

Polynomial Time Inference of Extended Regular Pattern Languages

Takeshi Shinohara

Computer Center, Kyushu University 91,
Fukuoka, 812 Japan

ABSTRACT

A pattern is a string of constant symbols and variable symbols. The language of a pattern p is the set of all strings obtained by substituting any non-empty constant string for each variable symbol in p . A regular pattern has at most one occurrence of each variable symbol. In this paper, we consider polynomial time inference from positive data for the class of extended regular pattern languages which are sets of all strings obtained by substituting any (possibly empty) constant string, instead of non-empty string. Our inference machine uses MINL calculation which finds a minimal language containing a given finite set of strings. The relation between MINL calculation for the class of extended regular pattern languages and the longest common subsequence problem is also discussed.

1. Introduction

Two kinds of inferences have been known, deductive inference and inductive inference. Many studies on deductive inference cover a wide range from theoretical problems to practical problems. Although some theories of inductive inference have been developed, few of them have reached practical applications to computer softwares. This study presents an approach to practical applications of inductive inference and gives its theoretical basis.

The direct motivation of this research is to develop a data entry system with learning function, proposed by Arikawa[4]. The following is a brief description of the learning system.

At the stage of data entry from a keyboard, a user types, for example, some bibliographic data:

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$
Author:  Angluin, D.
Title:   Inductive Inference of Formal Languages from Positive
         Data
Journal: Inform. Contr. 45
Year:    1980
$
Author:  Maier, D.
Title:   The Complexity of Some Problems on Subsequences and
         Supersequences
Journal: JACM 25
Year:    1978
$

```

where \$ is a special symbol to mark off records. Halfway through this entry process, the system will learn or infer a structure of records in the form

