

SCS&E Report 9404
March, 1994

On the Intrinsic Complexity of Language Identification

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Abstract

A new investigation of the complexity of language identification is undertaken using the notion of reduction from recursion theory and complexity theory. The approach, referred to as the intrinsic complexity of language identification, employs notions of ‘weak’ and ‘strong’ reduction between learnable classes of languages. The intrinsic complexity of several classes are considered and the results agree with the intuitive difficulty of learning these classes. Several complete classes are shown for both the reductions and it is also established that the weak and strong reductions are distinct.

An interesting result is that the self referential class of Wiehagen in which the minimal element of every language is a grammar for the language and the class of pattern languages introduced by Angluin are equivalent in the strong sense.

This study has been influenced by a similar treatment of function identification by Freivalds, Kinber, and Smith.

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1 Introduction

The present paper introduces a novel way to look at the difficulty of learning collections of languages from positive data. Most studies on feasibility issues in learning have concentrated on the complexity of the learning algorithm. The present paper describes a model which provides an insight into why certain classes are more easily learned than others. Our model adopts a similar study in the context of learning functions by Freivalds, Kinber, and Smith [8]. The main idea of the approach is to introduce reductions between collections of languages. If a collection \mathcal{L}_1 can be reduced to a collection \mathcal{L}_2 , then the learnability of \mathcal{L}_1 is no more difficult than that of \mathcal{L}_2 . We illustrate our ideas with the help of simple examples.

Consider the following collections of languages over N , the set of natural numbers.

$SINGLE = \{L \mid L \text{ is singleton} \}$.

$COINIT = \{L \mid (\exists n)[L = \{x \mid x \geq n\}]\}$.

$FIN = \{L \mid \text{cardinality of } L \text{ is finite} \}$.

So, $SINGLE$ is the collection of all singleton languages, $COINIT$ is the collection of languages that contain all natural numbers except a finite initial segment, and FIN is the collection of all finite languages. Clearly, each of these three classes is identifiable in the limit from only positive data. For example, a machine \mathbf{M}_1 that upon encountering the first data element, say n , keeps on emitting a grammar for the singleton language $\{n\}$ identifies $SINGLE$. A machine \mathbf{M}_2 that, at any given time, finds the minimum element among the data seen so far, say n , and emits a grammar for the language $\{x \mid x \geq n\}$ can easily be seen to identify $COINIT$. Similarly, a machine \mathbf{M}_3 that continually outputs a grammar for the finite set of data seen so far identifies FIN .

Now, although these three classes are identifiable, it can be argued that they present learning problems of varying difficulty. One way to look at the difficulty is to ask the question, “At what stage in the processing of the data can a learning machine confirm its success?” In the case of $SINGLE$, the machine can be confident of success as soon as it encounters the first data element. In the case of $COINIT$, the machine cannot always be sure that it has identified the language. However, at any stage after it has seen the first data element, the machine can provide an upper bound on the number of further mind changes that the machine will make before converging to a correct grammar. For example, if at some stage the minimum element seen is m , then \mathbf{M}_2 will make no more than m further mind changes because it changes its mind only if a smaller element appears. In the case of FIN , the learning machine can neither be confident about its success nor can it, at any stage, provide an upper bound on the number of further mind changes that it may have to undergo before it is rewarded with success. Clearly, these three collections of languages pose learning problems of varying difficulty where $SINGLE$ appears to be the least difficult to learn and FIN is seen to be the most difficult to learn with $COINIT$ appearing to be of intermediate difficulty. The model described in the present paper captures this gradation in difficulty of various identifiable collections of languages. Following Freivalds, Kinber, and Smith [8], we refer to such a notion of difficulty as “intrinsic complexity.”

We next present an informal description of reductions that are central to our analysis

of the intrinsic complexity of language learning. We discuss our results in the context of the identification in the limit paradigm [9]. The analysis can easily be applied to other paradigms like finite identification, behaviorally correct identification [13, 7], and vacillatory identification [13, 5] and will be presented in the full paper. We next introduce some technical notions about language learning in order to facilitate our discussion.

Informally, a *text* for a language L is just an infinite sequence of elements, with possible repetitions, of all and only the elements of L . A text for L is thus an abstraction of the presentation of positive data about L . Elements of a text are sequentially fed to a learning machine one element at a time. The machine, as it receives elements of the text, outputs an infinite sequence of grammars. There are numerous criteria for a learning machine to be successful on a text. If the infinite sequence of grammars converges to a single correct grammar for the content of the text, then the machine is said to **TextEx**-identify the text. A machine is said to **TextEx**-identify a language just in case it **TextEx**-identifies each text for the language. **TextEx**-identification is essentially *identification in the limit* paradigm introduced by Gold [9]. It is also useful to call a sequence of grammars, g_0, g_1, g_2, \dots , **TextEx**-admissible for a text T just in case the sequence of grammars converges to a single correct grammar for the language represented by text T .

Our reductions are based on the idea that for a collection of languages \mathcal{L} to be reducible to \mathcal{L}' , we should be able to transform texts T for languages in \mathcal{L} to texts T' for languages in \mathcal{L}' and further transform **TextEx**-admissible sequences for T' into **TextEx**-admissible sequences for T . This is achieved with the help of two enumeration operators. Informally, enumeration operators are algorithmic devices that map infinite sequences of objects (for example, texts and infinite sequences of grammars) into infinite sequences of objects. The first operator, Θ , transforms texts for languages in \mathcal{L} into texts for languages in \mathcal{L}' . The second operator, Ψ , behaves in the following way. Suppose g_0, g_1, g_2, \dots is a **TextEx**-admissible sequence for a text for a language in \mathcal{L}' which was formed by applying the operator Θ to a text T for a language $L \in \mathcal{L}$. Then Ψ transforms the sequence g_0, g_1, g_2, \dots into a **TextEx**-admissible sequence for the text T .

To see that the above satisfies the intuitive notion of reduction consider collections \mathcal{L} and \mathcal{L}' such that \mathcal{L} is reducible to \mathcal{L}' . We now argue that if \mathcal{L}' is identifiable then so is \mathcal{L} .

Let \mathbf{M}' **TextEx**-identify \mathcal{L}' . Let enumeration operators Θ and Ψ witness the reduction of \mathcal{L} to \mathcal{L}' . Then we describe a machine \mathbf{M} that **TextEx**-identifies \mathcal{L} . \mathbf{M} , upon being fed a text T for some language $L \in \mathcal{L}$, uses Θ to construct a text T' for a language in \mathcal{L}' . It then simulates machine \mathbf{M}' on text T' and feeds conjectures of \mathbf{M}' to the operator Ψ to produce its conjectures. It is easy to verify that the properties of Θ , Ψ , and \mathbf{M}' guarantee the success of \mathbf{M} on each text for each language in \mathcal{L} .

