Optimal redundancy in computations from random oracles *

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Abstract. A classic result in algorithmic information theory is that every infinite binary sequence is computable from a Martin-Löf random infinite binary sequence. Proved independently by Kučera [Kuč85] and Gács [Gác86], this result answered a question by Charles Bennett and has seen numerous applications in the last 30 years. The optimal redundancy in such a coding process has, however, remained unknown. If the computation of the first $n$ bits of a sequence requires $n + g(n)$ bits of the random oracle, then $g$ is the redundancy of the computation. Kučera implicitly achieved redundancy $n \log n$ while Gács used a more elaborate block-coding procedure which achieved redundancy $\sqrt{n} \log n$. Different approaches to coding such as the one by Merkle and Mihailović [MM04] have not improved this redundancy bound.

In this paper we devise a new coding method that achieves optimal logarithmic redundancy. In particular, for any computable non-decreasing function $g$ such that $\sum_i 2^{-g(i)} < 1$ we show that there is a uniform coding process that codes every infinite binary sequence into a Martin-Löf infinite binary sequence with redundancy $g$. This redundancy bound is exponentially smaller\(^1\) than the previous bound of $\sqrt{n} \log n$ and is known to be the best possible by recent work [BLPT15], where it was shown that if $\sum_i 2^{-g(i)}$ diverges then there exists an infinite binary sequence $X$ which cannot be computed by any Martin-Löf random infinite binary sequence with redundancy $g$. It follows that redundancy $\epsilon \cdot \log n$ in computation from a random oracle is possible for every stream, if and only if $\epsilon > 1$.

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\(^1\)When $f$ and $g$ are unbounded, we say $g$ is exponentially smaller than $f$ if there exists a constant $c$ such that $2^c \cdot g(n) < f(n)$ for all $n$. 
1 Introduction

If a binary stream is algorithmically random, one does not expect to be able to extract any useful information from it. Although this reasonable intuition can be verified in many formal contexts, it fails for the most accepted and robust notion of algorithmic randomness, which is Martin-Löf randomness\[^{[ML66]}\], also formulated as incompressibility in terms of Kolmogorov complexity by Chaitin\[^{[Cha75]}\] and Levin\[^{[Lev73]}\]. Indeed, Kučera\[^{[Kuč85]}\], and independently Gács\[^{[Gác86]}\], showed that any infinite binary sequence is computable from a Martin-Löf random sequence. Both authors constructed a uniform process that codes every infinite binary sequence into some Martin-Löf random stream. The Kučera-Gács theorem, as it is known in algorithmic information theory, has been studied and extended in numerous ways in the last 30 years\[^{[3]}\] and has become a standard prominent topic in most textbooks and presentations of this area.\[^{[4]}\] In the context of Martin-Löf randomness, this result says that any type of information that can be coded into a binary stream, no matter how structured that might be, can be obfuscated into an algorithmically random stream, from which it is effectively recoverable.

Here information could be the solution to a problem of interest, such as the halting problem, the word problem for finite groups, or any of the numerous and often algorithmically unsolvable problems whose solutions can be represented as a set of integers. Effectively recoverable means computable by means of a Turing reduction, without any restrictions on time or memory. However, as we discuss below, the coding constructed in both\[^{[Kuč85]}\] and\[^{[Gác86]}\] gives a Turing reduction with a computable upper bound on the length of the initial segment of the oracle that is used in the computation on any given argument—the oracle use.\[^{[5]}\]

It is hardly surprising that such a coding process occasionally introduces an overhead on the codes of the initial segments of certain streams. More specifically, if we code a stream $X$ into a Martin-Löf random stream $Y$, then it is very possible that for some $n$, in order to recover the first $n$ bits of $X$ (denoted $X|^n$), we need $Y|^n + g(n)$, i.e., $g(n)$ more bits of $Y$. Such a function $g$ that bounds from above the number of extra bits needed in the decoding process is known as the redundancy in the computation of $X$ from $Y$. For example, it is known from\[^{[BLPT15]}\] that certain streams $X$ are not computable from any Martin-Löf random stream with redundancy $\log n$. Such restrictions can be intuitively understood if one considers that information introduces structure, and in order to obfuscate the structure of a given $X$ into a random $Y$, extra bits amplifying the complexity of the code might be necessary.

In the context of information theory, it is important to

(a) determine the optimal redundancy in coding into Martin-Löf random streams;

(b) construct a coding process that achieves the optimal redundancy.