```
Author:  <w>Title:  <x>Journal:  <y>Year:  <z>
```

where each of <w>, <x>, <y>, and <z> stands for any string, and it will successively emit the constant parts "Author: ", "Title: ", "Journal: ", and "Year: " as prompts. Then the user may only type the data corresponding to <w>, <x>, <y>, and <z>.

The information, the system can use, is only the input data. Hence we should consider inference from positive data. Another important problem is computational complexity. The system should respond quickly. For these reasons we consider a polynomial time inference from positive data.

The inference from positive data has been considered of little interest to study, since Gold[4] proved that any class of languages over an alphabet is not inferrable from positive data if it contains all finite languages and at least one infinite languages, and hence that even the class of regular sets is not inferrable from positive data.

Recently Angluin[2,3] gave new life to the study of inference by characterizing the class of languages inferrable from positive data and presenting interesting classes. The class of pattern languages is one of her classes.

A pattern is a string of constant symbols and variable symbols, and the language of a pattern p is the set of all strings obtained by substituting any non-empty constant string for each variable

by substituting any non-empty constant string for each variable symbol in p . A regular pattern is a pattern in which each variable symbol occurs at most once. Note that the structure obtained by the learning data entry system is a regular pattern. The class of regular pattern languages has been shown to be inferrable from positive data [Shinohara, 9].

In this paper we first point out some problems of our previous version of inference method and then we give a solution to them by considering polynomial time inferrability of extended regular pattern languages. Our extension is to allow the substitutions to erase some variable symbols. More precisely, an extended language of a pattern p is a set of all strings obtained by substituting any (possibly empty) constant string, instead of non-empty string, for each variable symbol in p . For example, the extended language of a pattern "0x1" can contain string "01" while the language by Angluin can not. The erasing variables requires a new discussion.

The inference, we deal with here, is carried out by using MINL calculation introduced by Angluin[2]. Hence our main attention is paid to the time complexity of MINL calculation for the class of extended regular pattern languages. MINL for extended regular pattern languages finds a regular pattern which represents a minimal extended regular pattern language containing a given non-empty finite set of strings. We also refer to the relation between MINL calculation for extended regular pattern languages and the longest common subsequence problem. We propose an algorithm which calculates MINL for extended regular pattern languages in polynomial time. By using this fact, the class of extended regular pattern languages is shown to be polynomial time inferrable from positive data.

2. Preliminaries

We begin with a brief review of previous results.

2.1. Patterns and Their Languages

Let Σ be a finite set of symbols containing at least two symbols and let $X = \{x_1, x_2, \dots\}$ be a countable set of symbols disjoint from Σ . Elements in Σ are called constants and elements in X are called variables. A pattern is any string over $\Sigma \cup X$. The set $(\Sigma \cup X)^*$ of all patterns is denoted by P .

We say that a pattern p is regular if each variable in p occurs exactly once in p .

Let f be a non-erasing homomorphism from P to P . If $f(a) = a$ for any constant a , then f is called a substitution. If f is a substitution, $f(x)$ is in X , and $f(x) = f(y)$ implies $x = y$ for any variables x and y , then f is called a renaming of variables. We use a notation $[a_1/v_1, \dots, a_k/v_k]$ for the substitution which maps each variable symbol v_i to a_i and every other symbol to itself. We define two binary relations on P as follows:

- 1) $p \equiv' q$ iff $p = f(q)$ for some renaming of variables f ,
- 2) $p \leq' q$ iff $p = f(q)$ for some substitution f .

The language of a pattern p , denoted by $L(p)$, is the set $\{w \in \Sigma^* \mid w \leq' p\}$. These syntactic relations \equiv' and \leq' are characterized by the following lemma.

Lemma 1. [2]

- 1) For all patterns p and q , $p \equiv' q$ iff $L(p) = L(q)$.
- 2) For all patterns p and q , if $p \leq' q$ then $L(p) \subseteq L(q)$, but the converse is not true in general.
- 3) If p and q are patterns such that $|p| = |q|$, then $p \leq' q$ iff $L(p) \subseteq L(q)$.

2.2 Polynomial Time Inference from Positive Data

Inference machine is an effective procedure which requires inputs from time to time and produces outputs from time to time. Let $s = s_1, s_2, \dots$ be an arbitrary infinite sequence, and let $g = g_1, g_2, \dots$ be a sequence of outputs produced by an inference machine M when inputs in s are successively given to M on request. Then we say that M on input s converges to g_0 iff g is a finite sequence ending with g_0 or all but finitely many elements of g are equal to g_0 .

Let $L = L_1, L_2, \dots$ be a class of recursive languages, and let $s = s_1, s_2, \dots$ be an arbitrary enumeration of some language L_1 . Then we say that a machine M infers L from positive data if M on input s converges to an index j with $L_j = L_1$. We say that a class L is inferred from positive data if there exists a machine which infers L from positive data.

Theorem 1. [2] If a class $L = L_1, L_2, \dots$ satisfies the following condition, then L is inferred from positive data.

Condition: For any non-empty finite set S of strings, the set $\{L \mid S \subseteq L, L = L_i \text{ for some index } i\}$ has finite cardinality.

Lemma 2. [2] The class of pattern languages satisfies Condition of Theorem 1.

Hereafter we omit the phrase "from positive data", hence for example "inference" means "inference from positive data."

An inference by a machine M is consistent iff a language L_{g_i} contains all inputs given so far whenever M produces output g_i . An inference is conservative iff an output g_i from M is never changed unless L_{g_i} fails to contain some of the inputs. These two properties are natural and valuable in inference problem. It is, however, known that inferrability does not always mean consistency and conservativeness [2].

A class L is polynomial time inferrable iff there exists an inference machine M which infers L consistently and conservatively, and requests a new input in polynomial time (with respect to the length of the inputs read so far) after the last input has been received.

MINL calculation for a class $L = L_1, L_2, \dots$ is defined by Angluin[2] as follows:

$\text{MINL}(S) =$ "Given non-empty finite set S of strings, find an index i such that $S \subseteq L_i$ and for no index j , $S \subseteq L_j \not\subseteq L_i$."

The following theorem shows the importance of MINL calculation.

Theorem 2. [2] If a class $L = L_1, L_2, \dots$ satisfies Condition of Theorem 1 and MINL for L is computable, then the procedure Q below infers L consistently and conservatively.

