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OPTIMAL CACHE-AWARE SUFFIX SELECTION

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Abstract. Given string $S[1..N]$ and integer $k$, the suffix selection problem is to determine the $k$th lexicographically smallest amongst the suffixes $S[i..N]$, $1 \leq i \leq N$. We study the suffix selection problem in the cache-aware model that captures two-level memory inherent in computing systems, for a cache of limited size $M$ and block size $B$. The complexity of interest is the number of block transfers. We present an optimal suffix selection algorithm in the cache-aware model, requiring $\Theta \left( \frac{N}{B} \right)$ block transfers, for any string $S$ over an unbounded alphabet (where characters can only be compared), under the common tall-cache assumption (i.e. $M = \Omega \left( B^{1+\epsilon} \right)$, where $\epsilon < 1$). Our algorithm beats the bottleneck bound for permuting an input array to the desired output array, which holds for nearly any nontrivial problem in hierarchical memory models.

1. Introduction

Background: Selection vs Sorting. A collection of $N$ numbers can be sorted using $\Theta \left( N \log N \right)$ comparisons. On the other hand, the famous five-author result [2] from early 70’s shows that the problem of selection — choosing the $k$th smallest number — can be solved using $O \left( N \right)$ comparisons in the worst case. Thus, selection is provably simpler than sorting in the comparison model.

Consider a sorting vs selection question for strings. Say $S = S[1\cdots N]$ is a string. The suffix sorting problem is to sort the suffixes $S[i\cdots N]$, $i = 1,\ldots, N$, in the lexicographic order. In the comparison model, we count the number of character comparisons. Suffix sorting can be performed with $O \left( N \log N \right)$ comparisons using a combination of character sorting and classical data structure of suffix arrays or trees [11, 9, 4]. There is a lower bound of $\Omega \left( N \log N \right)$ since sorting suffixes ends up sorting the characters. For the related suffix selection problem where the goal is to output the $k$th lexicographically smallest suffix of $S$, the result in [6] recently gave an optimal $O \left( N \right)$ comparison-based algorithm, thereby showing that suffix selection is provably simpler than suffix sorting.

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The Model. Time-tested architectural approaches to computing systems provide two (or more) levels of memory: the highest one with a limited amount of fast memory; the lowest one with slow but large memory. The CPU can only access input stored on the fastest level. Thus, there is a continuous exchange of data between the levels. For cost and performance reasons, data is exchanged in fixed-size blocks of contiguous locations. These transfers may be triggered automatically like in internal CPU caches, or explicitly, like in the case of disks; in either case, more than the number of computing operations executed, the number of block transfers required is the actual bottleneck.

Formally, we consider the model that has two memory levels. The cache level contains $M$ locations divided into blocks (or cache lines) of $B$ contiguous locations, and the main memory level can be arbitrarily large and is also divided into blocks. The processing unit can address the locations of the main memory but it can process only the data residing in cache. The algorithms that know and exploit the two parameters $M$ and $B$, and optimize the number of block transfers are cache-aware. This model includes the classical External Memory model [1] as well as the well-known Ideal-Cache model [7].

Motivation. Suffix selection as a problem is useful in analyzing the order statistics of suffixes in a string such as the extremes, medians and outliers, with potential applications in bioinformatics and information retrieval. A quick method for finding say the suffixes of rank $i(n/10)$ for each integer $i$, $0 \leq i \leq 10$, may be used to partition the space of suffixes for understanding the string better, load balancing and parallelization. But in these applications, such as in bioinformatics, the strings are truly massive and unlikely to fit in the fastest levels of memory. Therefore it is natural to analyze them in a hierarchical memory model.

Our primary motivation however is really theoretical. Since the inception of the first block-based hierarchical memory model ([1],[10]), it has been difficult to obtain “golden standard” algorithms i.e., those using just $O(N/B)$ block transfers. Even the simplest permutation problem (PERM henceforth) where the output is a specified permutation of the input array, does not have such an algorithm. In the standard RAM model, PERM can be solved in $O(N)$ time. In both the Ideal-Cache and External Memory models, the complexity of this problem is denoted $\text{PERM}(N) = \Theta \left( \min \left\{ N, \frac{N}{B} \log_{M/B} \frac{N}{B} \right\} \right)$. Nearly any nontrivial problem one can imagine from list ranking to graph problems such as Euler tours, DFS, connected components etc., sorting and geometric problems have the lower bound of $\text{PERM}(N)$, even if they take $O(N)$ time in the RAM model, and therefore do not meet the “golden standard”. Thus the lower bound for PERM is a terrible bottleneck for block-based hierarchical memory models.

The outstanding question is, much as in the comparison model, is suffix selection provably simpler than suffix sorting in the block-based hierarchical memory models? Suffix sorting takes $\Theta \left( \left( \frac{N}{B} \right) \log_{M/B} \left( \frac{N}{B} \right) \right)$ block transfers [5]. Proving any problem to be simpler than suffix sorting therefore requires one to essentially overcome the PERM bottleneck.

Our Contribution. We present a suffix selection algorithm with optimal cache complexity. Our algorithm requires $\Theta(N/B)$ block transfers, for any string $S$ over an unbounded alphabet (where characters can only be compared) and under the common tall-cache assumption, that is $M = \Omega \left( B^{1+\epsilon} \right)$ with $\epsilon < 1$. Hence, we meet the “golden standard”; we
beat the perm bottleneck and consequently, prove that suffix selection is easier than suffix sorting in block-based hierarchical memory models.

**Overview.** Our high level strategy for achieving an optimal cache-aware suffix selection algorithm consists of two main objectives.

In the first objective, we want to efficiently reduce the number of candidate suffixes from \( N \) to \( O(N/B) \), where we maintain the invariant that the wanted \( k \)th smallest suffix is surely one of the candidate suffixes.

