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MIT/LCS/TM-123 AN IMPROVED PROOF OF THE RABIN-HARTMANIS-STEARNS CONJECTURE Harold M. Perry January 1979

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January 1979

This report was prepared with the support of a National Science Foundation Grant No. MCS77-19754 A02

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MASSACHUSETTS 02139

Keywords: Rabin-Hartmanis-Stearns conjecture real time language recognition, Turing machine, overlap argument

AN IMPROVED PROOF OF THE RABIN-HARTMANIS-STEARNS CONJECTURE

by

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Submitted to the Department of Electrical Engineering and Computer Science on May 12, 1978 in partial fulfillment of the requirements for the Degrees of Master of Science and Electrical Engineer.

ABSTRACT

We offer an improved presentation of Aanderaa's constructive proof of the Rabin-Hartmanis-Stearns conjecture:

For all $k \ge 2$, there exists a language L_k such that L_k can be recognized by a k-worktape real time Turing machine but cannot be recognized by any (k-1)-worktape real time Turing machine.

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INTRODUCTION

All Turing machine models recognize precisely the same class of languages. Intuition tells us however, that a well-equipped model, (e.g., a Turing machine with worktapes which have multiple heads) should have capabilities lacked by a less-equipped model, (e.g., a Turing machine with worktapes which have only one worktape head).

One way to measure such capabilities is by comparing the time it takes each machine model to recognize the same language. This thesis studies the effect of having extra worktapes in the computational process. It addresses such questions as:

 i) Does having more worktapes make a machine model more efficient?
ii) If so, how many extra worktapes are needed to achieve a given gain in speed?

We shall answer the first question affirmatively. Moreover, we will prove that only one extra worktape need be added to make the Turing machine model faster and hence conclude:

For all $k \ge 2$, the class of k-worktape Turing machines are more efficient than the class of (k-1)-worktape Turing machines.

Our proof will consist in specifying a language L_k and a k-worktape Turing machine \mathfrak{M} such that \mathfrak{M} can recognize L_k faster than any Turing machine with (k-1)-worktapes.

HISTORY OF THE PROBLEM

In 1960, Rabin [10] showed that the 2-worktape Turing machine model was more efficient than the 1-worktape model. His proof consisted in specifying a language L_2 and showing that it may be recognized faster by a 2-worktape Turing machine \mathfrak{M} than by any 1-worktape Turing machine. L_2 was specified as follows:

 $L_2^{def}(UV#U^R)U(UV#V^R)$ where $U \in (0, 1)^*$, $V \in \{a, b\}^*$ and R denotes reversal.

Rabin showed that \mathfrak{M} may recognize whether an input was an element of L_2 in time equal to the amount of time it took to read the input, i.e., in "real time". (Given an input, \mathfrak{M} need only store U on one worktape and V on the second worktape. After scanning #, \mathfrak{M} would then be able to simultaneously check whether U^R or V^R followed.) It was then shown that no 1-worktape Turing machine could recognize L_2 as fast. Intuitively, no 1-worktape Turing machine could operate as fast as \mathfrak{M} and still be able to achieve a configuration after scanning # that would allow it to check simultaneously whether U^R or V^R followed.

In 1965, Hartmanis and Stearns [6] conjectured a generalization of Rabin's result. This conjecture was finally proven by Aanderaa [1] fourteen years after the initial result.

Although Aanderaa's proof was impressive in its result, we found it difficult to read. The ideas and insight which led to the proof did not seem readily apparent. We offer what we believe are simplified proofs of several of the technical results as well as an improved exposition.

PRELIMINARY DEFINITIONS

We assume the reader is familiar with the multitape Turing machine model. (See [2,7].) We may characterize a multitape Turing machine computation as follows:

Initially, the input word appears with endmarkers on the input tape. The input head is scanning the left endmarker while all the worktapes are blank. The input word is accepted if and only if the multitape Turing machine started in a designated initial state, makes a sequence of moves at the end of which it enters an accepting state. The language accepted is the set of input words so accepted.

We shall find it convenient to number the squares of the worktapes and the input tape. A tape may be numbered by assigning consecutive integers to the squares such that zero denotes the square initially scanned by the tape head, while positive and negative integers correspond to those squares to the right and left, respectively, of the square denoted by zero.

For the following definitions, we fix a multitape Turing machine \mathfrak{M} , with input alphabet Γ , worktape alphabet Σ , set Q of internal states and k worktapes indexed 1 through k. It will be convenient to refer to the input tape as the 0th worktape. An integer j such that $0 \le j \le k$ will be called a tape index of \mathfrak{M} .

Let $W \in \Gamma^*$ denote an input word. The length of W shall be denoted by L(W). We may therefore denote W as $w_1 w_2 \dots w_{L(W)}$ where $w_i \in \Gamma$ for

 $1 \leq i \leq L \ (\texttt{W})$. The time of the computation of \mathfrak{M} on $\mathbb W$ shall be denoted by $T_{\mathfrak{M}} \ (\texttt{W})$.

For the rest of this section we let $s \in \mathbb{N}$ such that $0 \le s \le T_{\mathfrak{M}}(W)$. (We shall use s to refer to a step, i.e., an instantaneous description (i.d.) in the computation of \mathfrak{M} on W.)

DEFINITION 1

Let j be a tape index of \mathfrak{M} . POSN(j,W,s) is defined to be the number of the worktape square which is scanned by the jth tape head immediately after step s (i.e., the square scanned in the sth i.d. where the initial i.d. is counted as the zeroth) in the computation of \mathfrak{M} on W.

DEFINITION 2

Let j be a tape index of \mathfrak{M} . Let $z \in \mathbb{Z}$. CONT(j, z, W, s) is defined to be the contents of the z^{th} square of the jth worktape immediately after step s in the computation of \mathfrak{M} on W.

DEFINITION 3

STATE (W, s) is defined to be the internal state of \mathfrak{M} immediately after step s in the computation of \mathfrak{M} on W.

We shall let ID(W, s) denote the i.d. immediately after the sth step in the computation of \mathfrak{M} on W. We can easily specify ID(W, s) in terms of the previous definitions.

DEFINITION 4

The instantaneous description ID(W, s) of \mathfrak{M} on input W immediately after step s is specified as follows:

ID (W, s) $d\underline{e}^{f}(q, p_0, \ldots, p_k, \sigma, \delta_1, \ldots, \delta_k)$ where:

q=STATE(W,s), $p_j = POSN(j, W, s)$ for $0 \le j \le k$, $\sigma = CONT(0, p_0, W, s)$, and

for $1 \le j \le k$, $\delta_j : \mathbb{Z} \to \Sigma$ such that $\delta_j(z) = \text{CONT}(j, z, W, s)$ for $z \in \mathbb{Z}$. We remark that for almost all $z \in \mathbb{Z}$, $\delta_j(z)$ is equal to the blank symbol.

