# A New Approach to Formal Language Theory by Kolmogorov Complexity* 

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#### Abstract

We present a new approach to formal language theory using Kolmogorov complexity. The main results presented here are an alternative for pumping lemma(s), a new characterization for regular languages, and a new method to separate deterministic context-free languages and nondeterministic context-free languages. The use of the new 'incompressibility arguments' is illustrated by many examples. The approach is also successful at the high end of the Chomsky hierarchy since one can quantify nonrecursiveness in terms of Kolmogorov complexity. (This is a preliminary uncorrected version. The final version is the one published in SIAM J. Comput., 24:2(1995), 398-410.)


## 1 Introduction

It is feasible to reconstruct parts of formal language theory using algorithmic information theory (Kolmogorov complexity). We provide theorems on how to use Kolmogorov complexity as a concrete and powerful tool. We do not just want

[^0]to introduce fancy mathematics; our goal is to help our readers do a large part of formal language theory in the most essential, usually easiest, sometimes even obvious ways. In this paper it is only important to us to demonstrate that the application of Kolmogorov complexity in the targeted area is not restricted to trivialities. The proofs of the theorems in this paper may not be easy. However, the theorems are of the type that are used as a tool. Once derived, our theorems are easy to apply.

### 1.1 Prelude

The first application of Kolmogorov complexity in the theory of computation was in 19, 20. By re-doing proofs of known results, it was shown that static, descriptional (program size) complexity of a single random string can be used to obtain lower bounds on dynamic, computational (running time) complexity. None of the inventors of Kolmogorov complexity originally had these applications in mind. Recently, Kolmogorov complexity has been applied extensively to solve classic open problems of sometimes two decades standing, 16, 12, 9, 10. For more examples see the textbook 13.

The secret of Kolmogorov complexity's success in dynamic, computational lower bound proofs rests on a simple fact: the overwhelming majority of strings has hardly any computable regularities. We call such a string 'Kolmogorov random' or 'incompressible'. A Kolmogorov random string cannot be (effectively) compressed. Incompressibility is a noneffective property: no individual string, except finitely many, can be proved incompressible.

Recall that a traditional lower bound proof by counting usually involves all inputs of certain length. One shows that a certain lower bound has to hold for some 'typical' input. Since an individual typical input is hard (sometimes impossible) to find, the proof has to involve all the inputs. Now we understand that a typical input of each length can be constructed via an incompressible string. However, only finitely many individual strings can be effectively proved to be incompressible. No wonder the old counting arguments had to involve all inputs. In a proof using the new 'incompressibility method', one uses an individual incompressible string that is known to exist even though it cannot be constructed. Then one shows that if the assumed lower time bound would not hold, then this string could be compressed, and hence it would not be incompressible.

### 1.2 Outline of the Paper

The incompressibility argument above also works for formal languages and automata theory proper. Assume the basic notions treated in a textbook like [7].

The first result is a powerful alternative to pumping lemmas for regular languages. It is well known that not all nonregular languages can be shown to be
nonregular by the usual uvw-pumping lemma. There is a plethora of pumping lemmas to show nonregularity, like the 'marked pumping lemma', and so on. In fact, it seems that many example nonregular languages require their own special purpose pumping lemmas. Comparatively recently, \&, 22, 4, exhaustive pumping lemmas that characterize the regular languages have been obtained.

These pumping lemmas are complicated and complicated to use. The last reference uses Ramsey theory. In contrast, using Kolmogorov complexity we give a new characterization of the regular languages that simply makes our intuition of 'finite state'ness of these languages rigorous and is easy to apply. Being a characterization it works for all non-regular languages. We give several examples of its application, some of which were quite difficult using pumping lemmas.

To prove that a certain context-free language (cfl) is not deterministic contextfree (dcfl) has required laborious ad hoc proofs, [7] or cumbersome-to-state and also difficult-to-apply pumping lemmas or iteration theorems [5, 25]. We give necessary (Kolmogorov complexity) conditions for dcfl, that are very easy to apply. We test the new method on several examples in cfl - dcfl, which were hard to handle before. In certain respects the KC-DCFL Lemma may be more powerful than the related lemmas and theorems mentioned above. On the high end of the Chomsky hierarchy we present, for completeness, a known characterization of recursive languages, and a necessary condition for recursively enumerable languages.

## 2 Kolmogorov Complexity

From now on, let $x$ denote both the natural number and the $x$ th binary string in the sequence $0,1,00,01,10,11,000, \ldots$ That is, the representation ' 3 ' corresponds both to the natural number 3 and to the binary string 00 . This way we obtain a natural bijection between the nonnegative integers $\mathcal{N}$ and the finite binary strings $\{0,1\}^{*}$. Numerically, the binary string $x_{n-1} \ldots x_{0}$ corresponds to the integer

$$
\begin{equation*}
2^{n}-1+\sum_{i=0}^{n-1} x_{i} 2^{i} \tag{1}
\end{equation*}
$$

We use notation $l(x)$ to denote the length (number of bits) of a binary string $x$. If $x$ is not a finite binary string but another finite object like a finite automaton, a recursive function, or a natural number, then we use $l(x)$ to denote the length of its standard binary description. Let $\langle\cdot, \cdot\rangle: \mathcal{N} \times \mathcal{N} \rightarrow \mathcal{N}$ be a standard recursive, invertible, one-one encoding of pairs of natural numbers in natural numbers. This idea can be iterated to obtain a pairing from triples of natural numbers with natural numbers $\langle x, y, z\rangle=\langle x,\langle y, z\rangle\rangle$, and so on.

Any of the usual definitions of Kolmogorov complexity in [11, 20, 13] will do for the sequel. We are interested in the shortest effective description of a finite
object $x$. To fix thoughts, consider the problem of describing a string $x$ over 0's and 1's. Let $T_{1}, T_{2}, \ldots$ be the standard enumeration of Turing machines. Since $T_{i}$ computes a partial recursive function $\phi_{i}: \mathcal{N} \rightarrow \mathcal{N}$ we obtain the standard enumeration $\phi_{1}, \phi_{2}, \ldots$ of partial recursive functions. We denote $\phi(\langle x, y\rangle)$ as $\phi(x, y)$. Any partial recursive function $\phi$ from strings over 0's and 1's to such strings, together with a string $p$, the program for $\phi$ to compute $x$, such that $\phi(p)=x$, is a description of $x$. It is useful to generalize this idea to the conditional version: $\phi(p, y)=x$ such that $p$ is a program for $\phi$ to compute $x$, given a binary string $y$ for free. Then the descriptional complexity $C_{\phi}$ of $x$, relative to $\phi$ and $y$, is defined by

$$
C_{\phi}(x \mid y)=\min \left\{l(p): p \in\{0,1\}^{*}, \phi(p, y)=x\right\}
$$

or $\infty$ if no such $p$ exists.
For a universal partial recursive function $\phi_{0}$, computed by the universal Turing machine $U$, we know that, for each partial recursive function $\phi$, there is a constant $c_{\phi}$ such that for all strings $x, y$, we have $\phi_{0}(i, x, y)=\phi(x, y)$. Hence, $C_{\phi_{0}}(x \mid y) \leq C_{\phi}(x \mid y)+c_{\phi}$. We fix a reference universal function $\phi_{0}$ and define the conditional Kolmogorov complexity of $x$ given $y$ as $C(x \mid y)=C_{\phi_{0}}(x \mid y)$. ${ }^{\dagger}$