In the second objective, we want to achieve a cache optimal solution for the sparse suffix selection problem, where we are given a subset of \( O(N/B) \) suffixes including also the wanted \( k \)th suffix. To achieve this objective we first find a simpler approach to suffix selection for the standard comparison model. (The only known linear time suffix selection algorithm for the comparison model [6] hinges on well-known algorithmic and data structural primitives whose solutions are inherently cache inefficient.) Then, we modify the simpler comparison-based suffix selection algorithm to exploit, in a cache-efficient way, the hypothesis that \( O(N/B) \) (known) suffixes are the only plausible candidates.

**Map of the paper.** We will start by describing the new simple comparison-based suffix selection algorithm in Section 2. This section is meant to be intuitive. We will use it to derive a cache-aware algorithm for the sparse suffix selection problem in Section 3. We will present our optimal cache-aware algorithm for the general suffix selection problem in Section 4.

**2. A Simple(r) Linear-Time Suffix Selection Algorithm**

We now describe a simple algorithm for selecting the \( k \)th lexicographically smallest suffix of \( S \) in main memory. We give some intuitions on the central notion of work, and some definitions and notations used in the algorithm. Next, we show how to perform main iteration, called phase transition. Finally, we present the invariants that are maintained in each phase transition, and discuss the correctness and the complexity of our algorithm.

**Notation and intuition.** Consider the regular linear-time selection algorithm [2], hereafter called BFPRRT. Our algorithm for a string \( S = S[1...N] \) uses BFPRRT as a black box.\(^1\)

Each run of BFPRRT permits to discover a longer and longer prefix of the (unknown) \( k \)th lexicographically smallest suffix of \( S \). We need to carefully orchestrate the several runs of BFPRRT to obtain a total cost of \( O(N) \) time. We use \( S = \text{bbbabbbbbaa}$, where \( n = 12 \), as an illustrative example, and show how to find the median suffix (hence, \( k = n/2 = 6 \)).

**Phases and phase transitions.** We organize our computation so that it goes through phases, numbered \( t = 0, 1, 2, \ldots \) and so on. In phase \( t \), we know that a certain string, denoted \( \sigma_t \), is a prefix of the (unknown) \( k \)th lexicographically smallest suffix of \( S \). Phase \( t = 0 \) is the initial one: we just have the input string \( S \) and no knowledge, i.e., \( \sigma_0 \) is the empty string. For \( t \geq 1 \), a main iteration of our algorithm goes from phase \( t - 1 \) to phase \( t \) and is termed phase transition \((t - 1 \rightarrow t)\): it is built around the \( t \)th run of BFPRRT on a suitable subset of the suffixes of \( S \). Note that \( t \leq N \), since we ensure that the condition \( |\sigma_{t-1}| < |\sigma_t| \) holds, namely, each phase transition extends the known prefix by at least one symbol.

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\(^1\)In the following, we will assume that the last symbol in \( S \) is an endmarker \( S[N] = \$ \), smaller than any other symbol in \( S \).
Phase transition ($0 \rightarrow 1$). We start out with phase 0, where we run BFPRT on the individual symbols of $S$, and find the symbol $\alpha$ of rank $k$ in $S$ (seen as a multiset). Hence we know that $\sigma_1 = \alpha$, and this fact has some implications on the set of suffixes of $S$. Let $s_i$ denote the $i$th suffix $S[i \ldots N]$ of $S$, for $1 \leq i \leq N$, and $w_i$ be a special prefix of $s_i$ called work. We anticipate that the works play a fundamental role in attaining $O(N)$ time. To complete the phase transition, we set $w_i = S[i]$ for $1 \leq i \leq N$, and we call degenerate the works $w_i$ such that $w_i \neq \alpha$. (Note that degenerate works are only created in this phase transition.)

We then partition the suffixes of $S$ into two disjoint sets:

- The set of active suffixes, denoted by $A_1$—they are those suffixes $s_i$ such that $w_i = \sigma_1 = \alpha$.
- The set of inactive suffixes, denoted by $I_1$ and containing the rest of the suffixes—none of them is surely the $k$th lexicographically smallest suffix in $S$.

In our example ($k = 6$), we have $\sigma_1 = \alpha = b$ and, for $i = 1, 2, 3, 5, 6, 7, 8, 9$, $w_i = b$ and $s_i \in A_1$. Also, we have $s_j \in I_1$ for $j = 4, 10, 11, 12$, where $w_4 = w_{10} = w_{11} = a$ and $w_{12} = \$ are degenerate works.

A comment is in order at this point. We can compare any two works in constant time, where the outcome of the comparison is ternary $[<,=,>]$. While this observation is straightforward for this phase transition, we will be able to extend it to longer works in the subsequent transitions. Let us discuss the transition from phase 1 to phase 2 to introduce the reader to the main point of the algorithm.

Phase transition ($1 \rightarrow 2$). If $|A_1| = 1$, we are done since there is only one active suffix and this should be the $k$th smallest suffix in $S$. Otherwise, we exploit the total order on the current works. Letting $z_1$ be the number of works smaller than $\sigma_1$, our goal becomes how to find the $(k - z_1)$th smallest suffix in $A_1$. In particular, we want a longer prefix $\sigma_2$ and the new set $A_2 \subseteq A_1$.