We show that under the above reduction, *SINGLE* is reducible to *COINIT* but *COINIT* is not reducible to *SINGLE*. We also show that *COINIT* is reducible to *FIN* while *FIN* is not reducible to *COINIT*, thereby justifying our intuition about the intrinsic complexity of these classes. We also show that *FIN* is in fact complete with respect to the above reduction. Additionally, we study the status of numerous language classes with respect to this reduction and show several of them to be complete.

We also consider a stronger notion of reduction than the one discussed above. The

reader should note that in the above reduction, different texts for the same language may be transformed into texts for different languages by Θ . If we further require that Θ is such that it transforms all texts for a language into texts for some unique language then we have a stronger notion of reduction. In the context of function learning [8], these two notions of reductions are the same. However, surprisingly, in the context of language identification this stronger notion of reduction turns out to be different from its weaker counterpart as we are able to show that *FIN* is not complete with respect to the stronger reduction. We give an example of complete class with respect to the strong reduction.

We now discuss two interesting collections that are shown not to be complete with respect to either reduction.

The first one is a class of languages introduced by Wiehagen [15] which contains all those languages L such that the minimum element in L is a grammar for L . This self-referential class, which can be **TextEx**-identified, is a very interesting class as it contains a finite variant of every recursively enumerable language. We show that this class is not complete and is in fact equivalent to *COINIT* under the strong reduction.

The second class is the collection of pattern languages introduced by Angluin [1]. Pattern languages have been studied extensively in the computational learning theory literature since their introduction as a nontrivial class of languages that could be learned in the limit from only positive data. We show that pattern languages are also equivalent to *COINIT* in the strong sense, thereby implying that they pose a learning problem of similar difficulty to that of Wiehagen's class.

Finally, we also study intrinsic complexity for identification from both positive and negative data. As in the case of functions, the weak and strong reductions result in the same notion. We show that *FIN* is complete for identification from both positive and negative data, too.

We now proceed formally. In Section 2, we present notation and preliminaries from language learning theory. In Section 3, we introduce our reducibilities. Results are presented in Section 4. Finally, in Section 5, we look at the intrinsic complexity of language identification from both positive and negative data.

2 Notation and Preliminaries

Any unexplained recursion theoretic notation is from [14]. The symbol N denotes the set of natural numbers, $\{0, 1, 2, 3, \dots\}$. Unless otherwise specified, $e, g, i, j, k, l, m, n, q, r, s, t, w, x, y$, with or without decorations¹, range over N . Symbols $\emptyset, \subseteq, \subset, \supseteq, \supset$ denote empty set, subset, proper subset, superset, and proper superset, respectively. Symbols A and S , with or without decorations, range over sets. S , with or without decorations, ranges over finite sets. D_0, D_1, \dots , denotes a canonical recursive indexing of all the finite sets [14]. We assume that if $D_i \subseteq D_j$ then $i \leq j$ (indexing defined in [14] satisfies this property).

Cardinality of a set S is denoted by $\text{card}(S)$. The maximum and minimum of a set are denoted by $\max(\cdot), \min(\cdot)$, respectively, where $\max(\emptyset) = 0$ and $\min(\emptyset) = \infty$.

¹Decorations are subscripts, superscripts and the like.

Unless otherwise specified, letters f, F and h , with or without decorations, range over *total* functions with arguments and values from N . Symbol \mathcal{R} denotes the set of all total computable functions. A pair $\langle i, j \rangle$ stands for an arbitrary, computable, one-to-one encoding of all pairs of natural numbers onto N [14]. We define $\pi_1(\langle x, y \rangle) = x$ and $\pi_2(\langle x, y \rangle) = y$. $\langle \cdot, \cdot \rangle$ can be extended to n -tuples in a natural way.

By φ we denote a fixed *acceptable* programming system for the partial computable functions: $N \rightarrow N$ [14, 11]. By φ_i we denote the partial computable function computed by program i in the φ -system. The letter, p , in some contexts, with or without decorations, ranges over programs; in other contexts p ranges over total functions with its range being construed as programs. By Φ we denote an arbitrary fixed Blum complexity measure [3, 10] for the φ -system. By W_i we denote $\text{domain}(\varphi_i)$. W_i is, then, the r.e. set/language ($\subseteq N$) accepted (or equivalently, generated) by the φ -program i . We also say that i is a grammar for W_i . Symbol \mathcal{E} will denote the set of all r.e. languages. Symbol L , with or without decorations, ranges over \mathcal{E} . Symbol \mathcal{L} , with or without decorations, ranges over subsets of \mathcal{E} . We denote by $W_{i,s}$ the set $\{x \leq s \mid \Phi_i(x) < s\}$. \downarrow denotes defined. \uparrow denotes undefined.

We now present concepts from language learning theory. The definition below introduces the concept of a *sequence* of data.

Definition 1

- (a) A *sequence* σ is a mapping from an initial segment of N into $(N \cup \{\#\})$. Empty sequence is denoted by Λ .
- (b) The *content* of a sequence σ , denoted $\text{content}(\sigma)$, is the set of natural numbers in the range of σ .
- (c) The *length* of σ , denoted by $|\sigma|$, is the number of elements in σ . So, $|\Lambda| = 0$.
- (d) For $n \leq |\sigma|$, the initial sequence of σ of length n is denoted by $\sigma[n]$. So, $\sigma[0]$ is Λ .
- (e) The last element of a nonempty sequence σ is denoted $\text{last}(\sigma)$; the last element of Λ is defined to be 0. Formally, $\text{last}(\sigma) = \sigma[|\sigma| - 1]$ if $\sigma \neq \Lambda$, otherwise $\text{last}(\sigma)$ is defined to be 0.
- (f) The result of stripping last element from sequence σ is denoted $\text{prev}(\sigma)$. Formally, if $\sigma \neq \Lambda$, then $\text{prev}(\sigma) = \sigma[|\sigma| - 1]$, else $\text{prev}(\sigma) = \Lambda$.

Intuitively, $\#$'s represent pauses in the presentation of data. We let σ, τ , and γ , with or without decorations, range over finite sequences. We denote the sequence formed by the concatenation of τ at the end of σ by $\sigma \diamond \tau$. Sometimes we abuse the notation and use $\sigma \diamond x$ to denote the concatenation of sequence σ and the sequence of length 1 which contains the element x . SEQ denotes the set of all finite sequences.

Definition 2 A *language learning machine* is an algorithmic device which computes a mapping from SEQ into N .

