Computing Runs on a General Alphabet

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Abstract

We describe a RAM algorithm computing all runs (maximal repetitions) of a given string of length n over a general ordered alphabet in $O(n \log^{\frac{2}{3}} n)$ time and linear space. Our algorithm outperforms all known solutions working in $\Theta(n \log \sigma)$ time provided $\sigma = n^{\Omega(1)}$, where σ is the alphabet size. We conjecture that there exists a linear time RAM algorithm finding all runs.

Keywords: runs, general alphabet, maximal repetitions, linear time, repetitions

1. Introduction

Repetitions in strings are fundamental objects in both stringology and combinatorics on words. In some sense the notion of *run*, introduced by Main [13], allows to grasp the whole repetitive structure of a given string in a relatively simple form. Informally, a run of a string is a maximal periodic substring that is at least as long as twice its minimal period (the precise definition follows). In [9] Kolpakov and Kucherov showed that any string of length n contains $O(n)$ runs and proposed an algorithm computing all runs in linear time on an integer alphabet $\{0, 1, \ldots, n^{O(1)}\}$ and $O(n \log \sigma)$ time on a general ordered alphabet, where σ is the number of distinct letters in the input string. Recently, Bannai et al. described another interesting algorithm computing all runs in $O(n \log \sigma)$ time [1]. Modifying the approach of [1], we prove the following theorem.

Theorem. *For a general ordered alphabet, there is an algorithm that computes all runs in a string of length n in $O(n \log^{\frac{2}{3}} n)$ time and linear space.*

This is in contrast to the result of Main and Lorentz [14] who proved that any algorithm deciding whether a string over a general *unordered* alphabet has at least one run requires $\Omega(n \log n)$ comparisons in the worst case.

Our algorithm outperforms all known solutions when the number of distinct letters in the input string is sufficiently large (e.g., $\sigma = n^{\Omega(1)}$). It should be noted that the algorithm of Kolpakov and Kucherov can hardly be improved in a similar way since it strongly relies on a structure (namely, the Lempel–Ziv decomposition) that cannot be computed in $o(n \log \sigma)$ time on a general ordered alphabet (see [11]).

Based on some theoretical observations of [11], we conjecture that one can further improve our result.

Conjecture. *For a general ordered alphabet, there is a linear time algorithm computing all runs.*

2. Preliminaries

A *string of length* n over an alphabet Σ is a map $\{1, 2, \ldots, n\} \mapsto \Sigma$, where n is referred to as the length of w, denoted by $|w|$. We write $w[i]$ for the ith letter of w and $w[i..j]$ for $w[i]w[i+1] \ldots w[j]$. A string u is a *substring* (or a *factor*) of w if $u = w[i..j]$ for some i and j. The pair (i, j) is not necessarily unique; we say that i specifies an *occurrence* of u in w. A string can have many occurrences in another string. A substring $w[1..j]$ (respectively, $w[i..n]$) is a *prefix* (respectively, *suffix*) of w. An integer p is a *period* of w if $0 < p \le |w|$ and $w[i] = w[i+p]$ for all $i = 1, \ldots, |w|-p$; p is the *minimal period* of w if p is the minimal positive integer that is a period of w. For integers i and j, the set $\{k \in \mathbb{Z} \colon i \le k \le j\}$ (possibly empty) is denoted by $[i..j]$. Denote $[i..j) = [i..j-1]$ and $(i..j] = [i+1..j]$.

A *run* of a string w is a substring $w[i..j]$ whose period is at most half of the length of $w[i..j]$ and such that both substrings $w[i-1..j]$ and $w[i..j+1]$, if

defined, have strictly greater minimal periods than $w[i..j]$.

We say that an alphabet is *general* and *ordered* if it is totally ordered and the only allowed operation is comparing two letters. Hereafter, w denotes the input string of length n over a general ordered alphabet.

In the *longest common extension (LCE)* problem one has to preprocess w for queries $LCE(i,j)$ returning for given positions i and j of w the length of the longest common prefix of the suffixes $w[i..n]$ and $w[j..n]$. It is well known that one can perform the LCE queries in constant time after preprocessing w in $O(n \log \sigma)$ time, where σ is the number of distinct letters in w (e.g., see [7]). It turns out that the time consumed by the LCE queries is dominating in the algorithm of [1]; namely, one can prove the following lemma.