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\[^{3}\]If a stream is arithmetically random in the sense that it avoids all null sets of reals which are arithmetically definable, then it cannot compute any noncomputable stream that is definable in arithmetic. Similarly, if a stream is Martin-Löf random relative to the halting problem, then it cannot compute any noncomputable stream which definable in arithmetic with one unbounded quantifier. A less trivial example is a fact from Stephan\[^{[Ste06]}\] that incomplete Martin-Löf random binary streams (i.e., infinite sequences) cannot compute any complete extensions of Peano Arithmetic, and its extensions in Levin\[^{[Lev02, Lev13]}\].

\[^{4}\]For example see\[^{[Kuč89, Boo94, Her97, MM04, Dot06, DM06, BDN11]}\].

\[^{5}\]For example consider the standard textbooks\[^{[LV97, Cal94, Nie09, DH10]}\] and the surveys\[^{[DHNT06, MN06]}\].

\[^{5}\]In the terminology of computability theory, every binary stream is weak-truth-table computable from a Martin-Löf random stream. Bennett\[^{[Ben88]}\] observed that this is no longer true for truth-table computations, and used this fact in order to define the logical depth for infinite binary sequences. Other refined reducibilities were considered by Book\[^{[Boo94]}\].
The original work in [Kuč85, Gác86] did not achieve these goals, nor did the subsequent work of Merkle and Mihailović [MM04] and Doty [Dot06]. The goal of the present work is to give a definitive answer to challenges (a) and (b).

1.1 Previous, directly relevant work

Kučera [Kuč85] did not show an interest in optimizing the redundancy of the coding, other than observing that it can be computably bounded. An examination of his argument (see the survey [BLP16] for such a discussion) shows that his method produces redundancy $n \log n$. Gács [Gác86], on the other hand, has a clear interest in minimising the redundancy of his coding, which he bounds by $\sqrt{n \log n}$ by means of a more sophisticated block-coding, with carefully chosen block-lengths. Merkle and Mihailović [MM04] give an interpretation of Gács’ coding in terms of effective martingales, instead of the effective closed sets approach employed in the original argument. Although the latter analysis is rather elegant and geared toward obtaining a small redundancy $o(n)$, the resulting upper bound is identical with Gács’ bound of $\sqrt{n \log n}$.

Doty [Dot06] showed how to reduce the oracle-use when coding a stream $X$ into a Martin-Löf random $Y$, based on suitable bounds on the constructive dimension of $X$. In particular, partially extending previous work by Ryabko [Rya86], he showed that the asymptotic ratio between the optimal oracle-use in computing $X |_n$ and $n$ is directly related to the constructive dimension of $X$. Unfortunately, the arguments developed in this latter work do not shed light on our main goal, which asks for the actual optimal redundancy in coding into Martin-Löf random streams, and not the mere asymptotic behavior of the oracle-use in such a reduction. For example, for streams $X$ of dimension 1 the work in [Dot06] merely shows that they can be computed by a Martin-Löf random stream $Y$ with oracle-use $\ell_n$ such that $\liminf_n (\ell_n/n) = 1$. From the latter we cannot even deduce Gács upper bound $n + \sqrt{n \log n}$ on the oracle-use.

1.2 Our results

Our main contribution is a coding process that codes an arbitrary binary stream into a Martin-Löf binary stream, with optimal redundancy which is exponentially smaller than the previous known bound of $\sqrt{n \log n}$. In the following statement, ‘uniformly computable’ means that there is a single coding process that works for all binary streams, i.e. a single Turing functional that provides the promised reduction of each given stream to some Martin-Löf random stream.

**Theorem 1.1.** If $(\ell_i)$ is a computable increasing function such that $\sum_i 2^{-\ell_i+i} < 1$ then every binary stream is uniformly computable from a Martin-Löf random stream with oracle-use $(\ell_i)$.

In fact, Theorem 1.1 is part of the following slightly more general fact that we prove, regarding coding into effectively closed sets of positive measure.

**Lemma 1.2.** Let $g$ be a nondecreasing computable function and let $\mathcal{P}$ be a $\Pi^0_1$ class. If $\sum_i 2^{-\ell_i+i} < \mu(\mathcal{P})$ then every binary stream is uniformly computable from some member of $\mathcal{P}$ with oracle-use $(\ell_i)$.

Note that Theorem 1.1 follows directly from Lemma 1.2 since the class of Martin-Löf random binary streams is a $\Sigma^0_2$ set of measure 1. We are also able to establish the optimality of these two results. In

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In actual fact, Gács [Gác86] achieves redundancy $3 \sqrt{n \log n}$, but a careful examination of his argument (as this is discussed in the survey [BLP16]) shows that it can be reduced to $\sqrt{n \log n}$.
it was shown that if the sum in Theorem 1.1 is not bounded, then there exists a binary stream which is not computable by any Martin-Löf random stream with oracle-use \((\ell_i)\). If we combine this with Theorem 1.1 we get the following characterization.

**Corollary 1.3.** Let \(g\) be a nondecreasing computable function. Then the following are equivalent:

(i) every binary stream is computable from a Martin-Löf random stream with redundancy \(g\);

(ii) \(\sum_i 2^{-g(i)} < \infty\).

Note that in clause (i) for Corollary 1.3 we can replace ‘computable’ with ‘uniformly’ computable.