We let \mathbf{M} , with or without decorations, range over learning machines.

Definition 3

- (a) A *text* T for a language L is a mapping from N into $(N \cup \{\#\})$ such that L is the set of natural numbers in the range of T .
- (b) The *content* of a text T , denoted $\text{content}(T)$, is the set of natural numbers in the range of T .
- (c) $T[n]$ denotes the finite initial sequence of T with length n .

Thus, $\mathbf{M}(T[n])$ is interpreted as the grammar (index for an accepting program) conjectured by learning machine \mathbf{M} on initial sequence $T[n]$.

There are several criteria for a learning machine to be successful on a language. The one defined below was introduced by Gold [9] and is also known in the literature as “identification in the limit.”

Definition 4 [9]

- (a) \mathbf{M} **TxtEx**-identifies a text T just in case $(\exists i \mid W_i = \text{content}(T)) (\forall n)[\mathbf{M}(T[n]) = i]$.
- (b) \mathbf{M} **TxtEx**-identifies an r.e. language L (written: $L \in \mathbf{TxtEx}(\mathbf{M})$) just in case \mathbf{M} **TxtEx**-identifies each text for L .
- (c) $\mathbf{TxtEx} = \{\mathcal{L} \subseteq \mathcal{E} \mid (\exists \mathbf{M})[\mathcal{L} \subseteq \mathbf{TxtEx}(\mathbf{M})]\}$.

Other criteria of success are finite identification, behaviorally correct identification [13, 7], and vacillatory identification [13, 5]. In the present extended abstract, we only discuss results about **TxtEx**-identification; results relating to the remaining criteria will be presented in the full paper.

3 Weak and Strong Reductions

Before we present our reductions we introduce some technical machinery.

We write “ $\sigma \subseteq \tau$ ” if σ is an initial segment of τ , and “ $\sigma \subset \tau$ ” if σ is a proper initial segment of τ . Likewise, we write $\sigma \subset T$ if σ is an initial finite sequence of text T . Let finite sequences $\sigma^0, \sigma^1, \sigma^2, \dots$ be given such that $\sigma^0 \subseteq \sigma^1 \subseteq \sigma^2 \subseteq \dots$ and $\lim_{i \rightarrow \infty} |\sigma^i| = \infty$. Then there is a unique text T such that for all $n \in N$, $\sigma^n = T[|\sigma^n|]$. This text is denoted $\bigcup_n \sigma^n$. Let \mathcal{T} denote the set of all texts, that is, the set of all infinite sequences over $N \cup \{\#\}$.

We define an *enumeration operator*, Θ , to be an algorithmic mapping from SEQ into SEQ such that for all $\sigma, \tau \in \text{SEQ}$, if $\sigma \subseteq \tau$, then $\Theta(\sigma) \subseteq \Theta(\tau)$. We further assume that for all texts T , $\lim_{n \rightarrow \infty} |\Theta(T[n])| = \infty$. By extension, we think of Θ as also defining a mapping from \mathcal{T} into \mathcal{T} such that $\Theta(T) = \bigcup_n \Theta(T[n])$.

A final notation about the operator Θ . If for a language L , there exists an L' such that for each text T for L , $\Theta(T)$ is a text for L' , then we write $\Theta(L) = L'$, else we say

that $\Theta(L)$ is undefined. The reader should note the overloading of this notation because the type of the argument to Θ could be a sequence, a text, or a language; it will be clear from the context which usage is intended.

We also need the notion of an infinite sequence of grammars. We let G , with or without decorations, range over infinite sequences of grammars. From the discussion in previous section it is clear that infinite sequences of grammars are essentially infinite sequences over N . Hence, we adopt the machinery defined for sequences and texts over to finite sequences of grammars and infinite sequences of grammars. Hence, if $G = g_0, g_1, g_2, g_3, \dots$, then $G[3]$ denotes the sequence g_0, g_1, g_2 , $G(3)$ is g_3 , $\text{last}(G[3])$ is g_2 , and $\text{prev}(G[3])$ is the sequence g_0, g_1 .

Let \mathbf{I} be an identification criterion. We say that an infinite sequence of grammars G is \mathbf{I} -admissible for text T just in case G is an infinite sequence of grammars witnessing \mathbf{I} -identification of text T . So, if $G = g_0, g_1, g_2, \dots$ is a **TextEx**-admissible sequence for T , then there exists n such that for all $n' \geq n$, $g_{n'} = g_n$ and $W_{g_n} = \text{content}(T)$.

We now introduce our first reduction.

Definition 5 Let $\mathcal{L}_1 \subseteq \mathcal{E}$ and $\mathcal{L}_2 \subseteq \mathcal{E}$ be given. Let identification criteria \mathbf{I}_1 and \mathbf{I}_2 be given. Let $\mathcal{T}_1 = \{T \mid T \text{ is a text for } L \in \mathcal{L}_1\}$. Let $\mathcal{T}_2 = \{T \mid T \text{ is a text for } L \in \mathcal{L}_2\}$. We say that $\mathcal{L}_1 \leq_{\text{weak}}^{\mathbf{I}_1, \mathbf{I}_2} \mathcal{L}_2$ just in case there exist operators Θ and Ψ such that for all $T \in \mathcal{T}_1$ and for all infinite sequences of grammars $G = g_0, g_1, \dots$ the following hold:

- (a) $\Theta(T) \in \mathcal{T}_2$ and
- (b) if G is an \mathbf{I}_2 -admissible sequence for $\Theta(T)$, then $\Psi(G)$ is an \mathbf{I}_1 -admissible sequence for T .

We say that $\mathcal{L}_1 \leq_{\text{weak}}^{\mathbf{I}} \mathcal{L}_2$ iff $\mathcal{L}_1 \leq_{\text{weak}}^{\mathbf{I}, \mathbf{I}} \mathcal{L}_2$. We say that $\mathcal{L}_1 \equiv_{\text{weak}}^{\mathbf{I}} \mathcal{L}_2$ iff $\mathcal{L}_1 \leq_{\text{weak}}^{\mathbf{I}} \mathcal{L}_2$ and $\mathcal{L}_2 \leq_{\text{weak}}^{\mathbf{I}} \mathcal{L}_1$.

We have deliberately made the above reduction general. In the present extended abstract, we present results about $\leq_{\text{weak}}^{\mathbf{I}}$ reductions only. We now define the corresponding notions of hardness and completeness for the above reduction.

Definition 6 Let \mathbf{I} be an identification criterion. Let $\mathcal{L} \subseteq \mathcal{E}$ be given.