Lemma 1 (see [1, Alg. 1 and Sect. 4.2]). *Suppose we can answer in an online fashion any sequence of $O(n)$ LCE queries on w in $O(f(n))$ time for some function $f(n)$; then we can find all runs of w in $O(n + f(n))$ time.*

In what follows we describe an algorithm that computes $O(n)$ LCE queries in $O(n \log^{\frac{2}{3}} n)$ time and thus prove Theorem using Lemma 1. The key notion in our construction is a *difference cover*. Let $k \in \mathbb{N}$. A set $D \subset [0..k)$ is called a difference cover of $[0..k)$ if for any $x \in [0..k)$, there exist $y, z \in D$ such that $y - z \equiv x \pmod{k}$. Clearly $|D| \geq \sqrt{k}$. Conversely, for any $k \in \mathbb{N}$, there is a difference cover of $[0..k)$ with $O(\sqrt{k})$ elements: for example, the difference cover $[0..\lfloor\sqrt{k}\rfloor] \cup \{2\lfloor\sqrt{k}\rfloor, 3\lfloor\sqrt{k}\rfloor, \ldots\}$, which is depicted in Fig. 1. For further discussions and estimations of minimal difference covers, see [4, 15, 16].

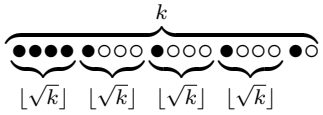

Figure 1: Simple difference cover of $[0..k)$ with $k = 18$.

Example. The set $D = \{1, 2, 4\}$ is a difference cover of $[0..5)$.

x	0	1	2	3	4
y, z	1, 1	2, 1	1, 4	4, 1	1, 2

Our algorithm utilizes the following interesting property of difference covers.

Lemma 2 (see [3]). *Let D be a difference cover of $[0..k)$. For any integers i, j, there exists $d \in [0..k)$ such that $(i+d) \bmod k \in D$ and $(j+d) \bmod k \in D$.*

3. Longest Common Extensions

At the beginning, our algorithm fixes an integer τ (the precise value of τ is given below). Let D be a difference cover of $[0..\tau^2)$ of size $O(\tau)$. Denote $M = \{i \in [1..n] \colon (i \bmod \tau^2) \in D\}$. Obviously, we have $|M| = O(\frac{n}{\tau})$. Our algorithm builds in $O(\frac{n}{\tau}(\tau^2 + \log n)) = O(\frac{n}{\tau} \log n + n\tau)$ time a data structure that can calculate $LCE(x, y)$ in constant time for any $x, y \in M$. To compute $LCE(x, y)$ for arbitrary $x, y \in [1..n]$, we simply compare $w[x..n]$ and $w[y..n]$ from left to right until we reach positions $x + d$ and $y + d$ such that $x + d \in M$ and $y + d \in M$, and then we obtain $LCE(x, y) = d + LCE(x+d, y+d)$ in constant time. By Lemma 2, we have $d < \tau^2$ and therefore, the value $LCE(x, y)$ can be computed in $O(\tau^2)$ time. Thus, our algorithm can execute any sequence of $O(n)$ LCE queries in $O(\frac{n}{\tau} \log n + n\tau^2)$ time. Putting $\tau = \lceil \log^{\frac{1}{3}} n \rceil$, we obtain $O(\frac{n}{\tau} \log n + n\tau^2) = O(n \log^{\frac{2}{3}} n)$. Now it suffices to describe the data structure answering the LCE queries on the positions from M.

Let $i_1, i_2, \ldots, i_m$ be the sequence of all positions from M in the increasing lexicographical order of the corresponding suffixes $w[i_1..n], w[i_2..n], \ldots, w[i_m..n]$. Our algorithm builds a *longest common prefix array* $\mathsf{lcp}[1..m-1]$ such that $\mathsf{lcp}[j] = LCE(i_j, i_{j+1})$ for $j \in [1..m)$ and a *sparse suffix array* $\mathsf{sa}[1..n]$ such that $i_{\mathsf{sa}[x]} = x$ for $x \in M$ and $\mathsf{sa}[x] = 0$ for $x \notin M$. Obviously $LCE(i_j, i_k) = \min\{\mathsf{lcp}[j], \mathsf{lcp}[j+1], \ldots, \mathsf{lcp}[k-1]\}$ for $j < k$. Based on this observation, we equip the lcp array with the *range minimum query (RMQ)* structure [5] that allows to compute $\min\{\mathsf{lcp}[j], \mathsf{lcp}[j+1], \ldots, \mathsf{lcp}[k-1]\}$ for any $j < k$ in $O(1)$ time. Now, to answer $LCE(x, y)$ for $x, y \in M$, we first obtain $j = \mathsf{sa}[x]$ and $k = \mathsf{sa}[y]$ and then answer $LCE(i_j, i_k)$ using the RMQ structure on the lcp array. Since the RMQ structure can be built in $O(n)$ time [5], it remains to describe how to construct lcp and sa.