1.3 Terminology, methodology and novelty

1.3.1 Terminology

The Cantor space \(2^\omega\) is the class of all binary streams (i.e. infinite binary strings). Unless explicitly stated otherwise, it is to be assumed that binary strings are finite, so that strings are finite objects while streams are infinite. We let \(|\sigma|\) denote the length of the string \(\sigma\). If \(Q\) is a computably enumerable set of binary strings, we let \(\llbracket Q \rrbracket\) denote the class of binary strings which are prefixed by some string in \(Q\). If \(Q = \{\nu\}\), then we simply write \(\llbracket \nu \rrbracket\) for \(\llbracket Q \rrbracket\). In this way \(\Sigma^0_1\) subsets of \(2^\omega\) can be represented by c.e. sets of strings \(Q\). The Lebesgue measure of \(\llbracket Q \rrbracket\) may be denoted simply by \(\mu(Q)\).

A tree \(T\) in the Cantor space is a downward closed set of binary strings with respect to the prefix relation. A branch of \(T\) is simply a string in \(T\) and a path through \(T\) is a stream for which all finite initial segments are branches of \(T\). A \(\Pi^0_1\) class in the Cantor space can be represented as a \(\Pi^0_1\) tree or as \(2^\omega - \llbracket Q \rrbracket\) for some c.e. set of strings \(Q\). The \(n\)-th level of \(T\) consists of the strings in \(T\) of length \(n\). A leaf of a tree is branch of the tree with no proper extensions in the tree. For any set of strings \(T\), we let \([T]\) denote the set of all streams with infinitely many prefixes in \(T\).

Now suppose we are given the \(\Pi^0_1\) class \(P\) and the increasing computable sequence \((\ell_i)\) of Lemma 1.2. Our task is to construct a Turing functional \(\Phi\) with uniform oracle-use \((\ell_i)\) on all oracles, with the property that for every binary stream \(X\) there exists some \(Y_{X} \in P\) such that \(X = \Phi^{Y_X}\). It is convenient to define \(\Phi\) by assigning labels for strings of each length \(n\) to strings of each length \(\ell_n\) and we assign the label \(x_\sigma\) to \(\tau\), then this is equivalent to defining \(\Phi^\tau = \sigma\). Of course these assignments need to be consistent, in the sense that if \(\tau' \supseteq \tau\), \(\Phi^\tau = \sigma\) and \(\Phi^{\tau'} = \sigma'\) then \(\sigma' \supseteq \sigma\). In our analysis we thus present \(\Phi\) as a partially labelled tree, by which we mean the full binary tree \(2^{<\omega}\) along with a partial labelling of it. Given a partially labelled tree \(T\) and \(\ell \in \mathbb{N}\), we let \(T \upharpoonright \ell\) denote the restriction of \(T\) to the strings of length at most \(\ell\).

1.3.2 The departure from existing approaches

There are a number of different presentations of the Kučera-Gács theorem in the literature. Kučera [Kuč85] uses the recursion theorem and the universality properties of the class of Martin-Löf random streams. His coding method may be seen as being of the following inductive form. Working within a \(\Pi^0_1\) class of Martin-Löf randoms \(P\), which is the set of all infinite paths through the computable tree \(T\), let us suppose that we have already determined \(2^n\) strings of length \(\ell_n\) in \(T\) which are extendable (i.e. have infinite extensions in
such that for each string $\sigma$ of length $n$ there is precisely one of these extendable strings $\tau$ for which we have defined $\Phi^\tau = \sigma$. From properties of the class $\mathcal{P}$, we are then able to determine a length $\ell_{n+1}$ such that each of these $2^n$ strings $\tau$ must have at least two incompatible and extendable extensions in $\mathcal{T}$ of length $\ell_{n+1}$.

If $\Phi^\tau = \sigma$, then for two of these extendable extensions $\tau'$ and $\tau''$ of length $\ell_{n+1} + 1$, we can define $\Phi^{\tau'} = \sigma * 0$ and $\Phi^{\tau''} = \sigma * 1$. The coding may therefore be thought of as occurring bit-by-bit, and actually takes place inside a subclass $\mathcal{P}' \subseteq \mathcal{P}$ defined by the tree $\mathcal{T}'$ with the property that for all $n$:

$$\text{every branch of } \mathcal{T}' \text{ at level } \ell_n \text{ has at least two extensions at level } \ell_{n+1} \text{ in } \mathcal{T}'. \tag{1.3.1}$$

As we proceed to code $X$, the manner in which we code $\sigma * i \subset X$ may also be seen to satisfy a strong independence property: our code for the initial segment of $X$ which is $\sigma * i$ depends only on $\mathcal{P}$, $i$, and the code for $\sigma$ (and not, for example, on $X(n)$ for $n > |\sigma * i|$).

In Gács’ approach, he does not code bit-by-bit, but rather breaks the streams to be coded into finite blocks of appropriately chosen lengths, and then codes each block rather than each bit one at a time. Coding in blocks in this way allows for a substantial reduction in the redundancy. Nevertheless, it is easily seen that weaker versions of the independence property and condition (1.3.1) still hold. If the $(n+1)$st block is of length $m_n$ then (1.3.1) will hold with two replaced by $2^m$. Similarly, the way in which we code the $(n+1)$st block will depend only on $\mathcal{P}$ and the coding of previous blocks. In order to achieve an exponentially smaller redundancy bound with our coding, we shall need to develop more general techniques, for which neither of these strong restrictions apply.

