

A FULLY COMPRESSED PATTERN MATCHING ALGORITHM FOR SIMPLE COLLAGE SYSTEMS

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ABSTRACT

We study the *fully compressed pattern matching problem (FCPM problem)*: Given \mathcal{T} and \mathcal{P} which are descriptions of text T and pattern P respectively, find the occurrences of P in T without decompressing \mathcal{T} or \mathcal{P} . This problem is rather challenging since patterns are also given in a compressed form. In this paper we present an FCPM algorithm for *simple collage systems*. Collage systems are a general framework representing various kinds of dictionary-based compressions in a uniform way, and simple collage systems are a subclass that includes LZW and LZ78 compressions. Collage systems are of the form $(\mathcal{D}, \mathcal{S})$, where \mathcal{D} is a dictionary and \mathcal{S} is a sequence of variables from \mathcal{D} . Our FCPM algorithm performs in $O(\|\mathcal{D}\|^2 + mn \log |\mathcal{S}|)$ time, where $n = |\mathcal{T}| = \|\mathcal{D}\| + |\mathcal{S}|$ and $m = |\mathcal{P}|$. This is faster than the previous best result of $O(m^2n^2)$ time.

Keywords: string processing, text compression, fully compressed pattern matching, collage systems, algorithm

1. Introduction

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The *compressed pattern matching problem* (CPM problem) [1] is a challenging problem in Stringology such that, given compressed text \mathcal{T} and uncompressed pattern P , find the pattern occurrences *without decompressing \mathcal{T}* . This problem has been intensively studied for a variety of text compression schemes, e.g. [2, 4, 3, 17].

An ultimate extension of the CPM problem is the *fully compressed pattern matching problem* (FCPM problem) [10] where *both* text T and pattern P are given in compressed forms \mathcal{T} and \mathcal{P} respectively, and the objective is to find all occurrences of P in T *without decompressing \mathcal{T} or \mathcal{P}* . Miyazaki et al. [18] presented an algorithm to solve the FCPM problem for *straight line programs*, in $O(m^2n^2)$ time using $O(mn)$ space, where $m = |\mathcal{P}|$ and $n = |\mathcal{T}|$. For LZW compressed text \mathcal{T} and pattern \mathcal{P} , Gąsieniec and Rytter [7] addressed a pattern matching algorithm running in $O((m+n)\log(m+n))$ time but this one *explicitly decompresses* part of \mathcal{T} or \mathcal{P} when the decompressed size does not exceed n . Hence their algorithm does not really solve the FCPM problem where pattern matching *without* any decompressing is strictly required. Therefore, the best known result for the FCPM problem on LZW is $O(m^2n^2)$ time and $O(mn)$ space by Miyazaki et al. [18].

In this paper, we consider the FCPM problem on *simple collage systems* which are a subclass of *collage systems* [11]. Collage systems are a general framework that represents various compression schemes such as LZ family [22, 20, 23, 21], run-length encoding, BPE [5], RE-PAIR [15], SEQUITUR [19], grammar transform [12, 14, 13], and straight line programs. A collage system is a pair $\langle \mathcal{D}, \mathcal{S} \rangle$ where \mathcal{D} is a dictionary and \mathcal{S} is a sequence of variables from \mathcal{D} . Simple collage systems [16] are a subclass of collage systems including LZ78 [23] and LZW [21]. Simple collage systems are very attractive in terms of accelerating CPM [16] despite of their generally weaker compression ratio.

In this paper, we present an efficient FCPM algorithm for simple collage systems, which runs in $O(|\mathcal{D}|^2 + mn \log |\mathcal{S}|)$ time with $O(|\mathcal{D}|^2 + mn)$ space, where $|\mathcal{D}|$ denotes the size of the dictionary \mathcal{D} , and $|\mathcal{S}|$ the length of the sequence \mathcal{S} . A preliminary version of this work has appeared in [8]. Although our algorithm requires more space than the algorithm by Miyazaki et al. [18], it consumes less time. In addition, since simple collage systems are a general framework, for our algorithm a text and a pattern may be compressed by different compression schemes. Namely, our algorithm is so flexible that it can deal with an LZ78-compressed text and an LZW-compressed pattern, and vice versa. Since it is natural to assume that a text and a pattern can be chosen from different sources, this feature can be a practical advantage of our algorithm.

2. Preliminary

Let \mathcal{N} be the set of natural numbers, and \mathcal{N}^+ be the set of positive integers. Let Σ be a finite *alphabet*. An element of Σ^* is called a *string*. The length of a string T is denoted by $|T|$. The i -th character of a string T is denoted by $T[i]$ for $1 \leq i \leq |T|$, and the substring of a string T that begins at position i and ends at position j is denoted by $T[i : j]$ for $1 \leq i \leq j \leq |T|$. A *period* of a string T is an integer p ($1 \leq p \leq |T|$) such that $T[i] = T[i + p]$ for any $i = 1, 2, \dots, |T| - p$.