```

procedure Q;
  begin
     $g_1 := \text{"none"} ; S := \emptyset ;$ 
    for each input  $s_i$  do
      begin
         $S := S \cup \{s_i\} ;$ 
        if  $s_i \in L_{g_i}$  then
           $g_{i+1} := g_i$ 
        else
          begin
             $g_{i+1} := \text{MINL}(S) ;$ 
            output  $g_{i+1}$ 
          end
        end
      end
    end
  end

```

Corollary 1. If a class $L = L_1, L_2, \dots$ satisfies Condition of Theorem 1, and the membership decision and MINL calculation for L are computable in polynomial time, then the class L is polynomial time inferrable from positive data.

Angluin [2] showed that the membership decision of pattern languages is NP-complete and ℓ -MINL calculation, a special case of MINL for pattern languages, is NP-hard.

ℓ -MINL(S) = "Given non-empty finite set S of strings, find a pattern of maximum possible length which represents a minimal pattern language containing S ."

The following summarize the results of our previous study [9].

Lemma 3. For any regular pattern p and any string w , whether $w \in L(p)$ or not is decidable in $O(|p| + |w|)$ time.

Theorem 3. The following procedure computes ℓ -MINL(S) for regular pattern languages in $O(m^2n)$ time, where $m = \max\{|w|; w \in S\}$, $n = \text{card}(S)$, and $w = a_1 \dots a_k$ ($a_i \in \Sigma$) is one of the shortest strings in S .

