# On a Question of Nearly Minimal Identification of Functions

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#### Abstract

Suppose  $\mathcal{A}$  and  $\mathcal{B}$  are classes of recursive functions.  $\mathcal{A}$  is said to be an *m*-cover (\*-cover) for  $\mathcal{B}$ , iff for each  $g \in \mathcal{B}$ , there exsits an  $f \in \mathcal{A}$  such that f differs from g on at most m inputs (finitely many inputs).  $\mathcal{C}$ , a class of recursive functions, is *a*-immune iff  $\mathcal{C}$  is infinite and every recursively enumerable subclass of  $\mathcal{C}$  has a finite *a*-cover.  $\mathcal{C}$  is *a*-isolated iff  $\mathcal{C}$  is finite or *a*-immune.

Chen [Che81] conjectured that every class of recursive functions that is  $\mathbf{MEx}_m^*$ -identifiable is \*-isolated. We refute this conjecture.

### 1 Introduction

Formal definitions of notions informally discussed below are given in Section 3. Gold's [Gol67] criterion of identification of functions may be described as follows: A learning machine  $\mathbf{M}$  is said to *identify* (or learn) a function f just in case  $\mathbf{M}$ , when presented with the graph of f, outputs a sequence of programs that converges (in the limit) to a program for f. The above criterion of identification is called **Ex**-*identification* (**Ex** stands for explains). Freivalds [Fre75] (see also [Che81, Che82]) introduced the notion of nearly minimal identification, by placing an additional restriction on size of the final programs. In this criterion, the learning machine is required to converge to a program whose size is within a recursive factor of the size of the smallest program for the input function.

The above notions of identification can be extended in the following two directions:

- Error Bound ([BB75, CS83]): The above model may be relaxed by allowing the learning machine to converge to a program which may make some errors in computing the input function. An error bound of a natural number m means that the final program makes at most m errors in computing the input function. An error bound of \* means that the final program makes at most finitely many errors in computing the input function.
- Mind-Change Bound ([CS83, BF74]): The above model may be restricted by placing a bound on the number of mind changes allowed by the learning machine. A mind change bound of a natural number *m* means that the learning machine may make at most *m* mind changes before converging to the final program. A mind change bound of \* means that the learning machine may make at most finitely many mind changes before converging to the final program (note that \*-mind change bound is equivalent to the Gold's notion of identification in the limit).

Chen [Che81] showed that the recursively enumerable (r.e.) classes of functions that can be identified in the nearly minimal sense with *m*-errors and with a mind change bound of *n* (where *m* and *n* are natural numbers) are not very complex — they can be "approximated" with at most *m*-errors using a finite class of functions. For a recursively enumerable classes, this latter notion of being approximated with at most *m*-errors, by a finite class of functions, is referred to as being *m*-isolated. Chen [Che81] also showed that classes of functions which can be nearly-minimallyidentified with \*-errors, but with only 0-mind changes, are \*-isolated. The question of \*-errors, but with mind change bound of n > 0, was left open by Chen. He conjectured that such classes would also be \*-isolated.

In this paper we refute Chen's conjecture. Thus complex r.e. classes can be identified in the nearly-minimal sense with \*-errors and a nonzero mind change bound.

We now proceed formally.

# 2 Notation

Recursion-theoretic concepts not explained below are treated in [Rog67]. N denotes the set of natural numbers,  $\{0, 1, 2, \ldots\}$ . All conventions regarding range of variables apply, with or without decorations<sup>1</sup>, unless otherwise specified. The symbols i, j, k, l, m, n, s, t, u, x, y, and z, range over natural numbers unless otherwise specified. card(S) denotes the cardinality of a set S. \* denotes a nonmember of N and is assumed to satisfy  $(\forall n \in N)[n < * < \infty]$ . Thus, card $(S) \leq *$  means that cardinality of the set S is finite. a and b range over  $N \cup \{*\}$ . max $(), \min()$  denote the maximum and minimum of a set, respectively. By convention  $\max(\emptyset) = 0$  and  $\min(\emptyset) = \infty$ .

 $\mathcal{R}$  denotes the set of all total recursive functions.  $\mathcal{A}$ ,  $\mathcal{B}$ ,  $\mathcal{C}$ , and  $\mathcal{S}$  range over subsets of  $\mathcal{R}$ . h, f, and g range over total recursive functions.  $\eta$  ranges over partial functions. domain( $\eta$ ) denotes the domain of  $\eta$ . For  $a \in N \cup \{*\}$ , we say that  $\eta_1 =^a \eta_2$  (read:  $\eta_1$  is an a-variant of  $\eta_2$ ) iff  $\operatorname{card}(\{x \mid \eta_1(x) \neq \eta_2(x)\}) \leq a$ . Thus,  $\eta_1 =^* \eta_2$  means that  $\eta_1$  and  $\eta_2$  are finite variants of each other.

