Kolmogorov Complexity, Circuits, and the Strength of Formal Theories of Arithmetic

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Abstract: Can complexity classes be characterized in terms of efficient reducibility to the (undecidable) set of Kolmogorov-random strings? Although this might seem improbable, a series of papers has recently provided evidence that this may be the case. In particular, it is known that there is a class of problems \( \mathcal{C} \) defined in terms of polynomial-time truth-table reducibility to \( R_K \) (the set of Kolmogorov-random strings) that lies between BPP and PSPACE [5, 4].

The results in this paper were obtained, as part of an investigation of whether this upper bound can be improved, to show

\[
\text{BPP} \subseteq \mathcal{C} \subseteq \text{PSPACE} \cap \text{P/poly}. \quad (\ast)
\]

In fact, we conjecture that \( \mathcal{C} = \text{BPP} = \text{P} \), and we close this paper with a discussion of the possibility this might be an avenue for trying to prove the equality of BPP and P.

In this paper, we present a collection of true statements in the language of arithmetic, (each provable in ZF) and show that if these statements can be proved in certain extensions...
of Peano arithmetic (PA), then (*) holds. Although it was subsequently proved that infinitely many of these statements are, in fact, independent of those extensions of PA [1], we present these results in the hope that related ideas may yet contribute to a proof of \( \mathcal{C} = \text{BPP} \), and because this work did serve as a springboard for subsequent work in the area [1].

1 Introduction

Kolmogorov complexity provides a mathematically precise definition of the set \( R \) of “random” strings. Actually, it provides at least two distinct but closely-related notions of randomness that we will need to discuss here, one defined in terms of the prefix Kolmogorov complexity function \( K \), and one defined in terms of the plain Kolmogorov complexity function \( C \).[1] This yields the two sets that lie at the center of this paper:

\[ R_K = \{ x : K(x) \geq |x| \} \quad \text{and} \quad R_C = \{ x : C(x) \geq |x| \}. \]

When it is not important to distinguish between \( K \) and \( C \) we will simply refer to \( R \).

It is known that \( \text{PSPACE} \subseteq \text{PR} \) [3], but it is unknown if any larger class such as \( \text{EXP} \) is in \( \text{PR} \).

In this paper we will focus especially on polynomial-time truth-table reductions (also known as non-adaptive reductions) \( \leq_{\text{p tt}} \); our motivation comes in part from a theorem of Buhrman et al., showing \( \text{BPP} \subseteq \{ A : A \leq_{\text{p tt}} R \} \) [5].

Because no larger complexity classes have been shown to be reducible to \( R \) in this way, we are interested in the question of whether these inclusions are optimal in some sense. It was observed earlier [2] that the class \( \{ A : A \leq_{\text{p}} R_C \} \) contains arbitrarily complex decidable sets: that is, for every computable time bound \( t \) there is a decidable set \( A \notin \text{DTIME}(t(n)) \) such that \( A \leq_{\text{p}} R_C \). Thus this does not look very much like a complexity class. But the same paper also suggested that a more promising avenue was to investigate the classes of problems that are always reducible to \( R \), no matter which universal Turing machine was used to define the Kolmogorov functions \( C \) and \( K \). This gives rise to the following classes:

**Definition 1.1.** As usual, let \( \Delta^0_1 \) denote the class of decidable sets. Let \( C_U \) denote the plain Kolmogorov complexity function as given by universal Turing machine \( U \), and let \( K_U \) denote the prefix complexity function as given by universal prefix Turing machine \( U \). Define

\[ C_C = \Delta^0_1 \cap \bigcap_U \{ A : A \leq_{\text{p}} R_{C_U} \}. \]

\[ C_K = \Delta^0_1 \cap \bigcap_U \{ A : A \leq_{\text{p}} R_{K_U} \}. \]

In each case, the intersection is taken over all universal Turing machines \( U \).2 See Section 2 for more background and definitions relating to Kolmogorov complexity.

The first upper bound on the complexity of sets in \( C_K \) was provided recently: \( C_K \subseteq \text{PSPACE} \) [4]. (We conjecture that similar bounds hold for \( C_C \), but at present it is still unknown whether \( C_C = \Delta^0_1 \).) Thus, in particular, we have

\[ \text{BPP} \subseteq C_K \subseteq \text{PSPACE} \subseteq \text{PR}. \]

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1We provide a brief introduction to some basic notions of Kolmogorov complexity in Section 2. For a more comprehensive introduction to the topic, we refer the reader to standard texts such as [13, 6].

2It was recently announced [14] that \( C_K \) is equal to \( \bigcap_U \{ A : A \leq_{\text{p}} R_{K_U} \} \). That is, any set in this intersection is decidable.
We conjecture that $C_K$ is actually equal to BPP. The results in this paper were proved as part of an attempt to establish a slightly weaker conjecture:

$$BPP \subseteq C_K \subseteq \text{PSPACE} \cap \text{P/poly}.$$ 

This, in turn, would follow from the following, which lies at the center of the present investigation:

**Conjecture 1.2.** $A = \{ A \in \Delta^0_1 : A \leq^{p,\text{tt}R}_U \} \subseteq \text{P/poly}$. That is, no matter which universal machine $U$ is used to define $C$ or $K$, $\{ A \in \Delta^0_1 : A \leq^{p,\text{tt}R}_C \} \subseteq \text{P/poly}$ and $\{ A \in \Delta^0_1 : A \leq^{p,\text{tt}R}_K \} \subseteq \text{P/poly}$.

