, and hence by a DNF (and a CNF) of constant size. Plugging in these DNFs (or CNFs) for the bits of f into the DNF formula from Claim 3.11 yields the required depth-3 AC^0 circuit for PARITY on inputs of length at least n/2.

Next, since PARITY on *m*-bit strings requires depth-3 AC^0 circuits of size at least $2^{\Omega(\sqrt{m})}$ [Hås86], we get that $v \ge n^{1/5}$. Hence, on input 0^n , our *uniform* NC⁰ reduction produces (f, s) where f is the truth table of a Boolean function on *r*-bit inputs that has *A*-oracle circuit complexity at least $v \ge n^{1/5} \ge 2^{\epsilon r}$, for some $\epsilon > 0$.

Finally, since the NC^0 reduction is (uniform AC^0)-uniform, we get that the Boolean function whose truth table is f is computable in LTH.

Proof of Theorem 3.9. Towards a contradiction, suppose that $MCSP^A$ is TC^0 -hard under uniform AC^0 reductions. Then, by Lemma 3.10, there is a language $L \in PH$ that requires A-oracle circuit complexity $2^{\Omega(n)}$ almost everywhere. However, since A is PH-hard under P/poly reductions, we get that $L \in SIZE^A[poly]$. A contradiction.

Corollary 3.12. $MCSP^{\oplus P}$ is not TC^0 -hard under uniform AC^0 reductions.

Proof. By Toda's theorem [Tod91], $\mathsf{PH} \subseteq \mathsf{BPP}^{\oplus \mathsf{P}}$, which in turn is contained in the class of problems P/poly -reducible to the standard complete problem for $\oplus \mathsf{P}$. The result then follows by Theorem 3.9.

Corollary 3.13. Suppose that, for some oracle A, $MCSP^A$ is TC^0 -hard under uniform AC^0 reductions. Then $NP^A \not\subseteq SIZE^A[poly]$.

Proof. If $NP^A \subseteq SIZE^A[poly]$, then $PH^A \subseteq SIZE^A[poly]$. Now the result follows from Lemma 3.10.

Remark 3.14. Murray and Williams [MW14] prove results similar to (and implied by) our Lemma 3.10 and Corollary 3.13 for the case of the empty oracle $A = \emptyset$. Namely, they show that if MCSP is NP-hard under uniform AC^0 reductions, then NP $\not\subseteq$ P/poly and E $\not\subseteq$ io-SIZE[$2^{\Omega(n)}$].

Finally, we observe that the ideas in our proof of Lemma 3.10 yield an alternate proof of the result by Murray and Williams [MW14] that PARITY is not reducible to MCSP via "local" $O(n^{1/2-\epsilon})$ -time reductions. We prove the version for polylogtime-uniform NC⁰ reductions, but the same argument applies also to the "local" reductions of [MW14].

Theorem 3.15 ([MW14]). There is no polylogtime-uniform NC⁰ reduction from PARITY to MCSP.

Proof. Suppose there is such a reduction. Similarly to the proof of Lemma 3.10, we conclude that this NC^0 reduction maps 0^n to an MCSP instance (f, s) where f is the truth table of a Boolean function on $r := O(\log n)$ inputs that requires exponential circuit size $s \ge 2^{\Omega(r)}$. On the other hand, since our NC^0 reduction is polylogtime-uniform, the Boolean function with the truth table f is computable in P, and hence in SIZE[poly]. A contradiction.

3.3 Gap MCSP

For $0 < \epsilon < 1$, we consider the following gap version of MCSP, denoted ϵ -gap MCSP: Given (f, s), output 'Yes' if f requires circuits of size at least s, and output 'No' if f can be computed by a circuit of size at most $(1 - \epsilon)s$.

For $\alpha : \mathbb{N} \to \mathbb{R}^+$, call a mapping $R : \{0,1\}^n \to \{0,1\}^m \alpha$ -stretching if $m \leq \alpha(n) \cdot n$. We will prove that there is no n^{δ} -stretching nonuniform AC⁰ reduction from PARITY to ϵ -gap MCSP, for certain parameters $0 < \epsilon, \delta < 1$. First, we rule out nonuniform NC⁰ reductions.

Theorem 3.16. For every $n^{-1/6} < \epsilon < 1$ and for every constant $\delta < 1/30$, there is no n^{δ} -stretching (nonuniform) NC⁰ reduction from PARITY to ϵ -gap MCSP.

