# Suffix Array 構成アルゴリズムの比較

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文書データベースから一致文字列を検索するためのコンパクトなデータ構造として suffix array がある。これは文字列の全ての接尾辞のポインタを辞書順に格納した配列で、木構造よりも必要なメモリが小さいため大きなデータベースに対して有効である。また suffix array の構成はブロックソート圧縮法でも必要である。この suffix array を構成するいくつかのアルゴリズムについて必要メモリと速度を比較し、それらを組み合わせた 省メモリで高速なアルゴリズムを提案する。このアルゴリズムは文字列中に繰り返しが 多い場合に有効である。

# Comparison among Suffix Array Construction Algorithms

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Suffix array is a compact data structure for searching matched strings from text databases. It is an array of pointers and stores all suffixes of a text in lexicographic order. Because its memory requirement is less than tree structures, it is effective for large databases. Moreover, constructing the suffix array is used in the Block Sorting compression scheme. We compare algorithms for constructing suffix arrays on speed and required memory and propose a fast and memory efficient algorithm by combining them. It is effective when the length of repeated strings in a text is large.

### 1 Introduction

#### 1.1 Suffix array

Today large databases become available, such as full text of newspapers or Web pages, and Genome sequences, therefore it is important to store them on memory for quick queries. The suffix array [9] is a compact data structure for on-line string searches. For such purposes, suffix tree is used because it enables us to find the longest substring of a large text that matches the query string in linear time. However, the suffix tree requires huge memory and therefore it is impossible to use it for large text.

Though the suffix tree can be constructed in linear time [10, 12, 8], the time complexity depends on alphabet size. Searching time also depends on alphabet size, it becomes slow if

the alphabet size is large. On the other hand, construction and searching time of the suffix array does not depend the size of alphabet. Therefore searching time of the suffix array may be comparable with the suffix tree though it is  $O(P \log n)$  time where P is the length of a query string and n is the length of a large text.

## 1.2 Block Sorting data compression

The Block Sorting is a lossless data compression scheme [4]. Though its compression ratio is comparable with variants of the PPM [6], its compression speed is faster than them and decompression speed is much faster than compression. Moreover, the required memory is smaller than the PPM, therefore the block sorting is a promising scheme.

The Burrows-Wheeler transformation used in the Block Sorting is a permutation of a text to be compressed and it requires sorting of all suffixes of the text. Though the Block Sorting is faster than the PPM, it is slower than the gzip because sorting of long text is required. Therefore speeding-up of sorting suffixes is important.

### 1.3 Our results

We propose a fast and memory efficient algorithm for sorting all suffixes of a text. It requires 9n bytes for sorting a text of length n. Note that word size of integers is four. We define a property of texts: average match length (AML). Our algorithm is faster than other sorting algorithms when AML is large. Its speed is compared with a suffix tree construction algorithm and it requires less than half of memory for the suffix tree. The AML becomes large if a text is long, therefore our algorithm is practical for large texts.

## 1.4 Definitions

Assume that we make a suffix array of a text  $X = x[0..n-1] = x_0x_1...x_{n-1}$ . The size of alphabet  $\Sigma$  is constant. We add a special symbol '\$' at the tail of the text X, that is,  $x_n = \$$ . It is not in the alphabet and smaller than any other symbols in the alphabet. Suffixes of the text are represented by  $S_i = x[i..n]$ . The suffix array I[0..n] is an array of indexes of suffixes  $S_i$ . All suffixes in the array is sorted in lexicographic order. A suffix  $S_i$  is lexicographically less than a suffix  $S_i$  if  $\exists l \geq 0 \ x[i..i+l-1] = x[j..j+l-1]$  and x[i+l] < x[j+l].

We use another ordering  $<_k$ .

- $S_i <_k S_j \iff 0 \le \exists l < k \ x[i..i+l-1] = x[j..j+l-1] \text{ and } x[i+l] < x[j+l]$
- $S_i =_k S_j \iff x[i..i+k-1] = x[j..j+k-1]$
- $S_i >_k S_j \iff 0 \le \exists l < k \ x[i..i+l-1] = x[j..j+l-1] \text{ and } x[i+l] > x[j+l]$

### 2 Related works

It is necessary to sort all suffixes of a string in lexicographic order for making a suffix array, Several algorithms below are available to sort suffixes. First two algorithms are general algorithms for sorting strings and last two algorithms are for sorting suffixes of a string.