Our main technical contribution is to build a set of formulas $\{ \Psi_A(n,j,k) \}_{A \in A}$ in the language of Peano Arithmetic, and for each $A \in A$ present a proof (which can be formalized in certain extensions of Zermelo-Frankel, or ZF) of the statement $\forall n \forall j \forall k \Psi_A(n,j,k)$. We then show that if for each $A \in A$, and each fixed tuple $(n,j,k)$, the true statement $\Psi_A(n,j,k)$ is provable in certain extensions of Peano Arithmetic, then Conjecture 1.2 holds.

At the time that the results in this paper were originally submitted for publication, we believed that a plausible approach to prove Conjecture 1.2 would be to show that proofs of $\Psi_A(n,j,k)$ exist in these extensions of PA. However, it is now known that infinitely many of the statements $\Psi_A(n,j,k)$ are independent of these extensions of PA [1]. In light of these developments, the reasons for publishing these results are:

- They served as a springboard for the subsequent work of [1], including results having a flavor similar to Conjecture 1.2, but in the context of Kolmogorov complexity with (large) time bounds,
- The connection between Conjecture 1.2 and proof theory is of some independent interest, and
- Perhaps a more sophisticated argument, based on our general approach, may yet serve to prove Conjecture 1.2.

Recall that, if Conjecture 1.2 holds, then $BPP \subseteq C_K \subseteq \text{PSPACE} \cap \text{P/poly}$. Thus we think that it is very reasonable to conjecture that $C_K = BPP$. But in fact we conjecture more. We believe that $C_C = C_K = P$. In fact, for limited classes of truth-table reductions, equalities of this form are known. In particular, it has been shown that $\Delta^1_1 \cap \bigcap_U \{ A : A \leq^{p,\text{tt}R}_U \} = \Delta^1_1 \cap \bigcap_U \{ A : A \leq^{p,\text{tt}R}_C \} = \text{P}$ [2].

In Section 7 we speculate about the possible advantages of pursuing this avenue toward the goal of proving $BPP = P$. At a minimum, we believe that our results raise the possibility that various mathematical techniques (e.g., from proof theory) might be relevant to the BPP vs. P problem, where such a connection may have seemed less likely before. Certainly the connection surprised some of the authors.

## 2 A Warm-Up Result

In this section we start with some basic definitions, and then present an easy theorem that provides intuition for Conjecture 1.2 and whose proof will help motivate some additional definitions.

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3We use the familiar convention that (bound or free) variables are represented by italic characters, as in “$\Psi_A(n,j,k)$” whereas bold-face characters are used, when a variable is replaced by a term of the form $1 + 1 + \cdots + 1$, as in “$\Psi_A(n,j,k)$”. 

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We say that a language $A$ polynomial-time truth-table reduces to a language $B$, denoted by $A \preceq^p_B$, if there exists a polynomial-time machine $M$ that computes $A$ when given $B$ as an oracle, with the additional requirement that, on input $x$, $M$ must compute the query set $Q(x)$ of all queries it will ask the oracle $B$ before receiving answers to any of its queries.

We will consider only truth-table reductions in this paper; as such we will write $M^A$ to indicate that machine $M$ is using a set $A$ as an oracle, and it will be implicit that the oracle access is non-adaptive.

The plain Kolmogorov complexity of a string $x$ with respect to a Turing machine $M$ is defined as $C_M(x) = \min \{|y| : M(y) = x\}$. A universal Turing machine is a machine $U$ such that for all $M$ and all $x$, $C_U(x) \leq C_M(x) + c_M$, where $c_M$ is a constant depending only on $M$. At times the choice of reference machine is not important as long as we choose a universal machine; when this is the case we fix some universal machine $U$ and write $C(x)$ in place of $C_U(x)$. We then define the Kolmogorov random strings to be the set $R_C = \{x : C(x) \geq |x|\}$.

In many settings where Kolmogorov complexity arises, it is more appropriate to use what is known as prefix complexity. A Turing machine $M$ is called a prefix machine, if, for any string $x$ on which $M$ halts, it is the case that $M$ does not halt on any string of the form $xy$ for any non-empty string $y$. That is, the domain of the machine must form a prefix code. Given such a prefix machine $M$, we define $K_M(x) = \min \{|y| : M(y) = x\}$. A universal prefix Turing machine is a prefix machine $U$ such that for all prefix machines $M$ and all $x$, $K_U(x) \leq K_M(x) + c_M$, where $c_M$ is a constant depending only on $M$. Similar to the case with plain complexity, we fix some universal prefix machine $U$ and write $K(x)$ in place of $K_U(x)$. We refer to the set of random strings under this version of Kolmogorov complexity as $R_K$.

All our theorems about random strings from this paper work for both $R_C$ and $R_K$; we will prove them with respect to $R_C$ and simply write $R$ for the set of random strings, but in Section 5 we indicate how to adjust the proofs to work for $R_K$ as well.