Proof. Towards a contradiction, suppose there is an n^{δ} -stretching NC⁰ reduction from PARITY on inputs $x \in \{0,1\}^n$ to ϵ -gap MCSP instances (f,s). Fix to zeros all $O(\log n)$ bit positions in the string x that determine the value of the size parameter s. As in the proof of Lemma 3.10, we get an NC⁰ reduction from PARITY on at least n/2 bits y to the ϵ -gap MCSP instance with the size parameter fixed to some value $v \ge n^{1/5}$.

By our assumption, $|f| \leq n \cdot n^{\delta}$. Since each bit of f is computable by an NC⁰ circuit, we get that each bit of f depends on at most c bits in the input y. The total number of pairs (i, j) where f_i depends on bit y_j is at most $c \cdot |f|$. By averaging, there is a bit y_j , $1 \leq j \leq n/2$, that influences at most $c|f|/(n/2) \leq 2cn^{\delta}$ bit positions in the string f.

Fix y so that all bits are 0 except for y_j (which is set to 1). This y is mapped by our NC^0 reduction to the truth table f' that is computable by a circuit of size at most $(1-\epsilon)v$. On the other hand, flipping the bit y_j to 0 forces the reduction to output a truth table f'' of circuit complexity at least v. But, y_j influences at most $2cn^{\delta}$ positions in f', and so the circuit complexity of f'' differs from that of f' by at most $O(n^{\delta} \log n)$ gates (as we can just construct a "difference" circuit of that size that is 1 on the at most $2cn^{\delta}$ affected positions of f'). We get $\epsilon v \leq O(n^{\delta} \log n)$, which is impossible when $\delta < 1/30$.

Now we extend Theorem 3.16 to the case of nonuniform AC^0 reductions.

Theorem 3.17. For every $n^{-1/7} < \epsilon < 1$ and for every constant $\delta < 1/31$, there is no n^{δ} -stretching (nonuniform) AC⁰ reduction from PARITY to ϵ -gap MCSP.

Proof. Towards a contradiction, suppose there is a n^{δ} -stretching AC^0 reduction from PARITY on *n*-bit strings to the ϵ -gap MCSP. We will show that this implies the existence of an NC⁰ reduction with parameters that contradict Theorem 3.16 above.

Claim 3.18. For every constant $\gamma > 0$, there exist a constant a > 0 and a restriction of our AC⁰ circuit satisfying the following: (1) each output of the restricted circuit depends on at most a inputs, and (2) the number of unrestricted variables is at least $n^{1-\gamma}$.

Proof of Claim 3.18. Recall that a random *p*-restriction of *n* variables x_1, \ldots, x_n is defined as follows: for each $1 \le i \le n$, with probability *p*, leave x_i unrestricted, and with probability 1 - p, set x_i to 0 or 1 uniformly at random. By Håstad's Switching Lemma [Hås86], the probability that a given CNF on *n* variables with bottom fan-in at most *t* does not become a decision tree of depth at most *r* after being hit with a random *p*-restriction is at most $(5pt)^r$.

For an AC⁰ circuit of size n^k and depth d, set $p := (5a)^{-1}n^{-2k/a}$ for some constant a > 0 to be determined. Applying this random p-restriction d times will reduce the original circuit to a decision tree of depth a with probability at least $1 - dn^k(5pa)^a > 3/4$. The expected number of unrestricted variables at the end of this process is $p^d n \ge \Omega(n/n^{2kd/a}) = \Omega(n/n^{\gamma'})$, for $\gamma' := 2kd/a$. By Chernoff bounds, the actual number of unrestricted variables is at least 1/2 of the expectation with probability at least 3/4.

Thus, with probability at least 1/2, we get a restriction that makes the original AC^0 circuit into an NC^0 circuit on at least $n/n^{2\gamma'}$ variables, where each output of the new circuit depends on at most *a* input variables. Setting $\gamma := 2\gamma'$, we get that $a = (4kd)/\gamma$.

We get an NC⁰ reduction from PARITY on $n' := n^{1-\gamma}$ variables to ϵ -gap MCSP. This reduction is at most $(n')^{(\delta+\gamma)/(1-\gamma)}$ -stretching. Choose $0 < \gamma < (1/31)^2$ so that $(\delta + \gamma)/(1-\gamma) < 1/30$, and $\epsilon > n^{-1/7} > (n')^{-1/6}$. Finally, appeal to Theorem 3.16 for contradiction.

