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Časovno učinkoviti algoritmi
ujemanja nizov in metoda grobe sile

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UNIVERZITETNI ŠTIĐIJSKI PROGRAM
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Faculty of Computer and Information Science issues the following thesis:

Thesis subject:

Describe the main ideas, workings, computational complexity, and applicability of modern algorithms for solving the string matching problem. Describe their relation with the naive string matching based on the brute-force method.
Fakulteta za računalništvo in informatiko izdaja naslednjo nalogo:

Tematika naloge:

Opisite osnovne zamisli, delovanje, računsko zahtevnost in uporabnost sodobnih algoritmov za reševanje problema iskanja nizov. Opišite njihovo morebitno povezavo z naivnim iskanjem nizov z metodo grobe sile.
Minimalna zahvala je majhen način, da izrazim izjemno hvaležnost in ljubezen do moje družine in vseh mojih prijateljev, ki so me podpirali že od prvega začetka mojega akademskega potovanja. Nenazadnje bi se še posebej zahvalila mojem mentorju, ki me je vodil pri sestavljanju te diplomske naloge v odlično delo.
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<tr>
<td>ebook</td>
<td>electronic book</td>
<td>elektronska knjiga</td>
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<tr>
<td>regex</td>
<td>regular expressions</td>
<td>regularni izrazi</td>
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<td>KMP</td>
<td>Knuth-Morris-Pratt</td>
<td>Knuth-Morris-Pratt</td>
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<tr>
<td>2D</td>
<td>two dimensional</td>
<td>dvodimenzionalna</td>
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<tr>
<td>BCR</td>
<td>bad character rule</td>
<td>pravilo slabega znaka</td>
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<td>GSR</td>
<td>good suffix rule</td>
<td>pravilo dobre pripone</td>
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<td>RK</td>
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Abstract

Title: Time-Efficient String Matching Algorithms and the Brute-Force Method

Author: Lina Lumburovska

One of the most researched areas of computer science is the string matching problem. In everyday life, people read, write, and encounter character strings all the time. Very often they want to find substrings (e.g. words) that match parts of the original text and have higher probability of matching. Finding a new efficient algorithm for the String Matching Problem involves a tremendous number of testing, just to slightly improve on the existing algorithms. In this, the algorithm based on the Brute-Force Method is of considerable help, as many current algorithms are founded on it.

My bachelor thesis explores different algorithms for the String Matching Problem and comes to a conclusion that each such algorithm has advantages and disadvantages, and is suitable for solving a particular version of the String Matching Problem and type of situations. Nevertheless, the most used algorithm in practice is the Knuth-Morris-Pratt algorithm.

Keywords: string, matching, algorithm.
Povzetek

Naslov: Časovno učinkoviti algoritmi ujemanja nizov in metoda grobe sile

Avtor: Lina Lumburovska

Eno izmed najbolj raziskanih področij na področju računalništva je problem ujemanja nizov. V vsakodnevnem življenjem ljudje ves čas berejo, pišijo, srečujejo nize in pogosto želijo najti nekaj podnizov ali besed, ki se ujemajo z izvirnim besedilom in imajo večjo verjetnost ujemanja. Razvoj mnogih algoritmov za problem ujemanja nizov zahteva ogromno preizkušanja, če želimo le malenkostno izboljšati kak obstoječi algoritem. Pri tem nam je velikokrat v pomoč algoritem, ki deluje po metodi grobe sile, saj na njem temelji veliko novejših algoritmov.


Ključne besede: niz, ujemanje, algoritem.
Daljši povzetek

Naslov: Časovno učinkoviti algoritmi ujemanja nizov in metoda grobe sile

Avtor: Lina Lumburovska


Najpogosteje uporabljen algoritem je metoda grobe sile za iskanje podnizov, ki jo pogosto imenujemo tudi naivni algoritem. Za skoraj vsak računalniški problem obstaja pripadajoča metoda grobe sile, ki pa običajno ni najbolj učinkovita. V primeru iskanja in ujemanja nizov ta metoda preverja prav vsak možen položaj vzorca znova besedila. To pomeni preverjanje vsakega položaja v besedilu, na katerem se vzorec lahko ujema. Ker ni nujno, da bo