The reader may consult [2,7] for an explanation of how a Turing machine computation may be formalized as a sequence of instantaneous descriptions and how the time of such a computation may be defined as the index of the last instantaneous description in this sequence.

DEFINITION 5

Consider the computations of \mathfrak{M} on inputs W and \hat{W} . Let $s, \hat{s} \in \mathbb{N}$ be such that $0 \le s \le T_{\mathfrak{M}}(W)$ and $0 \le \hat{s} \le T_{\mathfrak{M}}(\hat{W})$. Let $n \in \mathbb{N}$ and let j > 0 be a tape index of \mathfrak{M} . We say that ID(W, s) and ID(\hat{W}, \hat{s}) are n-equivalent on worktape j if the following conditions hold:

- i) STATE $(W, s) = STATE(\widehat{W}, \widehat{s})$,
- ii) $POSN(j, W, s) = POSN(j, \hat{W}, \hat{s})$, and
- iii) $(\forall m (-n \le m \le n))$

 $[CONT(j, POSN(j, W, s) + m, W, s) = CONT(j, POSN(j, \hat{W}, \hat{s}) + m, \hat{W}, \hat{s})].$

That is, immediately after step s in the computation of \mathfrak{M} on W and immediately after step \hat{s} in the computation of \mathfrak{M} on \hat{W} we have that:

i) the internal states are equal,

ii) the jth worktape head is scanning the same tape square, and

iii) the information accessible to the j^{th} worktape head in the next n steps is equivalent.

We say that ID(W, s) and $ID(\widehat{W}, \widehat{s})$ are n-equivalent if ID(W, s) and $ID(\widehat{W}, \widehat{s})$ are n-equivalent on all worktapes.

We now offer a formalism by which specific parts of an input word may be identified.

DEFINITION 6

Let $1, m \in \mathbb{N}$ such that $0 \le 1 \le m$. An interval I is defined to be the set of natural numbers $(n|1 \le n \le m)$, which we also denote as [1,m].

For convenience, we shall define MIN(I) to equal 1 and MAX(I) to equal m. The length of interval I, |I|, is defined to equal MAX(I)-MIN(I)+1. For completeness, we shall specify an empty interval equal to ϕ such that $|\phi|=0$ and MAX(ϕ) and MIN(ϕ) are undefined.

DEFINITION 7

Let I and J be intervals.

i) If MIN(J) = MAX(I) + 1 then I and J are said to be **adjacent**. In this case, IUJ is equal to the interval, [MIN(I), MAX(J)].

ii) J is said to be a subinterval of I if JcI.

DEFINITION 8

i) The interval I is said to be an input interval of $W=W_1...W_{L(W)}$ if $1 \le MIN(I)$ and $MAX(I) \le L(W)$.

ii) Let I be an input interval of W. W at I (written W@I) is defined to be the word $W_{MIN(I)}$ $W_{MAX(I)}$. We note that $W@\phi = \lambda$ (where λ denotes the empty string.)

We conclude this section by presenting an information-theoretic concept by which we may characterize the retrieval of stored information. Consider the computation of \mathfrak{M} on input W. Stored information will be retrieved precisely when some worktape head revisits a tape square. We shall refer to such a happening as an **overlap event.** Formally, we define an overlap event as follows:

DEFINITION 9

Let s,t $\in\mathbb{N}$ such that $0\leq s < t\leq T_{\mathfrak{M}}(W)$. Let j be a tape index of \mathfrak{M} . Let $\sigma \in \Sigma \cup \Gamma$ and let $z \in \mathbb{Z}$. The 5-tuple (s, t, j, z, σ) is said to be an **overlap** event in the computation of \mathfrak{M} on W if the following conditions hold:

- i) POSN(j, W, s) = z and $POSN(j, W, s+1) \neq z$,
- ii) $CONT(j, z, W, s+1) = \sigma$, and
- iii) (min $x | s < x \le T_{\mathfrak{M}}(W)$) [POSN(j, W, x) = z] = t.

Informally, the overlap event (s, t, j, z, σ) occurs if: immediately after step s, the z^{th} square of the j^{th} worktape is being scanned. At the next step, the tape head moves off this square leaving symbol σ written. Step t is the first subsequent step where the z^{th} square on the j^{th} worktape is revisited.

We call σ the overlap value and (j,z) the overlap location associated with the overlap event (s,t,j,z, σ). Finally, t is called the step at which the overlap event occurred.

We shall denote the set of overlap events in the computation of \mathfrak{M} on W by OVERLAP(W).

REAL TIME LANGUAGE RECOGNITION

DEFINITION 10

 \mathfrak{M} is said to be a real time Turing Machine if it scans a new input symbol during each step of its operation. (\mathfrak{M} must halt once the right endmarker is scanned on its input tape.) A language is said to be recognized in real time if it is recognizable by a real time Turing machine.

For the rest of this section we shall assume that \mathfrak{M} is a fixed k-worktape real time Turing machine with input alphabet Γ . Let W be an input word to \mathfrak{M} such that L(W) > 1.

We make the following trivial observations:

i) The head on the input tape of \mathfrak{M} must be one-way. Thus, no overlap events can occur on the input tape.

ii) $T_{99}(W) = L(W) + 1$.

iii) $(\forall s \mid 0 \le s \le T_{\mathfrak{M}}(W))$ [POSN(0, W, s) = s]. Thus if I is an input interval of W, the set of indices of steps where \mathfrak{M} is scanning some part of W@I equals I.

We remind the reader that the definitions presented in the previous section apply to all mutitape Turing machine models. The following definitions become relevant under the real time constraint.

DEFINITION 11

Let I be an input interval of W. Let j be a tape index of M.

DSPM(j, W, I) is defined to be the displacement incurred by the jth tape head while \mathfrak{M} was processing W@I. Formally, DSPM(j, W, I) is defined to equal |POSN(j, W, MIN(I)) - POSN(j, W, MAX(I) + 1)|. Notice for example, that DSPM(0, W, I) = |I|.

We now formalize our notion of the information retrieval activity of M.

DEFINITION 12

Let I, J be adjacent input intervals of W. We define the adjacent overlap events of W during I and J to be the set

 $\{(s, t, j, z, \sigma) \in OVERLAP(W) | MIN(I) \leq s \leq MAX(I) < t \leq MAX(J) \}$. We shall denote this set as ADJ-OVERLAP(W, I, J). Notice that ADJ-OVERLAP(W, I, J) consists of those overlap events where information was first stored while \mathfrak{M} was processing W0I and then retrieved while \mathfrak{M} was processing W0J.