### 1.4 Background and organization

We assume a basic working knowledge of computability theory and its main concepts. Other than that, the proof of Lemma 1.2 is self-contained. In particular, knowledge of previous proofs of the Kučera-Gács theorem is not assumed. The reader who is interested in a more detailed analysis of the different approaches to the task of coding into random streams, is referred to the recent survey [BLP16]. For background on Martin-Löf randomness we refer to the textbooks Li and Vitányi [LV97], Downey and Hirschfeldt [DH10] or Nies [Nie09]. The latter two books also contain background in computability theory.

As we discussed in Section 1.3.1, the promised reduction of Lemma 1.2 will be achieved by means of a labelling of the full binary tree. Section 2 is devoted to the construction of this labelling and the statement of its key properties. In Section 3 and Section 4 we verify the properties of the labelling construction and complete the proof of Lemma 1.2.

### 2 Partial labelling of the full binary tree

The reduction needed for the proof of Lemma 1.2 is constructed via the enumeration of a partially labelled tree $\mathcal{T}$ with certain properties, which we construct in this section. Recall that we are given a $\Pi^0_1$ class $\mathcal{P}$ and an increasing computable sequence $(\ell_i)$ such that:

$$\sum_i 2^{-\ell_i+i} < \mu(\mathcal{P}). \tag{2.0.1}$$

The partially labelled tree $\mathcal{T}$ will be determined as the limit of a computable sequence $\{(\mathcal{T}_i)\}$ of partially labelled trees. We call such $(\mathcal{T}_i)$ a labelling process for $\mathcal{T}$. Let $Q$ be a c.e. set of binary strings such that
\( P = 2^\omega - \|Q\| \). Let \((Q_s)\) be a computable enumeration of \( Q \). Before we give the construction of \((T_s)\), we state a number of key properties that \((T_s)\) will have and define some relevant notions.

### 2.1 Basic properties of the labelling

The partially labelled tree \( T \) that we construct will be structured in the following sense.

**Definition 2.1** (Structured partially labelled trees). A partially labelled tree \( T \) is structured with respect to an increasing sequence \((\ell_i)\), if the following properties are met.

1. **Restriction**: only strings at levels \( \ell_i, i \in \mathbb{N} \) of \( T \) can have a label;
2. **Layering**: the labels placed on the level \( \ell_i \) of \( T \) are of the type \( x_\sigma \) where \( |\sigma| = i \);
3. **Completeness**: if label \( x_\sigma \) exists in \( T \) then all labels \( x_\rho, \rho \in 2^{\leq |\sigma|} \) exist in \( T \);
4. **Uniqueness**: each string in \( T \) can have at most one label;
5. **Consistency**: if \( \rho \) of level \( \ell_k \) in \( T \) has label \( x_\sigma \), then for each \( i < k \), \( \rho \upharpoonright \ell_i \) has label \( x_\sigma \upharpoonright \ell_i \).

\( T \) will be determined as the limit of a computable labelling process \((T_s)\) which is canonical with respect to the given \((\ell_i)\), in the following sense.

**Definition 2.2** (Canonical labelling process). A labelling process \((T_s)\) is canonical with respect to an increasing sequence \((\ell_i)\) if the following properties hold for all \( s \).

1. **Structure**: the tree \( T_s \) is structured with respect to \((\ell_i)\);
2. **Finiteness**: only strings of length at most \( \ell_s \) can have a label in \( T_s \);
3. **Persistence**: if \( \rho \) has label \( x_\sigma \) in \( T_s \), then it has the same label in \( T_t \) for all \( t > s \).

Clearly a canonical labelling process \((T_s)\) has a limit, which is a structured partially labelled tree. From now on we suppress the qualification ‘with respect to an increasing sequence \((\ell_i)\)’ when we use the notions of Definitions 2.1 and 2.2, and always assume the fixed sequence \((\ell_i)\) that is given in Lemma 1.2.

Note that Definition 2.1 and Definition 2.2 allow the possibility that a single label \( x_\sigma \) may have many copies at some level \( \ell_k \) of some \( T_s \).

### 2.2 Definitions for the labelling construction

The following notation will be handy.

**Definition 2.3** (Labelled subset and size). The set of labelled strings of a structured partially labelled tree \( T \) is denoted by \( T^* \). The length of the longest \( \sigma \) such that a label \( x_\sigma \) has been placed on a string in \( T^* \) is denoted \( \|T^*\| \).

The purpose of the labelling process is to ensure that eventually for every string \( \sigma \) there is a string \( \rho \) in \( T \) which is extendible in \( P \) and which has label \( x_\sigma \). In this sense, the enumeration \((Q_s)\) of \( Q \) is the main driver of the process, and determines the placement of additional copies of already existing labels. Timing is a crucial aspect of the verification of the success of the labelling, however, and for this reason we will not use
the arbitrary enumeration \((Q_s)\) directly in the construction. Instead, we use the following filtered version which takes into account the existing labelling at each stage.

