# A Note on the Karp-Lipton Collapse for the Exponential Hierarchy 

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#### Abstract

We extend previous collapsing results involving the exponential hierarchy by using recent hardness-randomness trade-off results. Specifically, we show that if the second level of the exponential hierarchy has polynomialsized circuits, then it collapses all the way down to MA.


## Introduction

Much consideration has been given to the proposition that certain complexity classes may be Turing reducible to sparse sets. Equivalently, what happens if certain complexity classes have polynomially-sized (non-uniform) circuits? Such research has proven fruitful in giving evidence that such reductions and circuits do not exist for many interesting complexity classes.

The first such result, the Karp-Lipton collapse [6], showed that if NP $\subset P /$ poly then the entire polynomial hierarchy collapses to the second level $\left(\Sigma_{2}^{P} \cap \Pi_{2}^{P}\right)$. This collapse has since been improved ( $[7,2,3]$ Köbler \& Watanabe improved it to ZPP ${ }^{N P}$, Cai, with Sengupta, improved it to $S_{2}^{P}$, and Chakaravarthy \& Roy improved it further to $\mathrm{O}_{2}^{\mathrm{P}}$ ). In the same paper, they showed a stronger hypothesis results in a stronger collapse; that if EXP $\subset \mathrm{P} /$ poly then EXP $=\Sigma_{2}^{\mathrm{P}} \cap \Pi_{2}^{\mathrm{P}}$.

But what if even larger classes have polynomially sized circuits-do similar collapses occur? In fact, they do. Buhrman and Homer [1] strengthened the Karp-Lipton collapse to one higher level of the exponential hierarchy. This hierarchy is a natural exponential analog of the polynomial hierarchy inductively defined with NP oracles. That is, EXP, EXP ${ }^{N P}, E^{N} P^{N P^{N P}}$, etc; and NEXP, NEXP ${ }^{N P}, \operatorname{NEXP}^{N P^{N P}}$ along with their complements.

Theorem 1 (Buhrman \& Homer [1]).

$$
\mathrm{EXP}^{N P} \subset \mathrm{P} / \text { poly } \Rightarrow \mathrm{EXP}^{N P}=\Sigma_{2}^{\mathrm{P}} \cap \Pi_{2}^{P}
$$

In contrast, however, Kannan [5] was able to provably separate the exponential hierarchy from $P /$ poly. Specifically, he showed that any level of the exponential hierarchy above EXP ${ }^{N P}$ is not contained in $\mathrm{P} /$ poly. Improving this separation may be exceedingly difficult, as the oracle construction of Wilson [9] shows that EXP ${ }^{N P}$ has polynomial-sized (in fact linear-sized) circuits (relative to this oracle).

Research in the area of derandomization has used similar hypotheses to get conditional hardness-randomness trade-off results. That is, assuming the existence of a hard Boolean function (e.g. EXP $\not \subset \mathrm{P} /$ poly), one can construct pseudorandom generators from their truth table and derandomize some probabilistic complexity class like BPP or MA. A more recent result of Impagliazzo, Kabanets and Wigderson, shows that for the case of MA, such circuit complexity lower bounds are actually necessary for derandomization.

Theorem 2 (Impagliazzo, Kabanets \& Wigderson [4]).

$$
\mathrm{NEXP} \subset \mathrm{P} / \text { poly } \Leftrightarrow \mathrm{NEXP}=\mathrm{MA}
$$

We observe that the containments, $\mathrm{MA} \subseteq \Sigma_{2}^{P} \cap \Pi_{2}^{P} \subseteq E X P$, mean that the collapse also implies NEXP $=$ EXP.

## Main Result

The collapse in Theorem 2 is, in a sense, incomplete. In particular, it does not immediately follow that an inclusion in P / poly one higher level in the exponential hierarchy would cause a similar collapse. We extend this result by showing that such a collapse does indeed hold.

Definition 1. A language $L \subseteq\{0,1\}^{*}$ is in $\operatorname{EXP}^{N P}$ if it is decidable in deterministic exponential time with an oracle for NP. Additionally, $L \in \operatorname{EXP}^{\operatorname{NP}[z(n)]}$ if $L \in \operatorname{EXP}^{\mathrm{NP}}$ and $L$ is computable using at most $z(n)$ NP queries on inputs of length $n$.