DEFINITION 13

Let I be an input interval of W such that |I|>1. We define the **DIVIDING POINT of W during I** (denoted by DP(W, I)) to be the least natural number 1 (MIN(I) $\leq 1 < MAX(I)$), such that |ADJ-OVERLAP(W, [MIN(I), 1], [1+1, MAX(I)])| is maximized. For convenience, we shall denote the set ADJ-OVERLAP(W, [MIN(I), DP(W, I)], [DP(W, I)+1, MAX(I)]) as SP-OVERLAP(W, I).

In the argument to follow, |SP-OVERLAP(W, I)| will serve to capture our notion of the information retrieval activity of \mathfrak{M} while it processes W@I. It is for this reason that we shall refer to the elements of SP-OVERLAP(W,I) as special overlap events.

LEMMA 1

Let I be an input interval of W. Then

i) $|OVERLAP(W)| \leq kL(W)$, and

ii) $|SP-OVERLAP(W, I)| \le k |I|$.

Proof:

i) For any given worktape and step of \mathfrak{M} , at most one overlap event can occur. Since there are at most L(W) steps for which an overlap event can occur we have that $|OVERLAP(W)| \leq kL(W)$.

ii) For any given worktape of \mathfrak{M} , there are at most |I| steps for which an overlap event can occur while the input head is scanning some part of W@I. This implies that $|SP-OVERLAP(W, I)| \le k|I|.\Box$

We next demonstrate that for any set $S \subset \Gamma^{\mathbb{N}}$ for sufficiently large $\mathbb{N} \in \mathbb{N}$, while \mathfrak{M} is processing the words of S there must exist some large subinterval I of [1, N] for which:

mean |SP-OVERLAP(W, I)| is a small fraction of |I|.

That is, on the average \mathfrak{M} will exhibit relatively small information retrieval activity while processing the words of S at I. We shall argue that if mean |SP-OVERLAP(W, I)| was a large fraction of |I| for all large subintervals I of [1,N], then mean |OVERLAP(W)| would be greater $W \in S$ than kN and thus contradict lemma 1.

Let N=pn²⁽ⁿ⁻¹⁾ for some integer p≥1 and some integer n>1. Let W∈ Γ^N .

We define a set \Im of input intervals of W as follows:

for $0 \le i \le 2(n-1)$ and $0 \le j < N/(pn^i)$, we define $I_{ij} = [jpn^i+1, (j+1)pn^i]$.

 \Im is then defined to be the set of input intervals

 $\{I_{ii} \mid 0 \le i \le 2 (n-1) \text{ and } 0 \le j < N/(pn^i)\}.$

For convenience, we define for $0 \le i \le 2(n-1)$ the set \Im_i of input intervals at level i to be $(I_{ij} \in \Im | 0 \le j < N/(pn^i))$.

We observe the following:

i) Level i is a partition of [1, N] into N/(pnⁱ) intervals, each of size pnⁱ.

ii) Each interval at level i+1 consists of n consecutive intervals at level i, viz., $I_{i+1,j}$ is the disjoint union of I_{im} such that $nj \le m < n(j+1)$.

iii) Any two input intervals are either disjoint or one must contain the other.

To aid the reader, we offer an example which illustrates the set \Im of input intervals.

Let n=3. Then for $W \in \Gamma^N$, where N=81p, we may specify the set \Im of input intervals of W as follows:

$$\Im = \bigcup_{i=0}^{4} \Im_i$$
 where:

 $\Im_0 = (I_{00}, I_{01}, \dots, I_{080}) = ([1, p], [p+1, 2p], \dots, [80p+1, N]),$

 $\Im_1 = \{I_{10}, I_{11}, \dots, I_{126}\} = \{[1, 3p], [3p+1, 6p], \dots, [78p+1, N]\},\$

 $\Im_2 = \{I_{20}, I_{21}, \dots, I_{28}\} = \{[1, 9p], [9p+1, 18p], \dots, [72p+1, N]\},\$

 $\Im_3 = \{I_{30}, I_{31}, I_{32}\} = \{[1, 27p], [27p+1, 54p], [54p+1, N]\}, and$

 $\Im_{a} = (I_{a0}) = ([1, N]).$

Notice that level i consists of $81/3^{i}$ distinct input intervals, each of size $p3^{i}$.

THEOREM 1

Let \mathfrak{M} be a k-worktape real time Turing machine with input alphabet Γ . Let N=pn²⁽ⁿ⁻¹⁾ for some integer p≥1 and some integer n>1. Suppose Sc $\Gamma^{\mathbb{N}}$. Then for some i (n-1≤i≤2(n-1)), there exists an interval I∈ \mathfrak{F}_i such that

> mean $|SP-OVERLAP(W, I)| \leq 3k |I| / (n-1)$. Wes

Proof:

Let WeS. We proceed by first proving the following lemma.

$$\sum_{I \in \mathfrak{F}} |SP-OVERLAP(W, I)| \leq 3kN.$$

Proof:

Let $\widehat{S} = (I \in \widehat{S} | (\forall J \in \widehat{S}) [|J| \rangle | I| \Rightarrow DP(W, J) \notin I])$, i.e., $I \in \widehat{S}$ only if I does not contain the dividing point of any input interval at a greater level. Lemma 2 shall follow immediately from claim 1 and claim 2.

CLAIM 1

$$\sum_{I \in \widehat{S}} |SP-OVERLAP(W, I)| \le kN.$$

Proof:

Let I, $J \in \widehat{S}$ such that $I \neq J$. We shall show that **SP-OVERLAP** (W, I) \cap **SP-OVERLAP** (W, J) = ϕ from which we may conclude

$$\sum_{I \in \widehat{S}} |SP-OVERLAP(W, I)| \le |OVERLAP(W)|$$
$$\le kN \quad (by \ lemma \ 1.)$$

case 1

Suppose $I \cap J = \phi$. Clearly then, SP-OVERLAP (W, I) \cap SP-OVERLAP (W, J) = ϕ .

case 2

Suppose $I \subseteq J$. Since $I \in \widehat{S}$, $DP(W, J) \notin I$. Thus either $I \subset [MIN(J), DP(W, J)]$ or $I \subseteq [DP(W, J) + 1, MAX(J)]$. Suppose the first alternative holds. But then the special overlap events of W at I occur before step DP(W, J), while the special overlap events of W at J occur after step DP(W, J). Hence, $SP-OVERLAP(W, I) \cap SP-OVERLAP(W, J) = \phi$. The second alternative may be dealt with similarly.