For a set $S$ of binary strings, let $S^{\leq k}$ be the set of all strings in $S$ that have length at most $k$; i.e., $S^{\leq k} = \bigcup_{i \leq k} S \cap \{0,1\}^i$. Let $V_k$ be the set of all sets of binary strings that only contain strings of length at most $k$; i.e., $V_k = \mathcal{P}\{0,1\}^{\leq k}$, where $\mathcal{P}$ denotes the powerset operation and $\{0,1\}^{\leq k}$ is shorthand for $\{(0,1)^*\}^{\leq k}$.

The complement of $R$ is computably-enumerable; therefore there is a Turing machine $E$ that outputs an enumeration $x_1,x_2,x_3,\ldots$ of all nonrandom strings. We define $R_{k,0} = \{0,1\}^{\leq k}$, and $R_{k,i}$ to be $R_{k,i-1} \setminus \{x_j\}$, where $x_j$ is the $i$th non-random string of length at most $k$ in the enumeration. One can view $R_{k,i}$ as an updated approximation to $R^{\leq k}$ after $i$ nonrandom strings of length at most $k$ have been discovered. Note that for some $i^*$, $R_{k,i^*} = R^{\leq k}$, and that for all $i > i^*$, $R_{k,i}$ is undefined, since there are no further nonrandom strings of length at most $k$ to be discovered. Even though $R_{k,i}$ is undefined for all $i > i^*$, in order to make the following proposition easier to read we state “$\forall i \in \mathbb{N} R_{k,i} \subseteq R_{k,i}$...” as a shorthand for “for all $i$ for which $R_{k,i}$ is defined, there exists a $V \subseteq R_{k,i}$...” We refer the reader to the Appendix, Section 8 for additional details regarding how to make precise certain details that we present at a more intuitive level.

**Proposition 2.1.** Let $A \in \mathcal{A}$, and let $M$ be a polynomial-time Turing machine running in time $f(n)$ computing a truth-table reduction from $A$ to $R$. Then

1. If $\exists d \forall n \exists V_n \in \mathcal{V}_{d+\log f(n)} \forall x \in \{0,1\}^n M^{V_n}(x) = A(x)$, then $A \in \mathcal{P}/\mathsf{poly}$.

2. $\exists d \forall n \forall x \in \{0,1\}^n \forall i \exists V \subseteq R_{d+\log f(n),i} M^V(x) = A(x)$. 


This proposition resembles an earlier observation from [2], but adds the condition in Part 2 that $V \subseteq R_{d+\log f(n),i}$.\footnote{Adding the condition “$V_n \subseteq R_{d+\log f(n),i}$” to Part 1 results in a false statement [1]! However, the hypothesis of Part 1, as stated, still seems plausible. Additional discussion of this point can be found in [1].} The following is an informal interpretation of the proposition. Part 1 states that if for each $n$ there is some oracle that (a) says that all “long” queries are nonrandom and (b) makes the reduction work for all $x$ of length $n$, then $A \in \text{P/poly}$. Part 2 says something similar to the hypothesis of Part 1, but weaker: although there might not be a single such oracle that works for all $x$, for every $x$ there is some such oracle that works for that $x$ (and furthermore is a subset of $R_{d+\log f(n),i}$). Thus, in some sense it is consistent for an oracle to say that all long queries are nonrandom, although this might entail giving incorrect answers to short queries; see Section 6 for more on this topic.

**Proof.** Part 1 is easy. On inputs of size $n$ the advice string is just an encoding of $V_n$. Because $|V_n| \leq 2^{d+\log f(n)+1}$, the advice can be encoded using $n^{O(1)}$ bits.

Now, we prove Part 2. Suppose, for the sake of contradiction, that $\forall d \exists n \exists x \in \{0,1\}^n \exists \forall V \subseteq R_{d+\log f(n),i} M^V(x) \neq A(x)$.

Let $Q(x')$ be the set of queries that $M$ asks on an input $x'$. Note that because $M$ runs in time $f(n)$, $|Q(x')| \leq f(|x'|)$. Let $T$ be the Turing machine that, on input $(d,r)$, does a dovetailing search until it finds some tuple $(n',x',i')$ such that for all $V \subseteq R_{d+\log f(n'),i'}, M^V(x') \neq A(x')$. (This is where we make use of the assumption that $A$ is decidable.) By our assumptions, it is guaranteed that $T$ will find such a tuple. $T$ then outputs the $r$th element of $Q(x')$.

The machine $T$ demonstrates that for all queries $z \in Q(x')$, $C(z) \leq 2\log d + \log f(n') + c_T$, where $c_T$ is some constant large enough to encode all the information needed to describe $T$, including $f,E,M$ and the algorithm $N$ that decides membership in $A$.

However, for the tuple $(n',x',i')$ that $T$ finds, the oracle $V^* = R_{d+\log f(n')}$ which agrees with $R$ on all short queries and says that all long queries are nonrandom causes $M$ to give the wrong answer on input $x'$, in the following sense:

- $V^* = R_{d+\log f(n')} \subseteq R_{d+\log f(n'),i'}$, and
- $M^{V^*}(x') \neq A(x') = M^R(x')$.

Because $M^{V^*}(x') \neq M^R(x')$, there must be some query $z \in Q(x')$ such that $z \in R$ and $|z| > d + \log f(n')$. However, we know that the Kolmogorov complexity of this $z$ is low, so for sufficiently large $d$ this is a contradiction: when $d$ is large enough, we have that $2\log d + \log f(n') + c_T < d + \log f(n')$. 