The unconditional Kolmogorov complexity of $x$ is $C(x)=C(x \mid \epsilon)$, where $\epsilon$ denotes the empty string $(l(\epsilon)=0)$.

Since there is a Turing machine that just copies its input to its output we have $C(x \mid y) \leq l(x)+O(1)$, for each $x$ and $y$. Since there are $2^{n}$ binary strings of length $n$, but only $2^{n}-1$ possible shorter descriptions $d$, it follows that $C(x) \geq l(x)$ for some binary string $x$ of each length. We call such strings incompressible or random. It also follows that, for any length $n$ and any binary string $y$, there is a binary string $x$ of length $n$ such that $C(x \mid y) \geq l(x)$. Considering $C$ as an integer function, using the obvious one-one correspondence between finite binary words and nonnegative integers, it can be shown that $C(x) \rightarrow \infty$ for $x \rightarrow \infty$. Finally, $C(x, y)$ denotes $C(\langle x, y\rangle)$.

Example 1 (Self-Delimiting Strings) A prefix code is a mapping from finite binary code words to source words, such that no code word is a proper prefix of any other code word. We define a particular prefix code.

For each binary source word $x=x_{1} \ldots x_{n}$, define the code word $\bar{x}$ by

$$
\bar{x}=1^{l(x)} 0 x .
$$

Define

$$
x^{\prime}=\overline{l(x)} x
$$

The string $x^{\prime}$ is called the self-delimiting code of $x$.
Set $x=01011$. Then, $l(x)=5$, which corresponds to binary string ' 10 ', and $\overline{l(x)}=11010$. Therefore, $x^{\prime}=1101001011$ is the self-delimiting code of '01011'.

[^1]The self-delimiting code of a positive integer $x$ requires $l(x)+2 \log l(x)+1$ bits. It is easy to verify that $l(x)=\lfloor\log (x+1)\rfloor$. All logarithms are base 2 unless otherwise noted. For convenience, we simply denote the length $l(x)$ of a natural number $x$ by ' $\log x$ '.

Example 2 (Substrings of incompressible strings) Is a substring of an incompressible string also incompressible? A string $x=u v w$ can be specified by a short description for $v$ of length $C(v)$, a description of $l(u)$, and the literal description of $u w$. Moreover, we need information to tell these three items apart. Such information can be provided by prefixing each item with a self-delimiting description of its length. Together this takes $C(v)+l(u w)+O(\log l(x))$ bits. Hence,

$$
C(x) \leq C(v)+O(\log l(x))+l(u w)
$$

Thus, if we choose $x$ incompressible, $C(x) \geq l(x)$, then we obtain

$$
C(v) \geq l(v)-O(\log l(x))
$$

It can be shown that this is optimal - a substring of an incompressible string of length $n$ can be compressible by an $O(\log n)$ additional term. This conforms to a fact we know from probability theory: every random string of length $n$ is expected to contain a run of about $\log n$ consecutive zeros (or ones). Such a substring has complexity $O(\log \log n)$.

## 3 Regular Sets and Finite Automata

Definition 1 Let $\Sigma$ be a finite nonempty alphabet, and let $Q$ be a (possibly infinite) nonempty set of states. A transition function is a function $\delta: \Sigma \times Q \rightarrow$ $Q$. We extend $\delta$ to $\delta^{\prime}$ on $\Sigma^{*}$ by $\delta^{\prime}(\epsilon, q)=q$ and

$$
\delta^{\prime}\left(a_{1} \ldots a_{n}, q\right)=\delta\left(a_{n}, \delta^{\prime}\left(a_{1} \ldots a_{n-1}, q\right)\right)
$$

Clearly, if $\delta^{\prime}$ is not $1-1$, then the automaton 'forgets' because some $x$ and $y$ from $\Sigma^{*}$ drive $\delta^{\prime}$ into the same memory state. An automaton $A$ is a quintuple $\left(\Sigma, Q, \delta, q_{0}, q_{f}\right)$ where everything is as above and $q_{0}, q_{f} \in Q$ are distinguished initial state and final state, respectively. We call $A$ a finite automaton (fa) if $Q$ is finite.

We denote 'indistinguishability' of a pair of histories $x, y \in \Sigma^{*}$ by $x \sim y$, defined as $\delta^{\prime}\left(x, q_{0}\right)=\delta^{\prime}\left(y, q_{0}\right)$. 'Indistinguishability' of strings is reflexive, symmetric, transitive, and right-invariant $\left(\delta^{\prime}\left(x z, q_{0}\right)=\delta^{\prime}\left(y z, q_{0}\right)\right.$ for all $\left.z\right)$. Thus, 'indistinguishability' is a right-invariant equivalence relation on $\Sigma^{*}$. It is a simple matter to ascertain this formally.

Definition 2 The language accepted by automaton $A$ as above is the set $L=$ $\left\{x: \delta^{\prime}\left(x, q_{0}\right)=q_{f}\right\}$. A regular language is a language accepted by a finite automaton.

It is a straightforward exercise to verify from the definitions the following fact (which will be used later).

Theorem 1 (Myhill, Nerode) The following statements about $L \subseteq \Sigma^{*}$ are equivalent.
(i) $L \subseteq \Sigma^{*}$ is accepted by some finite automaton.
(ii) $L$ is the union of equivalence classes of a right-invariant equivalence relation of finite index on $\Sigma^{*}$.
(iii) For all $x, y \in \Sigma^{*}$ define right-invariant equivalence $x \sim y$ by: for all $z \in \Sigma^{*}$ we have $x z \in L$ iff $y z \in L$. Then the number of $\sim$-equivalence classes is finite.