CLAIM 2

$$\sum_{I \in (\mathfrak{F} - \hat{\mathfrak{F}})} | SP-OVERLAP(W, I) | \le 2kN.$$

Proof:

Let i be such that $0 \le i < 2 (n-1)$. The maximum number of input intervals at level i such that each contains the dividing point of an interval at a greater level is bounded, a fortiori, by the total number of intervals at a greater level than level i. Thus,

$$|\mathfrak{F}_{i} \cap (\mathfrak{F}_{i})| \leq \sum_{j > i} |\mathfrak{F}_{j}|$$

$$= \sum_{i < j \le 2} N/(pn^j)$$

= $(N/p) ((1/n^{i}) - (1/n^{2(n-1)})) / (n-1)$ < $(N/p) (1/n^{i}) / (n-1)$ = $|\Im_{i}| / (n-1)$.

By lemma 1, we have therefore that

 $\sum_{I \in (\mathfrak{F}_{i} \cap (\mathfrak{F}))} |SP-OVERLAP(W, I)| \le |\mathfrak{F}_{i}| kpn^{i} / (n-1)$ = kN/(n-1).

Thus, by summing over levels 0 through 2(n-1)-1 we have that

$$\sum_{\mathbf{I} \in (\mathfrak{T} - \widehat{\mathfrak{T}})} | SP-OVERLAP(W, \mathbf{I}) | \le 2kN.$$

Lemma 2 follows immediately from claim 1 and claim 2.

Lemma 2 trivially implies that

(*) mean
$$\left[\sum_{I \in \mathfrak{S}} |SP-OVERLAP(W, I)|\right] \leq 3kN.$$

We now proceed to prove theorem 1.

Suppose to the contrary that for every $I \in \Im_i$ such that $(n-1 \le i \le 2(n-1))$, mean |SP-OVERLAP(W, I)| > 3k |I| / (n-1).

Then mean $\begin{bmatrix} \sum_{I \in \mathfrak{Y}} |SP-OVERLAP(W, I)| \end{bmatrix}$

 $\geq \underset{W \in S}{\text{mean}} \left[\sum_{(n-1) \leq i \leq 2 \text{ (n-1)}} \sum_{0 \leq j \leq (N/(pn^{i})) - 1} | \text{SP-OVERLAP (W, I_{ij})} | \right]$

$$= \sum_{(n-1) \leq i \leq 2 (n-1)} \sum_{\substack{0 \leq j \leq (N/(pn^{i})) - 1 \\ W \in S}} [mean | SP-OVERLAP(W, I_{ij}) |]$$

> 3k/(n-1) $\sum_{(n-1)\leq i\leq 2(n-1)}$ $\sum_{0\leq j\leq (N/(pn^i))-1} |I_{ij}|$

= 3k/(n-1) $\sum_{(n-1) \le i \le 2(n-1)} N$

> 3kN

which contradicts (★).□

THE LANGUAGE L.

In this section, we specify the language L_k for $k \ge 2$ which can be recognized by a k-worktape real time Turing machine but cannot be recognized as fast by any (k-1)-worktape Turing machine. L_k may be derived as follows:

Let \mathfrak{S} be a machine with k pushdown stacks; we use $i \in \{1, \ldots, k\}$ to index the stacks. There are three canonical operations which \mathfrak{S} may perform:

1) push symbol α onto stack i,

2) push symbol β onto stack i, and

3) pop stack i.

Such actions may be encoded by supplying specific inputs to G. We shall let:

 $\{a_i \mid 1 \leq i \leq k\}$ denote operation 1),

 $\{b, | 1 \le i \le k\}$ denote operation 2), and

 $\{c_1 \mid 1 \leq i \leq k\}$ denote operation 3).

Thus for example, the input $a_1c_3b_2c_1$ would cause \mathfrak{S} to push α onto stack 1, pop stack 3, push β onto stack 2, and finally pop stack 1. If stack i is empty when c_i is scanned, then by convention the encoded action is ignored.

Let $(a_i|1 \le i \le k) U(b_i|1 \le i \le k) U(c_i|1 \le i \le k)$ be denoted by Γ_k . We define L_k to be those words $W \in \Gamma_k^*$ such that the last operation encoded by W does <u>NOT</u> involve popping symbol α from some stack.

Obviously, L_k is accepted by a k-worktape real time Turing machine \mathfrak{M} , which simulates the behavior of \mathfrak{S} by simulating a different stack on each worktape. The input alphabet of \mathfrak{M} is Γ_k while the worktape alphabet is $(\mathfrak{B}, \alpha, \beta)$ where \mathfrak{B} denotes the blank symbol. When \mathfrak{M} scans a_i or b_i , it writes α or β respectively on the ith worktape and then shifts the ith tape head one square to the right. When c_i is scanned, \mathfrak{M} writes \mathfrak{B} on the ith worktape (thus simulating a pop) and then shifts the ith tape head one square to the left. When the right endmarker is scanned, if no worktape head is scanning α , the input word is accepted.

 L_k may be termed an information retrieval language. As long as the input symbols that \mathfrak{M} scans are a_i or b_i it will store information. When \mathfrak{M} scans c_i it will retrieve information.

DEFINITION 14

Let $i \in \{1, ..., k\}$ and let $W \in \Gamma_k^*$. PROJ(W, i) is defined to be that word which is obtained from W be deleting every a_j, b_j, c_j in W such that $j \neq i$, i.e., the projection of W on the ith index.

For example, $PROJ(c_{2}a_{1}b_{2}a_{2}a_{3}c_{4}, 2) = c_{2}b_{2}a_{2}$. We note that |PROJ(W, i)| equals the number of symbols in W with index i.

We now define a set ScL, with the following properties:

i) Let $W \in S$ and let $1 \le i < j \le k$. Then the number of symbols with index i in W is significantly less than the number of symbols with index j.

ii) All words of ${\bf S}$ encode actions which reference the stacks in the same order.

For integer m>1, let \widehat{m} denote $\sum_{i=1}^{k} m^{i} = (m^{k+1}-m)/(m-1)$ and let **PSTRING** (pattern-string) denote $\prod_{i=1}^{k} a_{i}^{m^{i}} \in (a_{i}|1 \le i \le k)^{\widehat{m}}$. Thus we have that:

i) $(\forall i \mid 1 \leq i \leq k) [\mid PROJ(PSTRING, i) \mid =m^{i}]$, and

ii) $(\forall i, j | 1 \le i < j \le k) [| PROJ (PSTRING, j) | = m^{j-i} | PROJ (PSTRING, i) |].$

Therefore by choosing m appropriately large, we may cause the relative frequencies of symbols occurring in **PSTRING** with a specific index to vary as greatly as desired.

Let $N = q \widehat{m}$ for some integer q > 0 and let SEED denote the string $(PSTRING)^{q} \in \{a_{i} | 1 \le i \le k\}^{N}$. All the words of S shall be derived from SEED.