Collage systems [11] are a general framework that enables us to capture the structure of different types of dictionary-based compressions. *Regular collage systems*, which are a subclass of collage systems, are pair $\langle \mathcal{D}, \mathcal{S} \rangle$ such that \mathcal{D} is a sequence of assignments

$$X_1 = \text{expr}_1, X_2 = \text{expr}_2, \dots, X_h = \text{expr}_h,$$

where X_k are variables and expr_k are expressions of either of the form

$$\begin{array}{ll} a & \text{where } a \in (\Sigma \cup \varepsilon), \quad (\text{primitive assignment}) \\ X_i X_j & \text{where } i, j < k, \quad (\text{concatenation}) \end{array}$$

and \mathcal{S} is a sequence of variables $X_{i_1}, X_{i_2}, \dots, X_{i_s}$ obtained from \mathcal{D} . The size of \mathcal{D} is h and is denoted by $\|\mathcal{D}\|$, and the size of \mathcal{S} is s and is denoted by $|\mathcal{S}|$. The total size of the collage system $\langle \mathcal{D}, \mathcal{S} \rangle$ is $n = \|\mathcal{D}\| + |\mathcal{S}| = h + s$.

A regular collage system is said to be *simple* if, for any variable $X = X_\ell X_r$, either $|X_\ell| = 1$ or $|X_r| = 1$ [16]. LZW [21] and LZ78 [23] are simple collage systems formalized as follows.

LZW. $\mathcal{S} = X_{i_1}, X_{i_2}, \dots, X_{i_s}$ and \mathcal{D} is the following:

$$\begin{array}{l} X_1 = a_1; X_2 = a_2; \dots; X_q = a_q; \\ X_{q+1} = X_{i_1} X_{\sigma(i_2)}; X_{q+2} = X_{i_2} X_{\sigma(i_3)}; \dots; X_{q+s-1} = X_{i_{s-1}} X_{\sigma(i_s)}, \end{array}$$

where the alphabet is $\Sigma = \{a_1, a_2, \dots, a_q\}$, $1 \leq i_1 \leq q$, and $\sigma(j)$ denotes the integer k ($1 \leq k \leq q$) such that a_k is the first symbol of X_j .

LZ78. $\mathcal{S} = X_1, X_2, \dots, X_s$ and \mathcal{D} is the following:

$$X_0 = \varepsilon; X_1 = X_{i_1} b_1; X_2 = X_{i_2} b_2; \dots; X_s = X_{i_s} b_s;$$

where b_j is a symbol in Σ .

In this paper, we study the *fully compressed pattern matching problem for simple collage systems*: Given two simple collage systems that are the descriptions of text T and pattern P , find all occurrences of P in T . Namely, we compute the following set:

$$\text{Occ}(T, P) = \{i \mid T[i : i + |P| - 1] = P\}.$$

We emphasize that our goal is to solve this problem *without decompressing either of the two simple collage systems*. Our result is the following:

Theorem 1 *Given two simple collage systems $\langle \mathcal{D}, \mathcal{S} \rangle$ and $\langle \mathcal{D}', \mathcal{S}' \rangle$ that are the description of T and P respectively, $\text{Occ}(T, P)$ can be computed in $O(\|\mathcal{D}\|^2 + mn \log |\mathcal{S}|)$ time using $O(\|\mathcal{D}\|^2 + mn)$ space, where $n = \|\mathcal{D}\| + |\mathcal{S}|$ and $m = \|\mathcal{D}'\| + |\mathcal{S}'|$.*

3. Overview of algorithm

3.1. Translation to straight line programs

Consider a regular collage system $\langle \mathcal{D}, \mathcal{S} \rangle$. Note that $\mathcal{S} = X_{i_1}, X_{i_2}, \dots, X_{i_s}$ can be translated in linear time to a sequence of assignments of size s . For instance, $\mathcal{S} = X_1, X_2, X_3, X_4$ can be rewritten to $X_5 = X_1 X_2$; $X_6 = X_5 X_3$; $X_7 = X_6 X_4$, and $S = X_7$. Therefore, a regular collage system, which represents string $T \in \Sigma^*$, can be seen as a context free grammar of the Chomsky normal form that generates only T , and thus correspond to *straight line programs (SLPs)*. In the sequel, for string $T \in \Sigma^*$, let \mathcal{T} denote the SLP representing T . The size of \mathcal{T} is denoted by $\|\mathcal{T}\|$, and $\|\mathcal{T}\| = \|\mathcal{D}\| + |\mathcal{S}| = h + s = n$.

Now we introduce *simple* straight line programs (SSLP) that correspond to simple collage systems.