Theorem 3. For any time-constructible function $z(n)$,

$$
\mathrm{EXP}^{\mathrm{NP}[z(n)]} \subset \mathrm{P} / \text { poly } \Rightarrow \mathrm{EXP}^{\mathrm{NP}[z(n)]}=\mathrm{EXP}
$$

It follows from Theorem 2 and standard hierarchy inclusions that this implies an even stronger collapse.

## Corollary 4.

$$
\mathrm{EXP}^{\mathrm{NP}[z(n)]} \subset \mathrm{P} / \text { poly } \Rightarrow \mathrm{EXP}^{\mathrm{NP}[z(n)]}=\mathrm{MA}
$$

Clearly, EXP $\subseteq \operatorname{EXP}^{N P[z(n)]}$ so it suffices to show that the assumption implies $\operatorname{EXP}^{\mathrm{NP}[z(n)]} \subseteq$ EXP. We proceed by proving a series of lemmas.

## Lemma 5.

$$
\mathrm{EXP}^{\mathrm{NP}[z(n)]} \subset \mathrm{P} / \text { poly } \Rightarrow \mathrm{NEXP} \subset \mathrm{P} / \text { poly }
$$

Proof. This follows from a simple padding argument; any set $A \in$ NEXP can be decided by an EXP machine with a single (though exponentially long) query to NP, i.e. we pad out the input $\left\langle x, 1^{2^{|x|}}\right\rangle$ in EXP time, then a query to NP (say to SAT) runs in polynomial time with respect to $\left|\left\langle x, 1^{|x|}\right\rangle\right|$. Thus NEXP $\subseteq \operatorname{EXP}^{N P[1]}$ and by the assumption, NEXP $\subset P /$ poly .
Lemma 6.

$$
\operatorname{EXP}^{\mathrm{NP}[z(n)]} \subset \mathrm{P} / \text { poly } \Rightarrow \mathrm{NEXP}=\mathrm{EXP}
$$

Proof. It follows from Lemma 5 and Theorem 2.
We are now able to mimic the argument of Krentel [8] who showed that any OptP function is computable by a $P$ machine with access to an NP oracle (i.e. OptP $=\mathrm{FP}^{\mathrm{NP}}$ ). For completeness, we give the following definitions which are also analogous to those presented in [8].

Definition 2. A NEXP metric Turing machine $N$ is a non-deterministic, exponentially time-bounded Turing machine such that every branch writes a binary number and accepts. For each $x \in \Sigma^{*}$ we write $\operatorname{Opt}_{N}(x)$ for the largest value on any branch of $N(x)$
Definition 3. A function $f: \Sigma^{*} \rightarrow \mathbb{N}$ is in OptEXP if there is a NEXP metric Turing machine such that $f(x)=\operatorname{Opt}_{N}(x)$ for all $x \in \Sigma^{*}$. The function $f$ is in $\operatorname{OptEXP}[z(n)]$ if $f \in \operatorname{OptEXP}$ and the length of $f(x)$ is bounded by $z(|x|)$ for all $x \in \Sigma^{*}$.
Lemma 7. Any $f \in \operatorname{EXP}^{\mathrm{NP}[z(n)]}$ can be computed as $f(x)=h(x, g(x))$ where $g \in \operatorname{OptEXP}[z(n)]$ and $h$ is computable in EXP time with respect to $|x|$. That is, $E X P^{N P}=O p t E X P$.