We define $S \subset (\{a_i | 1 \le i \le k\} | U\{b_i | 1 \le i \le k\})^N$ by the condition that $W \in S$ if and only if W is obtainable from SEED by replacing arbitrarily selected occurrences of a_i by b_i for $1 \le i \le k$. Formally, we may specify S as follows:

Let h: $\Gamma_k^* \rightarrow \Gamma_k^*$ be a homomorphism defined for $\gamma \in \Gamma_k$ as:

h(γ) = a_i if γ = b_i for $1 \le i \le k$, h(γ) = γ otherwise.

Then we define S to be h^{-1} (SEED). We offer the following remarks:

i) S⊂L_k.
ii) |S|=2^N.

iii) For a given N, PSTRING completely specifies the indices of all the symbols of any word in S.

iv) Let X, Y \in S such that X \neq Y. Then there exists n, $1 \le n \le N$, such that X@[n, n] and Y@[n, n] are different symbols with the same stack index i. Thus by the definition of L_k , there exists a string $V \in \Gamma_k^*$ for which $XV \in L_k$ if and only if $YV \notin L_k$. Specifically, we can choose $V = c_i^r$ where r = |PROJ(SEED@[n, N], i)|.

Suppose \mathfrak{N} is a (k-1)-worktape Turing machine which can simulate the behavior of \mathfrak{S} in real time. Consider how \mathfrak{N} might behave on XV such that X \in S and V $\in \Gamma_k^*$. We would expect \mathfrak{N} to use a single worktape 1 to simulate, at least in part, the operation of two stacks indexed i, j (i < j) of \mathfrak{S} . By the construction of S, \mathfrak{N} must store considerably more information obtained from processing input symbols with index j, (which we shall refer to as "j-information") than from processing input symbols with index i, i.e., "i-information".

Suppose in order to store new j-information, tape head 1 incurs significant displacement away from old i-information. After processing X, \Re must be able to retrieve old i-information fast in case V necessitates the retrieval of i-information. Since \Re operates in real time, tape head 1 would not have enough time to traverse the new j-information in order to retrieve the old i-information. Hence in order to operate properly, tape head 1 must "carry along" old i-information as it stores new j-information. But this can only be achieved if tape head 1 can sustain substantial information retrieval activity.

In the next section we shall invoke Theorem 1 and show there is an

input interval of the words in S for which the information retrieval activity is small enough to prevent \mathfrak{N} from proper operation. This shall imply that no (k-1)-worktape Turing machine can recognize L_k in real time.

A Proof of the Rabin-Hartmanis-Stearns Conjecture

Let L_k be defined as in the previous section for fixed $k \ge 2$. We now proceed to prove by contradiction that there exists no (k-1)-worktape Turing machine which can recognize L_k in real time.

Assume that \mathfrak{N} is a (k-1)-worktape Turing machine with input alphabet Γ_k , worktape alphabet Σ , and set Q of internal states which recognizes L_k in real time.

Let $m \in \mathbb{N}$ be such that $m > max[4(\log|Q| + \log(2k)), 32k^2\log|\Sigma|]$. As in the previous section, we let $\widehat{m}^{\deg f} \sum_{\substack{i=1 \\ i=1}}^{k} m^i = (m^{k+1} - m) / (m-1)$. Let $n \in \mathbb{N}$ be such that $n > 12k(k-1)\widehat{m} + 1$ and let $N = k\widehat{m}n^{2(n-1)}$.

By the methods outlined previously, we may construct a set $S \subset \Gamma_k^N$ such that $S \subset L_k$. The reader should consult the previous section for the special properties of S.

By Theorem 1, there must exist some subinterval I of [1,N] with the following properties:

i) $N^{1/2} < |I| \le N$,

ii) |I| is a multiple of km, and

iii) mean $|SP-OVERLAP(W, I)| \le 3(k-1)|I|/(n-1)$ $W \in S$

 $\langle |I|/(4k\hat{m})$ (by our choice of n.)

(* all logs are to the base 2.)

Since \mathfrak{N} has only (k-1) worktapes, we expect \mathfrak{N} to use a single worktape 1 to store, at least in part, the information obtained from processing input symbols with index i and index j, for some i, j $(1 \le i < j \le k)$. We shall refer to the information that is stored as a result of processing input symbols with index i and index j as "i-information" and "j-information", respectively. As implied by the discussion at the end of the previous section, we expect that \mathfrak{N} can only process correctly words of the form X0[1, MAX(I)]V for X \in S and $V \in \Gamma_k^*$, if the computation proceeds with substantial information retrieval activity while X0[1, MAX(I)] is processed.

Let $S_0^{d \in f} \{ w \in S \mid | SP-OVERLAP(w, I) \mid > | I | / (k \hat{m}) \}$.

LEMMA 3

|S₀|<|S|/4.

Proof:

Suppose to the contrary that $|S_0| \ge |S|/4$. Then mean $|SP-OVERLAP(W, I)| \ge |S_0||I|/(|S|km)$ $W \in S$ > |I|/(4km)

which contradicts property iii) of subinterval $I.\,\square$

Since \mathfrak{N} may incur moderate information retrieval activity while processing the words of S_0 at I, we expect that \mathfrak{N} might be able to process correctly words of the form, X0[1, MAX(I)]V for X $\in S_0$ and $V \in \Gamma_k^*$. On the other hand, we shall argue that \mathfrak{N} cannot process properly all the words of the form X0[1, MAX(I)]V if X \in (S-S₀). Intuitively, since \mathfrak{N} must incur little information retrieval activity while processing X at I, it will not be able to store the i-information and j-information on worktape 1 in such a manner that would allow it to process correctly any suffix V in real time. This will imply that \mathfrak{N} cannot process correctly all the words of $(S-S_0)$ and therefore contradict the assumption that \mathfrak{N} can recognize L, in real time.

We proceed by formalizing our intuition. Let $X \in (S-S_0)$. We suppose that $\mathfrak N$ can only "store properly" either i-information or j-information (but not both!) on worktape 1 while it processes X@I. (By store properly, we mean, storing the information in such a way that would allow $\mathfrak N$ to retrieve any part of it fast enough so that it may process any suffix in real time.) Since \mathfrak{N} incurs little information retrieval activity while processing X@I, tape head 1 cannot revisit too many tape squares while $\mathfrak N$ stores information on worktape 1. Therefore, we expect \mathfrak{N} to use some scheme for storing information on the 1th worktape which approaches that of transcribing information onto consecutive tape squares such that the tape head is always close to the most recent information that is stored. Thus while processing X@I, if ${\mathfrak N}$ predominantly uses worktape 1 to store properly i-information, we expect DSPM(1, X, I) to be a large fraction of |PROJ(X@I, i)| which is the amount of i-information needed to be stored in order to process correctly X@I. Otherwise, we expect DSPM(1, X, I) to be a large fraction of |PROJ(X@I, j)|.