Definition 1 An SSLP \mathcal{T} is a sequence of assignments such that

$$X_1 = \text{expr}_1; X_2 = \text{expr}_2; \dots; X_n = \text{expr}_n,$$

where X_i are variables and expr_i are expressions of any of the form

$$\begin{array}{lll} a & \text{where } a \in \Sigma & \text{(primitive),} \\ X_\ell X' & \text{where } \ell < i \text{ and } X' = a & \text{(right simple),} \\ X' X_r & \text{where } r < i \text{ and } X' = a & \text{(left simple),} \\ X_\ell X_r & \text{where } \ell, r < i & \text{(complex),} \end{array}$$

and $\mathcal{T} = X_n$. Moreover, each type of variable satisfies the following properties:

- For any right simple variable $X_i = X_\ell X'$, X_ℓ is either simple or primitive.
- For any left simple variable $X_i = X' X_r$, X_r is either simple or primitive.
- For any complex variable $X_i = X_\ell X_r$, X_r is either simple or primitive.

An example of an SSLP \mathcal{T} for string $T = \text{abaabababb}$ is as follows:

$$\begin{aligned} X_1 = \mathbf{a}, X_2 = \mathbf{b}, X_3 = X_1 X_2, X_4 = X_1 X_3, X_5 = X_3 X_1, X_6 = X_2 X_2, \\ X_7 = X_3 X_4, X_8 = X_7 X_5, X_9 = X_8 X_6, \end{aligned}$$

and $\mathcal{T} = X_9$. See also Fig. 1 that illustrates the derivation tree of \mathcal{T} .

X_1 and X_2 are primitive variables, X_3, X_4, X_5 and X_6 are simple variables, and X_7, X_8 and X_9 are complex variables.

For any simple collage system $\langle \mathcal{D}, \mathcal{S} \rangle$, let \mathcal{T} be its corresponding SSLP. Let $\|\mathcal{D}\| = h$ and $|\mathcal{S}| = s$. Then the total number of primitive and simple variables in \mathcal{T} is h , and the number of complex variables in \mathcal{T} is s .

In the sequel, we consider computing $\text{Occ}(T, P)$ for given SSLPs \mathcal{T} and \mathcal{P} . We use X and X_i for variables of \mathcal{T} , and Y and Y_j for variables of \mathcal{P} . When not confusing, X_i (Y_j , respectively) also denotes the string derived from X_i (Y_j , respectively). Let $\|\mathcal{T}\| = n$ and $\|\mathcal{P}\| = m$.

Proposition 1 For any simple variable X , $|X| = \|X\|$, where $\|X\|$ denotes the number of variables in X .

3.2. Basic idea of algorithm

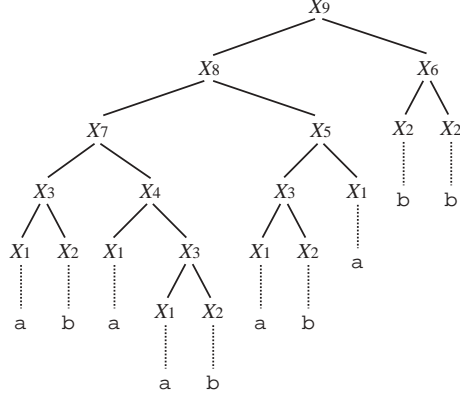


Fig. 1. Derivation tree of SSLP for string abaabababb.

In this section, we show a basis of our algorithm that outputs a compact representation of $Occ(T, P)$ for given SSLPs \mathcal{T}, \mathcal{P} .

For strings $X, Y \in \Sigma^*$ and integer $k \in \mathcal{N}$, we define the set of all occurrences of Y that cover or touch the position k in X by

$$Occ^\uparrow(X, Y, k) = \{i \in Occ(X, Y) \mid k - |Y| \leq i \leq k\}.$$

In the following, $[i, j]$ denotes the set $\{i, i + 1, \dots, j\}$ of consecutive integers. For a set U of integers and an integer k , we denote $U \oplus k = \{i + k \mid i \in U\}$ and $U \ominus k = \{i - k \mid i \in U\}$.

Observation 1 For any strings $X, Y \in \Sigma^*$ and integer $k \in \mathcal{N}$,

$$Occ^\uparrow(X, Y, k) = Occ(X, Y) \cap [k - |Y|, k].$$

Lemma 1 For any strings $X, Y \in \Sigma^*$ and integer $k \in \mathcal{N}$, $Occ^\uparrow(X, Y, k)$ forms a single arithmetic progression.