To this end, we need to extend some of the works of the active suffixes in $A_1$. Consider a suffix $s_i \in A_1$. In order to extend its work $w_i$, we introduce its prospective work. Recall that $w_i = \sigma_1 = \alpha = S[i]$. If $w_{i+1} = S[i+1] \neq \alpha$ (hence, $s_{i+1}$ is inactive in our terminology), the prospective work for $s_i$ is the concatenation $w_i w_{i+1}$, where $s_{i+1} \in I_1$. Otherwise, since $w_i = w_{i+1}$ (and so $s_{i+1} \in A_1$), we consider $i + 2, i + 3, \ldots$, until we find the first $i + r$ such that $w_i \neq w_{i+r}$ (and so $s_{i+r} \in I_1$). In the latter case, the prospective work for $s_i$ is the concatenation $w_i w_{i+1} \cdots w_{i+r}$, where $w_i = w_{i+1} = \cdots = w_{i+r-1} = \sigma_1 = \alpha$ and their corresponding suffixes are active, while $w_{i+r} \neq \sigma_1$ is different and corresponds to an inactive suffix.

In any case, each prospective work is a sequence of works of the form $\alpha' \beta = \sigma_1^r \beta$, where $r \geq 1$ and $\beta \neq \alpha$. The reader should convince herself that any two prospective works can be compared in $O(1)$ time. We exploit this fact by running BFPRT on the set $A_1$ of active suffixes and, whenever BFPRT requires to compare any two $s_i, s_j \in A_1$, we compare their prospective works. Running time is therefore $O(|A_1|)$ if we note that prospective works can be easily identified by a scan of $A_1$: if $w_i w_{i+1} \cdots w_{i+r}$ is the prospective work for $s_i$, then $w_{i+1} \cdots w_{i+r}$ is the prospective work for $s_{i+1}$, and so on. In other words, a consecutive run of prospective works forms a collision, which is informally a maximal concatenated sequence of works equal to $\sigma_1$ terminated by a work different from $\sigma_1$ (this notion will be described formally in Section 2).
After BFPRT completes its execution, we know the prospective work that is a prefix of the (unknown) kth suffix in S. That prospective work becomes \( \sigma_2 \) and \( A_2 \) is made up of the the suffixes in \( A_1 \) such that their prospective work equals \( \sigma_2 \) (and we also set \( z_2 \)).

In our example, \( z_1 = 3 \), and so we look for the third smaller suffix in \( A_1 \). We have the following prospective works: one collision is made up of \( p_1 = \text{bbba} \), \( p_2 = \text{bbba} \), and \( p_3 = \text{ba} \); another collision is made up of \( p_5 = \text{bbbbba} \), \( p_6 = \text{bbbbba} \), \( p_7 = \text{bbba} \), \( p_8 = \text{bbba} \), and \( p_9 = \text{ba} \). Algorithm BFPRT discovers that \( \text{bbba} \) is the third prospective work among them, and so \( \sigma_2 = \text{bbba} \) and \( A_2 = \{ s_2, s_8 \} \) (and \( z_2 = 5 \)).

**How to maintain the works.** Now comes the key point in our algorithm. For each suffix \( s_i \in A_2 \), we update its work to be \( w_i = \sigma_2 \) (whereas it was \( w_i = \sigma_1 \) in the previous phase transition, so it is now longer). For each suffix \( s_i \in A_1 - A_2 \), instead, we leave its work \( w_i \) unchanged. Note this is the key point: although \( s_i \) can share a longer prefix with \( \sigma_2 \), the algorithm BFPRT has indirectly established that \( s_i \) cannot have \( \sigma_2 \) as a prefix, and we just need to record a Boolean value for \( w_i \), indicating if \( w_i \) is either lexicographically smaller or larger than \( \sigma_2 \). We can stick to \( w_i \) unchanged, and discard its prospective work, since \( s_i \) becomes inactive and is added to \( T_2 \). In our example, \( w_3 = w_8 = \text{bbba} \), while the other works are unchanged (i.e., \( w_3 = \text{b} \) while \( p_3 = \text{ba} \), \( w_5 = \text{b} \) while \( p_5 = \text{bbbbba} \), and so on).

In this way, we can maintain a total order on the works. If two works are of equal length, we declare that they are equal according to the symbol comparisons that we have performed so far, unless they are degenerate—in the latter case they can be easily compared as single symbols. If two works are of different length, say \( |w_i| < |w_j| \), then \( s_i \) has been discarded by BFPRT in favor of \( s_j \) in a certain phase, so we surely know which one is smaller or larger. In other words, when we declare two works to be equal, we have not yet gathered enough symbol comparisons to distinguish among their corresponding suffixes. Otherwise, we have been able to implicitly distinguish among their corresponding suffixes. In our example, \( w_3 < w_2 \) because they are of different length and BFPRT has established this disequality, while we declare that \( w_3 = w_5 \) since they have the same length. Recall that the total order on the works is needed for comparing any two prospective works in \( O(1) \) time as we proceed in the phase transitions. The works exhibit some other strong properties that we point out in the invariants described in Section 2.

**Time complexity.** From the above discussion, we spend \( O(|A_1|) \) time for phase transition \((1 \rightarrow 2)\). We present a charging scheme to pay for that. Works come again into play for an amortized cost analysis. Suppose that, in phase 0, we initially assign each suffix \( s_i \) two kinds of credits to be charged as follows: \( O(1) \) credits of the first kind when \( s_i \) becomes inactive, and further \( O(1) \) credits of the second kind when \( s_i \) is already inactive but its work \( w_i \) becomes the terminator of the prospective work of an active suffix. Note that \( w_i \) is incapsulated by the prospective work of that suffix (which survives and becomes part of \( A_2 \)).

