

# 1 Introduction

## Definition 1.1

Let  $\Sigma = \{0, \dots, \sigma - 1\}$  be a finite, ordered set. The elements of  $\Sigma$  are called *characters* or *symbols* and  $\Sigma$  is called an *alphabet* of size  $\sigma$ .

## Definition 1.2

A *string*  $S$  is a sequence of characters from an alphabet  $\Sigma$ .

- We usually use  $n = |S|$  to be the length of the string.
- The  $i$ -th character of  $S$  is  $S[i]$ . Indices are 0-based.
- The substring from the  $i$ -th to the  $j$ -th character is  $S[i..j]$ .
- A substring with  $i = 0$  is called *prefix*. A substring with  $j = n - 1$  is called *suffix*.
- The  $i$ -th suffix is  $S[i..n - 1]$ .

## 1.1 Suffix Tries

### Definition 1.3

Let  $S = \{S_0, S_1, \dots, S_{N-1}\}$  be a set of strings over an alphabet  $\Sigma$ . A *trie*  $\mathcal{T}$  is a tree, where each node represents a different prefix in the set  $S$ . The root represents the empty prefix  $\varepsilon$ . Vertex  $u$  representing prefix  $Y$  is a child of vertex  $v$  representing prefix  $X$ , if and only if  $Y = Xc$  for some character  $c \in \Sigma$ . The edge  $(v, u)$  is then labeled  $c$ .

If  $S$  is the set of all suffixes of a string  $T$ , the trie is called *suffix trie*.

### Example 1.4

Figure 1.1 shows the suffix trie for the string "banana\$". The dollar sign "\$" is a sentinel that does not appear elsewhere in the text. This guarantees, that no suffix is a prefix of another suffix and the suffix trie therefore has  $n + 1$  leaves.

To construct a trie over string set  $S = \{S_0, \dots, S_{N-1}\}$ , we need  $\mathcal{O}(|S_0| + \dots + |S_{N-1}|)$  steps. This bound is tight: If all characters are pairwise distinct in all strings and no two strings share a character, then the number of different prefixes and therefore vertices is given by  $1 + \sum_{i=0}^{N-1} |S_i|$ , where the additional 1 represents the empty prefix  $\varepsilon$ .

The time needed to search for a string  $T$  of length  $m = |T|$  in the trie depends on the implementation of the tree. If the children of each vertex are stored in a list, the time is in  $\mathcal{O}(m\sigma)$ . If the children are stored in a sorted array (using the order of the characters

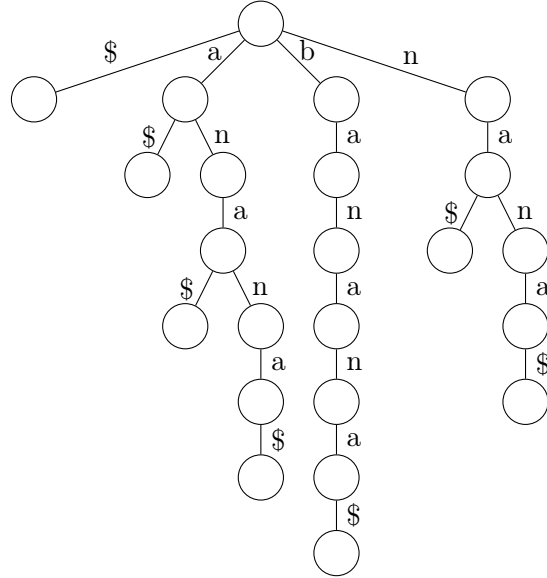


Figure 1.1: The suffix trie for the string "banana\$"

in the alphabet), the time is in  $\mathcal{O}(m \log \sigma)$ . By using a hash table and perfect hashing, the time is in  $\mathcal{O}(m)$ .

The space needed to store the suffix trie  $\mathcal{T}$  for a string of length  $n$  is in  $\mathcal{O}(n^2 \log \sigma + n^2 \log n)$  bits. The first summand is the space needed to store the  $\mathcal{O}(n^2)$  edge labels of one character  $c \in \Sigma$  each. The second summand is the space needed to store the pointers to the children of each node.