Naslednji algoritem, ki sem ga raziskala, je Knuth-Morris-Prattov algoritem. Osnovna ideja tega algoritma temelji na naslednji predpostavki: kadarkoli je zaznano neujemanje (znak vzorca ne sovpa z znakom teksta), so nekateri znaki v besedilu že znani, saj so se pred neujemanjem ujemali z nekaterimi vzorc ali nobenim od njih. To informacijo algoritem uporabi, da ne ponavlja preverjanja in s tem zmanjša časovno zahtevnost. Razlika med naivnim algoritmom in Knuth-Morris-Prattovim algoritmom je v tem, da se naivni algoritem vedno vrne na začetek in primerjanje začne pri prvem indeksu, kar pomeni, da se ne izogiba ponavljanju primerjanja. Knuth-Morris-Prattov algoritem je najbolj znan algoritem zaradi svoje linearne časovne zahtevnosti tudi v najslabšem primeru.

Rabin in Karp sta odkrila popolnoma drugačen pristop k iskanju podnizov, ki uporablja zgoščevalne funkcije. Zgoščevalna funkcija je metoda, ki se uporablja za preslikavanje podatkov poljubne velikosti v podatke fiksne velikosti. Funkcija vrne vrednosti, ki so včasih imenovane tudi vrednosti zgoščevalne funkcije ali kar zgoščevalne vrednosti. Funkcije se pogosto zamenjuje s pojmi kot so: prstni odtisi, kontrolne vsote, kontrolna števila, popravki napak ipd., kar poveča uporabljenoč v praksi. Različni problemi na splošno uporabljajo različne zgoščevalne funkcije. Za problem ujemanja nizov je, taka funkcija zgrajena in namenjena podnizom.


Knuth-Morris-Prattov algoritem se v praksi izkaže kot najboljši, na drugem mestu pa mu sledi Boyer-Mooreov algoritem. Glede na izbrana merila, omejene računalniške vire in jasno definiran problem, vedno lahko določimo tistega med znanimi algoritm, ki bo najučinkovitejši.

Ključne besede: niz, ujemanje, algoritem.
Chapter 1

Introduction

In computer science, in the field of algorithms, there are different criteria for dividing algorithms into groups. The main criterion is the kind of computational problem to be solved. Besides this leading division, there is a sub-division inside every group. This can be based on different kind of solutions for the problem, different time and space complexities, different quality of the solutions, more suitable examples they are solving etc.

One of the most useful groups of algorithms are the string matching algorithms. Those are algorithms for searching similar or equal strings of characters. Its dominant usage in practice is plagiarism. In my bachelor thesis, I will describe highly sophisticated string matching algorithms and explain/argue why they are better than the older ones.

Human communication [20] involves exchanging of strings of characters. Accordingly, numerous important and familiar applications are based on processing of character strings. Below are listed some well-known practical examples whose results are similar or equal strings:

- searching web pages that contain a given key word (Here, browser returns a list of matches based on the key word);
- sending a text message, email or downloading an ebook (Here, transmitting a string is transmitted from one to another place. This process is called communication system and applications that process strings
for this purpose were an original for the development of string matching algorithms.

- translating a program (Here, translators such as compilers, interpreters and assemblers convert programs from one form into another. Programming systems use highly sophisticated string processing techniques, where strings were used to construct formal written languages and alphabets.)
Chapter 2

Basic Concepts

2.1 Strings and Alphabets

A string [14, 18, 19, 20] is a sequence of characters, which is used as a constant or variable of some type in programming languages. It is considered as a data type and sometimes implemented as an array of bytes data structure. Each element of the array is in most of the cases a simple character. Depending on the programming language, strings can be statically allocated (with a fixed maximum length) or dynamically allocated (with variable number of characters).