\hfill $\Box$

Here is an idea for how we can improve on Proposition 2.1, by deriving additional conclusions about $V$ in part 2 of Proposition 2.1. The condition that $V \subseteq R_{d+\log f(n),i}$ can be viewed as restricting the set of $V$’s that need to be considered; the proof relies only on the fact that $R_{d+\log f(n)}$ ends up being one of the possible $V$’s. Thus, in the proof of Proposition 2.1, as the machine $T$ enumerates nonrandom strings as part of its dovetailing search, we can view this process as $T$ “proving” that certain sets $V$ cannot be $R_{d+\log f(n)}$. But enumerating a nonrandom string $z$ such that $z \in V$ is not the only way to prove that a set $V$ is not $R_{d+\log f(n)}$. For instance, one can prove that for each $k$, a constant fraction of strings of length $k$ are in $R$ (see, e.g., [12]). Therefore, if the cardinality of a set $V$ is too small, one can prove that $V \neq R_{d+\log f(n)}$.
We consider the first-order system Peano Arithmetic (PA) augmented with additional axioms. We will be concerned with languages from the set \( \mathcal{A} = \{ A \in A_0^1 : A \leq^p R \} \). A language \( A \in \mathcal{A} \) will be encoded as a finite string \( \langle M, N \rangle \), where \( N \) is a Turing machine that computes \( A \), and \( M \) is a clocked polynomial-time Turing machine computing the truth-table reduction from \( A \) to \( R \). Note that any \( A \in \mathcal{A} \) can be specified by two such machines; for all \( A \in \mathcal{A} \) we fix some such encoding. For a fixed \( A \), we let \( t_A(n) \) denote an upper bound on the running time of \( M \), which is bounded by \( n^c \) for some constant \( c \).

For a given \( A \in \mathcal{A} \) encoded by \( \langle M, N \rangle \), PA may not be able to prove that \( N \) halts on every input, or that for all \( x \), \( M^R(x) = N(x) \). Therefore we define a predicate \( \text{Hyp}(A) \), which is an encoding of the sentence “\( \forall x N \) halts on input \( x \) and \( M^R(x) = N(x) \)”, corresponding to the hypothesis \( A \in \mathcal{A} \). For each \( A \in \mathcal{A} \), we define the system \( \text{PA}^A \) to be PA augmented with the additional axiom \( \text{Hyp}(A) \). Since \( \text{Hyp}(A) \) is true, \( \text{PA}^A \) is consistent if \( \text{PA} \) is.

We also define hierarchies based on these \( \text{PA}^A \) systems as follows. We define \( \text{PA}^A_0 \) to be \( \text{PA}^A \), and for each \( k > 0 \), define \( \text{PA}^A_k \) to be \( \text{PA}^A_{k-1} \) augmented with an extra axiom \( \text{con}(\text{PA}^A_{k-1}) \) stating that \( \text{PA}^A_{k-1} \) is consistent. We also define \( \text{PA}^A_0 \) to be \( \text{PA}^A \) augmented with an extra axiom encoding “For all \( k \), \( \text{con}(\text{PA}^A_k) \)”.

The statement of Part 2 of Proposition 2.1 says “\( \exists d \ldots \)” – but in fact we will find it useful to be much more explicit about the value of \( d \). The analysis of Part 2 of Proposition 2.1 works as long as we pick \( d \) so that \( c_T + 2 \log d < d \). In subsequent arguments we will use a similar style of reasoning, using slightly more complicated machines \( T \), and of course the choice of universal Turing machine \( U \) that is used to define Kolmogorov complexity will also contribute, but in all cases \( 2^{|\langle M, N \rangle| + |U| + 2^{25}} \) is a conservative over-estimate on the size of \( c_T \). Thus, if we define \( d_A \) to be \( 8(|\langle M, N \rangle| + |U| + 2^{25}) \), and we define \( g_A(n) \) to be \( d_A + \log t_A(n) \), then we can restate Part 2 of Proposition 2.1 as follows:

\[
\forall A \in \mathcal{A}, \\
\forall n \forall x \in \{0, 1\}^n \exists i : R_{g_A(n), i} \subseteq R_{g_A(n), i} \text{ such that } M^V(x) = A(x).
\]

Note that the proposition remains true, even if we replace \( g_A \) by a somewhat larger function. For technical reasons, we will find it useful to define \( g_A(n) \) to be \( d_A + 2 \log n + \log t_A(n) \).

### 3.2 Other definitions

For a set \( V \) we define \( L_A(n, V) \doteq \{ x \in \{0, 1\}^n : M^V(x) = N(x) \} \), where \( A \) is encoded as \( \langle M, N \rangle \) as described in the previous section. That is, \( L_A(n, V) \) is the set of all \( x \)'s of length \( n \) for which \( M \) computes the correct answer when \( V \) is substituted in as the oracle in the truth-table reduction in place of \( R \).
Theorem 4.1. Let $k$ various vertex that the hypothesis of Part 2 of Theorem 4.1 implies the hypothesis of Part 1 of Proposition 2.1.

Before giving the proof, the reader may wish to understand the relationship between Theorem 4.1 and Proposition 2.1. Part 1 of Theorem 4.1 is stronger than Part 2 of Proposition 2.1, by adding the condition that there not be a short proof of the sentence $\forall n V \neq R_{g_A(n), i}$. Part 2 of Theorem 4.1 is proved by showing that the hypothesis of Part 2 of Theorem 4.1 implies the hypothesis of Part 1 of Proposition 2.1.