Now, when executing BFPRT on \( A_1 \) as mentioned above, we have that at most one prospective work survives in each collision and the corresponding suffix becomes part of \( A_2 \). We therefore charge the cost \( O(|A_1|) \) as follows. We take \( \Theta(|A_1| - |A_2|) \) credits of the first kind from the \( |A_1| - |A_2| \geq 0 \) active suffixes that become inactive at the end of the phase transition. We also take \( \Theta(|A_2|) \) credits from the \( |A_2| \) inactive suffixes whose work terminates the prospective work of the survivors. In our example, the \( \Theta(|A_1| - |A_2|) \) credits are taken from \( s_1, s_3, s_5, s_6, s_7, \) and \( s_9 \), while \( \Theta(|A_2|) \) credits are taken from \( s_4 \) and \( s_{10} \).
At this point, it should be clear that, in our example, the next phase transition \((2 \rightarrow 3)\) looks for the \((k - z_t)\)th smaller suffix in \(A_2\) by executing BFPT in \(O(|A_2|)\) time on the prospective works built with the runs of consecutive occurrences of the work \(\sigma_2 = \text{bbaa}\) into \(S\). We thus identify \(\text{bbaa}\$ (with \(\sigma_3 = \text{bbaa}\)) as the median suffix in \(S\).

**Phase transition \((t - 1 \rightarrow t)\) for \(t \geq 1\).** We are now ready to describe the generic phase transition \((t - 1 \rightarrow t)\) more formally in terms of the active suffixes in \(A_{t-1}\) and the inactive ones in \(I_{t-1}\), where \(t \geq 1\).

The input for the phase transition is the following: (a) the current prefix \(\sigma_{t-1}\) of the (unknown) \(k\)th lexicographically smallest suffix in \(S\); (b) the set \(A_{t-1}\) of currently active suffixes; (c) the number \(z_{t-1}\) of suffixes in \(I_{t-1}\) whose work is smaller than that of the suffixes in \(A_{t-1}\) (hence, we have to find the \((k - z_{t-1})\)th smallest suffix in \(A_{t-1}\)); and (d) a Boolean vector whose \(t\)th element is false (resp., true) iff, for suffix \(s_i \in I_{t-1}\), the algorithm BFPT has determined that its work \(w_i\) is smaller (resp., larger) than \(\sigma_{t-1}\). The output of the phase transition are data (a)–(d) above, updated for phase \(t\).

We now define collisions and prospective works in a formal way. We say that two suffixes \(s_i, s_j \in A_{t-1}\) collide if their works \(w_i\) and \(w_j\) are adjacent as substrings in \(S\), namely, \(|i - j| = |w_i| = |w_j|\). A collision \(C\) is the maximal subsequence \(w_i w_{i+1} \cdots w_{l-1} w_l\), such that \(w_1 = w_2 = \cdots = w_{l-1} = \sigma_t\), where the active suffixes \(s_i\) and \(s_{i+1}\) collide for any \(1 \leq f < r\). For our algorithm, a collision can also be a degenerate sequence of just one active suffix \(s_i\) (since its work does not collide with that of any other active suffix).

The prospective work of a suffix \(s_i \in A_{t-1}\), denoted by \(p_i\), is defined as follows. Consider the collision \(C\) to which \(s_i\) belongs. Suppose that \(s_i\) is the \(h\)th active suffix (from the left) in \(C\), that is, \(C = w_1 w_2 \cdots w_{h-1} w_h w_{h+1} \cdots w_{l-1} w_l\). Consider the suffix \(s_u \in I_{t-1}\) adjacent to \(w_l\) (because of the definition of collision, \(s_u\) must be an inactive suffix following \(w_l\)). We define the prospective work \(p_i\) of \(s_i\), to be the string \(p_i = \sigma_1 w_1 \cdots w_{h-1} w_h w_{h+1} \cdots w_{l-1} w_l w_u\). Note that \(w_i = w_{h+1} = \cdots = w_{l-1} = w_l = \sigma_{t-1}\) since their corresponding suffixes are all active, while \(w_u\) is shorter. In other words, \(p_i = \sigma_1^{1-h} w_u\), with \(|w_u| < |\sigma_{t-1}|\).

**Lemma 2.1.** For any two suffixes \(s_i, s_j \in A_t\), we can compare their prospective works \(p_i\) and \(p_j\) in \(O(1)\) time.

We now give the steps for the phase transition. Note that we can maintain \(A_{t-1}\) in a data structure with monotone order of suffix position (i.e., \(i < j\) implies that \(s_i\) comes first than \(s_j\) in \(A_{t-1}\)).

1. Scan the active set \(A_{t-1}\) and identify its collisions and the set \(\mathcal{I}\) containing all the suffixes \(s_u \in I_{t-1}\) such that \(w_u\) immediately follows a collision. For any suffix \(s_i\) in \(A_{t-1}\), determine its prospective work \(p_i\) using the collisions and \(\mathcal{I}\).
2. Apply algorithm BFPT to the set \(\{p_i\}_{s_i \in A_{t-1}}\) using the constant-time comparison as stated in Lemma 2.1. In this way, find the \((k - z_{t-1})\)th lexicographically smallest prospective work \(p\), and the corresponding set \(A_t = \{s_i \in A_{t-1} | p_i = p\}\) of active suffixes whose prospective works match \(p\).
   (a) If \(|A_t| = 1\), stop the computation and return the singleton \(s_i \in A_t\) as the \(k\)th smallest suffix in \(S\).
   (b) If \(|A_t| > 1\), set \(\sigma_t = p\) (and update \(z_t\) accordingly).
3. For each \(s_i \in A_t\): Let \(p = w_i w_{h+1} \cdots w_l w_u\) be its prospective work, where \(s_u \in \mathcal{I}\). Set its new work to be \(w_i = p = \sigma_t\).
4. For each \(s_j \in A_{t-1} - A_t\), leave its work \(w_j\) unchanged and, as a byproduct of running BFPT in step 2, update position \(j\) of the Boolean vector (d) given in input, so as to record the fact that \(w_j\) is lexicographically smaller or larger than \(\sigma_t\).
Lemma 2.2. Executing phase transition \((t - 1 \rightarrow t)\) with \(t \geq 1\), requires \(O(|A_{t-1}|)\) time in the worst case.

