String Algorithms

String-Matching:

Definition

Given a text $T \in \Sigma^*$ and a string $P \in \Sigma^+$, the string matching problem is to determine all $s \in \mathbb{N}_0$, satisfying:

$$(\exists v \in \Sigma^s, w \in \Sigma^{\star})(T = vPw).$$

The number s from this definition is named shift.

A shift is called *feasible*, if P is found at the respective place in T, otherwise s is called *infeasible*.

Naïve algorithm: Try all shifts $s \in [0, |T| - |P|]$ one by one.

Worst case running time: $\mathcal{O}(|P| \cdot |T|)$. E.g. if $P = a^m$, $T = a^n$, $m, n \in \mathbb{N}$, m < n.

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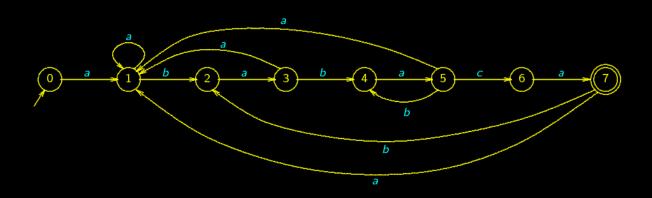
Reason of the slow running time: Knowledge about T gained in previous steps is not used. If e.g. P = aaab and s = 0 is a feasible shift, we already know that s = 1, s = 2 and s = 3 are infeasible. Thus algorithm is implemented in Java runtime library!

Efficient String Matching Algorithms

Here we only give an overview:

1) Using finite automata

Example: P = ababaca and T = abababacaba.



Fundamental definition:

Definition

The suffix function $\sigma_P: \Sigma^* \to \{0, 1, \dots, |P|\}$ of P is defined by

$$\sigma_P(X) := \max_{k \in \mathbb{N}_0} \{ k \mid P_{0,k} \sqsupset X \},$$



i.e. $\sigma_P(X)$ is the length of the longest prefix of P being a suffix of X (where $P \supseteq X$ denotes that P is a suffix of X and $P_{0,k}$ is the length k prefix of P).

Now with $\delta(q, a) := \sigma_P(P_{0,q}, a)$, $\forall q \in Q$ and $\forall a \in \Sigma$ a linear scan of the text is sufficient to find all feasible shifts.

Preprocessing: $\mathcal{O}(m^3 \cdot |\Sigma|)$ -algorithm to compute δ :

```
m:=|P|;

for q:=0 to m do begin

for a in Sigma do begin

k:=min(m+1,q+2); // P[0..k] should be

// suffix of P[0..q]+a

repeat

k:=k-1;

until (P[0..k] is suffix of (P[0..q]+a));

delta[q,a]:=k;
end;
```

- ▶ P[i..j] denotes substring $P_{i,j}$ of P,
- operator + on strings describes concatenation.

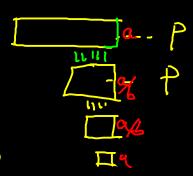
2) Knuth-Morris-Pratt Algorithm (KMP)

Definition

Let $P \in \Sigma^m$ a string. The prefix function $\Pi_P: \{1, 2, \dots, m\} \to \{0, 1, \dots, m-1\}$ of P is defined by

$$\Pi_P(q) := \max_{k \in \mathbb{N}_0} \{ k \mid k < q \land P_{0,k} \sqsupset P_{0,q} \}.$$

Example: For P = ababaca, $\Pi_P(4) = 2$ holds, since k = 2 is the maximum value for which $P_{0,k} \supseteq P_{0,4}$, k < 4, holds. This leads to the following situation:



```
m := |P|;
2
    Pi[1]:=0;
    k := 0;
4
    for q:=2 to m do begin
5
         while (k>0) and (P[k+1]<>P[q]) do
6
7
              k := Pi[k];
         if P[k+1]=P[q]
                          then
8
             k := k+1;
9
         Pi[q]:=k;
10
    end;
```

Running time: (amortized analysis using the potential method)

- ▶ Let the i-th operation Op_i be the i-th iteration of the for-loop. (Executing lines 7 through 9 yields constant cost c.).
- ▶ ⇒ Cost C_i of Op_i is c plus number of iterations of the while-loop.
- while-loop iterated often only if k is large. (Assignment in line 6 strictly decreasing). while-loop iterated often leaves k small.

Hence we choose pot(i) = k.