In order to discover what information has been stored properly, we shall view the computation of \mathfrak{N} on X@I in k stages, indexed 1 through k. For i such that $1 \le i \le k$, we shall use stages i, i+1,..., k to determine whether \mathfrak{N} has stored properly i-information on any worktape

j $(1 \le j \le k-1)$. We shall show that for some i, no worktape j will store properly the i-information during stages i, i+1,...,k.

We divide I into k consecutive equal sized subintervals I_1, \ldots, I_k , namely,

 $I_i^{def}[((i-1)|I|/k)+1,i|I|/k] \text{ for } 1 \leq i \leq k.$

Thus $I = \bigcup_{i=1}^{N} I_i$. Stage i shall correspond to the computation of \mathfrak{N} on X0I_i. Thus, stage i consists of the set of indices of steps equal to I_i .

LEMMA 4

Let $X \in S$ and let $h, i \in \{1, \ldots, k\}$. Then

 $|PROJ(X@I_{b}, i)| = m^{i} |I| / (k\widehat{m}).$

PROOF:

By the construction of S it suffices to show that the lemma holds for X=SEED.

 $|PROJ(SEED@I_{h}, i)| = |PROJ((PSTRING)^{|I|/(k\widehat{m})}, i)|$ $= |PROJ(PSTRING, i)||I|/(k\widehat{m})$ $= m^{i}|I|/(k\widehat{m}).\Box$

To aid the reader, we present the function NUM: $\{1, \ldots, k\} \rightarrow \mathbb{N}$ where NUM(i)^d $\mathbb{P}^{f} \mathbb{m}^{i} | I | / (k \widehat{\mathbb{m}})$.

Thus, NUM(i) equals the number of symbols with index i in X at any subinterval I_h (1 \leq h \leq k) of I.

Suppose $X \in (S-S_0)$. We note that i-information will not be stored properly on worktape j if either:

i) worktape j predominantly stores $\hat{1}$ -information for some $\hat{1} < i$, or

ii) worktape j predominantly stores $\hat{1}$ -information for some $\hat{1}$ >i.

We expect the first alternative to hold if

DSPM(j, X, I,) $\langle (4k/m) NUM(i) \rangle$.

That is, during stage i, if tape head j does not incur displacement greater than a small fraction of the displacement we expect necessary for the proper storage of i-information, we expect that \mathfrak{N} is primarily storing $\hat{1}$ -information for some $\hat{1}$ <i on worktape j.

We expect the second alternative to hold if

DSPM(j, X, $\bigcup_{i \leq 1} I_i$) >kNUM(i).

That is, during stages $i+1, \ldots, k$, if tape head j incurs displacement significantly greater than the displacement we expect necessary for the proper storage of i-information, we expect that \mathfrak{N} is primarily storing $\hat{1}$ -information for some $\hat{1}$ >i on worktape j.

This leads us to conjecture that i-information will not be stored properly during the computation of $\mathfrak N$ on X0I if

 $(\forall j \mid 1 \leq j \leq k-1)$ [DSPM (j, X, I_j) < (4k/m) NUM (i) or DSPM (j, X, U₁) > kNUM (i)].

For convenience, we shall define for $1 \le i \le k$,

 $y_i^{def}(4k/m) NUM(i)$, and

 $Z_i^{def} k NUM(i)$.

For 1≤i≤k,

 $\mathbf{S}_{i}^{\text{def}} \{ X \in (\mathbf{S} - \mathbf{S}_{0}) \mid (\forall j \mid 1 \leq j \leq k-1) \ [\text{DSPM}(j, X, \mathbf{I}_{i}) < \mathbf{y}_{i} \text{ or } \text{DSPM}(j, X, \bigcup_{i \leq 1} \mathbf{I}_{i}) > \mathbf{z}_{i}] \}.$

In particular, we note that $X \in S_k$ only if $DSPM(j, X, I_k) < y_k$ for all j $(1 \le j \le k-1)$.

Intuitively, we expect that $X \in S_i$ only if \mathfrak{N} does not store properly i-information while processing X@I.

Suppose $X \in (S-S_0)$. Since there is little information retrieval activity while \mathfrak{N} processes X@I, we expect for some i, that \mathfrak{N} does not store properly i-information while processing X@I, i.e., we expect that $X \in S_i$ for some i. We now confirm our expectations by proving the following two lemmas.

LEMMA 5

Let $X \in S$ and let J,L be input intervals of X such that JCL. Let j be a tape index of \mathfrak{N} . Suppose that $DSPM(j, X, J) \ge y$ and $DSPM(j, X, L) \le z$ for $y, z \in \mathbb{N}$ such that $y \ge 2z$. Then

| SP-OVERLAP (X, L) $| \ge (y/2) - (z+1)$.

PROOF:

Since DSPM $(j, X, J) \ge y$ we have that either:

i) $|POSN(j, X, MAX(J)+1) - POSN(j, X, MIN(L))| \ge y/2$, or

ii) $|POSN(j, X, MIN(J)) - POSN(j, X, MIN(L))| \ge y/2.$

Suppose the first alternative holds. Since DSPM(j, X, L) $\leq z$, the tape square denoted by POSN(j, X, MAX(L)) must be within (z+1) tape squares of the square denoted by POSN(j, X, MIN(L)). Therefore after step MAX(j)+1, tape head j must revisit at least (y/2)-(z+1) tape squares

which implies that $|SP-OVERLAP(X, L)| \ge (y/2) - (z+1)$. The second alternative may be dealt with similarly.

We note that an improved lower bound of (y-(z+1))/2 may be established by a more complicated proof. The weaker bound shall suffice for our purposes. We also remark that lemma 5 is a direct consequence of the linear structure of the worktapes of $\mathfrak{N}.$

LEMMA 6

$$(S-S_0) = \bigcup_{i=1}^k S_i$$

PROOF:

Clearly it suffices to show that $(S-S_0) \subseteq \bigcup_{i=1}^k S_i$. Let $X \in (S-S_0)$ and suppose to the contrary that $X \notin S_i$ for all i $(1 \le i \le k)$. Then by the definition of S_i for i>0, we have that

 $(\forall i \mid 1 \leq i \leq k) \quad (\exists j \mid 1 \leq j \leq k-1) \quad [DSPM(j, X, I_i) \geq y_i \text{ and } DSPM(j, X, \bigcup_{i \leq 1} I_i) \leq z_i].$ But then for some $1 \le i_1 < i_2 \le k$, there exists \hat{j} for which:

DSPM
$$(\hat{j}, X, I_{i_1}) \ge y_{i_1}$$
 and DSPM $(\hat{j}, X, \bigcup_{i_1 \le 1} I_1) \le z_{i_1}$, and
DSPM $(\hat{j}, X, I_{i_2}) \ge y_{i_2}$ and DSPM $(\hat{j}, X, \bigcup_{i_1 \le 1} I_1) \le z_{i_2}$.