For positive integers $p, d \in \mathcal{N}^+$ and non-negative integer $t \in \mathcal{N}$, we define $\langle p, d, t \rangle = \{p + (i - 1)d \mid i \in [1, t]\}$. Note that t denotes the cardinality of the set $\langle p, d, t \rangle$. By Lemma 1, $Occ^\uparrow(X, Y, k)$ can be represented as the triple $\langle p, d, t \rangle$ with the minimum element p , the common difference d , and the length t of the progression. By ‘computing $Occ^\uparrow(X, Y, k)$ ’, we mean to calculate the triple $\langle p, d, t \rangle$ such that $\langle p, d, t \rangle = Occ^\uparrow(X, Y, k)$.

Observation 2 Assume each of sets A_1 and A_2 of integers forms a single arithmetic progression, and is represented by a triple $\langle p, d, t \rangle$. Then, the union $A_1 \cup A_2$ can be computed in constant time.

Lemma 2 ([9]) Let $\langle p, d, t \rangle = Occ^\uparrow(X, Y, k)$ for strings $X, Y \in \Sigma^*$ and integer $k \in \mathcal{N}$. If $t \geq 1$, then d is the shortest period of $X[p : q + |Y| - 1]$ where $q = p + (t - 1)d$.

Lemma 3 For any strings $X, Y_1, Y_2 \in \Sigma^*$ and integers $k_1, k_2 \in \mathcal{N}$, the intersection $Occ^\uparrow(X, Y_1, k_1) \cap (Occ^\uparrow(X, Y_2, k_2) \ominus |Y_1|)$ can be computed in $O(1)$ time, provided that $Occ^\uparrow(X, Y_1, k_1)$ and $Occ^\uparrow(X, Y_2, k_2)$ are already computed.

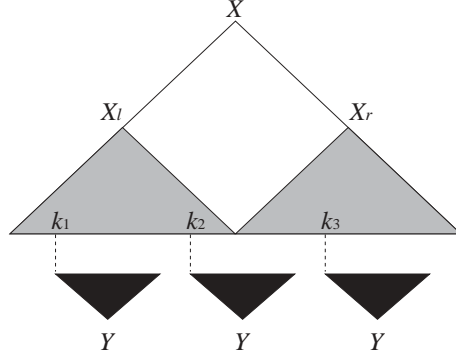


Fig. 2. $k_1, k_2, k_3 \in Occ(X, Y)$, where $k_1 \in Occ(X_\ell, Y)$, $k_2 \in Occ^\Delta(X, Y)$ and $k_3 \in Occ(X_r, Y)$.

For variables $X = X_\ell X_r$ and Y , we denote $Occ^\Delta(X, Y) = Occ^\uparrow(X, Y, |X_\ell| + 1)$. The following observation is explained in Fig. 2.

Observation 3 ([18]) For any variables $X = X_\ell X_r$ and Y ,

$$Occ(X, Y) = Occ(X_\ell, Y) \cup Occ^\Delta(X, Y) \cup (Occ(X_r, Y) \oplus |X_\ell|).$$

Observation 3 implies that $Occ(X_n, Y)$ can be represented by a combination of

$$\{Occ^\Delta(X_i, Y)\}_{i=1}^n = Occ^\Delta(X_1, Y), Occ^\Delta(X_2, Y), \dots, Occ^\Delta(X_n, Y).$$

Thus, the desired output $Occ(T, P) = Occ(X_n, Y_m)$ can be expressed as a combination of $\{Occ^\Delta(X_i, Y_m)\}_{i=1}^n$ that requires $O(n)$ space. Hereby, computing $Occ(T, P)$ is reduced to computing $Occ^\Delta(X_i, Y_m)$ for every $i = 1, 2, \dots, n$. In computing each $Occ^\Delta(X_i, Y_j)$ recursively, the same set $Occ^\Delta(X_{i'}, Y_{j'})$ might repeatedly be referred to, for $i' < i$ and $j' < j$. Therefore we take the dynamic programming strategy. We use an $m \times n$ table App where each entry $App[i, j]$ at row i and column j stores the triple for $Occ^\Delta(X_i, Y_j)$. We compute each $App[i, j]$ in a bottom-up manner, for $i = 1, \dots, n$ and $j = 1, \dots, m$. In the following sections, we will show that the whole table App can be computed in $O(h^2 + mn \log s)$ time using $O(h^2 + mn)$ space, where h is the number of simple variables in \mathcal{T} and s is the number of complex variables in \mathcal{T} . This leads to the result of Theorem 1.

4. Details of algorithm

In this section, we show how to compute each $Occ^\Delta(X_i, Y_j)$ efficiently. Our result is as follows:

Lemma 4 For any variables X_i of \mathcal{T} and Y_j of \mathcal{P} , $Occ^\Delta(X_i, Y_j)$ can be computed in $O(\log s)$ time, with extra $O(h^2 + mn)$ work time and space.