Amortized cost

increase of potential during O_{p_i} $C_i + pot(i) - pot(i-1)$.

To reach j iterations of the **while**-loop, $k \ge j$ is required. $\Rightarrow \ge j$ previous operations need to have gone without decreasing k during the **while**-loop but increasing k by 1 in line 8.

These operations have actual cost c, but are accounted with cost c+1 in our analysis.

(\sim Overcharging of j to account for the cost of j iterations of the while-loop).

On the other hand $C_i = c + j$ holds for the iteration, however the increase of potential is -j (k is reduced by j, thus pot(i) - pot(i-1) = -j) resp. -j + 1, if line 8 is evaluated after the loop.

Hence amortized costs are $\leq c+j-j+1=c+1$. (Here the previous overcharging and the cost of the **while**-loop are balanced, because in amortized analysis an operation including iterations of the **while**-loop is also rated with c+1 at most.)

Our discussion therefor leads to

$$C_i + \mathsf{pot}(i) - \mathsf{pot}(i-1) \le c+1 = \mathcal{O}(1).$$

Summing the amortized costs of all iterations of the **for**-loop, we get

$$\sum_{2 \leq i \leq m} \left(\mathcal{C}_i + \mathsf{pot}(i) - \mathsf{pot}(i-1)
ight) = \mathsf{total} \ \mathsf{cost} + \mathsf{pot}(m) - \mathsf{pot}(1).$$

Hence: $pot(m) - pot(1) \ge 0 \Rightarrow$ Summed amortized costs are upper bound of actual costs. This requirement is however fulfilled trivially as k never gets negative and starts with 0 in line 3. \Rightarrow Upper bound of

$$(m-1)\cdot \mathcal{O}(1)=\mathcal{O}(m)$$

for the running time of our algorithm to compute the prefix function.

Knuth-Morris-Pratt (KMP) algorithm

```
n := |T|;
    m := |P|;
    // Compute prefix function Pi here
    q := 0;
5
    for i:=1 to n do begin
6
         while (q>0) and (P[q+1]<>T[i]) do
             q := Pi[q];
7
         if P[q+1]=T[i] then q:=q+1;
8
9
         if q=m then do begin
              print('Occurrence at shift ',i-m);
10
             q := Pi[q];
11
12
         end;
13
    end;
```

Remarks:

- ▶ KMP has (optimal) running time in $\mathcal{O}(m+n)$ which can be proven by a similar analysis.
- The knowledge of Π_P makes it possible do compute δ of SMA(P) in linear time.
- Comparing the naïve method and the (optimized) KMP algorithm by dividing the expected number of comparisons both algorithms need on random texts we find

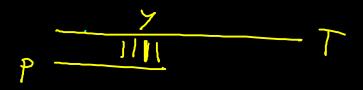
$$\mathsf{KMP/NAIVE} = 1 - \frac{1}{c} + \frac{1}{c^2} + \frac{c-1}{c^m}.$$

So if m and c are large enough both methods are almost equal.

3) The Boyer-Moore algorithm

Application: P long, Σ relatively large.

- ► Core: Naïve method: By setting s:=s+1 in lines 12 and 14 we get an implementation of the naïve method.
- ▶ **Notable:** *P* is compared to the text from right to left.
- ► **Speed-up:** In case of a mismatch two heuristics (bad character heuristic (lambda), good-suffix heuristic (gamma)) give an increment for s which does not miss a feasible shift and is usually greater than 1.



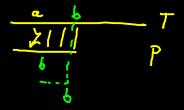
Worst case running time of the Boyer-Moore algorithm is in $\mathcal{O}((|T|-|P|+1)\cdot |P|+|\Sigma|)$ (and usually in $\Theta((|T|-|P|+1)\cdot |P|)$), as

- the computation of lambda takes $\mathcal{O}(|P| + |\Sigma|)$ time,
- the computation of gamma takes $\Theta(|P|)$ time and
- ▶ the algorithm does not use more than $\Theta(|P|)$ time on each of the at worst |T| |P| + 1 shifts.

Practise: BM often the best choice as the worst case rarely occurs and the two heuristics give relatively large increments on the considered shifts. ⇒ sublinear (in in length of text) running time. BM *faster* than optimized KMP algorithm.