- (a) If for all $\mathcal{L}' \in \mathbf{I}$, $\mathcal{L}' \leq_{\text{weak}}^{\mathbf{I}} \mathcal{L}$, then \mathcal{L} is $\leq_{\text{weak}}^{\mathbf{I}}$ -hard.
- (b) If \mathcal{L} is $\leq_{\text{weak}}^{\mathbf{I}}$ -hard and $\mathcal{L} \in \mathbf{I}$, then \mathcal{L} is $\leq_{\text{weak}}^{\mathbf{I}}$ -complete.

Intuitively, $\mathcal{L}_1 \leq_{\text{weak}}^{\mathbf{I}} \mathcal{L}_2$ just in case there exists an operator Θ that transforms texts for languages \mathcal{L}_1 into texts for languages in \mathcal{L}_2 and there exists another operator Ψ that transforms \mathbf{I} -admissible sequences for texts $\Theta(T)$ into \mathbf{I} -admissible sequences for T . It should be noted that there is no requirement that Θ map all the texts for a language in \mathcal{L}_1 into texts for a unique language in \mathcal{L}_2 . If we further place such a constraint on Θ , we get the following stronger notion.

Definition 7 Let $\mathcal{L}_1 \subseteq \mathcal{E}$ and $\mathcal{L}_2 \subseteq \mathcal{E}$ be given. We say that $\mathcal{L}_1 \leq_{\text{strong}}^{\mathbf{I}_1, \mathbf{I}_2} \mathcal{L}_2$ just in case there exist operators Θ, Ψ witnessing that $\mathcal{L}_1 \leq_{\text{weak}}^{\mathbf{I}_1, \mathbf{I}_2} \mathcal{L}_2$, and for all $L_1 \in \mathcal{L}_1$, there exists an $L_2 \in \mathcal{L}_2$, such that $(\forall \text{ texts } T \text{ for } L_1)[\Theta(T) \text{ is a text for } L_2]$.

We say that $\mathcal{L}_1 \leq_{\text{strong}}^{\mathbf{I}} \mathcal{L}_2$ iff $\mathcal{L}_1 \leq_{\text{strong}}^{\mathbf{I}, \mathbf{I}} \mathcal{L}_2$. We say that $\mathcal{L}_1 \equiv_{\text{strong}}^{\mathbf{I}} \mathcal{L}_2$ iff $\mathcal{L}_1 \leq_{\text{strong}}^{\mathbf{I}} \mathcal{L}_2$ and $\mathcal{L}_2 \leq_{\text{strong}}^{\mathbf{I}} \mathcal{L}_1$.

We can similarly define $\leq_{\text{strong}}^{\mathbf{I}}$ -hardness and $\leq_{\text{strong}}^{\mathbf{I}}$ -completeness.

It is easy to see that

Proposition 1 $\leq_{\text{weak}}^{\text{TxtEx}}, \leq_{\text{strong}}^{\text{TxtEx}}$ are reflexive and transitive.

The above proposition holds for most natural inference criteria. It is also easy to verify the following immediate proposition stating that strong reducibility implies weak reducibility.

Proposition 2 Let $\mathcal{L} \subseteq \mathcal{E}$ and $\mathcal{L}' \subseteq \mathcal{E}$ be given. Let \mathbf{I} be an identification criterion. Then $\mathcal{L} \leq_{\text{strong}}^{\mathbf{I}} \mathcal{L}' \Rightarrow \mathcal{L} \leq_{\text{weak}}^{\mathbf{I}} \mathcal{L}'$.

4 Results

Recall the three language classes, *SINGLE*, *COINIT*, and *FIN*, discussed in the introduction. Our first result uses the notion of weak reducibility to show that in the context of **TxtEx**-identification *SINGLE* presents a strictly weaker learning problem than *COINIT* which in turn is a strictly weaker learning problem than *FIN*. This is in keeping with our earlier intuitive discussion of these classes.

Theorem 1

- (a) $SINGLE \leq_{\text{strong}}^{\text{TxtEx}} COINIT \wedge COINIT \not\leq_{\text{weak}}^{\text{TxtEx}} SINGLE$.
- (b) $COINIT \leq_{\text{weak}}^{\text{TxtEx}} FIN \wedge FIN \not\leq_{\text{weak}}^{\text{TxtEx}} COINIT$.

PROOF. (a) We first construct a Θ such that $\Theta(\{n\}) = \{x \mid x \geq n\}$. Let $\tau_{m,n}$ denote a sequence such that $\text{content}(\tau_{m,n}) = \{x \mid m \leq x \leq n\}$. Note that $\text{content}(\tau_{n+1,n}) = \emptyset$. Consider operator Θ such that if $\text{content}(\sigma) = \emptyset$, then $\Theta(\sigma) = \sigma$, else $\Theta(\sigma) = \Theta(\text{prev}(\sigma)) \diamond \tau_{\min(\text{content}(\sigma)), |\sigma|}$. For $i \in N$, let $f(i)$ denote the index of a grammar (derived effectively from i) for the singleton language $\{i\}$. Let Ψ be defined as follows. Suppose G is a sequence of grammars, g_0, g_1, \dots . Then $\Psi(G)$ denotes the sequence of grammars g'_0, g'_1, \dots , where, for $n \in N$, $g'_n = f(\min(\{n\} \cup W_{g_n, n}))$. It is easy to verify that Θ and Ψ witness $SINGLE \leq_{\text{strong}}^{\text{TxtEx}} COINIT$.

Now suppose by way of contradiction that $COINIT \leq_{\text{weak}}^{\text{TxtEx}} SINGLE$ as witnessed by Θ and Ψ . By Smullyan's double recursion theorem [14], there exist $e_1 < e_2$ such that $W_{e_1} = \{x \mid x \geq e_1\}$ and $W_{e_2} = \{x \mid x \geq e_2\}$. Let σ be such that $\text{content}(\sigma) \subseteq W_{e_2}$ and $\text{content}(\Theta(\sigma)) \neq \emptyset$ (if no such σ exists then clearly Θ does not map any text for W_{e_2} to a text for a language in *SINGLE*). Let T_1 be a text for W_{e_1} and T_2 be a text for W_{e_2} such that $\sigma \subset T_1$ and $\sigma \subset T_2$. Now either $\text{content}(\Theta(T_1)) = \text{content}(\Theta(T_2))$ or

$\text{content}(\Theta(T_1)) \notin \text{SINGLE}$ or $\text{content}(\Theta(T_2)) \notin \text{SINGLE}$. It immediately follows that Θ and Ψ do not witness $\text{COINIT} \leq_{\text{weak}}^{\text{TxtEx}} \text{SINGLE}$.