The key to prove this lemma is, given integer k , to pre-compute $Occ^\uparrow(X_{i'}, Y_{j'}, k)$ for any $1 \leq i' < i$ and $1 \leq j' < j$. In case that X is simple, we have the following lemma:

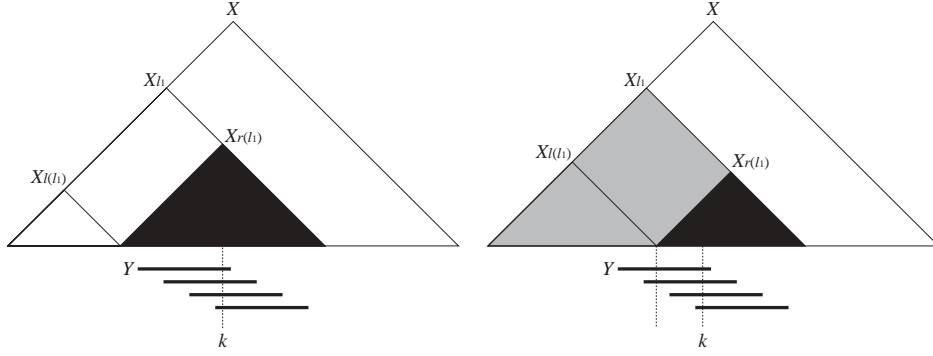


Fig. 3. In the left case, all the occurrences are covered by $Occ^\uparrow(X_{r(\ell_1)}, Y, k) \oplus |X_{\ell(\ell_1)}|$. In the right case, the first and second occurrences are covered by $Occ^\Delta(X_{\ell_1}, Y)$ and the third and fourth occurrences by $Occ^\uparrow(X_{r(\ell_1)}, Y, k) \oplus |X_{\ell(\ell_1)}|$.

Lemma 5 *Let X be any simple variable of \mathcal{T} and Y be any variable of \mathcal{P} . Given integer $k \in \mathcal{N}$, $Occ^\uparrow(X, Y, k)$ can be computed in $O(1)$ time, with extra $O(h^2 + mh)$ work time and space.*

As a counterpart to Lemma 5, we have the following lemma for X to be complex:

Lemma 6 *Let X be any complex variable of \mathcal{T} and Y be any variable of \mathcal{P} . Given integer $k \in \mathcal{N}$, $Occ^\uparrow(X, Y, k)$ can be computed in $O(\log s)$ time with extra $O(ms)$ work time and space.*

For any complex variable $X = X_\ell X_r$, let $range(X)$ denote the range $[r_1, r_2]$ such that $T[r_1, r_2] = X_r$. It is clear that for each complex variable its range is uniquely determined, since each complex variable appears in \mathcal{T} exactly once. In proving Lemma 6 above, Lemma 7 and Lemma 8 below are used.

Lemma 7 *Let $X = X_\ell X_r$ be any complex variable of \mathcal{T} and let Y be any variable of \mathcal{P} . Assume $Occ^\uparrow(X_\ell, Y, |X_\ell| - |Y| + 1)$ and $Occ^\Delta(X, Y)$ are already computed. Then $Occ^\uparrow(X, Y, |X| - |Y| + 1)$ can be computed in $O(1)$ time, with extra $O(ms)$ work space.*

Lemma 8 *Given integer $k \in \mathcal{N}$, we can retrieve in $O(\log s)$ time the complex variable X such that $range(X) = [r_1, r_2]$ and $r_1 \leq k \leq r_2$, after a preprocessing taking $O(s)$ time and space.*

Now the proof of Lemma 6 follows.

Proof. Let $A = Occ^\uparrow(X, Y, k)$. Let X_{ℓ_1} be the complex variable such that $k \in range(X_{\ell_1})$, and let $X_{\ell_1} = X_{\ell(\ell_1)} X_{r(\ell_1)}$. Let X_{ℓ_2} be the complex variable satisfying $k - |Y| \in range(X_{\ell_2})$, and let $X_{\ell_2} = X_{\ell(\ell_2)} X_{r(\ell_2)}$. There are the three following cases:

- (1) when $k - |Y| \geq |X_{\ell(\ell_1)}| + 1$ and $k + |Y| - 1 \leq |X_{\ell_1}|$ (Fig. 3, left).
In this case, we have $A = Occ^\uparrow(X_{r(\ell_1)}, Y, k) \oplus |X_{\ell(\ell_1)}|$.
- (2) when $k - |Y| < |X_{\ell(\ell_1)}| + 1$ and $k + |Y| - 1 \leq |X_{\ell_1}|$ (Fig. 3, right).