Subsequently, closure of finite automaton languages under complement, union, and intersection follow by simple construction of the appropriate $\delta$ functions from given ones. Details can be found in any textbook on the subject like [7]. The clumsy pumping lemma approach can now be replaced by the Kolmogorov formulation below.

### 3.1 Kolmogorov Complexity Replacement for the Pumping Lemma

An important part of formal language theory is deriving a hierarchy of language families. The main division is the Chomsky hierarchy, with regular languages, context-free languages, context-sensitive languages and recursively enumerable languages. The common way to prove that certain languages are not regular is by using 'pumping' lemmas, for instance, the uvw-lemma. However, these lemmas are quite difficult to state and cumbersome to prove or use. In contrast, below we show how to replace such arguments by simple, intuitive and yet rigorous, Kolmogorov complexity arguments.

Regular languages coincide with the languages accepted by finite automata. This invites a straightforward application of Kolmogorov complexity. Let us give an example. We prove that $\left\{0^{k} 1^{k}: k \geq 1\right\}$ is not regular. If it were, then the state $q$ of a particular accepting fa after processing $0^{k}$, together with the fa, is, up to a constant, a description of $k$. Namely, by running $A$, initialized in state $q$, on input consisting of only 1 's, the first time $A$ enters an accepting state is after precisely $k$ consecutive 1 's. The size of the description of $A$ and $q$ is bounded by a constant, say $c$, which is independent of $k$. Altogether, it follows that $C(k) \leq c+O(1)$. But choosing $k$ with $C(k) \geq \log k$ we obtain a contradiction for all large enough $k$. Hence, since the fa has a fixed finite number of states, there is a fixed finite number that bounds the Kolmogorov complexity of each natural number: contradiction. We generalize this observation as follows.

Definition 3 Let $\Sigma$ be a finite nonempty alphabet, and let $\phi: \mathcal{N} \rightarrow \Sigma^{*}$ be a total recursive function. Then $\phi$ enumerates (possibly a proper subset of) $\Sigma^{*}$ in order $\phi(1), \phi(2), \ldots$ We call such an order effective, and $\phi$ an enumerator.

The lexicographical order is the effective order such that all words in $\Sigma^{*}$ are ordered first according to length, and then lexicographically within the group of each length. Another example is $\phi$ such that $\phi(i)=p_{i}$, the standard binary representation of the $i$ th prime, is an effective order in $\{0,1\}^{*}$. In this case $\phi$ does not enumerate all of $\Sigma^{*}$. Let $L \subseteq \Sigma^{*}$. Define $L_{x}=\{y: x y \in L\}$.

Lemma 1 (KC-Regularity) Let $L \subseteq \Sigma^{*}$ be regular, and let $\phi$ an enumerator in $\Sigma^{*}$. Then there exists a constant $c$ depending only on $L$ and $\phi$, such that for each $x$, if $y$ is the nth string enumerated in (or in the complement of) $L_{x}$, then $C(y) \leq C(n)+c$.

Proof. Let $L$ be a regular language. The $n$th string $y$ such that $x y \in L$ for some $x$ can be described by

- this discussion, and a description of the fa that accepts $L$;
- a description of $\phi$; and
- the state of the fa after processing $x$, and the number $n$.

The statement "(or in the complement of)" follows, since regular languages are closed under complementation.

As an application of the KC-Regularity Lemma we prove that $\left\{1^{p}: p\right.$ is prime $\}$ is not regular. Consider the string $x y=1^{p}$ with $p$ the $(k+1)$ th prime. Set $x=1^{p^{\prime}}$, with $p^{\prime}$ the $k$ th prime. Then $y=1^{p-p^{\prime}}$, and $y$ is the lexicographical first element in $L_{x}$. Hence, by Lemma 1, $C\left(p-p^{\prime}\right)=O(1)$. But the difference between two consecutive primes grows unbounded. Since there are only $O(1)$ descriptions of length $O(1)$, we have a contradiction. We give some more examples from the well-known textbook of Hopcroft and Ullman that are marked * as difficult there:

Example 3 (Exercise 3.1(h)* in [7]) Show $L=\left\{x x^{R} w: x, w \in\{0,1\}^{*}-\right.$ $\{\epsilon\}\}$ is not regular. Set $x=(01)^{m}$, where $C(m) \geq \log m$. Then, the lexicographically first word in $L_{x}$ is $y$ with $y=(10)^{m} 0$. But, $C(y)=\Omega(\log m)$, contradicting the KC-Regularity Lemma.

Example 4 Prove that $L=\left\{0^{i} 1^{j}: i \neq j\right\}$ is not regular. Set $x=0^{m}$, where $C(m) \geq \log m$. Then, the lexicographically first word not in $L_{x} \bigcap\{1\}^{*}$ is $y=1^{m}$. But, $C(y)=\Omega(\log m)$, contradicting the KC-Regularity Lemma.

Example 5 (Exercise $3.6^{*}$ in [7]) Prove that $L=\left\{0^{i} 1^{j}: \operatorname{gcd}(i, j)=1\right\}$ is not regular. Set $x=0^{(p-1)!} 1$, where $p>3$ is a prime, $l(p)=n$ and $C(p) \geq \log n-\log \log n$. Then the lexicographically first word in $L_{x}$ is $1^{p-1}$, contradicting the KC-regularity lemma.

Example 6 (Section 2.2, Exercises 11-15, [5]) Prove that $\{p: p$ is the standard binary representation of a prime $\}$ is not regular. Suppose the contrary, and $p_{i}$ denotes the $i$ th prime, $i \geq 1$. Consider the least binary $p_{m}=u v$ $\left(=u 2^{l(v)}+v\right)$, with $u=\Pi_{i<k} p_{i}$ and $v$ not in $\{0\}^{*}\{1\}$. Such a prime $p_{m}$ exists since each interval [ $n, n+n^{11 / 20}$ ] of the natural numbers contains a prime, $\left.\| 6\right]$.

Considering $p_{m}$ now as an integer, $p_{m}=2^{l(v)} \Pi_{i<k} p_{i}+v$. Since integer $v>1$ and $v$ is not divided by any prime less than $p_{k}$ (because $p_{m}$ is prime), the binary length $l(v) \geq l\left(p_{k}\right)$. Because $p_{k}$ goes to infinity with $k$, the value $C(v) \geq C(l(v))$ also goes to infinity with $k$. But since $v$ is the lexicographical first suffix, with integer $v>1$, such that $u v \in L$, we have $C(v)=O(1)$ by the KC-Regularity Lemma, which is a contradiction.