Proof. Let $f \in \operatorname{EXP}^{\mathrm{NP}[z(n)]}$ and $M$ be the machine computing $f$. Note that $M$ is an EXP machine making $z(n)$ queries to an NP set (without loss of generality, say SAT). Algorithm 1 presents a NEXP metric Turing machine $N$.

```
Input : \(x \in\{0,1\}^{n}\)
Compute \(z(n)\)
Non-deterministically branch for each \(y \in\{0,1\}^{z(n)}\)
Let \(y=b_{1} b_{2} \cdots b_{z(n)}\)
Simulate \(M(x)\), constructing queries \(\varphi_{1}, \varphi_{2}, \cdots, \varphi_{z(n)}\)
foreach \(\varphi_{i}\) such that \(b_{i}=1\) do
    Guess a satisfying assignment for \(\varphi_{i}\)
    if \(\varphi_{i} \in\) SAT then
            Output \(b_{1} b_{2} \cdots b_{z(n)}\)
    end
end
```

Algorithm 1: A NEXP metric Turing machine computing $b_{1} b_{2} \cdots b_{z(n)}$

The claim that $\mathrm{Opt}_{N}(x)=b_{1} b_{2} \cdots b_{z(n)}$ are the true oracle answers for $M(x)$ is shown by induction. Let $\varphi_{1}$ be the first query for $M$. If $\varphi_{1} \in$ SAT then $N(x)$ on branch $100 \cdots 00$ will find a satisfying assignment and so $\mathrm{Opt}_{N}(x) \geq 100 \cdots 00$ and so it must be the case that $b_{1}=1$. Conversely, if $\varphi_{1} \notin$ SAT then no branch beginning with 1 will find a satisfying assignment and so $\mathrm{Opt}_{N}(x) \leq 011 \cdots 11$ and $b_{1}=0$. By induction on $i$, all of the $b_{i}$ 's must be correct oracle answers for the computation of $M(x)$.

Therefore, given oracle answers $\operatorname{Opt}_{N}(x)=b_{1} b_{2} \cdots b_{z(n)}$, $f$ can be computed in EXP time by simulating $M(x)$ using the bits of $\mathrm{Opt}_{N}(x)$ for oracle answers. It follows, then, that $f$ can be computed by $h(x, g(x))$ with $g \in$ OptEXP and $h$ computable in EXP time.

Proof of Theorem 3. Assume $\operatorname{EXP}^{\mathrm{NP}[z(n)]} \subset \mathrm{P} /$ poly and let $f \in$ OptEXP computed by an OptEXP machine $M_{f}$. By Lemma 7 it suffices to show that $f$ can be computed in deterministic exponential time. Define the language $L_{M_{f}}=\left\{\langle x, y\rangle \mid x, y \in\{0,1\}^{*}, M_{f}(x)=y\right\}$. Note that $L \in$ NEXP: one can simply guess a (exponentially long) computation path of $M_{f}$ and accept if and only if $y$ is equal to the computed function value. By Lemma 6, the assumption implies that EXP $=\mathrm{NEXP}$ thus $L_{M_{f}} \in \operatorname{EXP}$.

Now consider the procedure in Algorithm 2. Here, we take the view that the polynomial advice string is a circuit. The assumption thus entails the existence of a circuit of size $p(n)$ for some fixed polynomial that computes $f$. We simply have to cycle through all possible circuits to find the right one. For each such circuit $C_{i}$, we must check that $M_{f}(x)=C_{i}(x)$.

```
Input : \(x \in\{0,1\}^{*}\)
forall Circuits \(C_{i}\) of size \(\leq p(n)\) do
    Compute \(y=C_{i}(x)\)
    if \(\langle x, y\rangle \in L_{M_{f}}\) then
            Store \(y\)
    end
end
7 Among the stored strings \(y\), take the lexicographically
maximum, \(y_{\max }\)
8 Output \(\mathrm{Opt}_{N}(x)=y_{\text {max }}\)
```

Algorithm 2: An EXP machine computing $f(x)$
The loop in Line 1 cycles through all circuits of size $\leq p(n)$ which can be done in exponential time. Furthermore, the subroutine for deciding $L_{M_{f}}$ is an EXP procedure by assumption and again, Lemma 6 . Thus, $f$ can be computed in deterministic exponential time and the conclusion follows.

We conclude by asking if current techniques can be combined in a more clever way to get an even bigger collapse. Can we show that EXP ${ }^{N P} \subset P /$ poly collapses EXP ${ }^{\mathrm{NP}}$ to an even smaller class such as $\mathrm{O}_{2}^{\mathrm{P}}$ ?

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