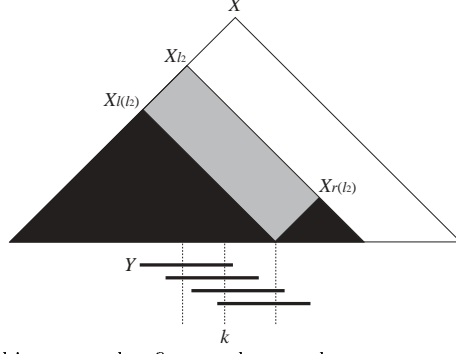


Fig. 4. In this case, the first and second occurrences are covered by $O_{cc}^\uparrow(X_{\ell(\ell_2)}, Y, |X_{\ell(\ell_2)}| - |Y| + 1)$ and the third and fourth occurrences are covered by $O_{cc}^\Delta(X_{\ell_2}, Y)$.

In this case, we have

$$A = (O_{cc}^\Delta(X_{\ell_1}, Y) \cap [k - |Y|, X_{\ell(\ell_1)} + 1]) \cup (O_{cc}^\uparrow(X_{r(\ell_1)}, Y, k) \oplus |X_{\ell(\ell_1)}|).$$

(3) when $k + |Y| - 1 > |X_{\ell_1}|$ (Fig. 4).

In this case, we have

$$A = (O_{cc}^\uparrow(X_{\ell(\ell_2)}, Y, |X_{\ell(\ell_2)}| - |Y| + 1) \cap [k - |Y|, |X_{\ell(\ell_2)}| - |Y| + 1]) \cup (O_{cc}^\Delta(X_{\ell_2}, Y) \cap [|X_{\ell(\ell_2)}| - |Y| + 1, k]).$$

Due to Lemma 8, X_{ℓ_1} and X_{ℓ_2} can be found in $O(\log s)$ time. Since $X_{r(\ell_1)}$ is simple, $O_{cc}^\uparrow(X_{r(\ell_1)}, Y, k)$ of cases (1) and (2) can be computed in $O(1)$ time by Lemma 5. According to Lemma 7, $O_{cc}^\uparrow(X_{\ell(\ell_2)}, Y, |X_{\ell(\ell_2)}| - |Y| + 1)$ of case (3) can be computed in $O(1)$ time. By Observation 2, the union operations can be done in $O(1)$ time. Thus, in any case $A = O_{cc}^\uparrow(X, Y, k)$ can be computed in $O(\log s)$ time. By Lemma 7 and Lemma 8, the extra work time and space are $O(ms)$. This completes the proof. \square

Now we have got Lemma 5 and Lemma 6 proved. Using these lemmas, we can prove Lemma 4 as follows:

Proof. Let $X_i = X_\ell X_r$ and $Y_j = Y_\ell Y_r$. Then, as seen in Fig. 5, we have

$$O_{cc}^\Delta(X_i, Y_j) = (O_{cc}^\Delta(X_i, Y_\ell) \cap (O_{cc}(X_r, Y_r) \oplus |X_\ell| \ominus |Y_\ell|)) \cup (O_{cc}(X_\ell, Y_\ell) \cap (O_{cc}^\Delta(X_i, Y_r) \ominus |Y_\ell|)).$$

Let $A = O_{cc}^\Delta(X_i, Y_\ell) \cap (O_{cc}(X_r, Y_r) \oplus |X_\ell| \ominus |Y_\ell|)$ and $B = O_{cc}(X_\ell, Y_\ell) \cap (O_{cc}^\Delta(X_i, Y_r) \ominus |Y_\ell|)$. Since $O_{cc}^\Delta(X_i, Y_j)$ forms a single arithmetic progression by Lemma 1, the union operation of $A \cup B$ can be done in constant time. Therefore, the key is how to compute A and B efficiently.

Now we show how to compute set A . Let $z = |X_\ell| - |Y_\ell|$. Let $\langle p_1, d_1, t_1 \rangle = O_{cc}^\Delta(X_i, Y_\ell)$ and $q_1 = p_1 + (t_1 - 1)d_1$. Depending on the value of t_1 , we have the following cases:

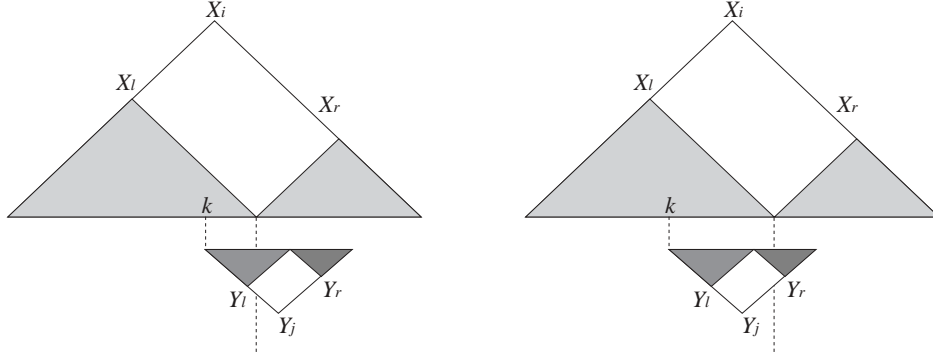


Fig. 5. $k \in \text{Occ}^\Delta(X, Y)$ if and only if either $k \in \text{Occ}^\Delta(X, Y_\ell)$ and $k + |Y_\ell| \in \text{Occ}(X, Y_r)$ (left case), or $k \in \text{Occ}(X, Y_\ell)$ and $k + |Y_\ell| \in \text{Occ}^\Delta(X, Y_r)$ (right case).