In general our construction is similar to that of [10]. We use the fact that the set M has "period" τ^2, i.e., for any $x \in M$, we have $x+\tau^2 \in M$ provided $x + \tau^2 \leq n$. For simplicity, assume that $w[n]$ is a special letter that is smaller than any other letter in w. Our algorithm iteratively inserts the suffixes

$\{w[x..n]\colon x \in M\}$ in the arrays lcp and sa from right to left. Suppose, for some $k \in M$, we have already inserted in lcp and sa the suffixes $w[x..n]$ for all $x \in M \cap (k..n]$. More precisely, denote by $i'_1, i'_2, \ldots, i'_{m'}$ the sequence of all positions $M \cap (k..n]$ in the increasing lexicographical order of the corresponding suffixes $w[i'_1..n], w[i'_2..n], \ldots, w[i'_{m'}..n]$; we suppose that $\mathsf{lcp}[j] = LCE(i'_j, i'_{j+1})$ for $j \in [1..m')$, $i'_{\mathsf{sa}[x]} = x$ for $x \in M \cap (k..n]$, and $\mathsf{sa}[x] = 0$ for $x \notin M \cap (k..n]$. We are to insert the suffix $w[k..n]$ in lcp and sa. In order to perform the insertions efficiently, during the construction, the arrays lcp and sa are represented by balanced search trees with some auxiliary structures as described below.

1. Balanced search tree for lcp. The lcp array is represented by an augmented balanced search tree so that any RMQ query and modification on lcp take $O(\log n)$ amortized time.

2. List L. We store all positions $M \cap (k..n]$ on a linked list L in the lexicographical order of the corresponding suffixes. We maintain on this list the order maintenance data structure of [2] that allows to determine whether a given node of L precedes another node of L in constant time. The insertion of a new node in L takes amortized constant time. To provide constant time access to the nodes of L, we maintain an array $\mathsf{nds}[1..n]$ such that $\mathsf{nds}[x]$ is the node of L corresponding to position x if $x \in M \cap (k..n]$, and $\mathsf{nds}[x] = \mathbf{nil}$ otherwise.

3. Balanced search tree for sa. It is straightforward that, for any $x \in (k..n]$, $\mathsf{sa}[x]$ is equal to one plus the number of nodes of L preceding $\mathsf{nds}[x]$. So, we store all nodes of L in an augmented balanced search tree allowing to calculate the number of nodes preceding $\mathsf{nds}[x]$ in $O(\log n)$ time (since the comparison of two nodes takes $O(1)$ time). This tree together with the list L and the array nds allows to compute $\mathsf{sa}[x]$ in $O(\log n)$ time.

4. Trie S. We maintain a compacted trie S that contains the strings $w[x..x{+}\tau^2]$ for all $x \in M \cap (k..n]$ (we assume $w[j] = w[n]$ for all $j > n$ and thus $w[x..x{+}\tau^2]$ is always well defined). We maintain on S the data structure of [6] supporting insertions in $O(\tau^2 + \log n)$ amortized time. Let a be the leaf of S corresponding to a string $w[x..x{+}\tau^2]$. We augment a with a balanced search tree B_a that contains nodes $\mathsf{nds}[y]$ for all positions $y \in M \cap (k..n]$ such that $w[y{-}\tau^2..y] = w[x..x{+}\tau^2]$ (see Figure 2). We use B_a to compute in $O(\log n)$ time the immediate predecessor and successor of any given node $\mathsf{nds}[z]$, where $z \in M \cap (k..n]$, in the set of nodes stored in B_a. It is easy to see that S together with the associated search trees occupies $O(\frac{n}{\tau})$ space in total.

Example. Let $\tau^2 = 4$. The set $D = \{0, 1, 3\}$ is a difference cover of $[0..\tau^2)$. Consider the string $w = \underline{a}bc\underline{ab}c\underline{aba}bc\underline{abb}\underline{\$}$; the underlined positions are from $M = \{i \in [1..n]\colon (i \bmod \tau^2) \in D\}$. Figure 2 depicts the compacted trie S; each leaf of S is augmented with a balanced search tree of certain positions from $M \cap (k..n]$ (we use positions rather than nodes in this example). Consider the leaf of S corresponding to the string *abcab*. The string *abcab* occurs at positions $4, 9, 1$ in w. Hence, the balanced search tree B_4 must contain three positions: $4{+}\tau^2 = 8, 9{+}\tau^2 = 13, 1{+}\tau^2 = 5$. Note that the positions are stored in the lexicographical order of the corresponding suffixes $w[8..n], w[13..n], w[5..n]$.