### 3.2 Kolmogorov Complexity Characterization of Regular Languages

While the pumping lemmas are not precise enough (except for the difficult construction in (4) to characterize the regular languages, with Kolmogorov complexity this is easy. In fact, the KC-Regularity Lemma is a direct corollary of the characterization below. The theorem is not only a device to show that some nonregular languages are nonregular, as are the common pumping lemmas, but it is a characterization of the regular sets. Consequently, it determines whether or not a given language is regular, just like the Myhill-Nerode Theorem. The usual characterizations of regular languages seem to be practically useful to show regularity. The need for pumping lemmas stems from the fact that characterizations tend to be very hard to use to show nonregularity. In contrast, the KC-characterization is practicable for both purposes, as evidenced by the examples.

Definition 4 Let $\Sigma$ be a nonempty finite alphabet, and let $y_{i}$ be the $i$ th element of $\Sigma^{*}$ in lexicographic order, $i \geq 1$. For $L \subseteq \Sigma^{*}$ and $x \in \Sigma^{*}$, let $\chi=\chi_{1} \chi_{2} \ldots$ be the characteristic sequence of $L_{x}=\{y: x y \in L\}$, defined by $\chi_{i}=1$ if $x y_{i} \in L$, and $\chi_{i}=0$ otherwise. We denote $\chi_{1} \ldots \chi_{n}$ by $\chi_{1: n}$.

Theorem 2 (Regular KC-Characterization) Let $L \subseteq \Sigma^{*}$, and assume the notation above. The following statements are equivalent.
(i) $L$ is regular.
(ii) There is a constant $c_{L}$ depending only on $L$, such that for all $x \in \Sigma^{*}$, for all $n, C\left(\chi_{1: n} \mid n\right) \leq c_{L}$.
(iii) There is a constant $c_{L}$ depending only on $L$, such that for all $x \in \Sigma^{*}$, for all $n, C\left(\chi_{1: n}\right) \leq C(n)+c_{L}$.
(iv) There is a constant $c_{L}$ depending only on $L$, such that for all $x \in \Sigma^{*}$, for all $n, C\left(\chi_{1: n}\right) \leq \log n+c_{L}$.

Proof. (i) $\rightarrow$ (ii): by similar proof as the KC-Regularity Lemma.
(ii) $\rightarrow$ (iii): obvious.
(iii) $\rightarrow$ (iv): obvious.
(iv) $\rightarrow$ (i):

Claim 1 For each constant $c$ there are only finitely many one-way infinite binary strings $\omega$ such that, for all $n, C\left(\omega_{1: n}\right) \leq \log n+c$.

Proof. The claim is a weaker version of Theorem 6 in [2]. It turns out that the weaker version admits a simpler proof. To make the treatment self-contained we present this new proof in the Appendix.

By (iv) and the claim, there are only finitely many distinct $\chi$ 's associated with the $x$ 's in $\Sigma^{*}$. Define the right-invariant equivalence relation $\sim$ by $x \sim x^{\prime}$ if $\chi=\chi^{\prime}$. This relation induces a partition of $\Sigma^{*}$ in equivalence classes $[x]=\{y$ : $y \sim x\}$. Since there is a one-one correspondence between the $[x]$ 's and the $\chi$ 's, and there are only finitely many distinct $\chi$ 's, there are also only finitely many $[x]$ 's, which implies that $L$ is regular by the Myhill-Nerode theorem.

Remark 1 The KC-regularity Lemma may be viewed as a corollary of the Theorem. If $L$ is regular, then clearly $L_{x}$ is regular, and it follows immediately that there are only finitely many associated $\chi$ 's, and each can be specified in at most $c$ bits, where $c$ is a constant depending only on $L$ (and enumerator $\phi)$. If $y$ is, say, the $n$th string in $L_{x}$, then we can specify $y$ as the string corresponding to the $n$th ' 1 ' in $\chi$, using only $C(n)+O(1)$ bits to specify $y$. Hence $C(y) \leq C(n)+O(1)$. Without loss of generality, we need to assume that the $n$th string enumerated in $L_{x}$ in the KC-regularity Lemma is the string corresponding to the $n$th ' 1 ' in $\chi$ by the enumeration in the Theorem, or that there is a recursive mapping between the two.

Remark 2 If $L$ is nonregular, then there are infinitely many $x \in \Sigma^{*}$ with distinct equivalence classes $[x]$, each of which has its own distinct associated characteristic sequence $\chi$. It is easy to see, for each automaton (finite or infinite), for each $\chi$ associated with an equivalence class $[x]$ we have

$$
C\left(\chi_{1: n} \mid n\right) \rightarrow \inf \{C(y): y \in[x]\}+O(1)
$$

for $n \rightarrow \infty$. The difference between finite and infinite automata is precisely expressed in the fact that only in the first case does there exist an a priori constant which bounds the lefthand term for all $\chi$.

We show how to prove positive results with the KC-Characterization Theorem. (Examples of negative results were given in the preceding section.)

Example 7 Prove that $L=\Sigma^{*}$ is regular. There exists a constant $c$, such that for each $x$ the associated characteristic sequence is $\chi=1,1, \ldots$, with $C\left(\chi_{1: n} \mid n\right) \leq c$. Therefore, $L$ is regular by the KC-Characterization Theorem.

Example 8 Prove that $L=\{x: x$ the number of ' 1 's in $x$ is odd $\}$ is regular. Obviously, there exists a constant $c$ such that for each $x$ we have $C\left(\chi_{1: n}\right) \leq$ $C(n)+c$. Therefore, $L$ is regular by the KC-Characterization Theorem.

## 4 Deterministic Context-free Languages

We present a Kolmogorov complexity based criterion to show that certain languages are not dcfl. In particular, it can be used to demonstrate the existence of witness languages in the difference of the family of context-free languages (cfls) and deterministic context-free languages (dcfls). Languages in this difference are the most difficult to identify; other non-dcfl are also non-cfl and in those cases we can often use the pumping lemma for context-free languages. The new method compares favorably with other known related techniques (mentioned in the Introduction) by being simpler, easier to apply, and apparently more powerful (because it works on a superset of examples). Yet, our primary goal is to demonstrate the usefulness of Kolmogorov complexity in this matter.

A language is a dcfl iff it is accepted by a deterministic pushdown automaton (dpda).

Intuitively, the lemma below tries to capture the following. Suppose a dpda accepts $L=\left\{0^{n} 1^{n} 2^{n}: n \geq 1\right\}$. Then the dpda needs to first store a representation of the all-0 part, and then retrieve it to check against the all-1 part. But after that check, it seems inevitable that it has discarded the relevant information about $n$, and cannot use this information again to check against the all-2 part. That is, the complexity of the all- 2 part should be $C(n)=O(1)$, which yields a contradiction for large $n$.