### 2.1 Bentley-Sedgewick algorithm

This is a practical algorithm for sorting strings [3]. It is similar to the quick sort, but it recursively partitions pointers of strings into three parts. A pivot used to partition is the first

symbol in a string and all strings are partitioned into less than, equal to and greater than the symbol according to their first symbols. A key idea of this algorithm is to use the equal part. Strings in it have same first symbols, therefore we can sort the strings without comparing the first symbols.

The Bentley-Sedgewick algorithm is used in a free software bzip2 [11], an implementation of the Block Sorting.

## 2.2 Andersson-Nilsson algorithm (Forward Radix Sort)

This is a general algorithm for sorting strings using radix sort [1]. First all strings are sorted and split into groups according to their first symbols by using a radix sort. Next all strings are moved to buckets according to second symbols. The buckets are traversed in alphabetical order and strings in them are returned to their own groups. Now all strings are sorted according to first two symbols, therefore we split the groups and iterate these operations until all strings are sorted. This algorithm is simple, but it theoretically has good time complexity.

# 2.3 Karp-Miller-Rosenberg (KMR) algorithm

This is an algorithm for finding repeated patterns in a string [7] and it can be used for sorting suffixes. First all suffixes  $S_i$  of a string are split into groups according to their first symbols and given numbers V[i]. All suffixes in a group have the same v = V[i] and the group is represented by the number v. All groups have different numbers. Next suffixes of each group are split into subgroups according to the V[i] of suffixes. As a result, all suffixes are split according to their first two symbols. This operation continues until all suffixes belong to groups of size one. If groups are split ted and ordered in alphabetical order, all suffixes can be sorted in lexicographic order.

The key of the algorithm is called doubling technique. Because all suffixes in a group have the same first k symbols x[i..i+k-1] after an iteration, we can split the groups according to x[i+k..n]. Since the numbers V[i+k] are already calculated in the last iteration, we can split the groups according to x[i+k..i+2k-1] in the next iteration. Consequently, all suffixes can be sorted in lexicographic order within  $\lceil \log n \rceil$  iterations.

### 2.4 Manber-Myers algorithm

This algorithm [9] uses the doubling technique in the KMR algorithm. First we sort and group all suffixes by their first symbols using bucket sort. Suffixes are given V[i] like the KMR algorithm. Next all groups are traversed in order of V[i] to sort suffixes in groups by the second symbols. If I[0] = i,  $S_{i-1}$  is the smallest suffix among the group V[i-1]. We can move suffixes to their correct position according to first two symbols. The number of iterations is less than  $\lceil \log n \rceil$  and each iteration can be done in O(n) time, therefore this algorithms works in  $O(n \log n)$  time.

### 2.5 Larsson's suffix tree construction algorithm

This algorithm can maintain a sliding window of a text in linear time and it can be used for PPM-style data compression [8]. Though the suffix tree requires more memory than the suffix array, it enables us to search substrings in linear time.

# 3 Proposed algorithm

#### 3.1 Idea

We propose an algorithm for sorting suffixes using the KMR algorithm and the Bentley-Sedgewick algorithm. Though the original KMR algorithm uses bucket sort for each group, it is not practical because the number of different values of  $V_i$  becomes large and the number of unsorted elements becomes small as iteration of the KMR algorithm proceeds and cost of initializing the bucket becomes large. The algorithm of Manber and Myers solved the problem of the KMR algorithm, but it traverses all suffixes in each iteration even if almost all suffixes were already sorted. We use a comparison-based algorithm for sorting each group and sort only unsorted strings in each iteration.

Our algorithm maintains  $<_k$  order of all suffixes in an array V[0..n] and iterate until all V[i] have different values. After iteration i, all suffixes are sorted according to their fist  $k=2^i$  symbols. Suffixes whose first k symbols are equal form a group. Groups are called *unsorted* if their size is more than one and they are called *sorted* if the size is one. We want to update V[i] and I[i] for only unsorted groups, therefore V[i] and I[i] of sorted groups must be consistent in all iterations, that is, if a suffix  $S_i$  is in a sorted group, V[i] and I[j] = i must be equal to the values in case of all suffixes are sorted.

**Definition 1** The array V[0..n] is consistent with  $<_k$  order if

- $S_i < S_j \rightarrow V[i] \leq V[j]$ ,
- $S_i > S_j \rightarrow V[i] \geq V[j]$ , and
- $S_i \neq_k S_j \rightarrow V[i] \neq V[j]$ .

Note that we can assign different values v1, v2 to V[i], V[j] even if  $S_i =_k S_j$ . It enables us to sort suffixes without temporary arrays.