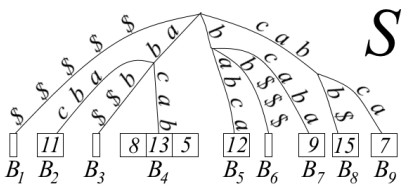

Figure 2: The balanced search trees $B_1, B_2, \ldots, B_9$ are augmented with some positions from M.

The construction of lcp *and* sa. To insert $w[k..n]$ in lcp and sa, we first insert $w[k..k{+}\tau^2]$ in S in $O(\tau^2 + \log n)$ time. If S did not contain the string $w[k..k{+}\tau^2]$ before, then, using auxiliary structures on S, we easily find in $O(1)$ time the position in lcp where the suffix $w[k..n]$ should be inserted; in the same way we obtain the LCE value between $w[k..n]$ and its immediate predecessor and successor in S. Then, we modify the balanced search tree representing lcp, insert a new node corresponding to $w[k..n]$ in L, insert this node in the balanced search tree supporting sa, and, finally, add a new empty tree B_a to the newly created leaf a of S. All these modifications take $O(\log n)$ amortized time.

Now suppose S contains $w[k..k{+}\tau^2]$. Denote by a the leaf of S corresponding to $w[k..k{+}\tau^2]$. In $O(\log n)$ time we obtain the immediate predecessor and successor of the node $\mathsf{nds}[k{+}\tau^2]$ (recall that $k{+}\tau^2 \in M$) in the search tree B_a; denote these nodes by $\mathsf{nds}[x]$ and $\mathsf{nds}[y]$, respectively. (We assume that the predecessor and successor both are

defined; the case when one of them is undefined is analogous). Note that $\mathsf{nds}[x]$ is the immediate predecessor only in the set of all nodes contained in B_a but it may not be the immediate predecessor in the whole list L; the situation with $\mathsf{nds}[y]$ is similar. Then we insert $\mathsf{nds}[k{+}\tau^2]$ between $\mathsf{nds}[x]$ and $\mathsf{nds}[y]$ in B_a. Since $w[x{-}\tau^2..x] = w[y{-}\tau^2..y] = w[k..k{+}\tau^2]$, it is straightforward that the suffixes $w[x{-}\tau^2..n]$ and $w[y{-}\tau^2..n]$ are, respectively, the immediate predecessor and successor of the suffix $w[k..n]$ in the set of all suffixes $\{w[x..n]\colon x \in M \cap (k..n]\}$. Hence, we insert a new node $\mathsf{nds}[k]$ in L between the nodes $\mathsf{nds}[x{-}\tau^2]$ and $\mathsf{nds}[y{-}\tau^2]$ (these nodes are certainly adjacent).

It is easy to see that $LCE(k, x{-}\tau^2) = \tau^2 + LCE(k{+}\tau^2, x)$ and $LCE(k, y{-}\tau^2) = \tau^2 + LCE(k{+}\tau^2, y)$. The values $LCE(k{+}\tau^2, x) = LCE(i'_{\mathsf{sa}[k+\tau^2]}, i'_{\mathsf{sa}[x]})$ and $LCE(k{+}\tau^2, y) = LCE(i'_{\mathsf{sa}[k+\tau^2]}, i'_{\mathsf{sa}[y]})$ can be computed in $O(\log n)$ time using the balanced search trees supporting access on sa and RMQ queries on lcp. All subsequent changes of other structures are the same as in the previous case and require $O(\log n)$ amortized time.

Finally, once the last suffix is inserted, we construct in an obvious way the plain arrays lcp and sa in $O(n)$ time.

Time and space. The insertion of a new suffix in the arrays lcp and sa takes $O(\tau^2 + \log n)$ amortized time. Thus, the construction of lcp and sa consumes overall $O(\frac{n}{\tau}(\tau^2 + \log n))$ time as required. The whole data structure occupies $O(n)$ space.

4. Conclusion

It seems that further improvements in the considered problem may be achieved by more efficient longest common extension data structures on a general ordered alphabet. One even might conjecture that there is a data structure that can execute any sequence of k *LCE* queries on a string of length n over a general ordered alphabet in $O(k + n)$ time. However, we do not yet have a theoretical evidence for such strong results.

Another interesting direction is a generalization of our result for the case of online algorithms (e.g., see [8] and [12]).

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