Invariants for phase \(t\). Before proving the correctness and the complexity of our algorithm, we need to establish some invariants that are maintained through the phase transitions. We say that \(w_i\) is maximal if there does not exist another suffix \(w_j\) such that \(w_j\) contains \(w_i\), namely, such that \(j < i\) and \(i + |w_i| \leq j + |w_j|\). For any \(t \geq 1\), the following invariants hold (where \(A_0\) is trivially the set of all the suffixes):

(i) [prefixes]: \(\sigma_{t-1}\) and \(\sigma_t\) are prefixes of the (unknown) \(k\)th smallest suffix of \(S\), and \(|\sigma_{t-1}| < |\sigma_t|\).

(ii) [work]: For any suffix \(s_i\), its work \(w_i\) is either degenerate (a single mismatching symbol) or \(w_i = \sigma_t\) for a phase \(t' \leq t\). Moreover, \(w_i = \sigma_t\) iff \(s_i \in A_t\).

(iii) [comparing]: For any \(s_i\) and \(s_j\), \(|w_i| \neq |w_j|\) implies that we know whether \(w_i < w_j\) or \(w_i > w_j\).

(iv) [nesting]: For any two suffixes \(s_i\) and \(s_j\), their works \(w_i\) and \(w_j\) do not overlap (either they are disjoint or one is contained within the other). Namely, \(i > j\) implies \(i + |w_i| \leq j + |w_j|\) or \(i \geq j + |w_j|\).

(v) [covering]: The works of the active suffixes are all maximal and, together with the maximal works generated by the inactive suffixes, form a non-overlapping covering of \(S\) (i.e. \(S = w_{i_1}w_{i_2} \cdots w_{i_r}\), where \(i_1 < i_2 < \cdots < i_r\) and either \(s_{i_j} \in A_t\), or \(s_{i_j} \in I_t\) and \(w_{i_j}\) is maximal, for \(1 \leq j \leq r\)).

Lemma 2.3. After phase transition \((t - 1 \rightarrow t)\) with \(t \geq 1\), the invariants (i)–(v) are maintained.

Theorem 2.4. The algorithm terminates in a phase \(t \leq N\), and returns the \(k\)th lexicographically smallest suffix.

Theorem 2.5. Our suffix selection algorithm requires \(O(N)\) time in the worst case.

This simpler suffix selection algorithm is still cache “unfriendly”. For example, it requires \(O(N)\) block transfers with a string \(S\) with period length \(\Theta(B)\) (if \(S\) is a prefix of \(g^i\) for some integer \(i\), then \(g\) is a period of \(S\)).

3. Cache-Aware Sparse Suffix Selection

In the sparse suffix selection problem, along with the string \(S\) and the rank \(k\) of the suffix to retrieve, we are also given a set \(\mathcal{S}\) of suffixes such that \(|\mathcal{S}| = O(N/B)\) and the \(k\)th smallest suffix belongs in \(\mathcal{S}\). We want to find the wanted suffix in \(O(N/B)\) block transfers using the ideas of the algorithm described in Section 2.

Consider first a particular situation in which the suffixes are equally spaced \(B\) positions each other. We can split \(S\) into blocks of size \(B\), so that \(S\) is conceptually a string of \(N/B\) metacharacters and each suffix starts with a metacharacter. This is a fortunate situation since we can apply the algorithm described in Section 2 as is, and solve the problem in the claimed bound. The nontrivial case is when the suffixes can be in arbitrary positions.

Hence, we revisit the algorithm described in Section 2 to make it more cache efficient. Instead of trying to extend the work of an active suffix \(s_i\) by just using the works of the following inactive suffixes, we try to batch these works in a sufficiently long segment, called reach. Intuitively, in a step similar to step 2 of the algorithm in Section 2, we could first apply the BFPPRT algorithm to the set of reaches. Then, after we select a subset of equal reaches, and the corresponding subset of active suffixes, we could extend their works using
their reaches. This could cause collisions between the suffixes and they could be managed in a way similar to what we did in Section 2. This yields the notion of super-phase transition.

**Super-phase transition.** The purpose of a super-phase is to group consecutive phases together, so that we maintain the same invariants as those defined in Section 2. However, we need further concepts to describe the transition between super-phases. We number the super-phases according to the numbering of phases. We call a super-phase \( m \) if the first phase in it is \( m \) (in the overall numbering of phases).

**Reaches, pseudo-collisions and prospective reaches.** Consider a generic super-phase \( m \). Recall that, by the invariant \((v)\) in Section 2, the phase transitions maintain the string \( S \) partitioned into maximal works. We need to define a way to access enough (but not too many) consecutive “lookahead” works following each active suffix, before running the super-phase. Since some of these active suffixes will become inactive during the phases that form the super-phase, we cannot prefetch too many such works (and we cannot predict which ones will be effectively needed). This idea of prefetching leads to the following notion.

For any active suffix \( s_i \in A_m \), the reach of \( s_i \), denoted by \( r_i \), is the maximal sequence of consecutive works \( w_{l_1}, w_{l_2}, \ldots, w_{l_f} \) such that

1. \( i < l_1 < l_2 < \cdots < l_f \) and \( l_f - l_1 < B \);
2. \( w_l \) and \( w_l \) are adjacent and, for \( 1 < x \leq f \), \( w_{l_{x-1}} \) and \( w_{l_x} \) are adjacent in \( S \);
3. if \( j \) is the leftmost active suffix in \( S[i + 1 \ldots N] \), then \( l_f < j \).

We call a reach full if \( l_f < j \) in condition (iii), namely, we do not meet an active suffix while loading the reach. Since we know how to compare two works, we also know how to compare any two reaches \( r_i, r_j \), seen as sequences of works. We have the following.

**Lemma 3.1.** For any two reaches \( r_i \) and \( r_j \), such that \( |r_i| < |r_j| \), we have that \( r_i \) cannot be a prefix of \( r_j \).