By lemma 5, the second and third of these bounds on DSPM imply that

$$|SP-OVERLAP(X, \bigcup_{i_{1} < 1} I_{1})| \ge (y_{i_{2}}/2) - (z_{i_{1}}+1)$$

= (2k/m) NUM (i_{2}) - kNUM (i_{1}) - 1
$$\ge 2kNUM(i_{1}) - kNUM(i_{1}) - 1$$

$$> NUM(1)$$

> |I|/(k俞).

But this implies that $X \in S_0$ and thus contradicts our choice of X. \Box

We call attention to the fact that lemma 6 is the only part of the proof which relies directly on \mathfrak{N} having only (k-1)-worktapes.

We have shown therefore that $S = \bigcup_{i=0}^{k} S_i$. We shall eventually prove that $\sum_{i=1}^{k} |S_i| < |S|/2$ which along with lemma 3 shall contradict the assumption that \Re can recognize L_k in real time.

DEFINITION 15

Let X, Y \in S and let i \in (1, ..., k). X = Y if and only if

 $(\forall n \in [1, N]) [X@[n, n] = Y@[n, n] \text{ or } (n \in I, and \{X@[n, n], Y@[n, n]\} = \{a_i, b_i\}\}].$

Thus $X \equiv_i Y$ if and only if X and Y are identical except for possible differences involving symbols with index i in $X \otimes I_i$ and $Y \otimes I_i$. Clearly then, \equiv_i is an equivalence relation on S.

We define E_i^X to be the equivalence class of X with respect to \equiv_i , that is,

 $E_i^{X \text{ def}} \{Y \in S \mid Y \equiv_i X\}.$

We note that $|E_i^X| = 2^{NUM(i)}$.

Suppose $X \equiv_{i} Y$ for some $i \in \{1, \ldots, k\}$ such that $X \neq Y$. By the definition of L_k there exists $V \in \Gamma_k^*$ such that $X \otimes [1, MAX(I)] V \in L_k$ if and only if $Y \otimes [1, MAX(I)] V \notin L_k$. Since X and Y are identical except for differences at I_i involving symbols with index i, we expect that \mathfrak{N} could distinguish between $X \otimes [1, MAX(I)] V$ and $Y \otimes [1, MAX(I)] V$ only if it stored properly the i-information obtained during stage i of the computations. Thus in order for \mathfrak{N} to operate properly, we expect that very few words of E_i^X are also contained in S_i . We now show that this is indeed the case.

LEMMA 7

Let $X \in S$ and let $i \in \{1, \ldots, k\}$. Then

$$|\mathbf{E}_{i}^{\mathsf{x}} \cap \mathbf{S}_{i}| < |\mathbf{E}_{i}^{\mathsf{x}}|/(2k).$$

PROOF:

Let X, Y \in S_i for i>0, be distinct and suppose X=_iY. Let V \in Γ_k^* . We shall denote X@[1, MAX(I)] by \hat{X} and Y@[1, MAX(I)] by \hat{Y} .

Consider the computations of \mathfrak{N} on $\widehat{\mathsf{XV}}$ and $\widehat{\mathsf{YV}}$. We shall show that if the i.d.'s immediately after step MAX(I_i)+1, i.e., immediately after stage i, are very equivalent then the subsequent i.d.'s immediately after step MAX(I)+1, i.e., immediately after stage k, become even more equivalent. If the equivalence after stage k has become too great, we shall show that \mathfrak{N} could not process correctly a suffix V which distinguishes between $\widehat{\mathsf{X}}$ and $\widehat{\mathsf{Y}}$. This shall imply that $|\mathbf{E}_i^{\mathsf{X}} \cap \mathbf{S}_i|$ is bounded by some number of non-equivalent i.d.'s that \mathfrak{N} can achieve immediately after stage i. We proceed by first presenting claim 1 and claim 2.

CLAIM 1

Suppose $ID(\hat{X}V, MAX(I_i)+1)$ and $ID(\hat{Y}V, MAX(I_i)+1)$ are $(y_i+|I|/(k\hat{m}))$ -equivalent. Then $ID(\hat{X}V, MAX(I)+1)$ and $ID(\hat{Y}V, MAX(I)+1)$ are $(y_i+z_i+|I|/(k\hat{m}))$ -equivalent.

PROOF:

Consider the computations of \mathfrak{N} on $\widehat{X}V$ and $\widehat{Y}V$. Since the input tape is one-way and $X \equiv_i Y$, the subsequent computations after stage i can differ only if some tape head scans different i-information which was first stored while \mathfrak{N} processed the inputs during stage i. Let j>0 be a tape index of \mathfrak{N} . We shall proceed by showing that worktape j must satisfy one of the following conditions:

i) The information stored on worktape j during stage i is identical for each computation, i.e.,

 $(\forall z \in \mathbb{Z})$ [CONT (j, z, $\hat{X}V$, MAX (I,)+1) = CONT (j, z, $\hat{Y}V$, MAX (I,)+1)].

ii) The information scanned on worktape j during stages i+1,...,k, is identical for each computation. In addition, immediately after stage k of each computation, tape head j is displaced at least $(y_i+z_i+|I|/(k\widehat{m}))$ tape squares away from any possibly different i-information.

Claim 1 will follow immediately from case 1 and case 2. We note that $DSPM(j, \hat{X}V, I_i)$ necessarily equals $DSPM(j, \hat{Y}V, I_i)$ and that the displacement is in the same direction.

case 1

Suppose DSPM(j, $\hat{x}V$, I_i) $\langle y_i$. We proceed by showing that worktape j must satisfy condition 1. (Intuitively, this case implies that during stage i, worktape j was used predominantly for storing $\hat{1}$ -information for some $\hat{1}\langle i$. Because of the substantial equivalence immediately after stage i, we shall be able to argue that no different

i-information could have been stored on worktape j during stage i.)

Suppose to the contrary that different i-information was stored on worktape j during stage i, that is,

 $(\exists z \in \mathbb{Z}) [CONT(j, z, \hat{X}V, MAX(I_i)+1) \neq CONT(j, z, \hat{Y}V, MAX(I_i)+1)].$ Since $ID(\hat{X}V, MAX(I_i)+1)$ and $ID(\hat{Y}V, MAX(I_i)+1)$ are $(\hat{y}_i+|I|/(k\hat{m}))$ -equivalent on worktape j and $DSPM(j, \hat{X}V, I_i) < \hat{y}_i$, this implies without loss of generality that $|SP-OVERLAP(\hat{X}V, I_i)| > |I|/(k\hat{m}).$ But then $|SP-OVERLAP(X, I)>|I|/(k\hat{m})$ which contradicts $X \in S_i$ for i>0.

case 2

Suppose DSPM $(j, \hat{X}V, I_i) \ge y_i$. We note that i k for this case to apply. (Intuitively, this case implies that during stage i, worktape j was used predominantly for storing î-information for some î>i. However, it is quite possible that different i-information was also stored. Because of the substantial equivalence immediately after stage i, and because \mathfrak{N} must store additional î-information during stages i+1,...,k, we shall be able to argue that worktape j must satisfy condition 2.)