4 Generalizations

Theorem 3.1 gives consequences of MCSP being hard for P. The property of P that is exploited in the proof is that the polynomial hierarchy collapses to P if NP = P. (This is required, so that we can efficiently a circuit that computes bits of the reduction, knowing only that it is in the polynomial hierarchy.)

The next theorem formalizes this observation:

Theorem 4.1. Let C be any class such that if NP = C, then PH = C. If there is a set $A \in PH$ that is hard for C under \leq_T^p reductions such that $MCSP^A$ (or $MKTP^A$) is hard for C under uniform AC^0 reductions, then $NP \neq C$.

Proof. Suppose that $\mathsf{NP} = \mathcal{C}$, and MCSP^A is hard for \mathcal{C} . Then, there exists a reduction from SAT to MCSP^A computable in AC^0 . As in the proof of Theorem 3.1, we can use this to construct a \leq_T^p reduction from succ.SAT to B-succ. MCSP^A for some B in PH; and thus B is in \mathcal{C} by our assumption. Thus $B \leq_T^p A$. By Corollary 2.8 this implies that succ.SAT is in $(\Sigma_2^p)^A$, which is in the polynomial hierarchy, and hence is in NP.

However, this implies that $NEXP \subseteq NP$, which contradicts the Nondeterministic Time Hierarchy Theorem [SFM78].

Corollary 4.2. Let A be any set in the polynomial hierarchy. If $MCSP^A$ (or $MKTP^A$) is hard for $AC^0[6]$ under AC^0 reductions, then $AC^0[6] \neq NP$.

Recall that SZK denotes the class of languages with Statistical Zero-Knowledge proofs.

Corollary 4.3. Let A be any set in the polynomial hierarchy that is hard for SZK under \leq_T^p reductions. If $MCSP^A$ is hard for SZK under AC^0 reductions, then $SZK \neq NP$.

Proof. SZK is closed under complementation [Oka00]. Thus if NP is equal to the class of languages in SZK, then coNP = NP = SZK and PH collapses to SZK. Thus SZK satisfies the hypothesis of Theorem 4.1.

Similarly, we can state the following theorem about TC^0 reductions.

Theorem 4.4. Let C be any class such that if PP = C, then CH = C. If there is a set $A \in CH$ that is hard for C under \leq_T^p reductions such that $MCSP^A$ (or $MKTP^A$) is hard for C under uniform TC^0 reductions, then $PP \neq C$.

Proof. Suppose that $\mathsf{PP} = \mathcal{C}$, and that MCSP^A is hard for \mathcal{C} . Then, there exists a reduction from Maj.SAT (the standard complete problem for PP) to MCSP^A computable in TC^0 . Similarly to Corollary 3.2, this gives us a \leq_T^p reduction from succ.MajSAT to B-succ.MCSP^A for some $B \in \mathsf{CH}$; and thus B is in \mathcal{C} . Then, $B \leq_T^p A$, and thus succ.MajSAT is in $(\Sigma_2^p)^A$, which is in CH, and hence is in PP . However, succ.MajSAT is complete for probabilistic exponential time, and hence is not in PP .

Fenner, Fortnow, and Kurtz [FFK94] introduced several complexity classes, including SPP and WPP that are "low for PP", in the sense that $PP = PP^{SPP} = PP^{WPP}$. Thus we obtain the following corollary:

Corollary 4.5. Let A be any set in the counting hierarchy that is hard for WPP under \leq_T^p reductions. If MCSP^A is hard for WPP (or SPP) under uniform TC⁰ reductions, then WPP \neq PP (respectively SPP \neq PP).

5 Discussion

The contrast between Theorem 3.1 and Corollary 3.7 is stark. Theorem 3.1 obtains a very unsurprising consequence from the assumption that MCSP is hard for P under a very restrictive class of reductions, while Corollary 3.7 obtains a very unlikely collapse from the assumption that the apparently much harder problem MCSP^{QBF} is hard for P under a much less restrictive class of reductions. Yet, the absence of any known efficient reduction from MCSP to MCSP^{QBF} means that we have been unable to obtain any *unlikely* consequences by assuming that MCSP is hard for P. We believe that it should be possible to provide evidence that MCSP is not hard for P, and we pose this as an open question for further research.

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