An alphabet has a finite number of elements, called symbols, and cannot be empty. Let Σ be an alphabet. A string (sometimes it is called word) is a sequence of characters from Σ. For example, if Σ={0,1}, a string is any sequence of zeros and ones. Therefore, 01001 is a string over the alphabet {0,1}. The number of all symbols in a given strings is called the length of the string. The length of a string is a non-negative integer, and is denoted by |s|. The empty string has zero elements and is usually denoted by ε.
2.2 String Concatenation and Kleene Closure

The string concatenation is the operation which joins two or more strings into one string by attaching one string to the end of another string. Here is a simple example: if $u = \text{tree}$ and $v = \text{house}$, then the concatenated string $uv$ is the $uv = \text{treehouse}$. Also, the string housetree is a string concatenation of the same words, but in different order $vu$. Note that $|uv| = |u| + |v|$. The concatenation $S_1S_2$ of two sets of strings, $S_1$ and $S_2$, is another set of strings and is defined by $S_1S_2 = \{uv : u \in S_1, v \in S_2\}$. Here, $uv$ is concatenation of strings $u$ and $v$.

The Kleene closure of an alphabet $\Sigma$ is the set of all strings over $\Sigma$, and it is denoted by $\Sigma^*$. Kleene closure of a set is a countably infinite set in which each string has a finite length. Formally,

$$\Sigma^* = \bigcup_{n \in \mathbb{N} \cup \{0\}} \Sigma^n$$  \hspace{1cm} (2.1)

Due to the fact that Kleene closure never has zero elements, it contains only an empty string if and only if the length of the alphabet is zero. For instance, if $\Sigma = \{0,1\}$, then

$\Sigma^* = \{\epsilon, 0, 1, 00, 01, 10, 11, 000, 001, 010, 011, 100, 101, 110, 111, 0000, 00001, \ldots\}$.

A formal language over an alphabet is any subset of the $\Sigma^*$.

2.3 Substrings

A contiguous sequence of characters within a string is a substring of that string [15, 18, 19, 20]. For example, the sequence the best of is a substring of the string It was the best of me. But note that, Itwasme is a subsequence of the same string, although not its substring. A substring $\hat{T}$ is called a factor of a string $T = t_1 \ldots t_n$ if $\hat{T} = t_{1+i} \ldots t_{m+i}$, where $0 \leq i$ and $m + i \leq n$. Substrings are used as patterns in string searching or string matching algorithms.

A string of length $n > 0$ has $\frac{n(n+1)}{2}$ non-empty substrings.
Often used terms when dealing with strings, substrings and string matching algorithms are also the following:

- **prefix**: a prefix of a string \( T = t_1 \ldots t_n \) is any string \( \hat{T} = t_1 \ldots t_m \), where \( m \leq n \).

- **suffix**: a suffix of a string \( T = t_1 \ldots t_n \) is any string \( \hat{T} = t_m \ldots t_n \), where \( m \leq n \).

- **rotation of a string**: rotation of \( \epsilon \) is \( \epsilon \), while rotation of \( au \) is \( ua \), where \( u \) is a string and \( a \) is a symbol.

- **reverse of a string**: reverse of \( a \) is \( a \), while reverse of \( uv \) is reverse of \( v \) concatenated with reverse of \( u \) (i.e. \( a^R = a \), \( (uv)^R = v^Ru^R \)).

The empty string \( \epsilon \) is both prefix and suffix of every string.

For instance, the string \( x = abbecca \) has a prefix \( w = ab \) and suffix \( y = cca \) (\( x = wy \)). It is useful to note that for any strings \( x, y \) and any character \( a \), \( y \) is a suffix of \( x \) if and only if \( ya \) is a suffix of \( xa \). Similarly for a prefix. This leads to the conclusion that the relations, "to be a prefix of" and "to be a suffix of" are transitive. This is proven by the following lemma.

**Lemma 2.1** (*Overlapping suffix lemma*) Suppose that \( x, y \) and \( z \) are strings and \( x \) and \( y \) are both suffixes of \( z \). If \( |x| \leq |y| \), then string \( x \) is a suffix of string \( y \); if \( |x| \geq |y| \) then \( y \) is a suffix of \( x \); and if \( |x| = |y| \) then \( x = y \).

*Proof.* We give a graphical proof. Suppose that \( x \) and \( y \) are suffixes of \( z \). The three parts of the figure describe the three cases of the lemma; part a) when \( |x| \leq |y| \); part b) when \( |x| \geq |y| \); and part c) when \( |x| = |y| \).