(b) $\text{COINIT} \leq_{\text{weak}}^{\text{TxtEx}} \text{FIN}$ follows from Corollary 2 below. $\text{FIN} \not\leq_{\text{weak}}^{\text{TxtEx}} \text{COINIT}$ follows from Theorem 2 below. (The reader should contrast this result with Theorem 11 later which implies that $\text{COINIT} \not\leq_{\text{strong}}^{\text{TxtEx}} \text{FIN}$.) ■

We next present a theorem that turns out to be very useful in showing that certain classes are not complete with respect to $\leq_{\text{weak}}^{\text{TxtEx}}$ reduction. The theorem states that if a collection of languages \mathcal{L} is such that each natural number x appears in only finitely many languages in \mathcal{L} , then FIN is not $\leq_{\text{weak}}^{\text{TxtEx}}$ reducible to \mathcal{L} . Since $\text{FIN} \in \text{TxtEx}$, this theorem immediately implies that COINIT is not $\leq_{\text{weak}}^{\text{TxtEx}}$ -complete.

Theorem 2 *Suppose \mathcal{L} is such that $(\forall x)[\text{card}(\{L \in \mathcal{L} \mid x \in L\}) < \infty]$. Then $\text{FIN} \not\leq_{\text{weak}}^{\text{TxtEx}} \mathcal{L}$.*

PROOF. Suppose by way of contradiction that Θ and Ψ witness that $\text{FIN} \leq_{\text{weak}}^{\text{TxtEx}} \mathcal{L}$. Let σ be such that $\text{content}(\Theta(\sigma)) \neq \emptyset$ (there exists such a σ , since otherwise clearly, Θ and Ψ do not witness the reduction from FIN to \mathcal{L}). Let $w = \min(\text{content}(\Theta(\sigma)))$. Let T_i be a text for $\text{content}(\sigma) \cup \{i\}$ such that $\sigma \subset T_i$. Thus for all i , $w \in \text{content}(\Theta(T_i))$. But since $\{\text{content}(T_i) \mid i \in \mathbb{N}\}$ contains infinitely many languages and $\{L \in \mathcal{L} \mid w \in L\}$ is finite, there exist i, j such that $\text{content}(T_i) \neq \text{content}(T_j)$ but $\text{content}(\Theta(T_i)) = \text{content}(\Theta(T_j))$. But then Θ and Ψ do not witness that $\text{FIN} \leq_{\text{weak}}^{\text{TxtEx}} \mathcal{L}$. ■

Our next example is a collection of languages first introduced by Wiehagen [15]. We define, $\text{WIEHAGEN} = \{L \in \mathcal{E} \mid L = W_{\min(L)}\}$.

WIEHAGEN is an interesting class because it can be shown that it contains a finite variant of every recursively enumerable language. It is easy to verify that $\text{WIEHAGEN} \in \text{TxtEx}$. It is also easy to see that there exists a machine which TxtEx -identifies WIEHAGEN and that this machine, while processing a text for any language in WIEHAGEN , can provide an upper bound on the number of additional mind changes required before convergence. In this connection this class appears to pose a learning problem similar in nature to COINIT above. This intuition is indeed justified by the following two theorems as these two classes turn out to be equivalent in the strong sense.

Theorem 3 $\text{WIEHAGEN} \leq_{\text{strong}}^{\text{TxtEx}} \text{COINIT}$.

PROOF. Suppose Θ is such that $\Theta(L) = \{x \geq y \mid y \in L\}$. Note that such a Θ can be easily constructed. Let Ψ be defined as follows. Suppose G is a sequence of grammars, g_0, g_1, \dots . Then $\Psi(G)$ denotes the sequence of grammars g'_0, g'_1, \dots , where, for $n \in \mathbb{N}$, $g'_n = \min(\{n\} \cup W_{g_n, n})$. It is easy to see that Θ and Ψ witness $\text{WIEHAGEN} \leq_{\text{strong}}^{\text{TxtEx}} \text{COINIT}$. ■

Theorem 4 $\text{COINIT} \leq_{\text{strong}}^{\text{TxtEx}} \text{WIEHAGEN}$.

PROOF. By operator recursion theorem [4] there exists a recursive 1-1 increasing function p such that for all i , $W_{p(i)} = \{x \mid x \geq p(i)\}$. Let Θ be such that $\Theta(L) = \{x \geq p(i) \mid i \in L\}$. Note that such a Θ can be easily constructed. Let Ψ be defined as follows. Let $f(i)$ denote

a grammar (effectively obtained from i) such that $W_{f(i)} = \{x \mid x \geq p^{-1}(i)\}$. Suppose G is a sequence of grammars, g_0, g_1, \dots . Then $\Psi(G)$ denotes the sequence of grammars g'_0, g'_1, \dots , where, for $n \in N$, $g'_n = f(\min(\{n\} \cup W_{g_n, n}))$. It is easy to see that Θ and Ψ witness $COINIT \leq_{\text{strong}}^{\mathbf{TxtEx}} WIEHAGEN$. \blacksquare

Corollary 1 $COINIT \equiv_{\text{strong}}^{\mathbf{TxtEx}} WIEHAGEN$.

We next consider the class, $PATTERN$, of pattern languages introduced by Angluin [1].

Suppose V is a set of variables and C is a nonempty *finite* set of constants. Any $w \in (V \cup C)^+$ is called a pattern. Suppose f is a mapping from $(V \cup C)^+$ to C^+ , such that, for $a \in C$, $f(a) = a$ and, for $w_1, w_2 \in (V \cup C)^+$, $f(w_1 \cdot w_2) = f(w_1) \cdot f(w_2)$, where \cdot denotes concatenation of strings. Let PatMap denote the collection of all such mappings f .

Let code denote a 1-1 onto mapping from strings in C^* to N .

The language associated with the pattern w is defined as $L(w) = \{\text{code}(f(w)) \mid f \in \text{PatMap}\}$. Then, $PATTERN = \{L(w) \mid w \text{ is a pattern}\}$.

Angluin [2] showed that $PATTERN \in \mathbf{TxtEx}$. However, we show that $PATTERN$ is not $\leq_{\text{weak}}^{\mathbf{TxtEx}}$ -complete.

Theorem 5 $FIN \not\leq_{\text{weak}}^{\mathbf{TxtEx}} PATTERN$.

The above theorem follows directly from Theorem 2, since for any string x , there are only finitely many patterns w such that $x \in L(w)$.

Actually, we are also able to establish the following surprising result.

Theorem 6 $COINIT \equiv_{\text{strong}}^{\mathbf{TxtEx}} PATTERN$.