Using reaches, we must possibly handle the collisions that may occur in an arbitrary phase that is internal to the current super-phase. We therefore introduce a notion of collision for reaches that is called pseudo-collision because it does not necessarily implies a collision.

For any two reaches \( r_i, r_j \) such that \( i < j \), we say that \( r_i \) and \( r_j \) pseudo-collide if \( r_i = r_j \) and the last work of \( r_i \) is \( w_j \) itself (not just equal to \( w_j \)). Thus, the last work of \( r_j \) is active and equal to \( w_i \) and \( w_j \). Certainly, the fact that \( r_i \) and \( r_j \) pseudo-collide during a super-phase does not necessarily imply that the works \( w_i \) and \( w_j \) collide in one of its phases. A pseudo-collision PC(l) is a maximal sequence \( r_{l_1}r_{l_2}\cdots r_{l_a} \) such that \( r_{l_1} \) and \( r_{l_{f+1}} \) pseudo-collide, for any \( 1 \leq f < a \). For our algorithm, a degenerate pseudo-collision is a sequence of just one reach.

Let us consider an active suffix \( s_i \) and the pseudo-collision to which \( r_i \) belongs. Let us suppose that the pseudo-collision is \( r_{l_1}r_{l_2}\cdots r_{l_{f-1}}r_f r_{l_{f+1}}\cdots r_{l_a} \) (i.e. \( r_i \) is the \( f \)th reach). Also, let us consider the reach \( r_u \) of the last work \( w_u \) that appears in \( r_u \) (by the definition of pseudo-collision, we know that the last work \( w_u \) of \( r_u \) is equal to its first work, so \( s_u \) is active and has a reach). The prospective reach of an active work \( w_i \), denoted by \( pr_i \), is the sequence \( r_f r_{l_{f+1}}\cdots r_u \) tail \( (pr_i) \), where tail \( (pr_i) = lcp(r_i, r_u) \) is the tail of \( pr_i \) and denotes the longest initial sequence of works that is common to both \( r_i \) and \( r_u \). Analogously to prospective works, we can define a total order on the prospective reaches. The multiplicity of \( pr_i \), denoted by \( mult \ (pr_i) \), is \( a - f + 1 \) (that is the number of reaches following \( r_i \) in the pseudo-collision plus \( r_i \)).
Lemma 3.2. If the invariants for the phases hold for the current super-phase then, for any two reaches \( r_i \) and \( r_j \) such that \( r_i = r_j \), we have that their prospective reaches \( pr_i \) and \( pr_j \) can be compared in \( O(1) \) time, provided we know the lengths of tail \( (pr_i) \) and tail \( (pr_j) \).

Super-phase transition \((m \rightarrow m')\). The transition from a super-phase \( m \) to the next super-phase \( m' \) emulates what happens with phases \( m, m+1, \ldots, m' \) in the algorithm of Section 2, but using \( O(N/B) \) block transfers.

1. For each active suffix \( s_i \), we create a pointer to its reach \( r_i \).
2. We find the \((k - z_m)\)th lexicographically smallest reach \( \rho \) using BFPT on the \( O(N/B) \) pointers to reaches created in the previous step. The sets \( \mathcal{R}_\rho = \{ s_i | s_i \text{ is active and } r_i = \rho \}, \mathcal{R}_< = \{ s_i | s_i \text{ is active and } r_i < \rho \}, \) and \( \mathcal{R}_\geq = \{ s_i | s_i \text{ is active and } r_i > \rho \} \) are thus identified, and, for any \( s_i \in \mathcal{R}_\rho \cap \mathcal{R}_\geq \), the length of \( lcp(r_i, \rho) \).
3. If \( |\mathcal{R}_\rho| = 1 \), we stop and return \( s_i \), such that \( s_i \in \mathcal{R}_\rho \), as the \( k \)th smallest suffix in \( S \).
4. For any \( s_i \in \mathcal{R}_\geq \), we compute its prospective reach \( pr_i \).
5. We find the \((k - z_m - |\mathcal{R}_<|)\)th lexicographically smallest prospective reach \( \pi \) among the ones in \( \{ pr_i | s_i \in \mathcal{R}_\geq \} \), thus obtaining \( \mathcal{P}_\pi = \{ s_i | s_i \text{ is active and } pr_i = \pi \}, \mathcal{P}_< = \{ s_i | s_i \text{ is active and } pr_i < \pi \}, \mathcal{P}_\geq = \{ s_i | s_i \text{ is active and } pr_i > \pi \} \), and, for any \( s_i \in \mathcal{P}_\geq \), the length of \( lcp(pr_i, \pi) \). If \( |\mathcal{P}_\pi| = 1 \), we stop and return \( s_i \), such that \( s_i \in \mathcal{P}_\pi \), as the \( k \)th smallest suffix in \( S \).

Theorem 3.3. The sparse suffix selection problem can be solved using \( O(N/B) \) block transfers in the worst case.

4. Optimal Cache-Aware Suffix Selection

The approach in Sec. 3 does not work if the number of input active suffixes is \( \omega(N/B) \). The process would cost \( O\left(\frac{N}{B} \log B\right) \) block transfers (since it would take \( \Omega(\log B) \) transitions to finally have \( O(N/B) \) active suffixes left). However, if we were able to find a set \( \mathcal{K} \) of \( O(N/B) \) suffixes such that one of them is the \( k \)th smallest, we could solve the problem with \( O(N/B) \) block transfers using the algorithm in Sec. 3. In this section we show how to compute such a set \( \mathcal{K} \).