\( \square \)
Figure 2.1: Graphical proof of Lemma 2.1
Chapter 3

The String Matching Problem and its Algorithms

A frequently arising problem in text-editing programs is finding all occurrences of a pattern in a text. Typically, the text represents a document being edited and the pattern is a word given by the user. Well-designed and efficient algorithms for this problem can significantly improve the responsiveness of the text-editing programs.

As we will see, the string matching problem has been researched thoroughly and there were many attempts during the history to find better algorithms for this problem in terms of smaller time and space complexity.
3.1 Definition of the Problem

We assume that the **text** is an array with \( n \) elements, \( T[1, 2, ..., n] \), and that the **pattern** is again an array with \( m \) elements, \( P[1, 2, ..., m] \), where \( m \leq n \). Let \( \Sigma \) be an alphabet, and let all characters of the text and the pattern be elements of this alphabet. The alphabet can contain characters such as binary numbers 0, 1; letters \( a, b, c, ..., z \); special characters \( *, ?; \) numbers of other number systems etc. The arrays \( P \) and \( T \) are of course called strings (over \( \Sigma \)).

The pattern \( P \) takes places/occurs with shift \( s \) in the text \( T \) if and only if there is an \( s \) such that \( 0 \leq s \leq n - m \) and \( T[s + 1...s + m] = P[1...m] \). This can only happen if \( T[s + j] = P[j] \) for \( 1 \leq j \leq m \). In other words, the pattern \( P \) occurs (beginning) at the position \( s + 1 \) in the text \( T \). If the pattern \( P \) occurs with the shift \( s \) in the text \( T \), then we say that \( s \) is a valid shift. Based on this terminology, the string matching problem and the associated algorithms have been defined and designed. The string matching

![Image](image-url)
problem [13, 18, 19, 20] is to find all valid shifts with which a given pattern \( P \) occurs in a given text \( T \). For example, in Figure 3.2, \( T = ABCABAACAB \), \( P = ABAA \), and shift is three because there are three characters from the beginning of the text where \( P \) does not match \( T \).

![STRING MATCHING PROBLEM](image)

Figure 3.2: Graphical presentation of the string-matching problem

The string matching problem is also called the string searching problem because, in computer science, besides matching a given pattern, also searching substrings, prefixes, suffixes, and subsequences within a large text is often performed. Thus, the problem is not to search/match only one string, but to search/match many (or all) of them. Nevertheless, in most of the cases the two definitions are used as synonyms.

Different algorithms are used for solving the string matching problem. There are older and there are newer highly sophisticated algorithms, which will be described separately in the next chapters. How quickly a particular algorithm solves the problem also depends on how the strings are encoded. Basically, if a variable-length encoding is used, it is harder and more time consuming to find the \( N \)-th occurrence of a pattern or a character; this can reflect in a relatively slow searching algorithm.
3.2 Usage and Examples

Often a large text $T$ is called the "haystack"; in this case the pattern $P$ is called the "needle". For example, in the following sentence, by using a pattern $P$ equal to the word "to", the goal is to find several (or all) occurrences of the needle $P$ within the haystack $T$. The text/sentence is $T = "To be or not to be is a question that requires you to think"$. Certain searches might request the first occurrence of the needle, which is the first word in the haystack. A different search might request all occurrences, which are three. Yet another search might ask for the last occurrence of the needle in the haystack, which is the second word from the end.

When a pattern contains more than one word, we often use a process called normalization. In the above example, if we take the pattern "to be", the normalization enables a string matching algorithm to succeed even if there is something between "to" and "be", such as:

- more than one space;
- line-breaks, tabs and non-breaking spaces;
- hyphen;
- tags, footnotes, list-numbers, embedded images, and so on.

String matching algorithms generally must succeed in all these cases.