**Definition 2.4** (Filtered enumeration of \(Q\)). During the construction we define a c.e. set of strings \(D\) inductively:

- at stage 0 let \(D_0 = \emptyset\);
- at stage \(s + 1\), if there exists a leaf of \(T_s^\ast\) which does not belong in \(D_s\) and has a prefix in \(Q_s\), pick the lexicographically least such leaf and enumerate it into \(D\);

where \(D_s\) denotes the set of strings enumerated in \(D\) by the end of stage \(s\).

Clearly \(\|D_s\| \subseteq \|Q_s\|\) while the converse is not generally true. Note that for a string \(\rho\) to enter \(D\) at stage \(s\) it is not enough to have a prefix in \(Q_s\). Hence \((D_s)\) is a filtered version of \((Q_s)\), in the sense that only previously labelled strings can be enumerated into \(D_s\).

As remarked previously, Definition 2.2 crucially allows for the possibility that a single label \(x_\rho\) may have many copies at some level \(\ell_n\) of some \(T_s\). Amongst all of the strings with the same label \(x_\rho\), however, we shall ensure that at any given time there is precisely one of these strings which is given the special status of being active. Roughly speaking, the active strings are those above which it presently seems there is still room for further coding at the next level. If \(|\rho| = n\) and the label \(x_\rho\) is placed on \(\tau\), then while \(\tau\) is active we may place labels for one element extensions of \(\sigma\) on the extensions of \(\tau\) of level \(\ell_{n+1}\). We shall do so as the demands of the construction require, working from left to right. As each of these labels are placed, we do not have to be concerned initially as to whether they are placed on strings with prefixes in \(Q\) – we simply place the labels and then wait for the enumeration of \(D\) to subsequently alert us if we have placed labels on strings which do not have extensions in \(P\). Once labels have been placed on all extensions of \(\tau\) of level \(\ell_{n+1}\), \(\tau\) is said to be saturated, and will no longer be active. It should be noted that at any given point, if \(\sigma \subset \sigma'\), \(\tau\) and \(\tau'\) have labels \(x_\rho\), \(x_{\rho'}\) respectively and are both active, it will not necessarily hold that \(\tau \subset \tau'\). We make the following definitions.

**Definition 2.5** (Active strings). Given a canonical labelling process \((T_s)\), a string \(\rho\) in \(T_s^\ast\) is active if it has some label \(x_\rho\) and \(\rho\) was the last string to receive this label in the approximations \(T_0, \ldots, T_s\).

A string in \(T_s^\ast\) that is not active is called inactive.

**Definition 2.6** (Saturated strings). A string \(\rho\) of level \(\ell_k\) of a structured partially labelled tree \(T\) is saturated if all of its extensions at level \(\ell_{k+1}\) of \(T\) are labelled.

Note that if \((T_s)\) is a canonical labelling process and a string of \(T_s\) is saturated, then the same string will also be saturated in \(T_t\) for all \(t > s\). Similarly, by Definition 2.5, if a string in \(T_s^\ast\) is inactive then the same string will also be inactive in \(T_t^\ast\) for all \(t > s\).

Each stage of the construction of \((T_s)\) after stage 0, will be one of the following two kinds.

**Definition 2.7** (Expansionary and adaptive stages). A stage \(s + 1\) is called expansionary if \(\|T_{s+1}\| > \|T_s^\ast\|\). Otherwise \(s + 1\) is called an adaptive stage.

It will be immediate from the construction that \(s + 1\) is expansionary if and only if \(D_{s+1} = D_s\).

In the labelling construction we will explicitly deactivate strings in order to emphasize the newly inactive strings. It will be evident that this is compatible with Definition 2.5.
**Definition 2.8** (Cloning a branch). Given \( \delta, \beta \in \mathcal{T}_s^* \) such that \( \delta \) is a leaf, suppose that:

(i) if \( x_\sigma, x_\tau \) are the labels of \( \beta, \delta \) respectively then \( \sigma \subset \tau \); also \( \beta \) has length \( \ell_k \) for some \( k \);

(ii) \( \eta \) is the leftmost string of length \( |\delta| \) which extends \( \beta \) and \( \eta \uparrow_{\ell_{k+1}} \) is not labelled.

Cloning \( \delta \) above \( \beta \) means to label \( \eta \uparrow_{\ell_i} \) with the label of \( \delta \uparrow_{\ell_i} \) for each \( i \) such that \( \ell_i \in (|\beta|, |\delta|] \), making each of these strings active.

In Definition 2.8, we allow the case that \( \beta \) is the empty string \( \lambda \), in which case there is no label placed on \( \beta \) and we must regard \( \beta \) as being of length \( \ell_{-1} \). Given a labelled string \( \rho \) in \( \mathcal{T}_s \), the active clone of \( \rho \) in \( \mathcal{T}_s \) is the unique active string in \( \mathcal{T}_s \) which has the same label as \( \rho \). For uniformity, we define the active clone of the empty string \( \lambda \) to be \( \lambda \).