```

begin
   $p_1 := x_1 \dots x_k$  ;
  for  $i := 1$  to  $k$  do
    begin
       $q := p_i[a_i/x_i]$  ;
      if  $S \subseteq L(q)$  then  $p_{i+1} := q$ 
      else  $p_{i+1} := p_i$ 
    end ;
  return  $p_{k+1}$ 
end
```

Theorem 4. The class of regular pattern languages is polynomial time inferrable from positive data.

3. Some Problems on ℓ -MINL Calculation for Regular Pattern Languages

There are some difficulties in the ℓ -MINL calculation when the polynomial time inference of regular pattern languages is applied to practical use. The main reasons are

- 1) restriction on the length of pattern, and
- 2) prohibition against substituting empty string for any variable.

We present some examples to explain these problems.

Example 1. Let S be the set $\{ABCdeFGh, ABCiFGjk\}$. Then every answer of ℓ -MINL is eight symbols long because the length of the shortest words in S is eight. Let p be any pattern of the form p_1FGp_2 , where p_1 and p_2 are any regular patterns. Assume $S \subseteq L(p)$. Then, clearly, $ABCi \in L(p_1)$ and $h \in L(p_2)$, therefore

$$|p| = |p_1| + |FG| + |p_2| \leq 4 + 2 + 1 = 7 < 8.$$

Hence the string "FG" does not appear in any answer of ℓ -MINL(S). However the pattern $q = ABCx_1FGx_2$ is a possible answer of MINL(S).

Thus MINL(S) may have an answer which contains more constant symbols than any answer of ℓ -MINL(S).

Example 2. Let $S = \{aBcdf, GHcdBiii\}$. Then both patterns

$$p_1 = x_1Bx_2x_3x_4 \text{ and}$$

$$p_2 = x_1x_2cdx_3$$

are correct answers of ℓ -MINL(S). Our ℓ -MINL algorithm of Theorem 2 returns p_1 for S . If we change the order of substitutions in the algorithm, we can get p_2 as the answer of ℓ -MINL(S).

Example 3. Let $S = \{ABC, AC\}$. Then MINL(S) does not have any answer containing both symbols A and C because we can not substitute empty string for any variable.

To solve these problems, we extend the definition of pattern languages to allow erasing substitutions.

4. Extension of Pattern Languages

We give new definitions on pattern languages to allow substitutions to erase variables and we show some of their properties. The definitions of patterns and regular patterns are the same ones as in Section 2.

A substitution is any (possibly erasing) homomorphism from P to P which maps each constant symbol to itself. A special substitution which maps each variable to empty string is denoted by c . For example, if $\Sigma = \{0, 1, 2\}$ and $X = \{x, y, \dots\}$, then $c(0x1y2) = 012$. We define two binary relations \leq' and \equiv' as follows:

- 1) $p \leq' q$ iff $p = f(q)$ for some substitution f ,
- 2) $p \equiv' q$ iff $p \leq' q$ and $q \leq' p$.

The language of a pattern p , denoted by $L(p)$, is the set $\{ w \in \Sigma^* \mid w \leq' p \}$. Hereafter we use the term "pattern languages" in the sense just defined above.

Proposition 1.

- 1) $p \leq' q \implies L(p) \subseteq L(q)$.
- 2) $p \equiv' q \implies L(p) = L(q)$.

We say that a pattern \hat{p} is in canonical form iff

$\hat{p} \equiv' q \implies |\hat{p}| \leq |q|$ for any pattern q , and \hat{p} contains exactly k variables x_1, x_2, \dots, x_k for some integer k and the leftmost occurrence of x_i is to the left of the leftmost occurrence of x_{i+1} for $i = 1, \dots, k-1$.

Note that the condition $|\hat{p}| \leq |q|$ in the definition above is necessary. For example, a pattern $0x_1x_21$ is not in canonical form because $0x_1x_21 \leq' 0x_11 \leq' 0x_1x_21$.

Theorem 5. There exists a unique canonical pattern \hat{p} equivalent (\equiv') to p for any regular pattern p .

Proof. Let $p = w_0v_1w_1v_2\dots w_{n-1}v_nw_n$ ($w_0, w_n \in \Sigma^*$, $w_i \in \Sigma^+$ ($i=1, \dots, n-1$), $v_j \in X^+$ ($j=1, \dots, n$)) be any regular pattern. Then $\hat{p} = w_0x_1w_1x_2\dots w_{n-1}x_nw_n$ is in canonical form and $\hat{p} \equiv' p$. Any regular pattern equivalent to p is in the form $w_0u_1w_1u_2\dots w_{n-1}u_nw_n$ ($u_i \in X^+$). Therefore the uniqueness of such canonical pattern is obvious.

Lemma 4. If p is a canonical regular pattern, then $|p| \leq 2|c(p)| + 1$.

Theorem 6. The class of extended regular pattern languages satisfies Condition of Theorem 1 and it is inferrable from positive data.

Proof. Let $S \subseteq \Sigma^*$ be any non-empty finite set of strings and let w be one of the shortest strings in S . Assume $S \subseteq L(p)$, where p is any canonical regular pattern. Then $|w| \geq |c(p)|$ because $w \in L(p)$. By Lemma 4, $|p| \leq 2|c(p)| + 1 \leq 2|w| + 1$. Therefore the number of such patterns p is finite.

Theorem 7. For any regular pattern p and any string w , whether $w \in L(p)$ or not is decided in $O(|p|+|w|)$ time.