Moreover, sometimes there are characters that should be treated as synonymous. Such characters are usually used in the following systems/alphabets:

- In Latin-based alphabet, where it is expected that string matching algorithms will ignore difference between lower-case and upper-case letter (although in this system they are distinguished);
- Languages that contain ligatures, where one complex character contains two or more simple characters joined together (e.g. æ);
• Writing systems that contain diacritical marks like language accents or vowel points;

• Languages with specific rules such as having different characters or forms of characters where they appear at the start, within or the end of a word; (e.g. Arabic or Hebrew)

• Strings which represent natural languages or aspects of these languages. Finding all occurrences of a string despite it having alternate spellings, prefixes or suffixes, etc.

![Google search example](image)

Figure 3.3: Example where string matching algorithms must match the case, such as a mistake while google searching.

In addition to the above examples, the **regular expressions** are an important example where string matching algorithms are used. A sequence of characters that define a pattern is a strict definition for regex.

The question mark ("?") is used for zero or one occurrences of the preceding element. For example, in order to catch both American and English versions of the word colour/color, we may use a regular expression such as "colou?r", in which the question mark means zero occurrences of u (if it is an American version) or one occurrence of u (if it is an English version).

The asterisk indicates zero or more occurrences of the preceding element, such as the regex ab*cd matches acd, abcd, abbcd, etc. The plus sign indicates one or more occurrences of the preceding element such as the regex
ab+cd matches abcd or abbed but compared with the asterisk it does not match acd, because in it, the letter b has to appear at least once.
Chapter 4

A Short History and the Most Known String Matching Algorithms

String matching algorithms have an interesting history, which we here summarize to help placing the various methods in the right perspective [18, 19, 20].

The most frequently used algorithm for substring search has always been the simple brute-force algorithm, which is also known as the naive string searching algorithm. Almost every computational problem has a corresponding brute-force algorithm that solves it and usually this algorithm is not the most efficient one. Interestingly, the brute-force string matching algorithm is more or less efficient, except for the pathological cases where the brute-force algorithm becomes too slow. Namely, in worst case scenarios, the time complexity can rise to $O(m*n)$ where $m$ is pattern’s length and $n$ is text’s length. (The ”usual” running time is $O(m+n)$, which is consider less than $O(m*n)$.) Furthermore, the brute-force string matching algorithm is well-suited to standard architectural features of most computer systems, so an optimized version of it can provide a standard benchmark, which is almost unbreakable even with much more intelligent algorithm.
In 1970s, Stephen Cook published a theoretical solution for a specific type of an abstract machine which implied the existence of an algorithm capable of solving the string-matching problem with worst-case time complexity $O(m+n)$. Donald Ervin Knuth and Vaughan Pratt laboriously followed through the Cook’s construction, which was not intended to be practical. They transformed Cook’s theorem into a relatively simple and practical algorithm. The algorithm was a rare and satisfying example of a theoretical result with an immediate practical applicability. Unexpectedly, it turned out that James Hiran Morris had already discovered the same (equivalent) algorithm as a solution to an annoying problem that arose when he was implementing a text editor. The fact that the same algorithm is useful for solving two different problems, indicated that there was something fundamental to this algorithm.

Knuth, Morris and Pratt did not publish their research on the algorithm until 1976. In the meantime, two computer scientists, Robert S. Boyer and Jeffrey S. Moore, constructed a string matching algorithm that was faster in many applications, since it often examined only a fraction of the characters in the text string. Due to this algorithm, many text editing programs achieved noticeable speedups in substring search.

Eventually, it turned out that there are two different algorithms which solve the same problem. Both Knuth-Morris-Pratt and Boyer-Moore algorithm require additional time for preprocessing of the pattern. Unfortunately, the preprocessing is hard to understand and has limited usefulness. In fact, an unknown system programmer declared Morris’s algorithm too difficult to understand and simply replaced it with the brute-force algorithm.

In 1980s, Michael O. Rabin and Richard M. Karp used hashing to solve the string matching problem and developed an algorithm that is almost as simple as the brute-force algorithm but has time complexity $O(m+n)$. Moreover, their algorithm can be used for two-dimensional patterns and texts, which makes it suitable for image processing.
This is only a brief history with only the most significant events of course. There have been many unsuccessful attempts. In the next section, the most important string matching algorithms will be listed, and each of them will be presented separately in the following chapters.