4 Main Results

Our main focus in this paper is Conjecture 1.2, which we restate below.

Conjecture 1.2 $\{A \in \Delta^0_1 : A \leq^P R\} \subseteq P/poly$.

Before stating and proving our main theorem, which concerns a hierarchy of proof systems $PA_k^A$ for various $k$, we state and prove a simpler version that focuses on $PA^A$ and $PA_1^A$.

Theorem 4.1. Let $\Psi_A(n, j)$ be the formula $\forall i S_A(n, B_A(n, j, 0), R_{g_A(n), i}) = \{0, 1\}^n$.

1. For all $A \in \mathcal{A}$, the sentence $\forall n \forall j \Psi_A(n, j)$ is true and provable in $PA^A_1$.

2. If for all $A \in \mathcal{A}$, and each fixed pair $(n, j)$, $PA^A$ proves $\Psi_A(n, j)$, then Conjecture 1.2 is true.
Proof. (of Part 2)
Let \( \Phi_A(n, j, V) \) be the formula “If \( S_A(n, B_A(n, j, 0), V) \neq \{0, 1\}^n \) then \( \forall i \, V \neq R_{g_A(n), i} \)”. Note that
\[
\Psi_A(n, j) \rightarrow \Phi_A(n, j, V)
\]
and this implication is provable in \( PA^A \).

Suppose that for each \( A \in \mathcal{A} \) and each fixed pair \((n, j)\), \( PA^A \) proves \( \Psi_A(n, j) \).

Let \( A \in \mathcal{A} \) and \( \langle M, N \rangle \) be the encoding of \( A \). To prove Conjecture 1.2 we must show that \( A \in \text{P/poly} \).
Suppose for contradiction that \( A \not\in \text{P/poly} \). Then, for some \( n \), by Proposition 3.1 there does not exist a set \( V \in \mathcal{V}_{g_A(n)} \) such that \( L_A(n, V) = \{0, 1\}^n \).

Choose an \( n \) with this property. We define a directed graph \( G_n \) as follows. For each \( V \in \mathcal{V}_{g_A(n)} \) there is a node in \( G_n \). The graph \( G_n \) is leveled, with level \( h \) containing all \( V \)'s of cardinality \( h \). There is an edge from a node \( V \) to a node \( V' \) in \( G_n \) if and only if \( V \subset V' \) and \( |V'| = |V| + 1 \). Thus \( G_n \) is a rooted, layered, directed graph with the empty set as root.

We make use of the following claim:

**Claim 4.2.** For every \( V \in \mathcal{V}_{g_A(n)} \) there is a \( PA^A \) proof of the sentence \( \forall i \, V \neq R_{g_A(n), i} \).

**Proof.** The proof is by induction on \( |V| \).

For the basis case, when \( V = \emptyset \), a simple counting argument that can be formalized in \( PA^A \) proves that there are random strings of every length, and hence \( PA^A \) proves \( \forall i \, \emptyset \neq R_{g_A(n), i} \).

Now assume inductively that for all \( V' \in \mathcal{V}_{g_A(n)} \) such that \( |V'| < h \) there is a \( PA^A \) proof of the sentence \( \forall i \, V' \neq R_{g_A(n), i} \). Let \( V \in \mathcal{V}_{g_A(n)} \) with \( |V| = h \). To prove the claim, it suffices to show that there is a \( PA^A \) proof of the sentence \( \forall i \, V \neq R_{g_A(n), i} \).

By the inductive hypothesis, for some \( j' \), we have that \( \{V' : V' \in \mathcal{V}_{g_A(n)} \land |V'| < h\} \subseteq B_A(n, j', 0). \)
Since, in the graph \( G_n - B_A(n, j', 0) \), \( V \) has indegree zero, it follows from the definition of \( S_A(\cdot, \cdot, \cdot) \) that \( PA^A \) proves
\[
S_A(n, B_A(n, j', 0), V) = L_A(n, V),
\]
and by the choice of \( n \) we have \( L_A(n, V) \neq \{0, 1\}^n \). Hence \( PA^A \) proves that \( S_A(n, B_A(n, j', 0), V) \neq \{0, 1\}^n \).
By assumption we have that \( PA^A \) proves \( \Psi_A(n, j') \), so by (4.1) we have that \( PA^A \) proves
\[
\text{“If } S_A(n, B_A(n, j', 0), V) \neq \{0, 1\}^n \text{ then } \forall i \, V \neq R_{g_A(n), i} \text{”}.
\]
Therefore \( PA^A \) proves \( \forall i \, V \neq R_{g_A(n), i} \). \( \square \)

Therefore, by Claim 4.2 we have that \( PA^A \) proves \( \forall i \{0, 1\} \leq g_A(n) \neq R_{g_A(n), i} \). However, by definition, \( \{0, 1\} \leq g_A(n) = R_{g_A(n), 0} \), which implies that \( PA^A \) is inconsistent. By the consistency of \( PA^A \) (which is provable in, say, \( ZF^A \)), we therefore get a contradiction. Thus we conclude that \( A \) is in \( \text{P/poly} \). \( \square \)

**Proof.** (of Part 1)
Let \( A \in \mathcal{A} \) be encoded by \( \langle M, N \rangle \), and suppose for contradiction that there exists \( (n, j) \) such that \( \neg \Psi_A(n, j) \).