Definition 5 A one-way infinite string $\omega=\omega_{1} \omega_{2} \ldots$ over $\Sigma$ is recursive if there is a total recursive function $f: \mathcal{N} \rightarrow \Sigma$ such that $\omega_{i}=f(i)$ for all $i \geq 1$.

Lemma 2 (KC-DCFL) Let $L \subseteq \Sigma^{*}$ be recognized by a deterministic pushdown machine $M$ and let $c$ be a constant. Let $\omega=\omega_{1} \omega_{2} \ldots$ be a recursive sequence over $\Sigma$ which can be described in $c$ bits. Let $x, y \in \Sigma^{*}$ with $C(x, y)<c$ and let $\zeta=\ldots \zeta_{2} \zeta_{1}$ be a (reversed) recursive sequence over $\Sigma$ of the form ...yyx. Let $n, m \in \mathcal{N}$ and $w \in \Sigma^{*}$ be such that Items (i) to (iii) below are satisfied.
(i) For each $i(1 \leq i \leq n)$, given $M$ 's state and pushdown store contents after processing input $\zeta_{m} \ldots \zeta_{1} \omega_{1} \ldots \omega_{i}$, a description of $\omega$, and an additional description of at most c bits, we can reconstruct $n$ by running $M$ and observing only acceptance or rejection.
(ii) Given M's state and pushdown store contents after processing input $\zeta_{m} \ldots \zeta_{1} \omega_{1} \ldots \omega_{n}$, we can reconstruct $w$ from an additional description of at most c bits.
(iii) $K\left(\omega_{1} \ldots \omega_{n}\right) \geq 2 \log \log m$.

Then there is a constant $c^{\prime}$ depending only on $L$ and $c$ such that $C(w) \leq c^{\prime}$.
Proof. Let $L$ be accepted by $M$ with input head $h_{r}$. Assume $m, n, w$ satisfy the conditions in the statement of the lemma. For convenience we write

$$
u=\zeta_{m} \ldots \zeta_{1}, \quad v=\omega_{1} \ldots \omega_{n}
$$

For each input $z \in \Sigma^{*}$, we denote with $c(z)$ the pushdown store contents at the time $h_{r}$ has read all of $z$, and moves to the right adjacent input symbol. Consider the computation of $M$ on input $u v$ from the time when $h_{r}$ reaches the end of $u$. There are two cases:

Case 1. There is a constant $c_{1}$ such that for infinitely many pairs $m, n$ satisfying the statement of the lemma if $h_{r}$ continues and reaches the end of $v$, then all of the original $c(u)$ has been popped except at most the bottom $c_{1}$ bits.

That is, machine $M$ decreases its pushdown store from size $l(c(u))$ to size $c_{1}$ during the processing of $v$. The first time this occurs, let $v^{\prime}$ be the processed initial segment of $v$, and $v^{\prime \prime}$ the unprocessed suffix (so that $v=v^{\prime} v^{\prime \prime}$ ) and let $M$ be in state $q$. We can describe $w$ by the following items. 2

- A self-delimiting description of $M$ (including $\Sigma$ ) and this discussion in $O(1)$ bits.
- A self-delimiting description of $\omega$ in $(1+\epsilon) c$ bits.
- A description of $c\left(u v^{\prime}\right)$ and $q$ in $c_{1} \log |\Sigma|+O(1)$ bits.
- The 'additional description' mentioned in Item (i) of the statement of the lemma in self-delimiting format, using at most $(1+\epsilon) c$ bits. Denote it by $p$.
- The 'additional' description mentioned in Item (ii) of the statement of the lemma in self-delimiting format, using at most $(1+\epsilon) c$ bits. Denote it by $r$.

[^2]By Item (i) in the statement of the lemma we can reconstruct $v^{\prime \prime}$ from $M$ in state $q$ and with pushdown store contents $c\left(u v^{\prime}\right)$, and $\omega$, using description $p$. Subsequently, starting $M$ in state $q$ with pushdown store contents $c\left(u v^{\prime}\right)$, we process $v^{\prime \prime}$. At the end of the computation we have obtained $M$ 's state and pushdown store contents after processing $u v$. According to Item (ii) in the statement of the lemma, together with description $r$ we can now reconstruct $w$. Since $C(w)$ is at most the length of this description,

$$
C(w) \leq 4 c+c_{1} \log |\Sigma|+O(1)
$$

Setting $c^{\prime}:=4 c+c_{1} \log |\Sigma|+O(1)$ satisfies the lemma.
Case 2. By way of contradiction, assume that Case 1 does not hold. That is, for each constant $c_{1}$ all but finitely many pairs $m, n$ satisfying the conditions in the lemma cause $M$ not to decrease its stack height below $c_{1}$ during the processing of the $v$ part of input $u v$.

Fix some constant $c_{1}$. Set $m, n$ so that they satisfy the statement of the lemma, and to be as long as required to validate the argument below. Choose $u^{\prime}$ as a suffix of $y y \ldots y x$ with $l\left(u^{\prime}\right)>2^{m}$ and

$$
\begin{equation*}
C\left(l\left(u^{\prime}\right)\right)<\log \log m . \tag{2}
\end{equation*}
$$

That is, $l\left(u^{\prime}\right)$ is much larger than $l(u)(=m)$ and much more regular. A moment's reflection learns that we can always choose such a $u^{\prime}$.

Claim 2 For large enough $m$ there exists a $u^{\prime}$ as above, such that $M$ starts in the same state and accesses the same top $l(c(u))-c_{1}$ elements of its stack during the processing of the $v$ parts of both inputs $u v$ and $u^{\prime} v$.

Proof. By assumption, $M$ does not read below the bottom $c_{1}$ symbols of $c(u)$ while processing the $v$ part of input $u v$.

We argue that one can choose $u^{\prime}$ such that the top segment of $c\left(u^{\prime}\right)$ is precisely the same as the top segment of $c(u)$ above the bottom $c_{1}$ symbols, for large enough $l(u), l\left(u^{\prime}\right)$.

To see this we examine the initial computation of $M$ on $u$. Since $M$ is deterministic, it must either cycle through a sequence of pushdown store contents, or increase its pushdown store with repetitions on long enough $u$ (and $\left.u^{\prime}\right)$. Namely, let a triple ( $q, i, s$ ) mean that $M$ is in state $q$, has top pushdown store symbol $s$, and $h_{r}$ is at $i$ th bit of some $y$. Consider only the triples $(q, i, s)$ at the steps where $M$ will never go below the current top pushdown store level again while reading $u$. (That is, $s$ will not be popped before going into $v$.) There are precisely $l(c(u))$ such triples. Because the input is repetitious and $M$ is deterministic, some triple must start to repeat within a constant number of steps and with a constant interval (in height of $M$ 's pushdown store) after $M$ starts reading $y$ 's. It is easy to show that within a repeating interval only a constant number of $y$ 's are read.