4.1 Time Complexity

During the history, several parameters had been researched in order to set a criterium for determining the best algorithm for a given computational problem. In the end, it turned out that the most useful and appropriate criterium is the algorithm’s running time. In the field of algorithms it is known as the algorithm’s time complexity, i.e. the amount of time an algorithm needs to process an instance of given computational problem.

We distinguish three cases of algorithm execution:

- best case (min. time needed for algorithm’s execution; notation: Ω)
- average case (average time needed for algorithm’s execution);
- worst case (max. time needed for algorithm’s execution; notation: O).

4.2 Algorithms

The most known algorithms for solving the string matching problem are:

- Brute-force algorithm (known as the naive algorithm),
- Knuth-Morris-Pratt algorithm,
- Boyer-Moore algorithm,
- Tuned Boyer-Moore algorithm,
- Rabin-Karp algorithm,
- Two-way string matching algorithm,
• String matching with finite automata,

• Quick-skip searching,

• Fast hybrid algorithm,

• Bitap algorithm (known as Baeza-Yates-Gonnet algorithm),

• BNDM (Backward Non-Deterministic Dawg Matching),

• Commentz-Walter algorithm,

• Aho–Corasick algorithm,

• Boyer–Moore–Horspool algorithm,

• Raita algorithm,

• BOM (Backward Oracle Matching),

• Apostolico–Giancarlo algorithm.
Chapter 5

The Brute-Force (Naive) Algorithm

An obvious method [18, 19, 20] for searching and matching substrings is to check every possible position of the pattern within the text. This means checking each position in the text at which the pattern might match. However, it is not necessary that a position is matchable, so the algorithm returns one of the two possible logical values: true or false. If there is a matching at a position, it returns true; otherwise, it returns false.

In my bachelor thesis, I will use pseudocode. This is an informal high-level description of the operating principle of an algorithm or computer program. Pseudocodes provide the easiest way to describe a method even for readers that have no knowledge of computer programming. Besides pseudocode, I will describe the most important algorithms in Java, one of the most used programming languages. In such cases, I will present some functions which already exist and describe how they are used.

Clearly, the naive algorithm for string matching accepts a pattern $P$ and a text $T$ as two arguments. The pseudocode for the naive string matching algorithm is presented below.
THE NAIVE STRING-MATCHING ALGORITHM \((P, T)\)

\[ m = \text{length}(P) \]
\[ n = \text{length}(T) \]

\text{for } s = 0 \text{ to } n-m
\hspace{1em} \text{if } P[1,\ldots,m] = T[s+1,\ldots,s+m]
\hspace{1em} \text{then print "The pattern occurs with shift" } s

The variable \( m \) is the length of the pattern \( P \) and the variable \( n \) is the length of the text \( T \).

The naive algorithm only uses one loop that checks the condition \( P[1,\ldots,M] = T[s + 1,\ldots, s + M] \) for each of the possible values of the shift \( s \) which shows from which position the pattern matches the text.

The algorithm can also be graphically described as sliding the pattern over the text. Using this, it can easily be seen at which shifts the pattern matches the corresponding characters in the text. This is depicted in Figure 5.1, where the body of the for loop considers each possible shift explicitly. The "if condition" checks whether the current shift is valid or not, and performs an implicit loop (which checks the characters of the pattern until either all positions match or there is a mismatch). Each valid shift is printed out by the last line (so there is an option to print more than one valid shift).

In the example presented in Figure 5.1, the pattern is \( P = aab \) and the text is \( T = acacaabc \). The pattern is in each picture shifted by one character to the right. Vertical lines in each picture connect the corresponding characters that have been found to match. In contrast, jagged lines connect the first mismatched character found, if any.
A simpler example using words from everyday life is given in Figure 5.2, where green color is used for matching characters and red for the mismatching ones. This example has only one occurrence of the pattern *simple* in the text *This is a simple example.*

String matching algorithms are intended to find all occurrences of a pattern. This is depicted in Figure 5.3, where the pattern *AABA* is found three times in the given text.