Suppose to the contrary that during stages i+1,...,k, worktape j scanned different information. Clearly then, this information was first stored while \mathfrak{N} was processing the inputs during stage i. But since $ID(\hat{x} \vee, MAX(I_i)+1)$ and $ID(\hat{y} \vee, MAX(I_i)+1)$ are $(y_i+|I|/(k\hat{m}))$ -equivalent, worktape j can only scan different information during stages i+1,...,k, if without loss of generality $|SP-OVERLAP(\hat{x} \vee, \bigcup_{l=i}^{k} I_l)| > (y_i+|I|/(k\hat{m}))$, which implies that $|SP-OVERLAP(X, I)| > |I|/(k\hat{m})$ which contradicts XeI, for i>0.

Since X, Y \in S_i, case 2 implies that DSPM(j, \hat{X} V, $\bigcup_{i \in I_1} \rangle > z_i$ and DSPM(j, \hat{Y} V, $\bigcup_{i \in I_1} | i_i \rangle > z_i$. The displacement incurred during stages i+1,...,k, of the computations must be in the same direction as the displacement incurred during stage i. (Otherwise, there would be too many special overlap events.) But since ID(\hat{X} V, MAX(I_i)+1) and ID(\hat{Y} V, MAX(I_i)+1) are (y_i+|I|/(km̂))-equivalent, immediately after stage k, tape head j must be at least y_i+z_i+|I|/(km̂)) tape squares away from any possibly different i-information.

CLAIM 2

 $|\mathbf{E}_{i}^{\mathsf{X}} \cap \mathbf{S}_{i}|$ is bounded by the number of non- $(\mathbf{y}_{i}+|\mathbf{I}|/(k\widehat{m}))$ -equivalent i.d.'s that \mathfrak{N} can achieve immediately after stage i, that is, immediately after step MAX(I,)+1.

PROOF:

Immediately after stage k, i.e., immediately after step MAX(I)+1, at most (k-i+1)NUM(i) pop operations are needed to retrieve the different i-information that \mathfrak{N} stored while processing \hat{X} and \hat{Y} during stage i. Thus for some $r \leq (k-i+1)NUM(i)$, we have that

(*) $\hat{X}c_{i}^{r} \in L_{k}$ if and only if $\hat{Y}c_{i}^{r} \notin L_{k}$.

Suppose that $ID(\hat{X}c_i, MAX(I_i)+1)$ and $ID(\hat{Y}c_i, MAX(I_i)+1)$ are $(y_i+|I|/(k\hat{m}))$ -equivalent. Then by claim 1, we have that $ID(\hat{X}c_i, MAX(I)+1)$ and $ID(\hat{Y}c_i, MAX(I)+1)$ are $(y_i+z_i+|I|/(k\hat{m}))$ -equivalent. But since $r \le (k-i+1) \text{ NUM } (i)$ $\langle y_i + z_i + |I| / (k \widehat{m}),$

we have that

 $\hat{X}c_{i} \in L_{k}$ if and only if $\hat{Y}c_{i} \in L_{k}$

which contradicts (*). Claim 2 immediately follows.

We now proceed to prove lemma 7. Claim 2 implies that

 $|E_{i}^{X} \cap S_{i}| \leq |Q| |\Sigma|^{(k-1)(2(y_{i}+|I|/(km))+1)} (2|I_{i}|+1)^{(k-1)}$

We shall prove that $|E_i^x \cap S_i| < |E_i^x|/(2k)$ by showing that

 $log |Q| + (k-1) (2(y_{i}+|I|/(k\hat{m}))+1) log |\Sigma| + (k-1) log (2|I_{i}|+1) + log (2k) < log |E_{i}^{k}|$ = NUM (i).

> 1) $\log |Q| + \log (2k) < m/4$ (by our choice of m) < NUM(i)/4.

> 2) $(k-1) (2(y_i+|I|/(k\hat{m}))+1) \log |\Sigma| < 4ky_i \log |\Sigma|$ = $16k^2 \log |\Sigma| NUM(i) / m$ < NUM(i)/2 (by our choice of m.)

3) $(k-1)\log(2|I_i|+1) < k\log|I|$ < $m^i |I|/(4k\hat{m})$ (since $\log|I|/|I| < m^i/(4k^2\hat{m})$ by our choice of n) = NUM(i)/4.

Lemma 7 follows from 1), 2), and 3). \Box

LEMMA 8

$$\sum_{i=1}^{k} |S_i| < |S|/2.$$

PROOF:

Let $i \in \{1, \ldots, k\}$. Since $(E_i^x | X \in S)$ partitions S, we have that:

 $|S_{i}| = |\bigcup_{x \in S} (E_{i}^{x} \cap S_{i})|$ < 1/(2k) | U E_{i}^{x}| (by lemma 7) x \in S = |S|/(2k).

Lemma 8 follows immediately.

Lemma 6 implies that:

$$|S| = |S_0| + \sum_{i=1}^{k} |S_i|$$

< $|S|/4 + |S|/2$ (by lemma 3 and lemma 8)

< |S|

which is a contradiction. Hence we conclude that \mathfrak{N} cannot recognize L_{L} in real time.

INDEX	
9N	8
Г	8
Σ	Ę
Q	8
tape index	8
W	8
L (W)	8
	9
s	
POSN(j, W, s)	g
CONT (j, z, W, s)	g
STATE (W, s)	9
ID (W, s)	10
n-equivalent on worktape j	10
n-equivalent	11
interval	11
[1, m]	11
MIN(I)	11
MAX(I)	11
III	11
ø	11
adjacent intervals	11
subinterval	12
input interval	12
Wei	12

overlap event			12
OVERLAP (W)			. 13
real time Turing machine			14
real time			14
DSPM (j, W, I)			15
adjacent overlap events			15
ADJ-OVERLAP (W, I, J)			15
dividing point			15
DP(W, 1)		×	15
SP-OVERLAP (W, I)			15
information retrieval activity			15
special overlap event			16
a _i			22
b,			22
C,			. 22
Γ _k			22
L _k			22
PROJ(W, i)			23
PSTRING			24
SEED	₿.		24
S			24
N			27
m			27
m			27
N			27
i-information	2		28
S			28

I	30
stage i	30
NUM (I)	30
y, ·	31
zi	31
Si	31
≡,	34
E ^x i	34

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