PROOF. We first show that $COINIT \leq_{\text{strong}}^{\mathbf{TxtEx}} PATTERN$. Let $L_i = L(a^i x)$, where $a \in C$ and $x \in V$. Let Θ be such that $\Theta(L) = \{\text{code}(a^l w) \mid w \in C^+ \wedge l \in L\}$. Note that such a Θ can be easily constructed. Note that $\text{code}(a^{l+1}) \in \text{content}(\Theta(L)) \Leftrightarrow l \geq \min(L)$.

Let $f(i)$ denote an index of a grammar (obtained effectively from i) for $\{x \mid x \geq i\}$. Let Ψ be defined as follows. Suppose $G = g_0, g_1, \dots$. Then $\Psi(G) = g'_0, g'_1, \dots$, such that, for $n \in N$, $g'_n = f(\min(\{l \mid \text{code}(a^{l+1}) \in W_{g_n, n}\}))$. It is easy to see that Θ and Ψ witness that $COINIT \leq_{\text{strong}}^{\mathbf{TxtEx}} PATTERN$.

We now show that $PATTERN \leq_{\text{strong}}^{\mathbf{TxtEx}} COINIT$. Note that there exists a recursive indexing L_0, L_1, \dots of pattern languages such that

- (1) $L_i = L_j \Leftrightarrow i = j$.
- (2) $L_i \subset L_j \Rightarrow i > j$.

(One such indexing can be obtained as follows. First note that for patterns w_1 and w_2 , if $L(w_1) \subseteq L(w_2)$ then length of w_1 is at least as large as that of w_2 . Also for patterns of the same length \subseteq relation is decidable [1]. Thus we can form the indexing as required using the following method. We consider only canonical patterns [1]. We place w_1 before w_2 if (a) length of w_1 is smaller than that of w_2 or (b) length of w_1 and w_2 are same, but $L(w_1) \supseteq L(w_2)$ or (c) length of w_1 and w_2 are same, $L(w_1) \not\subseteq L(w_2)$ and w_1 is lexicographically smaller than w_2 .)

Moreover, there exists a machine, \mathbf{M} , such that

(a) For all $\sigma \subseteq \tau$, such that $\text{content}(\sigma) \neq \emptyset$, $\mathbf{M}(\sigma) \geq \mathbf{M}(\tau)$.

(b) For all texts T for pattern languages, $\mathbf{M}(T) \downarrow = i$, such that $L_i = \text{content}(T)$.

(Angluin's method of identification of pattern languages essentially achieves this property).

Let $\tau_{m,n}$ be a sequence of length n , such that $\text{content}(\tau_{m,n}) = \{x \mid m \leq x \leq n\}$. If $\text{content}(\sigma) = \emptyset$, then $\Theta(\sigma) = \sigma$, else $\Theta(\sigma) = \Theta(\text{prev}(\sigma)) \diamond \tau_{\mathbf{M}(\sigma), |\sigma|}$.

Let $f(i)$ denote a grammar effectively obtained from i for L_i . Let Ψ be defined as follows. Suppose $G = g_0, g_1, \dots$. Then $\Psi(G) = g'_0, g'_1, \dots$, such that, for $n \in N$, $g'_n = f(\min(W_{g_n, n}))$. It is easy to see that Θ and Ψ witness that $\text{PATTERN} \leq_{\text{strong}}^{\text{TxtEx}} \text{COINIT}$. \blacksquare

Let $\text{INIT} = \{L \mid (\exists n)[L = \{x \mid x < n\}]\}$.

Theorem 7 $\text{INIT} \equiv_{\text{strong}}^{\text{TxtEx}} \text{FIN}$.

PROOF. Since $\text{INIT} \subseteq \text{FIN}$, we trivially have $\text{INIT} \leq_{\text{strong}}^{\text{TxtEx}} \text{FIN}$. We show that $\text{FIN} \leq_{\text{strong}}^{\text{TxtEx}} \text{INIT}$.

Note that our indexing D_0, D_1, \dots of finite sets satisfies the property that if $D_i \subseteq D_j$, then $i \leq j$. Let Θ be such that, $\Theta(D_i) = \{x \mid x \leq i\}$. Note that it is easy to construct such a Θ (since $D_i \subset D_j \Rightarrow i < j$). Let f be a function such that $W_{f(i)} = D_i$. Let Ψ be defined as follows. Suppose G is the sequence g_0, g_1, \dots . Then $\Psi(G)$ is the sequence g'_0, g'_1, \dots , where, for $n \in N$, $g'_n = f(\max(W_{g_n, n}))$. It is easy to see that Θ and Ψ witness that $\text{FIN} \leq_{\text{strong}}^{\text{TxtEx}} \text{INIT}$. \blacksquare

A few additional example classes are:

$\text{COSINGLE} = \{L \mid \text{card}(N - L) = 1\}$.

$\text{COFIN} = \{L \mid L \text{ is cofinite}\}$.

For $n \in N$, $\text{CONTON}_n = \{L \mid \text{card}(N - L) = n\}$.

Theorem 8 For all $n \in N^+$, $\text{COSINGLE} \equiv_{\text{strong}}^{\text{TxtEx}} \text{CONTON}_n$.

PROOF. Fix $n \in N^+$. First we show that $\text{COSINGLE} \leq_{\text{strong}}^{\text{TxtEx}} \text{CONTON}_n$. For $L \in \text{COSINGLE}$ let $L' = \{y \mid \lfloor \frac{y}{n} \rfloor \in L\}$. Let f be such that, for all i , $W_{f(i)} = \{x \mid (\exists y \in W_i)[\lfloor \frac{y}{n} \rfloor = x]\}$. Now consider Θ such that $\Theta(L) = L'$. Note that such a Θ can easily be constructed. Ψ is defined as follows. Suppose G is the sequence g_0, g_1, \dots . Then $\Psi(G)$ is the sequence $f(g_0), f(g_1), \dots$. It is easy to see that Θ and Ψ witness that $\text{COSINGLE} \leq_{\text{strong}}^{\text{TxtEx}} \text{CONTON}_n$.

Now we show that $\text{CONTON}_n \leq_{\text{strong}}^{\text{TxtEx}} \text{COSINGLE}$. For $L \in \text{CONTON}_n$, let $L' = \{\langle x_1, x_2, x_3, \dots, x_n \rangle \mid (\exists j \mid 1 \leq j \leq n)[x_j \in L] \vee (\exists i, j \mid 1 \leq i < j \leq n)[x_i = x_j]\}$. Let f be such that, for all $\langle x_1, x_2, \dots, x_n \rangle$, $W_{f(\langle x_1, x_2, \dots, x_n \rangle)} = \{x \mid (\forall j \mid 1 \leq j \leq n)[x \neq x_j]\}$. Let Θ be such that $\Theta(L) = L'$. Note that such a Θ can easily be constructed. Ψ is defined as follows. Suppose G is the sequence g_0, g_1, \dots . Then $\Psi(G)$ is the sequence g'_0, g'_1, \dots , where, for $i \in N$, $g'_i = f(\min(N - W_{g_i, i}))$. It is easy to see that Θ and Ψ witness that $\text{CONTON}_n \leq_{\text{strong}}^{\text{TxtEx}} \text{COSINGLE}$. \blacksquare

Since $\text{CONTON}_n \subseteq \text{COFIN}$, we trivially have $\text{CONTON}_n \leq_{\text{strong}}^{\text{TxtEx}} \text{COFIN}$ (note however that $\text{COFIN} \notin \text{TxtEx}$ [9]).