Basically, we consider all the substrings of length \( B \) of \( S \) and we select a suitable set of \( p > B \) pivot substrings that are roughly evenly spaced. Then, we find the pivot that is lexicographically “closest” to the wanted \( k \)-th and one of the following two situations arises:

- We are able to infer that the \( k \)th smallest suffix is strictly between two consecutive pivots (that is its corresponding substring of \( B \) characters is strictly greater and smaller of the two pivots). In this case, we return all the \( O(N/p) = O(N/B) \) suffixes that are contained between the two pivots.
- We can identify the suffixes that have the first \( B \) characters equal to those of the \( k \)th smallest suffix. We show that, in case they are still \( \Omega(N/B) \) in number, they must satisfy some periodicity property, so that we can reduce them to just \( O(N/B) \) with additional \( O(N/B) \) block transfers.

\(^2\)Given strings \( S \) and \( T \), their longest common prefix \( lcp(S, T) \) is longest string \( U \) such that both \( S \) and \( T \) start with \( U \).
4.1. Finding pivots and the key suffixes

Let $p = \sqrt{\frac{M}{c}}$, for a suitable constant $c > 1$. We proceed with the following steps.

First. We sort the first $M^c$ substrings of length $B$ of $S$ (that is substrings $S[1 \ldots B]$, $S[2 \ldots B + 1]$, $\ldots$, $S[M^c - 1 \ldots B + M^c - 2]$, $S[M^c \ldots B + M^c - 1]$). Then we sort the second $M^c$ substrings of length $B$ and so forth until all the $N$ positions in $S$ have been considered. The product of this step is an array $V$ of $N$ pointers to the substrings of length $B$ of $S$.

Second. We scan $V$ and we collect in an array $U$ of $N/p$ positions the $N/p$ pointers $V[p], V[2p], V[3p], \ldots$.

Third. We (multi)-select from $U$ the $p$ pointers to the substrings (of length $B$) $b_1, \ldots, b_p$ such that $b_i$ has rank $i \frac{N}{p}$ among the substrings (pointed by the pointers) in $U$. These are the pivots we were looking for. We store the $p$ (pointers to) pivots in an array $U'$.

Fourth. We need to find the rightmost pivot $b_x$ such that the number of substrings (of length $B$ of $S$) lexicographically smaller than $b_x$ is less than $k$ (the rank of the wanted suffix). We cannot simply distribute all the substrings of length $B$ according to all the $p$ pivots in $U'$, because it would be too costly. Instead, we proceed with the following refining strategy.

1. From the $p$ pivots in $U'$ we extract the group $G_1$ of $\delta M$ equidistant pivots, where $\delta < 1$ is a suitable constant, (i.e. the pivots $b_1, b_{2\delta}, \ldots$, where $t = \frac{N}{p\delta}$). Then, for any $b_i \in G_1$, we find out how many substrings of size $B$ are lexicographically smaller than $b_i$. After that we find the rightmost pivot $b_{x_1} \in G_1$ such that the number of substrings (of length $B$) smaller than $b_{x_1}$ is less than $k$.

2. From the $\frac{\delta M}{p}$ pivots in $U'$ following $b_{x_1}$ we extract the group $G_2$ of $\delta M$ equidistant pivots. Then, for any $b_j \in G_2$, we find out how many substrings of size $B$ are smaller than $b_j$. After that we find the rightmost pivot $b_{x_2} \in G_2$ such that the number of substrings smaller than $b_{x_2}$ is less than $k$.

More generally:

$f$. Let $G_f$ be the $\delta M$ pivots in $U'$ following $b_{x_{f-1}}$. Then, for any $b_j \in G_f$, we find out how many substrings of size $B$ are smaller than $b_j$. After that we find the rightmost pivot $b_{x_f} \in G_f$ such that the number of substrings smaller than $b_{x_f}$ is less than $k$.

The pivot $b_{x_f}$ found in the last iteration is the pivot $b_x$ we are looking for in this step.

Fifth. We scan $S$ and compute the following two numbers: the number $n_x^+$ of substrings of length $B$ lexicographically smaller than $b_x$; the number $n_x^-$ of substrings equal to $b_x$.

Sixth. In this step we treat the following case: $n_x^- < k \leq n_x^+ + n_x^-$. More specifically, this implies that the wanted $k$th smallest suffix has its prefix of $B$ characters equal to $b_x$.

We proceed as follows. We scan $S$ and gather in a contiguous zone $R$ (the indexes of) the suffixes of $S$ having their prefixes of $B$ characters equal to $b_x$. In this case we have already found the key suffixes (whose indexes reside in $R$). Therefore the computation in this section ends here and we proceed to discard some of them (sec. 4.2).

Seventh. In this step we treat the following remaining case: $n_x^+ + n_x^- < k$. In other words, in this case we know that the prefix of $B$ characters of the wanted $k$th smallest suffix is (lexicographically) greater than $b_x$ and smaller than $b_{x+1}$. Therefore, we scan $S$ and gather in a contiguous zone $R$ (the indexes of) the suffixes of $S$ having their prefix of $B$ characters greater than $b_x$ and smaller than $b_{x+1}$. Since there are less than $N/B$ such suffixes (see below Lemma 4.1), we have already found the set of sparse active suffixes (whose indexes reside in $R$) that will be processed in Sec. 3.
Lemma 4.1. For any $S$ and $k$, either the number of key suffixes found is $O(N/B)$, or their prefixes of $B$ characters are all the same.

Lemma 4.2. Under the tall-cache assumption, finding the key suffixes needs $O(N/B)$ block transfers in the worst case.

4.2. Discarding key suffixes

Finally, let us show how to reduce the number of key suffixes gathered in Sec. 4.1 to $\leq 2N/B$ so that we can pass them to the sparse suffix selection algorithm (Sec. 3). Let us assume that the number of key suffixes is greater than $2N/B$.