We let  $\varphi$  denote a standard acceptable programming system.  $\varphi_i$  denotes the partial recursive function computed by the  $i^{th}$  program in the standard acceptable programming system  $\varphi$ . We often refer to the  $i^{th}$  program as program *i*. *p* ranges over total functions, with its range being interpreted as programs. For a recursive function *f*, MinProg(*f*) denotes the minimal program for *f* (in the  $\varphi$  system), i.e., MinProg(*f*) = min({*i* |  $\varphi_i = f$ }).

A class S of recursive functions is said to be recursively enumerable iff there exists a recursive set Z such that  $S = \{\varphi_i \mid i \in Z\}$ .

 $\langle i, j \rangle$  stands for an arbitrary computable one to one encoding of all pairs of natural numbers onto N [Rog67].

The quantifiers ' $\exists$ ', ' $\forall$ ', ' $\forall^{\infty}$ ', and ' $\exists^{\infty}$ ' respectively denote 'there exists', 'for all', 'for all but finitely many', and 'there exist infinitely many'.

## 3 Learning Paradigms

For any partial function  $\eta$  and any natural number n such that, for each x < n,  $\eta(x)\downarrow$ , we let  $\eta[n]$  denote the finite initial segment  $\{(x, \eta(x)) \mid x < n\}$ . Let  $SEQ = \{f[n] \mid f \in \mathcal{R} \land n \in N\}$ .

<sup>&</sup>lt;sup>1</sup>Decorations are subscripts, superscripts, primes and the like.

**Definition 1** [Gol67] A *learning machine* is an algorithmic device which computes a mapping from SEQ into  $N \cup \{?\}$  such that, if  $\mathbf{M}(f[n]) \neq ?$ , then  $\mathbf{M}(f[n+1]) \neq ?$ .

We let **M**, with or without decorations, range over learning machines. In Definition 1 above, '?' denotes the situation when **M** outputs "no conjecture" on some member of SEQ.

In Definition 2 below we spell out what it means for a learning machine to converge in the limit.

**Definition 2** Suppose **M** is a learning machine and f is a computable function.  $\mathbf{M}(f) \downarrow$  (read:  $\mathbf{M}(f)$  converges) just in case  $(\exists i)(\forall^{\infty}n)$   $[\mathbf{M}(f[n]) = i]$ . If  $\mathbf{M}(f) \downarrow$ , then  $\mathbf{M}(f)$  is defined = the unique i such that  $(\forall^{\infty}n)[\mathbf{M}(f[n]) = i]$ , otherwise we say that  $\mathbf{M}(f)$  diverges (written:  $\mathbf{M}(f)\uparrow$ ).

#### 3.1 Explanatory Function Identification

We now formally define the criteria of inference considered in this paper.

**Definition 3** [Gol67, CS83, BB75, BF74] Suppose  $a, b \in N \cup \{*\}$ .

- (1) A learning machine **M** is said to  $\mathbf{Ex}_b^a$ -identify  $f \in \mathcal{R}$  (written:  $f \in \mathbf{Ex}_b^a(\mathbf{M})$ ) just in case  $(\exists i \mid \varphi_i = a f) \ (\forall^{\infty} n) [\mathbf{M}(f[n]) = i]$  and  $\operatorname{card}(\{n \mid ? \neq \mathbf{M}(f[n]) \neq \mathbf{M}(f[n+1])\}) \leq b$ .
- (2)  $\mathbf{E}\mathbf{x}_b^a = \{ \mathcal{C} \mid (\exists \mathbf{M}) [\mathcal{C} \subseteq \mathbf{E}\mathbf{x}_b^a(\mathbf{M})] \}.$

For a given f and  $\mathbf{M}$ , we refer to each instance of the case,  $? \neq \mathbf{M}(f[n]) \neq \mathbf{M}(f[n+1])$  as a *mind change* by  $\mathbf{M}$  on f. Intuitively, in  $\mathbf{E}\mathbf{x}_b^a$ , the superscript a refers to the error bound on the final program, and subscript b refers to the mind change bound. We often refer to  $\mathbf{E}\mathbf{x}_*^a$  as  $\mathbf{E}\mathbf{x}^a$ ,  $\mathbf{E}\mathbf{x}_b^0$  as  $\mathbf{E}\mathbf{x}_b$  and  $\mathbf{E}\mathbf{x}_*^0$  as  $\mathbf{E}\mathbf{x}$ .