Theorem 9

- (a) *COSINGLE* is $\leq_{\text{weak}}^{\mathbf{TxtEx}}$ -complete.
- (b) *COFIN* is $\leq_{\text{weak}}^{\mathbf{TxtEx}}$ -hard.
- (c) For all $n \in \mathbb{N}^+$, *CONTON* _{n} is $\leq_{\text{weak}}^{\mathbf{TxtEx}}$ -complete.

PROOF. We prove part (a). Other parts follow as corollaries. Suppose $\mathcal{L} \subseteq \mathbf{TxtEx}(\mathbf{M})$. We construct Θ and Ψ which witness that $\mathcal{L} \leq_{\text{weak}}^{\mathbf{TxtEx}} \text{COSINGLE}$. We define Θ inductively. It is helpful to simultaneously define a function F . $F(T[0]) = \langle \mathbf{M}(T[0]), 0 \rangle$. $\Theta(T[0]) = \Lambda$. Define $F(T[n+1])$ and $\Theta(T[n+1])$ as follows.

$$F(T[n+1]) = \begin{cases} F(T[n]), & \text{if } \mathbf{M}(T[n+1]) = \mathbf{M}(T[n]); \\ \langle \mathbf{M}(T[n]), j \rangle, & \text{otherwise; where } j \text{ is such that} \\ & \langle \mathbf{M}(T[n]), j \rangle > \max(\text{content}(\Theta(T[n]))) \end{cases}$$

$\Theta(T[n+1])$ is a proper extension of $\Theta(T[n])$ such that $\text{content}(T[n+1]) = \{x \mid x \leq n \wedge x \neq F(T[n+1])\}$.

We now define Ψ . Intuitively, Ψ is such that if G converges to a final grammar for a language in *COSINGLE*, then $\Psi(G)$ converges to the first component of the only element not in the language enumerated by the grammar to which G converges. We now formally define Ψ . Suppose G is a sequence of grammar g_0, g_1, \dots . Then $\Psi(G)$ is the sequence of grammars g'_0, g'_1, \dots , where, for $i \in \mathbb{N}$, $g'_i = \pi_1(\min(N - W_{g_i, i}))$.

It is easy to verify that, for $\text{content}(T) \in \mathbf{TxtEx}(\mathbf{M})$, if G is a \mathbf{TxtEx} -admissible sequence for $\Theta(T)$, then $\Psi(G)$ is a \mathbf{TxtEx} -admissible sequence for T .

Thus Θ and Ψ witness that $\mathcal{L} \leq_{\text{weak}}^{\mathbf{TxtEx}} \text{COSINGLE}$. ■

Theorem 10 *COSINGLE* $\leq_{\text{strong}}^{\mathbf{TxtEx}}$ *INIT*.

PROOF. For L , let $L' = \{x \mid (\forall y \leq x)[y \in L]\}$. Let Θ be such that $\Theta(L) = L'$. Note that such a Θ can be easily constructed. Let $f(i)$ denote a grammar effectively obtained from i , for $\{x \mid x \neq i\}$. Suppose G is the sequence g_0, g_1, \dots . Then $\Psi(G)$ is the sequence g'_0, g'_1, \dots , where for $n \in \mathbb{N}$, $g'_n = f(\min(N - W_{g_n, n}))$. It is easy to verify that Θ and Ψ witness that *COSINGLE* $\leq_{\text{strong}}^{\mathbf{TxtEx}}$ *INIT*. ■

Corollary 2 *INIT* and *FIN* are $\leq_{\text{weak}}^{\mathbf{TxtEx}}$ -complete.

Lemma 1 is useful in proving that certain classes are not strongly reducible to other classes.

Proposition 3 If $\Theta(L)$ is defined then, for all σ , such that $\text{content}(\sigma) \subseteq L$, $\text{content}(\Theta(\sigma)) \subseteq \Theta(L)$.

PROOF. Follows from the definition of $\Theta(L)$. ■

Lemma 1 Suppose $L \subseteq L'$. Then if both $\Theta(L)$ and $\Theta(L')$ are defined then $\Theta(L) \subseteq \Theta(L')$.

PROOF. Follows from Proposition 3 ■

Theorem 11 $COINIT \not\leq_{\text{strong}}^{\mathbf{TxtEx}} FIN$.

PROOF. Suppose by way of contradiction that $COINIT \leq_{\text{strong}}^{\mathbf{TxtEx}} FIN$, as witnessed by Θ and Ψ . Then by Lemma 1 it follows that $(\forall L \in COINIT)[\Theta(L) \downarrow \subseteq \Theta(N)]$. Since $COINIT$ is an infinite collection of languages, it follows that either $\Theta(N)$ is infinite or there exist L_1 and L_2 in $COINIT$ such that $\Theta(L_1) = \Theta(L_2)$. It follows that $COINIT \not\leq_{\text{strong}}^{\mathbf{TxtEx}} FIN$. ■

Corollary 3 FIN is not $\leq_{\text{strong}}^{\mathbf{TxtEx}}$ -complete.

Theorem 12 Suppose $L_1 \subset L_2$, then $\{L_1, L_2\} \not\leq_{\text{strong}}^{\mathbf{TxtEx}} COSINGLE$.

PROOF. Suppose by way of contradiction that $L_1 \subset L_2$ and Θ and Ψ witness that $\{L_1, L_2\} \leq_{\text{strong}}^{\mathbf{TxtEx}} COSINGLE$. Then by Lemma 1 we have that $\Theta(L_1) \subseteq \Theta(L_2)$. Since for all $L'_1, L'_2 \in COSINGLE$, $L'_1 \subseteq L'_2 \Rightarrow L'_1 = L'_2$, it must be the case that $\Theta(L_1) = \Theta(L_2)$. But then, Θ and Ψ do not witness that $\{L_1, L_2\} \leq_{\text{strong}}^{\mathbf{TxtEx}} COSINGLE$. ■

As an immediate corollary we have

Corollary 4 $COINIT \not\leq_{\text{strong}}^{\mathbf{TxtEx}} COSINGLE$.