The indexes of the key suffixes have been previously stored in an array $R$. Clearly, the $k$th smallest suffix is among the ones in $R$. We also know the number $n<q$ of suffixes of $S$ that are lexicographically smaller than each suffix in $R$. Finally, we know that there exists a string $q$ of length $B$ such that $R$ contains all and only the suffixes $s_i$ such that the prefix of length $B$ of $s_i$ is equal to $q$ (i.e. $R$ contains the indexes of all the occurrences of $q$ in $S$).

To achieve our goal we exploit the possible periodicity of the string $q$. A string $u$ is a period of a string $v$ ($|u| \leq |v|$) if $v$ is a prefix of $u^j$ for some integer $j \geq 1$. The period of $v$ is the smallest of its periods. We exploit the following:

Property 1 ([8]). If $q$ occurs in two positions $i$ and $j$ of $S$ and $0 < j - i < |q|$ then $q$ has a period of length $j - i$.

Let $u$ be the period of $q$. Since the number of suffixes in $R$ is greater than $2N/B$, there must be some overlapping between the occurrences of $q$ in $S$. Therefore, by Property 1, we can conclude that $|u| < |q|$. For the sake of presentation let us assume that $|q|$ is not a multiple of $|u|$ (the other case is analogous).

From how $R$ has been built (by left to right scanning of $S$) we know that the indexes in it are in increasing order, that is $R[i] < R[i+1]$, for any $i$ (i.e. the indexes in $R$ follow the order, from left to right, in which the corresponding suffixes may be found in $S$). Let us consider a maximal subsequence $R_i$ of $R$ such that, for any $1 \leq j < |R_i|$, $R_i[j+1] - R_i[j] \leq B/2$ (i.e. the occurrence of $q$ in $S$ starting in position $R_i[j]$ overlaps the one starting in position $R_i[j+1]$ by at least $B/2$ positions). Clearly, any two of these subsequences of $R$ do not overlap and hence $R$ can be seen as the concatenation $R_1R_2\cdots$ of these subsequences. From the definition of the partitioning of $R$ and from the periodicity of $q$ we have:

Lemma 4.3. The following statements hold:

(i) There are less than $2N/B$ such subsequences.

(ii) For any $R_i$, the substring $S[R_i[1] \ldots R_i[|R_i|] + B - 1]$ (the substring of $S$ spanned by the substrings whose indexes are in $R_i$) has period $u$.

(iii) The substring of length $B$ of $S$ starting in position $R_i[|R_i|] + |u|$ (the substring starting one period-length past the rightmost member of $R_i$) is not equal to $q$.

For any key suffix $s_j$, let us consider the following prefix: $ps_j = S[j \ldots R_i[|R_i|] + |u| + B - 1]$, where $R_i$ is the subsequence of $R$ where (the index of) $s_j$ belongs to. By Lemma 4.3, we know two things about $ps_j$: (a) the prefix of length $|ps_j| - |u|$ of $ps_j$ has period $u$; (b) the suffix of length $B$ of $ps_j$ is not equal to $q$.

In light of this, we associate with any key suffix $s_j$ a pair of integers $(\alpha_j, \beta_j)$ defined as follows: $\alpha_j$ is equal to the number of complete periods $u$ in the prefix of length $|ps_j| - |u|$ of $ps_j$; $\beta_j$ is equal to $|R_i| + |u|$ (that is the index of the substring of length $B$ starting one period-length past the rightmost member of $R_i$).
There is natural total order $\preceq$ that can be defined over the key suffixes. It is based on the pairs of integers $\langle \alpha_j, \beta_j \rangle$ and it is defined as follow. For any two key suffixes $s_{j'}, s_{j''}$:

- If $\alpha_{j'} = \alpha_{j''}$ then $s_{j'}$ and $s_{j''}$ are equal (according to $\preceq$).
- If $\alpha_{j'} < \alpha_{j''}$ then $s_{j'} \preceq s_{j''}$ iff $S[\beta_{j'} \ldots \beta_{j'} + B - 1]$ is lexicographically smaller than $q$.

By Lemma 4.3, we know that the suffix of length $B$ of $ps_{j'}$ (which is the substring $S[\beta_{j'} \ldots \beta_{j'} + B - 1]$) is not equal to $q$. Therefore the total order $\preceq$ is well defined.

We are now ready to describe the process for reducing the number of key suffixes. We proceed with the following steps.

First. By scanning $S$ and $R$, we compute the pair $\langle \alpha_j, \beta_j \rangle$ for any key suffix $s_j$. The pairs are stored in an array (of pairs of integers) $Pairs$.

Second. We scan $S$ and compute the array $Comp$ of $N$ positions defined as follows: for any $1 \leq i \leq N$, $Comp[i]$ is equal to $-1$, $0$ or $1$ if $S[i \ldots i + B - 1]$ is less than, equal to or greater than $q$, respectively (the array $Comp$ tells us what is the result of the comparison of $q$ with any substring of size $B$ different from it).

Third. By scanning $Pairs$ and $Comp$ at the same time, we compute the array $PComp$ of size $|Pairs|$, such that, for any $l$, $PComp[l] = Comp[Pairs[l], \beta]$ (where $Pairs[l], \beta$ is the second member of the pair of integers in position $l$ of $Pairs$).

Fourth. Using $Pairs$ and $PComp$, we select the $(k-n^\nu)$-th smallest key suffix $s_x$ and all the key suffixes equal to $s_x$ according to the total order $\preceq$ (where $n^\nu$ is the number of suffixes of $S$ that are lexicographically smaller than each suffix in $R$, known since Sec. 4.1). The set of the selected key suffixes is the output of the process.

**Lemma 4.4.** At the end of the discarding process, the selected key suffixes are less than $2N/B$ in number and the $k$th lexicographically smallest suffix is among them.

**Lemma 4.5.** The discarding process requires $O(N/B)$ block transfers at the worst case.

**Theorem 4.6.** The suffix selection problem for a string defined over a general alphabet can be solved using $O(N/B)$ block transfers in the worst case.

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