Our algorithm continues maintains V[0..n] is always consistent. If V[i] represents the number of suffixes which are less than a suffix  $S_i$  according to  $<_k$  order, the value is consistent and V[I[i]] = i if a suffix  $S_i$  is in a sorted group. This numbering comes from [9, 1].

Our algorithm proceeds as follows:

- 1. sort  $S_i$  by their first symbols using bucket sort initialize V[i]  $(1 \le i \le n)$  let k = 1
- 2. sort unsorted groups according to  $<_k$  order using a comparison-based sorting algorithm
- 3. split groups and update V
- 4. combine consecutive sorted groups into one sorted group
- 5. if the number of groups is n, exit
- 6.  $k = k \times 2$  and goto 2.

Sorting and splitting a group is done as follows. Assume that suffixes in a group G = I[s..e] are equal according to  $<_k$  order and now we calculate  $<_{2k}$  order and V[i] for  $i \in I[s..e]$ . First we sort the suffixes according to V[I[i]+k] ( $s \le i \le e$ ) by using a comparison-based algorithm, then we split the group and update V[i]. The group is split if  $V[I[i]+k] \ne V[I[i+1]+k]$ 

## 3.2 Implementation

Groups are represented by three arrays: I[0..n], V[0..n] and S[0..n]. After iteration k, they represents

- I[i]: the index of *i*-th suffix in  $<_k$  order,
- V[s]: the group of suffix  $S_s$ ,
- S[g]: the size of the group g.

To implement the algorithm without temporary arrays, we must update groups carefully. All groups are traversed in  $<_k$  order in iteration k. The first group begins at index 0 and its size is S[0]. We sort suffixes  $S_s \in I[0..size-1]$  according to V[s+k]. The next group begins at index g=size, therefore we sort  $S_s \in I[g..g+S[g]-1]$ . After all groups are traversed, they will be split to form  $<_{2k}$  order. If consecutive suffixes I[i] and I[i+1] in group g have different group V[I[i]+k] and V[I[i+1]+k], the group g is split between index i and i+1, that is, the size of the group is updated. Next we update the array V for having different value according to  $<_{2k}$  order.

If the size of a group g becomes one in iteration k, the group is sorted and it is unnecessary to sort the group after the iteration. Therefore we want to skip such groups. To do so, we combine consecutive sorted groups into one group. We use sign flag of the size of groups to indicate whether a group is sorted or not.

Figure 1 shows an example of the algorithm for sorting suffixes of "tobeornottobe\$."

	i	0	1	2	3	4	5	6	7	8	9	10	11	12	13
k	$x_i$	t	o	b	e	0	r	$\mathbf{n}$	0	t	t	O	b	e	\$
	I[i]	13	2	11	3	12	6	1	4	7	10	5	0	8	9
İ	S[i]	-1	2	•	2		-1	4				-1	3		]
	V[I[i]]	0	1	1	3	3	5	6	6	6	6	10	11	11	11
1	V[I[i]+k]		3	3	6	0		1	10	11	1		6	11	6
	I[i]		2	11	12	3		1	10	4	7		0	9	8
	S[i]	-1	2		-3			2		-3			2		-1
	V[I[i]]		1	1	3	4		6	6	8	9		11	11	13
2	V[I[i]+k]		8	0				4	3				1	1	
	I[i]		11	2				10	1				0 -	9	ĺ
	S[i]	-11											2		-1
	V[I[i]]		1	2				6	7				11	11	
4	V[I[i]+k]												8	0	
	I[i]												9	0	
	S[i]	-14													
	V[I[i]]												11	12	
	I[i]	13	11	2	12	3	6	10	1	4	7	5	9	0	8

Table 1: An example of the algorithm

## 3.3 Further improvements

#### 3.3.1 One pass implementation

The above implementation requires three passes: sorting, updating S, and updating V. Note that combining groups can be done in sorting pass of the next iteration. However, we can change the algorithm to work in one pass. To update V in the sorting pass, we must guarantee values of V to be consistent. If an element V[i] is updated, the value is always incremented and it may become larger than other values which are greater than  $S_i$  in  $<_k$  order. However, if we update the array V right to left, it is always consistent. Therefore we use quick sort or Bentley-Sedgewick algorithm and sort recursively from greater part to less part.

#### 3.3.2 Initial bucket sort

Because we use comparison-based sorting algorithms for each group, the first iteration requires  $O(n \log n)$  time. We can accelerate the iteration using bucket sort by first two symbols of suffixes. We use an array of size  $|\Sigma|^2$  and count the number of all patterns of two symbols. All suffixes are sorted according to first two symbols in O(n) time.