This implies that for some \( i, S_A(n, B_A(n, j, 0), R_{g_A(n), i}) \neq \{0, 1\}^n \).

Let \( T \) be the following machine. On input \( (n, r) \), \( T \) does a dovetailing search until it finds some tuple \((x, j', i')\) such that \( x \) is a string of length \( n \) that is not in \( S_A(n, B_A(n, j', 0), R_{g_A(n), j'}) \). \( PA^A \) can argue that
under the assumptions, $T$ is guaranteed to find such a tuple. $T$ then computes $Q(x)$, and outputs the $r$th element of $Q(x)$.

The input $(n, r)$ to $T$ has length at most $2 \log n + \log t_A(n)$. By the discussion at the end of Section 3.1, this implies that for all queries $z \in Q(x)$, $C(z) \leq g_A(n)$, so there can be no $z \in Q(x) \cap R$ such that $|z| > g_A(n)$. Thus $PA^A$ can argue that $M^{R}(x) = M^{R \circ g_A}(x)$, since these oracles answer all queries of length at most $g_A(n)$ identically, and by the previous sentence they answer queries from $Q(x)$ of length greater than $g_A(n)$ identically as well.

Therefore $PA^A$ can argue the following points:

- $\exists r < 2^{g_A(n)+1} R^{\leq g_A(n)} = R_{g_A(n), j'}$.
- $\exists V^* \in \forall g_A(n) V^* = R_{g_A(n), j'}$.
- $A(x) = M^R(x) = M^{R \circ g_A}(x) = M^{V^*}(x)$.
- If $V^* \not\in B_A(n, j', 0)$ then $x \in S_A(n, B_A(n, j', 0), R_{g_A(n), j'})$.

(The last item follows from the others together with the definition of $S_A(\cdot, \cdot, \cdot, \cdot, \cdot)$.)

Therefore, since from the way $x$ was obtained we also have that $x \not\in S_A(n, B_A(n, j', 0), R_{g_A(n), j'})$, $PA^A$ can conclude that $V^*$ is in $B_A(n, j', 0)$. From the definition of $B_A(n, j', 0)$ this means that $PA^A$ can conclude that there is a length $j'$ proof in $PA^A$ of the sentence $\forall i V^* \neq R_{g_A(n), j}$.

However, we have that $V^* = R_{g_A(n), j'}$, and as the relation $R(k, V, i)$ with intended meaning “$V = R_{k, i}$” can be defined by a $\Sigma^0_1$ formula, $PA^A$ can conclude that there is a $PA^A$ proof of $V^* = R_{g_A(n), j'}$. (See the Appendix, Section 8, for more details on this.) Therefore $PA^A$ proves that $PA^A$ is inconsistent. In $PA^A$ this gets us very little, but in $PA_1$ this is a contradiction. Thus $PA^A$ proves $\forall n \forall j \forall \Psi_A(n, j)$.

Already when we originally submitted this work for publication, we recognized that it was reasonably likely that the hypothesis of Part 2 of Theorem 4.1 is false. (Subsequently this was shown by [1].) But we held out hope that, by considering stronger theories, this obstacle could be avoided. This led us to our main theorem:

**Theorem 4.3.** Let $\Psi_A(n, j, k)$ be the formula $\forall i S_A(n, B_A(n, j, k), R_{g_A(n), i}) = \{0, 1\}^n$.

1. For all $A \in A$, the sentence $\forall n \forall j \forall k \Psi_A(n, j, k)$ is true and provable in $PA^A_{\omega_1}$.  

2. If for all $A \in A$ and each fixed tuple $(n, j, k)$ there exists an $l$ such that $PA^A_l$ proves $\Psi_A(n, j, k)$, then Conjecture 1.2 is true.

**Proof.** The proof of Part 2 is almost identical to that of Theorem 4.1, and we omit it here.  

The proof of Part 1 is very similar to that of Theorem 4.1 as well, but we include it here for completeness.

Let $A \in A$ be encoded by $\langle M, N \rangle$ and suppose for contradiction that there exists $(n, j, k)$ such that $-\Psi_A(n, j, k)$.

This implies that for some $i$, $S_A(n, B_A(n, j, k), R_{g_A(n), i}) \not= \{0, 1\}^n$.  


Let \( T \) be the following machine. On input \((n, r)\), \( T \) does a dovetailing search, until it finds some tuple \((x, j', k', \overline{t})\) such that \( x \) is a string of length \( n \) that is not in \( S_A(n, B_A(n, j', k'), R_{g_A(n), \overline{t}}) \). PA\(^A\) can argue that under the assumptions, \( T \) is guaranteed to find such a tuple. \( T \) then computes \( Q(x) \), and outputs the \( r \)th element of \( Q(x) \).

The input \((n, r)\) to \( T \) has length at most \( 2 \log n + \log t_A(n) \). By the discussion at the end of Section 3.1, this implies that for all queries \( z \in Q(x) \), \( C(z) \leq g_A(n) \), so there can be no \( z \in Q(x) \cap R \) such that \(|z| > g_A(n)\). Thus \( PA^A \) can argue that \( M^R(x) = M^{R \leq t_A(n)} \), since these oracles answer all queries of length at most \( g_A(n) \) identically, and by the previous sentence they answer queries from \( Q(x) \) of length greater than \( g_A(n) \) identically as well.