### 2.3 The labelling construction

At stage 0 we place a label \( x_1 \) on the leftmost string of length \( \ell_0 \) and make this string active. At stage \( s + 1 \) suppose that the labelled tree \( \mathcal{T}_s \) has been defined, and consider the following two cases:

**Expansionary stage:** If \( D_{s+1} = D_s \), then let \( \mathcal{T}_{s+1} \uparrow_{\ell_i} = \mathcal{T}_s \uparrow_{\ell_i} \) and for each active leaf \( \rho \) of \( \mathcal{T}_s^* \) with label some \( x_\sigma \), place labels \( x_{\sigma_0}, x_{\sigma_1} \) on the leftmost and rightmost extensions of \( \rho \) of level \( \ell_{s+1} \), making these strings active, then end stage \( s + 1 \).

**Adaptive stage:** If \( D_{s+1} \neq D_s \) then let \( \delta \) be the string in \( D_{s+1} \setminus D_s \) and let \( \alpha_j, j \leq k \) be the empty or labelled initial segments of \( \delta \) in order of magnitude, so that \( \alpha_0 = \lambda \) and \( \alpha_k = \delta \). Also let \( \beta_j, j \leq k \) be the active clones of \( \alpha_j, j \leq k \) respectively in \( \mathcal{T}_s \). Let \( j_0 \) be the largest number \( j < k \) such that \( \beta_j \) is not saturated and

- deactivate \( \beta_j \) for each \( j \in (j_0, k] \);
- clone \( \delta \) above \( \beta_{j_0} \).

If such \( j_0 \) does not exist, say that the construction terminates at stage \( s + 1 \); otherwise end stage \( s + 1 \).

### 3 Properties of the labelling algorithm

Note that since \( (\ell_i) \) is increasing, each string of length \( \ell_k \) has at least two distinct extensions of length \( \ell_{k+1} \). Hence the expansionary stages of the construction are well-defined. A straightforward induction on stages suffices to establish that \( (\mathcal{T}_s) \) is a canonical labelling process, according to Definition 2.2. In particular, the placing of labels satisfies the consistency condition required in order to define a valid functional. While Definition 2.5 specifies the active strings at each stage, during the construction we have also directly deactivated strings, as well as activating them during the process of cloning and at expansionary stages. It is clear that at any stage the strings which have been activated and not directly deactivated by the construction, are precisely those which are active according to Definition 2.5, since it is precisely when we place a new version of a given label that we deactivate the previously active string with that label. It also follows by a straightforward induction on stages, that at the end of each stage \( s \), any leaf of \( \mathcal{T}_s^* \) is either active, or else has already been enumerated into \( D_s \). In particular, when \( \delta \) is enumerated into \( D_{s+1} \) during stage \( s + 1 \), it was previously active and is deactivated during this adaptive stage. This means, in the notation of Section 2.3, that when we deactivate \( \beta_k \) without requiring that it be saturated, in fact \( \beta_k = \delta \), so that the only
strings which are deactivated during an adaptive stage are strings which are enumerated into $D$, or else are saturated:

Inactive labelled strings in $T_s$ are either saturated or else belong to $D_s$. \hfill (3.0.1)

The following is also established easily by induction on stages:

If $\delta \in D$ then for all $s$, no proper extension of $\delta$ is labelled in $T_s$. \hfill (3.0.2)

3.1 Non-termination

In order to show that the labelling construction does not terminate (i.e. that we do not run out of room for coding), it suffices to establish that $\lambda$ is never saturated (regarding $\lambda$ as of level $\ell - 1$ in the definition of saturation). The following definition will be useful.

**Definition 3.1** (Set of active strings). Let $U_s$ be the set of active strings in $T_s$. Moreover for each string $\rho$ let $U_s(\rho)$ be the set of strings $\gamma \supseteq \rho$ which are active in $T_s$.

We are interested in the weight of the active strings, where the weight of a set of strings $V$ is defined by

$$\text{wgt}(V) = \sum_{\eta \in V} 2^{-|\eta|}.$$ 

In order to show that $\lambda$ is never saturated we shall first establish:

$$\text{wgt}(U_s) + \mu(D_s) < 1 \text{ for all stages } s. \hfill (3.1.1)$$

The following claim will also be established by induction on stages:

Given any $s$ and any stream $Z$ which does not have a prefix in $D_s$, the largest labelled initial segment of $Z$ is active in $T_s$. \hfill (3.1.2)

Note that in this statement it is possible that $Z$ does not have a labelled initial segment, in which case the assertion is trivially true. An immediate consequence of (3.1.2) is that

For each $s$ and each $\nu$ which is labelled in $T_s$, we have $$[\nu] \subseteq [D_s] \cup [U_s(\nu)]. \hfill (3.1.3)$$

Now if $\lambda$ is saturated at stage $s$ then the entire Cantor space is covered by the labelled strings of length $\ell_0$. Hence by (3.1.3) we have $2^{\omega} \subseteq [D_s] \cup [U_s]$. Then $1 \leq \mu(D_s) + \text{wgt}(U_s)$, which contradicts (3.1.1).