(1) when $t_1 = 0$.

In this case we have $A = \emptyset$.

(2) when $t_1 = 1$.

In this case, $\text{Occ}^\Delta(X_i, Y_\ell) = \{p_1\}$. It stands that

$$\begin{aligned}
A &= \{p_1\} \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= (\{p_1 - z\} \cap \text{Occ}(X_r, Y_r)) \oplus z \\
&= (\{p_1 - z\} \cap [p_1 - z - |Y_r|, p_1 - z] \cap \text{Occ}(X_r, Y_r)) \oplus z \\
&= (\{p_1 - z\} \cap \text{Occ}^\uparrow(X_r, Y_r, p_1 - z)) \oplus z \quad (\text{By Observation 1}) \\
&= \begin{cases} \{p_1\} & \text{if } p_1 - z \in \text{Occ}^\uparrow(X_r, Y_r, p_1 - z), \\ \emptyset & \text{otherwise.} \end{cases}
\end{aligned}$$

Since X_r is simple, $\text{Occ}^\uparrow(X_r, Y_r, p_1 - z)$ can be computed in constant time by Lemma 5. Checking whether $p_1 - z \in \text{Occ}^\uparrow(X_r, Y_r, p_1 - z)$ or not can be done in constant time since $\text{Occ}^\uparrow(X_r, Y_r, p_1 - z)$ forms a single arithmetic progression by Lemma 1.

(3) when $t_1 > 1$.

There are two sub-cases depending on the length of Y_r with respect to $q_1 - p_1 = (t_1 - 1)d_1 \geq d_1$, as follows.

- when $|Y_r| \geq q_1 - p_1$ (see the left of Fig. 6). By this assumption, we have $q_1 - |Y_r| \leq p_1$, which implies $[p_1, q_1] \subseteq [q_1 - |Y_r|, q_1]$. Thus

$$\begin{aligned}
A &= \langle p_1, d_1, t_1 \rangle \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= (\langle p_1, d_1, t_1 \rangle \cap [p_1, q_1]) \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= (\langle p_1, d_1, t_1 \rangle \cap [q_1 - |Y_r|, q_1]) \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= \langle p_1, d_1, t_1 \rangle \cap ([q_1 - |Y_r|, q_1] \cap (\text{Occ}(X_r, Y_r) \oplus z))
\end{aligned}$$

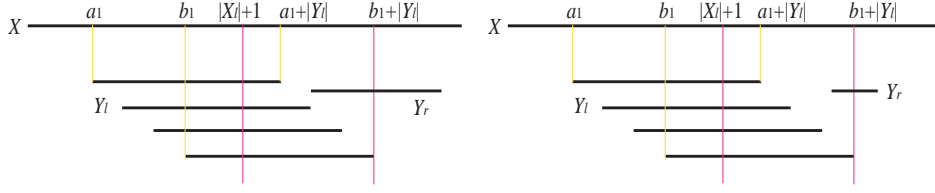


Fig. 6. Long case (left) and short case (right).

$$\begin{aligned}
&= \langle p_1, d_1, t_1 \rangle \cap (([q_1 - |Y_r| - z, q_1 - z] \cap \text{Occ}(X_r, Y_r)) \oplus z) \\
&= \langle p_1, d_1, t_1 \rangle \cap (\text{Occ}^\uparrow(X_r, Y_r, q_1 - z) \oplus z),
\end{aligned}$$

where the last equality is due to Observation 1. Since X_r is simple, due to Lemma 5, $\text{Occ}^\uparrow(X_r, Y_r, q_1 - z)$ can be computed in $O(1)$ time. By Lemma 3, $\langle p_1, d_1, t_1 \rangle \cap (\text{Occ}^\uparrow(X_r, Y_r, q_1 - z) \oplus |Y_\ell|)$ can be computed in constant time.