Theorem 13 $SINGLE \leq_{\text{strong}}^{\mathbf{TxtEx}} COSINGLE$.

PROOF. For n , let $L_n = \{x \mid x \neq n\}$. Let Θ be such that $\Theta(\{n\}) = L_n$. It is easy to construct such a Θ . Let $f(n)$ denote a grammar effectively obtained from n , for $\{n\}$. Let Ψ be defined as follows. If G is the sequence g_0, g_1, \dots , then $\Psi(G)$ is the sequence g'_0, g'_1, \dots , where, for $n \in N$, $g'_n = f(\min(N - W_{g_n, n}))$. It is easy to verify that Θ and Ψ witness that $SINGLE \leq_{\text{strong}}^{\mathbf{TxtEx}} COSINGLE$. ■

Clearly, $COINIT \leq_{\text{strong}}^{\mathbf{TxtEx}} COFIN$. However,

Theorem 14 $INIT \not\leq_{\text{strong}}^{\mathbf{TxtEx}} COFIN$.

PROOF. Suppose by way of contradiction that Θ and Ψ witness that $INIT \leq_{\text{strong}}^{\mathbf{TxtEx}} COFIN$. Let $L_n = \{x \mid x \leq n\}$. Now by Lemma 1, we have that for all n , $\Theta(L_n) \subseteq \Theta(L_{n+1})$. Moreover since $\Theta(L_n) \neq \Theta(L_{n+1})$ (otherwise Θ and Ψ cannot witness that $INIT \leq_{\text{strong}}^{\mathbf{TxtEx}} COFIN$), we have that $\Theta(L_n) \subset \Theta(L_{n+1})$. But since $\Theta(L_0) \in COFIN$, this is not possible (only finitely many additions can be done to $\Theta(L_0)$ before it becomes N). A contradiction. ■

We finally present a collection of languages that is complete with respect to strong reduction.

Suppose $\mathbf{M}_0, \mathbf{M}_1, \dots$ is an enumeration of the learning machines such that, $(\forall \mathcal{L} \in \mathbf{TxtEx})(\exists i)[\mathcal{L} \subseteq \mathbf{TxtEx}(\mathbf{M}_i)]$ (there exists such an enumeration, see for example [12]). For $j \in N$ and $L \in \mathcal{E}$, let $S_L^j = \{\langle x, j \rangle \mid x \in L\}$. Then, let $\mathcal{L}_{\mathbf{TxtEx}} = \{S_L^j \mid L \in \mathcal{E} \wedge j \in N \wedge L \in \mathbf{TxtEx}(\mathbf{M}_j)\}$.

Theorem 15 $\mathcal{L}_{\mathbf{TxtEx}}$ is $\leq_{\text{strong}}^{\mathbf{TxtEx}}$ complete for \mathbf{TxtEx} .

PROOF. Let $\mathcal{L}_j = \{S_L^j \mid L \in \mathbf{TxtEx}(\mathbf{M}_j)\}$.

If $\mathcal{L} \subseteq \mathbf{TxtEx}(\mathbf{M}_j)$, then it is easy to see that $\mathcal{L} \leq_{\text{strong}}^{\mathbf{TxtEx}} \mathcal{L}_j$. Since for all j , $\mathcal{L}_j \subseteq \mathcal{L}_{\mathbf{TxtEx}}$, it follows that $\mathcal{L}_{\mathbf{TxtEx}}$ is $\leq_{\text{strong}}^{\mathbf{TxtEx}}$ -complete for \mathbf{TxtEx} . ■

5 Language Identification from Informants

The concepts of weak and strong reduction can be adopted to language identification from informants. Informally, informants, first introduced by Gold [9], are texts which contain both positive and negative data. Thus if I_L is an informant for L , then $\text{content}(I_L) = \{\langle x, 0 \rangle \mid x \notin L\} \cup \{\langle x, 1 \rangle \mid x \in L\}$. Identification in the limit from informants is referred to as **InfEx**-identification (we refer the reader to [9] for details). The definition of weak and strong reduction can be adopted to language identification from informants in a straightforward way by replacing texts by informants in the Definitions 5 and 7.

Since a *canonical* informant can always be produced from any informant, we have the following:

Proposition 4 $\mathcal{L}_1 \leq_{\text{weak}}^{\text{InfEx}} \mathcal{L}_2 \iff \mathcal{L}_1 \leq_{\text{strong}}^{\text{InfEx}} \mathcal{L}_2$.

Theorem 16 *FIN* is $\leq_{\text{strong}}^{\text{InfEx}}$ complete.

PROOF. For a language L , let I_L be the canonical informant for L . Fix a machine \mathbf{M} . Let $S_L^{\mathbf{M}} = \{\langle \mathbf{M}(I_L[n+1]), n \rangle \mid \mathbf{M}(I_L[n]) \neq \mathbf{M}(I_L[n+1])\}$. Let Θ be such that for all L , $\Theta(I_L) = I_{S_L^{\mathbf{M}}}$. Note that such a Θ can easily be constructed. Suppose F is such that, for a finite set S , $F(S) = \min(\{i \mid (\exists j)[\langle i, j \rangle \in S \wedge j = \max(\{k \mid (\exists x)[\langle x, k \rangle \in S]\})]\})$. Let Ψ be defined as follows. Suppose G is a sequence g_0, g_1, \dots . Then $\Psi(G)$ is the sequence g_0, g'_1, \dots , where for $n \in \mathbb{N}$, $g'_n = F(W_{g_n, n})$. It is easy to verify that Θ and Ψ witness that $\text{InfEx}(\mathbf{M}) \leq_{\text{strong}}^{\text{InfEx}} \text{FIN}$. \blacksquare

6 Conclusion

A novel approach to studying the intrinsic complexity of language identification was undertaken using weak and strong reductions between classes of languages. The intrinsic complexity of several classes were considered. It was shown that the self referential class of Wiehagen [15] in which the least element of every language is a grammar for the language and the class of pattern languages introduced by Angluin [1] are equivalent in the strong sense. A number of complete classes were presented for both the reductions. It was also shown that the weak and strong reductions are distinct.

The results presented were for the widely studied identification in the limit criterion. These techniques have also been applied to other criteria of success. It is felt that the reductions studied in the present paper lay a foundation on which feasibility issues in language identification can be studied.

Acknowledgements

Our study has clearly been influenced by the work on intrinsic complexity of computable functions by Freivalds, Kinber, and Smith [8]. We would like to thank Efim Kinber for helpful discussion and for encouraging us to undertake the present study.

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