The pushdown store does not cycle through an a priori bounded set of pushdown store contents, since this would mean that there is a constant $c_{1}$ such that the processing by $M$ of any suffix of $y y \ldots y x$ does not increase the stack height above $c_{1}$. This situation reduces to Case 1 with $v=\epsilon$.

Therefore, the pushdown store contents grows repetitiously and unboundedly. Since the repeating cycle starts in the pushdown store after a constant number of symbols, and its size is constant in number of $y$ 's, we can adjust $u^{\prime}$ so that $M$ starts in the same state and reads the same top segments of $c(u)$ and $c\left(u^{\prime}\right)$ in the $v$ parts of its computations on $u v$ and $u^{\prime} v$. This proves the claim.

The following items form a description from which we can reconstruct $v$.

- This discussion and a description of $M$ in $O(1)$ bits.
- A self-delimiting description of the recursive sequence $\omega$ of which $v$ is an initial segment in $(1+\epsilon) c$ bits.
- A self-delimiting description of the pair $\langle x, y\rangle$ in $(1+\epsilon) c$ bits.
- A self-delimiting description of $l\left(u^{\prime}\right)$ in $(1+\epsilon) C\left(l\left(u^{\prime}\right)\right)$ bits.
- A program $p$ to reconstruct $v$ given $\omega$ and $M$ 's state and pushdown store contents after processing $u$. By Item (i) of the statement of the lemma, $l(p) \leq c$. Therefore, a self-delimiting description of $p$ takes at most $(1+\epsilon) c$ bits.

The following procedure reconstructs $v$ from this information. Using the description of $M$ and $u^{\prime}$ we construct the state $q_{u^{\prime}}$ and pushdown store contents $c\left(u^{\prime}\right)$ of $M$ after processing $u^{\prime}$. By Claim 22, the state $q_{u}$ of $M$ after processing $u$ satisfies $q_{u}=q_{u^{\prime}}$ and the top $l(c(u))-c_{1}$ elements of $c(u)$ and $c\left(u^{\prime}\right)$ are the same. Run $M$ on input $\omega$ starting in state $q_{u^{\prime}}$ and with stack contents $c\left(u^{\prime}\right)$. By assumption, no more than $l(c(u))-c_{1}$ elements of $c\left(u^{\prime}\right)$ get popped before we have processed $\omega_{1} \ldots \omega_{n}$. By just looking at the consecutive states of $M$ in this computation, and using program $p$, we can find $n$ according to Item (i) in the statement of the lemma. To reconstruct $v$ requires by definition at least $C(v)$ bits. Therefore,

$$
\begin{aligned}
C(v) & \leq(1+\epsilon) C\left(l\left(u^{\prime}\right)+4 c+O(1)\right. \\
& \leq(1+\epsilon) \log \log m+4 c+O(1)
\end{aligned}
$$

where the last inequality follows by Equation 2. But this contradicts Item (iii) in the statement of the lemma for large enough $m$.

Items (i) through (iii) in the KC-DCFL Lemma can be considerably weakened, but the presented version gives the essential idea and power: it suffices for many examples. A more restricted, but easier, version is the following.

Corollary 1 Let $L \subseteq \Sigma^{*}$ be a dcfl and let $c$ be a constant. Let $x$ and $y$ be fixed finite words over $\Sigma$ and let $\omega$ be a recursive sequence over $\Sigma$. Let $u$ be a suffix of $y y \ldots y x$, let $v$ be a prefix of $\omega$, and let $w \in \Sigma^{*}$ such that:
(i) $v$ can be described in $c$ bits given $L_{u}$ in lexicographical order;
(ii) $w$ can be described in $c$ bits given $L_{u v}$ in lexicographical order; and
(iii) $C(v) \geq 2 \log \log l(u)$.

Then there is a constant $c^{\prime}$ depending only on $L, c, x, y, \omega$ such that $C(w) \leq c^{\prime}$.
All the following context-free languages were proved to be not dcfl only with great effort before, [7, 5, 25]. Our new proofs are more direct and intuitive. Basically, if $v$ is the first word in $L_{u}$, then processing the $v$ part of input $u v$ must have already used up the information of $u$. But if there is not much information left on the pushdown store, then the first word $w$ in $L_{u v}$ cannot have high Kolmogorov complexity.

Example 9 (Exercise 10.5 (A)** in 7) Prove $L=\left\{x: x=x^{R}, x \in\{0,1\}^{*}\right\}$ is not dcfl. Suppose the contrary. Set $u=0^{n} 1$ and $v=0^{n}, C(n) \geq \log n$, satisfying Item (iii) of the lemma. Since $v$ is lexicographically the first word in $L_{u}$, Item (i) of the lemma is satisfied. The lexicographically first nonempty word in $L_{u v}$ is $10^{n}$, and so we can set $w=10^{n}$ satisfying Item (ii) of the lemma. But now we have $C(w)=\Omega(\log n)$, contradicting the KC-DCFL Lemma and its Corollary.

Approximately the same proof shows that the context-free language $\left\{x x^{R}\right.$ : $\left.x \in \Sigma^{*}\right\}$ and the context-sensitive language $\left\{x x: x \in \Sigma^{*}\right\}$ are not deterministic context-free languages.

Example 10 (Exercise 10.5 (b)** in 7], Example 1 in 25]) Prove $\left\{0^{n} 1^{m}\right.$ : $m=n, 2 n\}$ is not dcfl. Suppose the contrary. Let $u=0^{n}$ and $v=1^{n}$, where $C(n) \geq \log n$. Then $v$ is the lexicographically first word in $L_{u}$. The lexicographically first nonempty word in $L_{u v}$ is $1^{n}$. Set $w=1^{n}$, and $C(w)=\Omega(\log n)$, contradicting the KC-DCFL Lemma and its Corollary.

Example 11 (Example 2 in 25]) Prove $L=\{x y: l(x)=l(y), y$ contains a ${ }^{\prime} 1$ ', $\left.x, y \in\{0,1\}^{*}\right\}$ is not dcfl. Suppose the contrary. Set $u=0^{n} 1$ where $l(u)$ is even. Then $v=0^{n+1}$ is lexicographically the first even length word not in $L_{u}$. With $C(n) \geq \log n$, this satisfies Items (i) and (iii) of the lemma. Choosing $w=10^{2 n+3}$, the lexicographically first even length word not in $L_{u v}$ starting with a ' 1 ', satisfies Item (ii). But $C(w)=\Omega(\log n)$, which contradicts the KC-DCFL Lemma and its Corollary.