#### 3.2 Nearly Minimal Identification

We next consider nearly minimal identification criteria.

**Definition 4** [Fre75, Che82] Suppose  $a, b \in N \cup \{*\}$ .

- (1) Suppose h is a recursive function. A learning machine **M** is said to h-**ME** $\mathbf{x}_b^a$ -identify  $f \in \mathcal{R}$  (written  $f \in h$ -**ME** $\mathbf{x}_b^a$ (**M**)) iff **M E** $\mathbf{x}_b^a$ -identifies f and  $\mathbf{M}(f) \leq h(\operatorname{MinProg}(f))$ .
- (2)  $\mathbf{MEx}_{b}^{a} = \{ \mathcal{C} \mid (\exists \mathbf{M}) (\exists h \in \mathcal{R}) | \mathcal{C} \subseteq h \cdot \mathbf{MEx}_{b}^{a} (\mathbf{M}) ] \}.$

We often refer to  $\mathbf{MEx}_*^a$  as  $\mathbf{MEx}^a$ ,  $\mathbf{MEx}^0_b$  as  $\mathbf{MEx}_b$  and  $\mathbf{MEx}^0_*$  as  $\mathbf{MEx}$ .

**Theorem 5** [Che82, Fre75, Jai95] For all  $m, n \in N$ ,  $a \in N \cup \{*\}$ .

- (1)  $\mathbf{E}\mathbf{x} \mathbf{M}\mathbf{E}\mathbf{x}^m \neq \emptyset$ .
- (2)  $\mathbf{E}\mathbf{x}_0^0 \mathbf{M}\mathbf{E}\mathbf{x}_n^* \neq \emptyset$ .
- (3)  $\mathbf{Ex}_n^a \subseteq \mathbf{MEx}^a$ .
- (4)  $\mathbf{Ex}^* = \mathbf{MEx}^*$ .
- (5)  $\mathbf{MEx}_{n+1}^0 \mathbf{Ex}_n^* \neq \emptyset.$
- (6)  $\mathbf{MEx}_0^{m+1} \mathbf{Ex}^m \neq \emptyset.$

#### 3.3 Isolated Classes

**Definition 6** [Che81] Suppose  $\mathcal{A}$  and  $\mathcal{B}$  are classes of recursive functions.  $\mathcal{B}$  is an *a*-cover of  $\mathcal{A}$  iff for each  $g \in \mathcal{B}$ , there exists an  $f \in \mathcal{A}$ , such that  $f = {}^{a} g$ .

**Definition 7** [Che81, Rog67] C is *a-immune* iff (a) C is infinite and (b) every recursively enumerable subclass of C has a finite *a*-cover.

**Definition 8** [Che81] C is *a-isolated* iff C is finite or C is *a*-immune.

Chen [Che81] established the following two results.

**Theorem 9** [Che81] Suppose  $m, n \in N$ , and  $S \in \mathbf{MEx}_n^m$ . Then S is m-isolated.

**Theorem 10** [Che81] Suppose  $S \in MEx_0^*$ . Then S is \*-isolated.

Based on above results, Chen conjectured that, for  $n \in N$ , every  $S \in \mathbf{MEx}_n^*$  is \*-isolated. We surprisingly refute his conjecture.

### 4 Main Theorem

**Theorem 11** There exists an infinite recursively enumerable class  $S \in MEx_1^*$  such that S is not \*-isolated.

PROOF. Let

 $\begin{aligned} \mathcal{C}_1 &= \{ f \mid \varphi_{f(\langle 0, 0 \rangle)} =^* f \land f(\langle 0, 0 \rangle) \leq \operatorname{MinProg}(f) \land (\forall x) [f(\langle 1, x \rangle) = 0] \}, \\ \mathcal{C}_2 &= \{ f \mid (\forall^{\infty} x) [f(x) = 0] \land (\exists x) [f(\langle 1, x \rangle) \neq 0] \}, \\ \text{and } \mathcal{C} &= \mathcal{C}_1 \cup \mathcal{C}_2. \end{aligned}$ 

Intuitively,  $C_1$  is a class of (nearly) self-describing functions, where a small program for a finite variant of the function is coded into the function itself.  $C_2$  is a subclass of almost everywhere 0 functions. Additionally, we code into the functions (using  $\{\langle 1, x \rangle \mid x \in N\}$ ) whether it is from  $C_1$  or  $C_2$ .