Proof. We can construct a deterministic finite automaton

recognizing $L(p)$ in $O(|p|)$ time by using the method of pattern matching machines [Aho, et al., 1].

5. MINL Calculation for Regular Pattern Languages

To show polynomial time inferrability of regular pattern languages, we first discuss MINL calculation. We also refer to the relation between MINL calculation and the longest common subsequence (LCS for short) problem.

We start with the definitions on subsequences:

- 1) For any strings $w = a_1 \dots a_k$ ($a_i \in \Sigma$) and $s \in \Sigma^*$,
 $s \leq w$ (or $w \geq s$) iff $s = a_{i_1} \dots a_{i_m}$ ($1 \leq i_1 < \dots < i_m \leq k$).

We say that s is a subsequence of w (or w is a supersequence of s) if $s \leq w$ (or $w \geq s$).

- 2) The set of common subsequences of a set S of strings is
 $CS(S) = \{ s \in \Sigma^* \mid s \leq w \text{ for any string } w \in S \}.$

- 3) The set of maximal common subsequences of S is
 $MCS(S) = \{ s \in CS(S) \mid s = s' \text{ or } s \not\leq s' \text{ for any } s' \in CS(S) \}.$

- 4) The set of the longest common subsequences of S is
 $LCS(S) = \{ s \in CS(S) \mid |s| \geq |s'| \text{ for any } s' \in CS(S) \}.$

Proposition 2.

- 1) $w \in L(p) \implies w \geq c(p)$
- 2) $L(p) \subseteq L(q) \implies c(p) \geq c(q)$
- 3) $L(p) = L(q) \implies c(p) = c(q)$
- 4) $S \subseteq L(p) \implies c(p) \in CS(S)$

We need three notations in the discussions below:

- 1) For any string $w = a_1 \dots a_n$ and any integers i and j ,
 $w\langle i:j \rangle = \begin{cases} a_i \dots a_j & (\text{if } 1 \leq i \leq j \leq |w|) \\ \epsilon & (\text{otherwise}), \text{ and} \end{cases}$
 $w\langle i \rangle = a_i.$

- 2) For any symbol a and any integer i ,
 $a^i = \begin{cases} \epsilon & (\text{if } i=0) \\ aa^{i-1} & (\text{otherwise}). \end{cases}$

- 3) For any variables $v_1, \dots, v_k \in X$ and any constant strings $w_1, \dots, w_k \in \Sigma^*$, $[w_1/v_1, \dots, w_k/v_k]$ denotes the substitution which maps each variable v_i to w_i and every other variable to itself.

Theorem 8. Let p and q be any regular patterns and $\text{card}(\Sigma) \geq 3$. Then $L(p) \subseteq L(q)$ implies $p \leq' q$.

Proof. We may assume, without loss of generality, that p and q are canonical regular patterns. We also assume $\text{card}(\Sigma) \geq 3$, $L(p) \subseteq L(q)$, but $p \not\leq' q$. Let $q = w_0 x_1 w_1 \dots w_{n-1} x_n w_n$, where $w_0, w_n \in \Sigma^*$, and $w_i \in \Sigma^+$ ($i=1, \dots, n-1$). Since $c(p)$ is a supersequence of $c(q)$ and $p \not\leq' q$, there exist integers i, j , and k such that

$0 \leq i \leq n$, $1 \leq j < k \leq |p|$, and
 $p = p\langle 1:j \rangle p\langle j+1:k-1 \rangle p\langle k:|p| \rangle$, where
 $c(p\langle 1:j \rangle) \in L(w_0 x_1 \dots x_{i-1} w_{i-1})$,
 $c(p\langle 1:j' \rangle) \notin L(w_0 x_1 \dots x_{i-1} w_{i-1})$ for any integer $j' < j$,
 $p\langle j+1:k-1 \rangle = r w_i r'$ for any patterns r and r' ,
 $c(p\langle k:|p| \rangle) \in L(w_{i+1} x_{i+2} \dots x_n w_n)$, and
 $c(p\langle k':|p| \rangle) \notin L(w_{i+1} x_{i+2} \dots x_n w_n)$ for any integer $k' > k$.

Let $p_1 = p\langle 1:j \rangle$, $p_2 = p\langle j+1:k-1 \rangle$, and $p_3 = p\langle k:|p| \rangle$.

Then $L(p_2) \subseteq L(x_i w_i x_{i+1})$ because $L(p) \subseteq L(q)$ and $c(p_1) p_2 c(p_3) \leq' p$. Let a be any constant symbol except $w_i\langle 1 \rangle$ and $w_i\langle |w_i| \rangle$ and let v_1, \dots, v_m be all variables in p_2 . Then $p_2[a^{|w_i|}/v_1, \dots, a^{|w_i|}/v_m] \in L(p_2) - L(x_i w_i x_{i+1})$. This contradicts $L(p_2) \subseteq L(x_i w_i x_{i+1})$.

The following lemma says that the condition $\text{card}(\Sigma) \geq 3$ is necessary in Theorem 8.

Lemma 5. When $\text{card}(\Sigma) = 2$, there exist regular patterns p and q such that $L(p) \subseteq L(q)$, $p \not\leq' q$, and $q \not\leq' p$.

Proof. Let $\Sigma = \{0, 1\}$, $p = x_1 0 1 x_2 0 x_3$, and $q = x_1 0 x_2 1 0 x_3$. Then, clearly, $p \not\leq' q$, $q \not\leq' p$, but $L(p) = L(q)$.

Hereafter we assume that the constants alphabet Σ contains at least three symbols.

Theorem 9. For any maximal common subsequence $s \in \text{MCS}(S)$, there exists an answer p of $\text{MINL}(S)$ for regular pattern languages such that $c(p) = s$.

Proof. Let $s = a_1 \dots a_k \in \text{MCS}(S)$. Then the pattern q_{k+1} defined as follows is an answer of $\text{MINL}(S)$:

$$q_i := \begin{cases} x_1 a_1 \dots a_k x_{k+1} & (i=0) \\ \text{if } S \subseteq L(q_{i-1}[\varepsilon/x_i]) \text{ then } q_{i-1}[\varepsilon/x_i] \text{ else } q_{i-1} & (i=1, \dots, k+1). \end{cases}$$