It remains to establish (3.1.1) and (3.1.2). To see (3.1.1), note first that at each stage $s$ and for each $\sigma$ there is at most one active string in $T_s$ with label $x_\sigma$. Since for each $n$ there are only $2^n$ strings of length $n$, we have:

$$\text{wgt}(U_s) \leq \sum_{\rho \in U_s} 2^{-|\rho|} \leq \sum_{n} \left( \sum_{\rho \in U_s \cap 2^{\ell_0}} 2^{-|\rho|} \right) \leq \sum_{n} \left( 2^n \cdot 2^{-\ell_0} \right) = \sum_{n} 2^{n-\ell_0}. $$

If we combine this with our hypothesis (2.0.1) we get $\text{wgt}(U_s) < \mu(\mathcal{P}) = 1 - \mu(\mathcal{Q}_s)$. By the fact $[D_s] \subseteq [\mathcal{Q}_s]$ which we observed after Definition 2.4, we get $\text{wgt}(U_s) < 1 - \mu(D_s)$, from which (3.1.1) follows.

It remains to prove (3.1.2) by induction on the stages of the labelling construction. At stage 0 we have $D_0 = \emptyset$ and all labelled strings are active. It follows that in this case (3.1.2) holds. Inductively suppose
Recall that strings receive labels. Since construction does not terminate that there are infinitely many expansionary stages. Hence:

This completes the induction step and the proof of (3.1.2).

Now suppose that $s + 1$ is an adaptive stage and let $\delta$ be the unique element of $D_{s+1} - D_s$. Also let $Z$ be a stream which has at least one labelled initial segment in $T_s$, and let $\nu$ be the largest such initial segment. If $\nu = \delta$ there is nothing to prove, so assume otherwise. If no initial segment of $Z$ is deactivated during stage $s + 1$, the claim follows by the induction hypothesis. For the remaining case, let $\eta$ be the largest labelled prefix of $Z$ which is deactivated during stage $s + 1$. If $|\eta| = \ell_k$, let $\eta' = Z \mid \ell_{k+1}$. Since $\eta$ was deactivated at $s + 1$, it follows from (3.0.1) that it was saturated in $T_s$. This means that $\eta'$ must be labelled in $T_s$. Hence the largest labelled initial segment of $Z$ in $T_s$, which is also the largest in $T_{s+1}$, is active in $T_{s+1}$, just as it was active in $T_s$. Hence (3.1.2) holds for $Z$ at stage $s + 1$. Finally consider the case where $Z$ did not have a labelled initial segment in $T_s$, but it does in $T_{s+1}$. Since all the newly labelled strings at stage $s + 1$ are active in $T_{s+1}$, in this case also we can conclude that (3.1.2) holds for $Z$ at stage $s + 1$.

This completes the induction step and the proof of (3.1.2).

## 3.2 Growth of the tree and the enumeration of $\mathcal{D}$

Note that at each stage $s$ the set $T^*_s$ is finite, and that in every adaptive stage some previously unlabelled strings receive labels. Since $(T_s)$ is a canonical labelling process, it follows from the fact that the labelling construction does not terminate that there are infinitely many expansionary stages. Hence:

$$\lim_s \|T^*_s\| = \infty. \tag{3.2.1}$$

Recall that $T^*$ is the limit of all $T^*_s$. We wish to show that:

$$\text{If } \tau \in T^* \cap \|Q\| \text{ then there exists } s \text{ such that } \|\tau\| \subseteq \|D_s\|. \tag{3.2.2}$$

This will follow once we establish the following fact:

$$\text{If } \tau \in T^*_s \text{ is active, there is at least one leaf of } T^*_s \text{ extending } \tau \text{ which is not in } D_s. \tag{3.2.3}$$

In order to see that (3.2.2) follows from (3.2.3), suppose that $s_0$ is the least stage at which $\tau \in T^*_{s_0}$ and there exists $\tau' \subseteq \tau$ with $\tau' \in Q_{s_0}$. Let $s_1$ be the least expansionary stage $> s_0$. At the beginning of stage $s_1$, (3.2.3) implies that no string $\tau'' \supseteq \tau$ in $T^*_{s_1}$ can be active, meaning that all such strings must either be saturated or else belong to $D_{s_1}$, by (3.0.1). This implies that $\|\tau\| \subseteq \|D_{s_1}\|$ as required.