Thus \( PA^A \) can argue the following points:

- \( \exists i^* < 2^{g_A(n) + 1} \) \( R^{\leq g_A(n)} = R_{g_A(n), i^*}. \)
- \( \exists V^* \in V_{g_A(n)} V^* = R_{g_A(n), i^*}. \)
- \( A(x) = M^R(x) = M^{R \leq g_A(n)}(x) = M^{V^*}(x). \)
- \( V^* \notin B_A(n, j', k') \) implies \( x \in S_A(n, B_A(n, j', k'), R_{g_A(n), \overline{t}}). \)

(The last item follows directly from the definition of \( S_A(\cdot, \cdot, \cdot) \), along with the preceding items.)

Thus, since from the way \( x \) was obtained we also have that \( x \) is not in the set \( S_A(n, B_A(n, j', k'), R_{g_A(n), \overline{t}}) \), \( PA^A \) can conclude that \( V^* \) is in \( B_A(n, j', k') \). From the definition of \( B_A(n, j', k') \) this means that \( PA^A \) can conclude that there is a length \( j' \) proof in \( PA^A_{\overline{t}} \) of the sentence \( \forall i V^* \neq R_{g_A(n), i}. \)

However, we have that \( V^* = R_{g_A(n), i^*} \), and as the relation \( R(k, V, i) \) with intended meaning “\( V = R_{k, i} \)” can be defined by a \( \Sigma^0_1 \) formula, \( PA^A \) can conclude that there is \( PA^A_{\overline{t}} \) proof of \( V^* = R_{g_A(n), i^*}. \) Therefore \( PA^A \) proves that \( PA^A_{\overline{t}} \) is inconsistent. In \( PA^1_B \), for fixed \( l \), there is not much we can conclude from this, since it is not clear how to bound \( k' \) by any fixed number. But in \( PA^A_\omega \) this is a contradiction. Thus \( PA^A_\omega \) proves \( \forall n \exists j' \forall k \Psi_A(n, j, k). \)

\( \square \)

5 Adapting to Prefix Complexity

The results in the preceding section were proved with respect to the plain Kolmogorov complexity function \( C \). Here, we provide a few comments regarding how to adapt the arguments, so that they carry over to the prefix Kolmogorov complexity function \( K \).

Briefly, the descriptions of the elements \( y \) of \( Q(x) \) need to be presented as a prefix-free code. The descriptions that we used were of the form \((P, n, r)\) where we can think of \( P \) as being a “program”, and \( n \) and \( r \) are numbers. In the analysis for plain complexity, we gave an upper bound on the length of these descriptions, of the form \( g_A(n) = d_A + 2 \log n + \log t_A(n) \) (and we remarked that the analysis also would carry through if a slightly larger value of \( g_A(n) \) were used).

The term “\( 2 \log n \)” in this expression comes from the fact that we need to encode the “comma” between \( n \) and \( r \) in some way, and a very simple way to do this is to simply double each bit of the number \( n \), and then mark the end of “\( n \)” with a pair (either 01 or 10) that is not doubled.
If we similarly double each bit of $r$ and mark the end of $r$, then we will obtain a prefix-free encoding scheme, and the analysis will carry through if we just define $g_A(n)$ to be $d_A + 2\log n + 2\log t_A(n)$.

6 Epilogue

Two months after this work was originally submitted for publication, Buhrman and Loff proved some results that bear directly upon our investigation; see [1]. Buhrman and Loff had read a preliminary version of our paper, and sought to give an unconditional proof of Conjecture 1.2. Although this conjecture is still open, one of the theorems in [1] can be seen as lending additional support to the conjectured P/poly upper bound on the class of decidable sets polynomial-time truth-table reducible to $R$. For a polynomial-time reduction from a decidable set $A$ to the undecidable set $R$, it seems reasonable to hypothesize that the reduction would also work if one used a very high time-complexity approximation to $R$, such as $R_{K}^{r(n)}$ for some very rapidly-growing time bound $t(n)$. Buhrman and Loff have shown that, for each decidable set $A$ and polynomial-time truth-table reduction $M$, if it is the case that $M$ reduces $A$ to $R_{K}^{r(n)}$ for every large-enough time bound $t$, then $A \in$ P/poly.

In addition, however – the techniques used by Buhrman and Loff also immediately yield that many of the sentences $\Psi(n,j,k)$ considered here are, in fact, independent of $PA_\ell$ for every $\ell$. Moreover, they present a polynomial-time reduction $M_0$ with the property that it can not be directly replaced by a reduction that makes queries only of length $O(\log n)$, having as oracle a subset of $R$. Thus the general approach discussed in here will need to be revised substantially, if it is to be used to obtain a P/poly upper bound on $\{A \in \Delta^0_0 : A \leq^p R\}$.

The reduction $M_0$ alluded to in the preceding paragraph has the property that it obtains no useful information from the oracle. Thus it is still conceivable that one can formulate a notion of “useful” truth-table reductions, for which it still might hold that, for each length $n$, there is a set of short random strings $V$ that can be used as an oracle to cause the reduction to give the correct answer for all strings of length $n$. However, it is far from clear how to formulate such a definition.