We must show that $L(q_{k+1})$ is a minimal regular pattern language containing S . Assume that there exists a regular pattern q' such that $S \subseteq L(q') \subsetneq L(q_{k+1})$. Then $c(q') \geq c(q_{k+1}) = s$. Since s is a maximal common subsequence of S , $c(q') = c(q_{k+1}) = s$. By Theorem 8, $q' \leq q_{k+1}$ and $q' \neq q_{k+1}$. There exists a substitution f which maps q_{k+1} to q' . The substitution f maps at least one variables to empty string because $q' \neq q_{k+1}$. Let j be an integer such that x_j appears in q_{k+1} , $f(x_j) = \varepsilon$, and $f(x_{j'}) = x_{j'}$, for any integer $j' < j$. Then $q' \leq q_{j-1}[\varepsilon/x_j]$. Therefore $S \subseteq L(q_{j-1}[\varepsilon/x_j])$ and $q_j = q_{j-1}[\varepsilon/x_j]$. Hence the variable x_j can not appear in q_{k+1} . This contradicts the selection of j .

Here we should note that we can get an answer of $\text{MINL}(S)$ in $O(m^2n)$ time from any maximal common subsequence of a set S of strings, where $m = \max\{|w|; w \in S\}$ and $n = \text{card}(S)$.

We may prefer the longest common subsequences to the maximal common subsequence. However the problem to find one of the longest common subsequences of a set of strings is known to be NP-complete [Maier, 8]. Therefore finding an answer of $\text{MINL}(S)$, containing maximum number of constants, does not seem to be done in polynomial time. To find an answer of $\text{MINL}(S)$ for regular pattern languages, is it necessary to select one of the maximal common subsequences of S ? The following theorem asserts that it is not the case.

Theorem 10. There exists an answer p of $\text{MINL}(S)$ for regular pattern languages such that $c(p) \neq \text{MCS}(S)$ for some set S of strings.

Proof. Let $S = \{01020, 0212\}$. Then $02 \notin \text{MCS}(S)$ because $012 \in \text{CS}(S)$. However the pattern $p = x_102x_2$ represents a minimal regular pattern language containing S .

In the proof of Theorem 10, the pattern $q = 0x_11x_22x_3$ is a possible answer of $\text{MINL}(S)$, and $c(q) = 012 \in \text{LCS}(S)$. In some cases q is not always better answer of $\text{MINL}(S)$ than p because the pattern p contains a longer constant string "02" than q . Finally, from this observation, we get a MINL algorithm by using a method to find common strings in length decreasing order. The correctness is easily shown by Theorem 8 and the computing time is $O(m^4n)$, where $m = \max\{|w|; w \in S\}$ and $n = \text{card}(S)$.

In our MINL algorithm we use some notations for simplicity:

Let $\sigma = w_1, \dots, w_n$ be a sequence of strings. The notation $L(\sigma)$ denotes the regular pattern language $L(x_1w_1x_2\dots w_nx_{n+1})$, $|\sigma|$ denotes

the number of strings in σ , and $\|\sigma\|$ denotes the sum of lengths of strings in σ .

Procedure MINL(S);

(* Input S: non-empty finite set of strings *)

(* Output p: a pattern representing a minimal

extended regular pattern language containing S *)

begin

s := one of the shortest strings in S ;

$\sigma := \epsilon$; (* sequence of common strings *)

n := |s| ; (* length of candidate string *)

while n > 0 do begin

for i := 1 to |s| - n + 1 do

more: for j := 0 to | σ | do

if S \subseteq L(σ <1:j>, s<i:i+n-1>, σ <j+1:| σ |>)

then begin

$\sigma := \sigma$ <1:j>, s<i:i+n-1>, σ <j+1:| σ |> ;

go to more

end ;

n := min(|s| - $\|\sigma\|$, n-1)

end ;

p := x_1 σ <1> $x_2 \dots \sigma$ <| σ |> $x_{|\sigma|+1}$;

if S \subseteq L(p[ϵ/x_1]) then p := p[ϵ/x_1] ;

if S \subseteq L(p[$\epsilon/x_{|\sigma|+1}$]) then p := p[$\epsilon/x_{|\sigma|+1}$] ;

return p ;

end

Theorem 11. The class of extended regular pattern languages is polynomial time inferrable from positive data.

6. Concluding Remarks

We have discussed polynomial time inference for the class of the extended regular pattern languages and we have seen that MINL calculation for the class plays an important role in inference from positive data. We have also seen that finding a longest common subsequence is a special case of the MINL calculation for extended regular pattern languages. We believe that our discussion gives a new viewpoint of the LCS problem.

It should be noticed that our MINL algorithm for the extended regular pattern languages is consistent to the NP-completeness of the LCS problem. The MINL algorithm finds common strings to a set in

length decreasing order. It should also be noticed that our method in the algorithm is natural.

Since our evaluation of the time complexity is not so acute, the exponent of the maximum length of strings might be reduced. The MINL algorithm is originally designed for the learning data entry system, and it should have other practical applications. A little modification may be needed for some problems.

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