- when $|Y_r| < q_1 - p_1$ (see the right of Fig. 6). The basic idea is the same as the previous case, but computing $\text{Occ}^\uparrow(X_r, Y_r, q_1 - z)$ is not enough, since $|Y_r|$ is ‘too short’. However, we can fill up the gap as follows.

$$\begin{aligned}
A &= \langle p_1, d_1, t_1 \rangle \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= (\langle p_1, d_1, t_1 \rangle \cap [p_1, q_1]) \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= (\langle p_1, d_1, t_1 \rangle \cap ([p_1, q_1 - |Y_r| - 1] \cup [q_1 - |Y_r|, q_1])) \cap (\text{Occ}(X_r, Y_r) \oplus z) \\
&= \langle p_1, d_1, t_1 \rangle \cap (S \cup \text{Occ}^\uparrow(X_r, Y_r, q_1 - z)) \oplus z, \\
&\quad \text{where } S = [p_1 - z, q_1 - z - |Y_r| - 1] \cap \text{Occ}(X_r, Y_r).
\end{aligned}$$

By Lemma 2, d_1 is the shortest period of $X_i[p_1 : q_1 + |Y_\ell| - 1]$. For this string, we have

$$\begin{aligned}
&X_i[p_1 : q_1 + |Y_\ell| - 1] \\
&= X_\ell[p_1 : |X_\ell|]X_r[1 : q_1 + |Y_\ell| - 1 - |X_\ell|] \\
&= X_\ell[p_1 : |X_\ell|]X_r[1 : q_1 - z - 1] \\
&= X_\ell[p_1 : |X_\ell|]X_r[1 : p_1 - z - 1]X_r[p_1 - z : q_1 - z - 1] \\
&= X_i[p_1 : p_1 + |Y_\ell| - 1]X_r[p_1 - z : q_1 - z - 1].
\end{aligned}$$

Therefore, $X_r[p_1 - z : q_1 - z - 1] = u^{t'}$ where u is the suffix of Y_ℓ of length d_1 . Thus,

$$S = \begin{cases} \langle p_1 - z, d_1, t' \rangle & \text{if } p_1 - z \in \text{Occ}(X_r, Y_r), \\ \emptyset & \text{otherwise,} \end{cases}$$

where t' is the maximum integer satisfying $p_1 - z + (t' - 1)d_1 \leq q_1 - z - |Y_r| - 1$. According to Observation 2, the union operation of $S \cup$

$Occ^\uparrow(X_r, Y_r, q_I - z)$ can be done in constant time in both cases. By Observation 1, checking whether $p_1 - z \in Occ(X_r, Y_r)$ or not can be reduced to checking if $p_1 - z \in Occ^\uparrow(X_r, Y_r, p_I - z)$. Since X_r is simple, it can be done in $O(1)$ time by Lemma 1 and Lemma 5. Finally, the intersection operation can be done in constant time by Lemma 3.

Therefore, in any case we can compute A in constant time.

Now we consider computing $B = Occ(X_\ell, Y_\ell) \cap (Occ^\Delta(X_i, Y_r) \ominus |Y_\ell|)$. Let $\langle p_2, d_2, t_2 \rangle = Occ^\Delta(X_i, Y_r)$. We have to consider how to compute $Occ^\uparrow(X_\ell, Y_\ell, p_2 - |Y_\ell|)$ efficiently. When X_ℓ is simple, we can use the same strategy as computing A . In case where X_ℓ is complex, $Occ^\uparrow(X_\ell, Y_\ell, p_2 - |Y_\ell|)$ can be computed in $O(\log s)$ time by Lemma 6.

Due to Lemma 5 and Lemma 6, the total extra work time and space are $O(h^2 + mh) + O(ms) = O(h^2 + m(h + s)) = O(h^2 + mn)$. This completes the proof. \square

We have proven that each $Occ^\Delta(X, Y)$ can be computed in $O(\log s)$ time with extra $O(h^2 + mn)$ work time and space. Thus, the whole time complexity is $O(h^2 + mn) + O(mn \log s) = O(h^2 + mn \log s)$, and the whole space complexity is $O(h^2 + mn)$. This leads to the result of Theorem 1.

5. Conclusions

Miyazaki et al. [18] presented an algorithm to solve the FCPM problem for straight line programs in $O(m^2 n^2)$ time and with $O(mn)$ space. Since simple collage systems can be translated to straight line programs, their algorithm gives us an $O(m^2 n^2)$ time solution to the FCPM problem for simple collage systems. In this paper we developed an FCPM algorithm for simple collage systems which runs in $O(\|\mathcal{D}\|^2 + mn \log |\mathcal{S}|)$ time using $O(\|\mathcal{D}\|^2 + mn)$ space. Since $n = \|\mathcal{D}\| + |\mathcal{S}|$, the proposed algorithm is faster than the algorithm by Miyazaki et al. [18].

An interesting extension of this research is to consider the FCPM problem for *composition systems* [22]. Composition systems can be seen as collage systems without repetitions. Since it is known that LZ77 compression can be translated into a composition system of size $O(n \log n)$, an efficient FCPM algorithm for composition systems would lead to a better solution for the FCPM problem with LZ77 compression. We remark that the only known FCPM algorithm for LZ77 compression takes $O((n + m)^5)$ time [6], which is still very far from desired optimal time complexity.

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