7 Why Care?

It is popular these days to conjecture that $BPP = P$, and much of this popularity is owing to results such as those of Impagliazzo and Wigderson [11], who showed that $BPP = P$ if there is a problem in $E$ that requires circuits of exponential size. But note that a proof that $BPP = P$ that proceeds by first proving circuit size lower bounds yields much more than “merely” a proof that $BPP = P$. It also provides a recipe that one can follow, to start with an arbitrary probabilistic algorithm and replace it with an equivalent deterministic one of comparable complexity.

Indeed, Goldreich has recently argued that any proof of $BPP = P$ must proceed along these lines, in that any proof that these classes are equal yields pseudorandom generators that are suitable for derandomizing $BPP$ [8, 9].

But there is a catch! Goldreich’s proof requires that the $BPP = P$ question be phrased in terms of promise problems, rather than using the more traditional definition in terms of language classes, that we have used here.
We do not dispute Goldreich’s assertion that the formulation in terms of promise problems is in many ways more natural and useful than the traditional definition. And we certainly agree that it would be much more useful to have a recipe for obtaining derandomizations, rather than merely a proof that a derandomization must exist. But we find it intriguing that a proof that $\mathsf{C}_C = \mathsf{P}$ would prove that $\mathsf{BPP} = \mathsf{P}$ merely by showing that there would be a contradiction otherwise, and owing to the highly non-computable objects in the definition, it is not clear that such a proof would lend itself to an effective construction of a general-purpose derandomization algorithm. (In particular, it is not clear that it would yield the equality of the promise classes.) That is, since such a proof would deliver less than a proof that yields a derandomization, it is at least conceivable that it would be easier to obtain.

We do not wish to suggest that we have any idea of how to obtain such a proof. After all, we are currently unable even to prove $\mathsf{C}_K \subseteq \mathsf{P/poly}$.

Also, it is clear that such a proof must use nonrelativizing techniques. For instance, the work of [5] shows that, for any decidable oracle $B$, $\mathsf{BPP}^B$ is $\mathsf{P}^B$-truth-table reducible to $R_{K_U}$ for every $U$. (There is no need to add an oracle to the definition of $R_{K_U}$.) Thus it is not true that, for every decidable $B$, $\Delta^0_1 \cap \bigcap_U \{A : A \leq^P_{\mathsf{R}_{K_U}} R_{K_U}\} = \mathsf{P}^B$, because Heller [10] has presented such a $B$ relative to which $\mathsf{BPP}^B = \mathsf{NEXP}^B$.

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References


Appendix: Further Encoding Details

Throughout this paper, for the sake of readability, we have presented informally proofs meant to be in formal systems. In this section we attempt to clarify the formalization of a couple key definitions in these proofs.

An important definition, introduced in Section 2, is the definition of the set $R_{k,i}$. Formally, we define $R_{k,i}$ by means of a relation $R(V,k,i)$ that is TRUE if and only if the set $V$ is equal to $R_{k,i}$. (Of course $R$ takes as input an encoding $\langle V \rangle$ of the set $V$, but we will continue to abuse notation in this way). The quantifier complexity of the formula used to define this relation plays an important role. At the end of the proof of Part 1 of Theorem 4.1, we state that $PA^A$ proves the implication $\forall R(V^*,g_A(n),i^*) \rightarrow PA^A \vdash R(V^*,g_A(n),i^*)$ (a similar statement occurs in Theorem 4.3 as well). Here $PA^A \vdash R(V^*,g_A(n),i^*)$ is shorthand for a formula encoding that $R(V^*,g_A(n),i^*)$ is provable in $PA^A$. That this implication involving $PA^A$ actually is provable in $PA^A$ itself depends on $R(V,k,i)$ being definable by a $\Sigma^0_1$ formula; i.e., one that can be expressed as $\exists \bar{x} R'(\bar{x},V,k,i)$, where $R'(\bar{x},V,k,i)$ is a formula containing only bounded quantifiers. (See, for example, [7, Theorems 1.3.4 and 1.4.7] for a proof of this fact.)

Below we show that $R(V,k,i)$ can in fact be defined by a $\Sigma^0_1$ formula:
Here, $T(U, k, i, y)$ is a formula expressing that $y$ is the transcript of a halting execution of machine $U$ on input $(k, i)$, where $U$ is the Turing machine that takes as input $(a, b)$ and enumerates the first $b$ nonrandom strings of length at most $a$. (If there do not exist $b$ nonrandom strings of length at most $a$ then no $y$ will satisfy the formula). Also, $out(w, y)$ expresses that $w$ is the output of the execution with transcript $y$, and $w_j$ stands for the $j$th element of $w$ (viewing $w$ as a list of strings).

It is standard that the formula $T(U, k, i, y)$ can be defined by a formula containing only bounded quantifiers.

Note that with the definition $R(V, k, i)$ in hand, we can express a predicate $Z(V, k)$ with intended meaning “$V = R^{\leq k}$” as:

$$Z(V, k) \doteq \exists i^* \leq 2^{k+1} R(V, k, i^*) \land \forall V' \in V_k \neg R(V', k, i^* + 1).$$

Of course, this predicate $Z(V, k)$ is not $\Sigma_1^0$, but it is sufficient for our purposes that $R(V, k, i)$ is.
Kolmogorov Complexity, Circuits, and the Strength of Formal Theories of Arithmetic

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