# 3.3.3 Word size of the array S

Because elements of the array S is the size of groups, range of the values varies from 1 to n. However, only one byte is enough for each element.

If the size of a group g is s, S[g+1...g+s-1] are not used. Therefore we can store the size of the group to the unused array if the size cannot be represented by one byte.

# 4 Experimental results

We have experimented on sorting time of various algorithms. We used SparcStation20 with 128MB memory. Algorithms we experimented are Bentley-Sedgewick, Manber-Myers, Larsson, and our algorithm. All algorithm except the Larsson use initial bucket sorting. Larsson is not a sorting algorithm but a suffix tree construction algorithm. Children of a node are represented by a linked list and elements of the list is rearranged by move-to-front rule. Our algorithm uses doubling technique of the KMR algorithm and ternary partitioning of the Bentley-Sedgewick.

Table 2 shows memory requirements of the algorithms. Bentley-Sedgewick uses only one array I[0..n] and a text buffer. Larsson [8] uses five arrays of size n and a text buffer. Our algorithm uses two arrays I and V of 4n bytes and an array S of n bytes. Both Bentley-Sedgewick and our algorithm use stack. However, the depth of the stack of our algorithm is smaller than that of Bentley-Sedgewick because Bentley-Sedgewick is depth-first algorithm and ours is breadth-first algorithm. Though Larsson's suffix tree requires 21n bytes in the worst case, generally the number of internal nodes is about half of leaves and the tree requires about 14n bytes.

Table 3 shows time for sorting suffixes or making suffix tree. We use files in Calgary corpus [5] and Canterbury corpus [2] for benchmark. In the table, first, second and third columns show filename, size and average match length (AML) of the files. The AML is defined as follows.

$$AML = \frac{1}{n-1} \sum_{i=1}^{n-1} (\text{match length of } S_{I[i]} \text{ and } S_{I[i+1]})$$

Note that the AML is average number of symbols per suffix to verify that all suffixes are sorted correctly. Fourth to seventh columns show time of Bentley-Sedgewick, Manber-Myers, Larsson,

Table 2: memory requirements

algorithm	memory			
Bentley-Sedgewick	5n + stack			
Manber-Myers	13n			
Larsson	21n			
our algorithm	9n + stack			

and our algorithm respectively. Bentley-Sedgewick is fast if AML is small (geo, book1). Larsson

Table 3: sorting time and average match length

	sorting time (sec)					
name	size	AML	BS	MM	Lar	ours
geo	102400	3.5	0.6	0.9	3.3	0.7
book1	768771	7.3	8.2	18.0	14.3	9.5
progc	39611	8.2	0.5	0.3	0.3	0.2
book2	610856	9.6	7.6	14.1	8.9	6.9
bible.txt	4047392	13.9	83	136	65	70
E.coli	4638690	17.3	142	177	84	101
news	377109	18.1	12.9	10.1	6.4	3.5
world192.txt	2473400	22.9	70	80	38	38
progl	71646	24.6	3.0	1.0	0.6	0.4

and our algorithm are fast if AML is large. Larsson is much faster than our algorithm for E.coli because E.coli is a DNA sequence of a *colon bacillus* and linked lists in the tree is efficient for the small alphabet (ATCG). Our algorithm is fast if AML is large because it uses the doubling technique of the KMR algorithm. Roughly speaking, its sorting time is proportional to logarithm of the AML. On the other hand, sorting time of the Bentley-Sedgewick is proportional to the AML.

## 5 Conclusion

We examined speed and required memory of several algorithms for sorting suffixes and making a suffix tree and we proposed a practical algorithm for sorting suffixes by combining other algorithms. Our algorithm is faster than other sorting algorithms when average match length of suffixes is large. Its speed is compared with a suffix tree construction algorithm and it requires less than half of memory for the suffix tree. The average match length becomes large if a text is long, therefore our algorithm is practical for large texts.

Our algorithm is slower than the suffix tree construction algorithm if a text is large because it uses comparison-based sorting. As future works, we will therefore remove the comparison-based sorting like the algorithm of Manber and Myers.

The Block Sorting compression uses sorting of suffixes. However, sorting is not a necessary condition. For example, all preceding symbols  $x_{i-1}$  of suffixes  $S_i$  in a group are the same, sorting of the group is not required. We may accelerate compression speed of the Block Sorting by using the property.

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