It is easy to verify that C is in  $\mathbf{MEx}_1^*$ . We will construct the required S as an appropriate recursively enumerable subset of C. Intuitively, the idea is to use an appropriate subclass of  $C_1$  to ensure that S is \*-isolated.  $C_2$  is added to this subclass, to ensure that S is recursively enumerable. We now continue with the formal construction of S.

Using Operator Recursion Theorem [Cas74] we will define a recursive, 1–1, increasing function p such that the functions  $\varphi_{p(i)}$  satisfy the following four properties:

(A) For all x,  $\varphi_{p(i)}(\langle 0, x \rangle) = p(i);$ 

(B) For all x,  $\varphi_{p(i)}(\langle 1, x \rangle) = 0$ ;

(C)  $\varphi_{p(i)}$  is undefined on exactly one input; let this input be called  $u_i$ ;

(D) For all j < p(i), either  $\varphi_j$  is non-total, or there exists an  $x < u_i$  such that  $\varphi_j(x) \neq \varphi_{p(i)}(x)$ . Let  $f_i$  be defined as follows:

$$f_i(x) = \begin{cases} \varphi_{p(i)}(x), & \text{if } x \neq u_i; \\ 0, & \text{otherwise.} \end{cases}$$

Let  $\mathcal{S} = \{f_i \mid i \in N\} \cup \mathcal{C}_2.$ 

It is easy to verify that S is recursively enumerable. Moreover, by property (A) S is not  $\ast$ -isolated. It is also easy to verify (using properties (A)–(D)) that  $f_i \in C_1$ . Thus,  $S \subseteq C$  and  $S \in \mathbf{MEx}_1^*$ .

We now give the construction of  $\varphi_{p(i)}$  satisfying the properties (A) to (D) above. By operator recursion theorem [Cas74] there exists a 1–1, recursive, increasing function p such that  $\varphi_{p(i)}$  may be defined in stages as follows.

Let  $u_i^0 = \min(N - \{y, x\} \mid y \in \{0, 1\} \land x \in N)$ . Let  $Cancel_i^0 = \emptyset$ . Intuitively,  $u_i^s$  denotes the intended value of  $u_i$  as at the beginning of stage s.  $Cancel_i^s$  is used to keep track of programs < p(i), against which  $\varphi_{p(i)}$  has diagonalized against before stage s. Go to stage 0.

Stage s

- 1. Dovetail steps 2 and 3 until step 2 succeeds. If and when step 2 succeeds, go to step 4.
- 2. Search for a j < p(i), such that  $j \notin Cancel_i^s$ , and  $\varphi_j(u_i^s) \downarrow$ .
- 3. For z = 0 to  $\infty$  Do

If  $z \neq u_i^s$  and  $\varphi_{p(i)}(z)$  has not been defined upto now, Then Let  $\varphi_{p(i)}(z) = p(i)$ , if  $z = \langle 0, x \rangle$  for some  $x \in N$ ; Let  $\varphi_{p(i)}(z) = 0$ , if  $z = \langle y, x \rangle$  for some  $x \in N$  and  $y \neq 0$ ;

EndFor

4. If and when step 2 succeeds, then let j be as in step 2.

Let  $Cancel_i^{s+1} = Cancel_i^s \cup \{j\}.$ 

- Let  $\varphi_{p(i)}(u_i^s) = \varphi_j(u_i^s) + 1.$
- Let  $u_i^{s+1}$  be the minimum number z such that  $\varphi_{p(i)}(z)$  has not been defined up to now, and  $z \notin \{\langle y, x \rangle \mid y \in \{0, 1\} \land x \in N\}.$

Go to stage s + 1.

End Stage s

We now argue that  $\varphi_{p(i)}$  defined above satisfies properties (A) to (D) above. First note that there are only finitely many stages. This is so since each time a new stage > 0 is entered, step 4 in the previous stage must have diagonalized against a new program j < p(i). Since there are at most finitely many programs less than p(i), there are at most finitely many stages that are executed. Let s be the last stage that is entered but never finished. Let  $u_i = u_i^s$  and  $Cancel_i = Cancel_i^s$ . It is now easy to verify that (A), (B) and (C) are satisfied. Also, for all j < p(i), either  $j \in Cancel_i$ , or  $\varphi_j(u_i)\uparrow$ . In case  $j \in Cancel_i$ , then by step 4 of the construction, there exists a  $z < u_i$  such that  $\varphi_j(z)\downarrow \neq \varphi_{p(i)}(z)\downarrow$ . Thus, (D) is satisfied. This completes the proof of the theorem.

**Corollary 12** For all n > 0, there exists a recursively enumerable class  $S \in \mathbf{MEx}_n^*$  such that S is not \*-isolated.

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