We establish (3.2.3) by induction on stages. If $s_0$ is the first stage at which $\tau$ is active, then no extensions of $\tau$ are in $D_{s_0}$. At any subsequent stage $s > s_0$ at which $\tau$ is still active, if a leaf $\delta$ extending $\tau$ is enumerated into $D_s$, then that leaf will be cloned above some $\beta_{j_0} \supseteq \tau$, which completes the induction step.
4 The coding process and its verification

In this section we show how to determine the code \( Y \) of a given binary stream \( X \), so that \( X \upharpoonright_n \) can be uniformly computed by \( Y \upharpoonright_{\ell_n} \). We define this reduction based on the labelling process of Section 2 and its properties. Given \( X \) we define \( Y \) as the limit of an approximation \( (Y_s) \) which is computable relative to \( X \) and converges to a limit. We therefore have \( Y \leq_T X' \). The approximation \( (Y_s) \) will be driven by \( X \) and the labelling algorithm, which is in turn driven by the enumeration of the complement of \( \mathcal{P} \).

4.1 The coding process

Given \( X \), at stage \( s \) let \( Y_s \) be the leaf of \( T_s^+ \) with label the first \( \|T_s^+\| \) bits of \( X \). Then for each \( n \) define \( Y(n) = \lim_s Y_s(n) \). This determines the code \( Y \) of \( X \).

4.2 The coding verification

We verify that the code \( Y \) of \( X \) determined by the above construction has the required properties.

\( \blacktriangleright \) \( Y \) is well-defined and \( Y \leq_T X' \)

By (3.2.1) we have \( \lim_s \|T_s^+\| = \infty \), so \( \lim_s |Y_s| = \infty \). We also need to show that for each \( n \) the limit \( \lim_s Y_s(n) \) exists. To this end, it suffices to show that for each \( n \), \( \lim_s Y_s \upharpoonright_{\ell_n} \) exists. Fix \( n \) and let \( \sigma = X \upharpoonright_n \).

Suppose that at stage \( s + 1 \) we have \( Y_{s+1} \upharpoonright_{\ell_n} \neq Y_s \upharpoonright_{\ell_n} \). Then \( s + 1 \) must be an adaptive stage during which \( Y_s \) is enumerated into \( \mathcal{D} \). Clearly \( Y_{s+1} \upharpoonright_{\ell_n} \) and \( Y_s \upharpoonright_{\ell_n} \) are both strings with label \( x_\sigma \). We aim to show that \( Y_{s+1} \upharpoonright_{\ell_n} \) is labelled later than \( Y_s \upharpoonright_{\ell_n} \), meaning that at most \( 2^{\ell_n} \) such changes are possible. The proof is by induction on stages, and for all \( X \) (and their corresponding \( Y \)) simultaneously. Let \( \beta_0, \ldots, \beta_k \) and \( j_0 \) be defined as during the construction at stage \( s + 1 \), so that \( \beta_k = Y_s \). If \( j_0 < n \) then the result is immediate. If \( j_0 \geq n \), and \( Y_s \upharpoonright_{\ell_n} \) was labelled later than \( Y_{s+1} \upharpoonright_{\ell_n} \), then consider which of \( \beta_{j_0} \) and \( Y_s \upharpoonright_{\ell_n} \) was labelled first. If \( Y_s \upharpoonright_{\ell_{j_0}} \) is labelled first, then the stage at which \( \beta_{j_0} \) becomes active constitutes a contradiction to the induction hypothesis. On the other hand, if \( \beta_{j_0} \) is labelled first, then it becomes deactivated when \( Y_s \upharpoonright_{\ell_{j_0}} \) is labelled, contradicting the fact that it is active during stage \( s + 1 \). This shows that for each \( n \) there can be at most \( 2^{\ell_n} \) many stages \( s + 1 \) such that \( Y_{s+1}(n) \neq Y_s(n) \). Also, since \( Y \) is defined as the limit of a computable approximation relative to \( X \), we have \( Y \leq_T X' \). Note that not much more can be said with regard to this reduction, since there is no apparent way to compute \( Y(n) \) in a more efficient way, e.g. with a restricted use of \( X \).

\( \blacktriangleright \) \( Y \) belongs to \( \mathcal{P} \)

It suffices to show that \( Y \) does not have a prefix in \( \mathcal{Q} \). By the definition of \( Y \) it follows that \( Y \in [T^+] \). Hence by (3.2.2) it suffices to show that \( Y \notin [\mathcal{D}] \). This follows from (3.0.2).
We show that for each \( n \) we can compute \( X \upharpoonright n \) uniformly from \( Y \upharpoonright \ell_n \). Given \( n \) and \( Y \upharpoonright \ell_n \) we simply run the labelling construction until the first stage \( s_0 \) where the string \( Y \upharpoonright \ell_n \) is labelled in \( T_{s_0} \). Note that since \( Y \in [T^*] \) and \( T \) is a structured partially labelled tree, such a stage exists. Let \( x_\sigma \) be the label of \( Y \upharpoonright \ell_n \). Since the string \( Y \upharpoonright \ell_n \) is not going to change label at later stages, and since \( Y \) is the limit of all \( Y_s \), by our definition of \( Y \) given \( X \) in Section 4.1 it follows that \( X \upharpoonright n = \sigma \).

This concludes the verification of the coding process and the proof of Lemma 1.2.

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