Example 12 Prove $L=\left\{0^{i} 1^{j} 2^{k}: i, j, k \geq 0, i=j\right.$ or $\left.j=k\right\}$ is not dcfl. Suppose the contrary. Let $u=0^{n}$ and $v=1^{n}$ where $C(n) \geq \log n$, satisfying
item (iii) of the lemma. Then, $v$ is lexicographically the first word in $L_{u}$, satisfying Item (i). The lexicographic first word in $L_{u v} \cap\{1\}\{2\}^{*}$ is $12^{n+1}$. Therefore, we can set $w=12^{n+1}$ and satisfy Item (ii). Then $C(w)=\Omega(\log n)$, contradicting the KC-DCFL Lemma and its Corollary.

Example 13 (Pattern-Matching) The KC-DCFL Lemma and its Corollary can be used trickily. We prove $\left\{x \# y x^{R} z: x, y, z \in\{0,1\}^{*}\right\}$ is not dcfl. Suppose the contrary. Let $u=1^{n} \#$, and $v=1^{n-1} 0$ where $C(n) \geq \log n$, satisfying Item (iii) of the lemma. Since $v^{\prime}=1^{n}$ is the lexicographically first word in $L_{u}$, the choice of $v$ satisfies Item (i) of the lemma. (We can reconstruct $v$ from $v^{\prime}$ by flipping the last bit of $v^{\prime}$ from 1 to 0 .) Then $w=1^{n}$ is lexicographically the first word in $L_{u v}$, to satisfy Item (ii). Since $C(w)=\Omega(\log n)$, this contradicts the KC-DCFL Lemma and its Corollary.

## 5 Recursive, Recursively Enumerable, and Beyond

It is immediately obvious how to characterize recursive languages in terms of Kolmogorov complexity. If $L \subseteq \Sigma^{*}$, and $\Sigma^{*}=\left\{v_{1}, v_{2}, \ldots\right\}$ is effectively ordered, then we define the characteristic sequence $\lambda=\lambda_{1}, \lambda_{2}, \ldots$ of $L$ by $\lambda_{i}=1$ if $v_{i} \in L$ and $\lambda_{i}=0$ otherwise. In terms of the earlier developed terminology, if $A$ is the automaton accepting $L$, then $\lambda$ is the characteristic sequence associated with the equivalence class $[\epsilon]$. Recall Definition 5 of a recursive sequence. A set $L \in \Sigma^{*}$ is recursive iff its characteristic sequence $\lambda$ is a recursive sequence. It then follows trivially from the definitions:

Theorem 3 (Recursive KC Characterization) $A$ set $L \in \Sigma^{*}$ is recursive, iff there exists a constant $c_{L}$ (depending only on $L$ ) such that, for all $n$, $C\left(\lambda_{1: n} \mid n\right)<c_{L}$.
$L$ is r.e. if the set $\left\{n: \lambda_{n}=1\right\}$ is r.e. In terms of Kolmogorov complexity, the following theorem gives not only a qualitative but even a quantitative difference between recursive and r.e. languages. The following theorem is due to Barzdin', [1, 14 .

Theorem 4 (KC-R.E.) (i) If $L$ is r.e., then there is a constant $c_{L}$ (depending only on $L$ ), such that for all $n, C\left(\lambda_{1: n} \mid n\right) \leq \log n+c_{L}$.
(ii) There exists an r.e. set $L$ such that $C\left(\lambda_{1: n}\right) \geq \log n$, for all $n$.

Note that, with $L$ as in Item (ii), the set $\Sigma^{*}-L$ (which is possibly nonr.e.) also satisfies Item (i). Therefore, Item (i) is not a Kolmogorov complexity characterization of the r.e. sets.

Example 14 Consider the standard enumeration of Turing machines. Define $k=k_{1} k_{2} \ldots$ by $k_{i}=1$ if the $i$ th Turing machine started on its $i$ th program halts $\left(\phi_{i}(i)<\infty\right)$, and $k_{i}=0$ otherwise. Let $A$ be the language such that $k$ is its characteristic sequence. Clearly, $A$ is an r.e. set. In $\mathbb{1}]$ it is shown that $C\left(k_{1: n}\right) \geq \log n$, for all $n$.

Example 15 Let $k$ be as in the previous example. Define a one-way infinite binary sequence $h$ by

$$
h=k_{1} 0^{2} k_{2} 0^{2^{2}} \ldots k_{i} 0^{2^{i}} k_{i+1} \ldots
$$

Then, $C\left(h_{1: n}\right)=O(C(n))+\Theta(\log \log n)$. Therefore, if $h$ is the characteristic sequence of a set $B$, then $B$ is not recursive, but more 'sparsely' nonrecursive than is $A$.

Example 16 The probability that the optimal universal Turing machine $U$ halts on self-delimiting binary input $p$, randomly supplied by tosses of a fair coin, is $\Omega, 0<\Omega<1$. Let the binary representation of $\Omega$ be $0 . \Omega_{1} \Omega_{2} \ldots$ Let $\Sigma$ be a finite nonempty alphabet, and $v_{1}, v_{2}, \ldots$ an effective enumeration without repetitions of $\Sigma^{*}$. Define $L \subseteq \Sigma^{*}$ such that $v_{i} \in L$ iff $\Omega_{i}=1$. It can be shown, see for example 13 , that the sequence $\Omega_{1}, \Omega_{2}, \ldots$ satisfies

$$
C\left(\Omega_{1: n} \mid n\right) \geq n-\log n-2 \log \log n-O(1)
$$

for all but finitely many $n$.
Hence neither $L$ nor $\Sigma^{*}-L$ is r.e. It is not difficult to see that $L \in \Delta_{2}-$ $\left(\Sigma_{1} \cup \Pi_{1}\right)$, in the arithmetic hierarchy (that is, $L$ is not recursively enumerable), [23, 24].

## 6 Questions for Future Research

(1) It is not difficult to give a direct KC-analogue of the uvwxy Pumping Lemma (as Tao Jiang pointed out to us). Just like the Pumping Lemma, this will show that $\left\{a^{n} b^{n} c^{n}: n \geq 1\right\},\left\{x x: x \in \Sigma^{*}\right\},\left\{a^{p}: p\right.$ is prime $\}$, and so on, are not cfl. Clearly, this hasn't yet captured the Kolmogorov complexity heart of cfl. More in general, can we find a CFL-KC-Characterization?
(2) What about ambiguous context-free languages?
(3) What about context-sensitive languages and deterministic context-sensitive languages?