4) Boyer-Moore-Horspool algorithm Variation of BM with only one heuristic similar to the bad-character heuristic. (Negative movement is avoided.) Mismatch on comparing P with $T_{i-|P|+1,i} \Rightarrow P$ is moved to the right by $d(T_i)$ positions, where

$$d(x) := \min_{1 \le k \le |P|} \{k \mid k = |P| \lor P_{|P|-k} = x\}.$$



Intuition: T_i is brought to a match with a character of P (if possible). The minimizing guarantees that no potentially feasible shift is omitted.

Running time: Worst case $\Theta(|T| \cdot |P|)$, average case (sub)linear. The constant of the linear term in the average running time is asymptotical $(|T| \to \infty)$

$$\frac{1}{|\Sigma|} + \mathcal{O}\left(\frac{1}{|\Sigma|^2}\right).$$

5)Karp-Rabin algorithm

6) Algorithm of Aho and Corasick

This algorithm finds all occurrences of a set of search terms in a text (*set matching problem*) at the same time. This is achieved by organising the strings in a *search term tree*, a directed tree satisfying the following conditions:

- ightharpoonup Each edge is labeled with a symbol from Σ.
- Edges leaving the same node are labeled with different symbols.
- ► For each search term w there is exactly one node such that the path from the root to this node is labeled with w.
- ▶ Each leave is associated with a search term.



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Searching in the text:

- \triangleright Traverse the search term tree according to the letters of T.
- Reaching a node corresponding to a search term means we have found this term.
- If no outgoing vertex for the next symbol exists:
 ⇒ failure links: Link from node v to node w such that a path from the root to w is equal to the longest suffix of the path from the root to v.



- ▶ Determining these links: Refer to SMA(P), the search term tree is like a string matching automaton for a set of strings.
- ▶ **Difference:** failure links are not associated with symbols from the alphabet.
- Traversing a failure link does not consume a symbol of the text, but increase the current shift by the number of levels we went up in the tree.
- ▶ It is possible that multiple failure links are traversed in direct succession.
- If the current node is the root and there is no matching edge we stay at the root and advance to the next symbol of the text.

Suffix Trees

Idea: P appears in T, if and only if P is a prefix of a suffix of T.

Definition

Let $T \in \Sigma^n$ a text. A directed tree $B_T = (V, E)$ with root r is called a <u>simple</u> suffix tree for T, if it satisfies the following conditions:

- 1. B_T has exactly n leaves labeled with numbers 1 to n.
- 2. Every edge in B_T is labeled with a symbol from Σ .
- 3. All edges leaving an (internal) node are labeled differently.
- 4. The path from r to leaf i is labeled with $T_{i,n}$.

Method: Construction of a simple suffix tree B_T . Input: Text $T \in \Sigma^n$.

Step 1: Let $T' = T \cdot \$$, $\$ \notin \Sigma$; let $\Sigma' = \Sigma \cup \{\$\}$.

Step 2: Initialize B_T with childless root r.

Step 3: For *i* from 1 to *n* repeat:

- ▶ Traverse B_T starting at r along the path $T_{i,n} \cdot \$$ until node x, reached by symbol T_k , has no leaving edge matching T_{k+1} .
- ▶ Append to x a linear list of nodes, the corresponding edges labeled with $T_{k+1,n} \cdot \$$.
- ▶ Label the new leaf with i.

String-Matching: Deciding with running time $\Theta(|P|)$. Finding all matches: Additional effort proportional to the size of the subtree reached by P.

Problem: A simple suffix tree may have size in $\Omega(|T|^2 \cdot |\Sigma|)$.

Reason: Nodes with only one successor.

⇒ Allow each (nonempty) word as label and eliminate unary nodes. Words are represented by start- and end-position in the text.

Definition

Let $T \in \Sigma^n$ a text. A directed tree $B_T = (V, E)$ with root r is called <u>compact</u> suffix tree for T, if it satisfies the following conditions:

- 1. B_T has exactly n leaves, labeled with numbers 1 to n.
- 2. Each internal node of B_T has at least two successors.
- 3. The edges of B_T are labeled with substrings of T.
- 4. Labels of edges leaving the same node <u>start with pairwise</u> different symbols.
- 5. The path from the root to leaf i is labeled with $T_{i,n}$, $1 \le i \le n$.

Lemma

Let $T \in \Sigma^n$ a text. A compact suffix tree B_T for T has $\mathcal{O}(n)$ nodes. Labeling all edges takes $\mathcal{O}(n \log(n))$ bits.