## 1.2 Suffix Trees

### Definition 1.5

A *suffix tree*  $\mathcal{T}$  for a string  $S$  is the suffix trie of  $S$  where each unary path is converted into a single edge. Those edges are labeled with the concatenation of the characters from the replaced edges. The leaves of the suffix tree store the text position where the corresponding suffix starts.

### Example 1.6

Figure 1.2 shows the suffix tree for the string "banana\$". It contains only 11 vertices compared to the 23 vertices of the suffix trie.

The suffix array can be constructed in time  $\mathcal{O}(n)$  with algorithms by WEINER[4], MCCREIGHT[2] or UKKONEN[3]. It needs  $\mathcal{O}(n \log n + n \log \sigma)$  bits. The first summand is the space needed for the pointers to the children and the indices stored in the leaves. The second summand is the space needed for the edge labels. To achieve this space, the edge

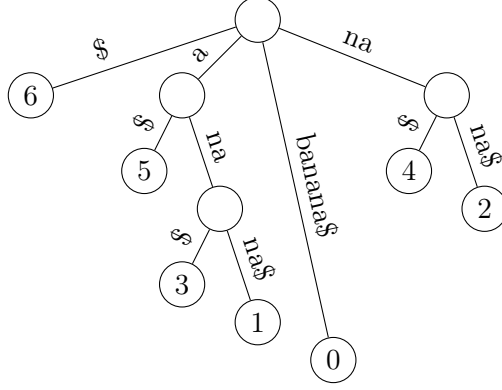


Figure 1.2: The suffix tree for the string "banana\$".

labels must not be stored explicitly. Instead we can store pointers to the first and last position of the label in the text.

In practice, a suffix tree needs more than 20 times the space of the original text. Based on the required functionality, this can even be worse.

### 1.3 Bitvectors

#### Definition 1.7

A *bitvector*  $B$  is an array of bits that are compactly stored. A bitvector supports the following operations:

- $\text{access}(i)$  returns the  $i$ -th element in  $B$ .
- $\text{rank}(i)$  returns the number of set bits in the prefix  $[0..i-1]$  of  $B$ .

#### Theorem 1.8

*The rank-Operation on bitvectors can be done constant time and  $o(n)$  additional space.*[1]

PROOF Let  $B$  be a bitvector of length  $n$ . Precompute the following information:

1. Divide  $B$  into *superblocks*  $SB_1, \dots, SB_{\lceil \frac{n}{L} \rceil}$  of size  $L$ .
2. For each superblock  $SB_j$  now store  $\sum_{i=0}^{(j-1)L-1} B[i]$ . This is the number of set bits in the first  $j$  superblocks. For each superblock this needs  $\mathcal{O}(\log n)$  bits of space.
3. Now further divide each superblock  $B_j$  into blocks  $B_1^j, \dots, B_{\lceil \frac{L}{S} \rceil}^j$  of size  $S$ .
4. For each block  $B_k^j$  of superblock  $SB_j$  store  $\sum_{i=(j-1)L}^{(j-1)L+kS-1} B[i]$ . This is the number of set bits in the first  $k$  blocks of superblock  $SB_j$ . For each block this needs  $\mathcal{O}(\log L)$  bits.

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For a given  $\text{rank}(i)$  query we now calculate the corresponding superblock  $SB_j$  and block  $B_k$ . We use the precomputed sums to get the number of set bits in all superblocks before  $SB_j$  and all blocks before  $B_k$ . All that's missing now is the number of set bits in block  $B_k$  until position  $i$ . If the blocks are small enough, this information can be precomputed efficiently for all possible blocks and positions (Four-Russians-Trick).

We still need to choose  $L$  and  $S$ :

$$L = \log^2 n \tag{1.1}$$

$$S = \frac{1}{2} \log n \tag{1.2}$$

The total amount of space needed is in  $\mathcal{O}(\frac{n}{\log n} + \frac{n \log \log n}{\log n} + \sqrt{n} \log n \log \log n)$  bits. ■

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