## Appendix: Proof of Claim $\square$

A recursive real is a real number whose binary expansion is recursive in the sense of Definition 5. The following result is demonstrated in 15 and attributed to A.R. Meyer. For each constant $c$ there are only finitely many $\omega \in\{0,1\}^{\infty}$ with $C\left(\omega_{1: n} \mid n\right) \leq c$ for all $n$. Moreover, each such $\omega$ is a recursive real.

In [2] this is strengthened to a version with $C\left(\omega_{1: n}\right) \leq C(n)+c$, and strengthened again to a version with $C\left(\omega_{1: n}\right) \leq \log n+c$. Claim 11 is weaker than the latter version by not requiring the $\omega$ 's to be recursive reals. For completeness sake, we present a new direct proof of Claim il avoiding the notion of recursive reals.

Recall our convention of identifying integer $x$ with the $x$ th binary sequence in lexicographical order of $\{0,1\}^{*}$ as in Equation 11.

Proof. [of Claim Let $c$ be a positive constant, and let

$$
\begin{align*}
A_{n} & =\left\{x \in\{0,1\}^{n}: C(x) \leq \log n+c\right\}  \tag{3}\\
A & =\left\{\omega \in\{0,1\}^{\infty}: \forall_{n \in \mathcal{N}}\left[C\left(\omega_{1: n}\right) \leq \log n+c\right]\right\}
\end{align*}
$$

If the cardinality $d\left(A_{n}\right)$ of $A_{n}$ dips below a fixed constant $c^{\prime}$, for infinitely many $n$, then $c^{\prime}$ is an upper bound on $d(A)$. This is because it is an upper bound on the cardinality of the set of prefixes of length $n$ of the elements in $A$, for all $n$.

Fix any $l \in \mathcal{N}$. Choose a binary string $y$ of length $2 l+c+1$ satisfying

$$
\begin{equation*}
C(y) \geq 2 l+c+1 \tag{4}
\end{equation*}
$$

Choose $i$ maximum such that for division of $y$ in $y=m n$ with $l(m)=i$ we have

$$
\begin{equation*}
m \leq d\left(A_{n}\right) \tag{5}
\end{equation*}
$$

(This holds at least for $i=0=m$.) Define similarly a division $y=s r$ with $l(s)=i+1$. By maximality of $i$, we have $s>d\left(A_{r}\right)$. From the easily proven $s \leq 2 m+1$, it then follows that

$$
\begin{equation*}
d\left(A_{r}\right) \leq 2 m \tag{6}
\end{equation*}
$$

We prove $l(r) \geq l$. Since by Equations 5 and 3 we have

$$
m \leq d\left(A_{n}\right) \leq 2^{c} n
$$

it follows that $l(m) \leq l(n)+c$. Therefore,

$$
2 l+c+1=l(y)=l(n)+l(m) \leq 2 l(n)+c
$$

which implies that $l(n)>l$. Consequently, $l(r)=l(n)-1 \geq l$.
We prove $d\left(A_{r}\right)=O(1)$. By dovetailing the computations of the reference universal Turing machine $U$ for all programs $p$ with $l(p) \leq \log n+c$, we can
enumerate all elements of $A_{n}$. We can reconstruct $y$ from the $m$ th element, say $y_{0}$, of this enumeration. Namely, from $y_{0}$ we reconstruct $n$ since $l\left(y_{0}\right)=n$, and we obtain $m$ by enumerating $A_{n}$ until $y_{0}$ is generated. By concatenation we obtain $y=m n$. Therefore,

$$
\begin{equation*}
C(y) \leq C\left(y_{0}\right)+O(1) \leq \log n+c+O(1) . \tag{7}
\end{equation*}
$$

From Equation 1 we have

$$
\begin{equation*}
C(y) \geq \log n+\log m . \tag{8}
\end{equation*}
$$

Combining Equations $\mathrm{T}^{7}$ and 8 , it follows that $\log m \leq c+O(1)$. Therefore, by Equation 6 ,

$$
d\left(A_{r}\right) \leq 2^{c+O(1)} .
$$

Here, $c$ is a fixed constant independent of $n$ and $m$. Since $l(r) \geq l$ and we can choose $l$ arbitrarily, $d\left(A_{r}\right) \leq c_{0}$ for a fixed constant $c_{0}$ and infinitely many $r$, which implies $d(A) \leq c_{0}$, and hence the claim.

We avoided establishing, as in the cited references, that the elements of $A$ defined in Equation 3 are recursive reals. The resulting proof is simpler, and sufficient for our purpose, since we only need to establish the finiteness of $A$.

Remark 3 The difficult part of the Regular KC-Characterization Theorem above consists in proving that the KC-Regularity Lemma is exhaustive, i.e., can be used to prove the nonregularity of all nonregular languages. Let us look a little more closely at the set of sequences defined in Item (iii) of the KCCharacterization Theorem. The set of sequences $A$ of Equation 3 is a superset of the set of characteristic sequences associated with $L$. According to the proof in the cited references, this set $A$ contains finitely many recursive sequences (computable by Turing machines). The subset of $A$ consisting of the characteristic sequences associated with $L$, satisfies much more stringent computational requirements, since it can be computed using only the finite automaton recognizing $L$. If we replace the plain Kolmogorov complexity in the statement of the theorem by the so-called 'prefix complexity' variant $K$, then the equivalent set of $A$ in Equation 3 is

$$
\left\{\omega \in\{0,1\}^{\infty}: \forall_{n \in \mathcal{N}}\left[K\left(\omega_{1: n}\right) \leq K(n)+c\right]\right\},
$$

which contains nonrecursive sequences by a result of R.M. Solovay, 21 .

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[^1]:    ${ }^{1}$ Similarly, we define the complexity of the $x$ th partial recursive function $\phi$ conditional to the $y$ th partial recursive function $\psi$ by $C(\phi \mid \psi)=C(x \mid y)$.

[^2]:    ${ }^{2}$ Since we need to glue different binary items in the encoding together, in a way so that we can effectively separate them again, like $\langle x, y\rangle=x^{\prime} y$, we count $C(x)+2 \log C(x)+1$ bits for a self-delimited encoding $x^{\prime}=1^{l(l(x))} 0 l(x) x$ of $x$. We only need to give self-delimiting forms for all but one constituent description item.