The worst case input is when both, the pattern and the text have the same form (for example, $P = a^m$ and $T = a^n$), because it is necessary to check $m$ characters $n - m + 1$-times. This leads to the worst case running time $O((n-m+1)m)$, and this upper bound is tight. Fortunately, such scenarios hardly ever occurs. Most of the strings find a mismatch at the first character of the pattern. That is why, in practice, the average running time
Text : A A B A A C A A D A A B A A B A

Pattern : A A B A

A A B A A A B A

A A B A A A B A A A B A

Pattern Found at 0, 9 and 12

Figure 5.3: An example where more than one occurrence of the pattern is found

is more appropriate for estimating the running time. The naive algorithm does not need preprocessing because due to its simplicity there is nothing to be prepared before the algorithm starts. The naive algorithm is not optimal; indeed, so it has disadvantages, such as ignoring the information gained about the text for one value of the shift $s$ on considering other values of $s$.

In Java, the function $charAt$ is used to point at a specific character or position. Using this function, a comparison between the current characters of the pattern and the text is performed. If there is a mismatch, the pattern is shifted to the next character; otherwise, it returns the suitable shift of the pattern matching.
Chapter 6

Knuth-Morris-Pratt Algorithm

The basic idea of this algorithm, discovered by Knuth, Morris and Pratt [9, 8, 18, 19, 20], stems from the following assumption: whenever a mismatch is detected, some of the characters in the text are already known, since they matched some or none of the pattern characters prior to the mismatch. This information is used as an advantage to avoid backing up the text pointer over all those known characters.

Before presenting the pseudocode, we give an example. For observation purposes we will allow the text to be $T = ABAAAAABAAAAA$, the searching pattern to be $P = BAAAAA$ and the alphabet to have only two characters, $A$ and $B$. Suppose that there is a match on the first five characters of the pattern and mismatch on the sixth one. When the mismatch is detected, it is known that sixth character is not $A$ but $B$. Text pointer is now pointing at $B$. The key observation is that there is no need to back up the text pointer $i$, since it is already known that the previous characters are $A$s and do not match the first character of the pattern. Knowing the information about the currently pointed value, the index can be easily incremented by one and compared to the next character of the pattern. The difference between the brute-force algorithm and KMP is that the brute-force algorithm always returns to the beginning and starts comparing from the first index. Since KMP algorithm remembers the text pointer, there is no need to repeat it
all the time, and this results in a shorter running time and better time complexity. The value of the text pointer \( i \) does not decrement and does not change within the for loop and this method performs at most \( n \) character comparisons.

It is important to note an exception when KMP algorithm does not work properly. This is when the pattern could match itself at any position overlapping the point of the mismatch. For example, when searching for the pattern \( P = AABAAA \) in the text \( T = AABAABAAAA \), first the mismatch is detected on the fifth position, but there is a better restart at the third position to continue the search, in order to miss the mismatch. The insight of the KMP algorithm is that the user of the algorithm can decide ahead of time exactly how to miss the mismatch and restart the search. Otherwise, the algorithm will make an error and will not return the correct value.

Part of the pseudocode is the computation of the prefix function \( \pi \)

![Constructing prefix table for Pattern ABCDABD](image)

Figure 6.1: Implementation for a random pattern, how its prefix function is computed

for a pattern \( P \). The function \( \pi \) encapsulates the knowledge about how \( P \) matches against shifts or itself. This information is used (as described in the previous paragraphs), to avoid testing useless shifts as in the naive string matching algorithm. The mathematical analysis of the correctness of the prefix function \( \pi \) is not of big importance in my thesis, and therefore I will omit it. A simple example of how the prefix function is computed for the
pattern $P = ABCDABD$ is presented in Figure 6.1, where it is demonstrated through six iterations how the longest prefix is determined and what is its length.

To sum up, given a pattern $P[1,\ldots,m]$, the prefix function for the pattern $P$ is the function $\pi : \{1,2,\ldots,m\} = \{0,1,\ldots,m-1\}$ such that $\pi[q] = \max\{k : k \leq q \text{ and } P_k \text{ is a proper suffix of } P_q\}$. Informally, $\pi[q]$ is the length of the longest prefix of $P$ which is $P_k$ and $q$ is the current character in the text $T$. Below is the description in pseudocode:

**KNUTH-MORRIS-PRATT ALGORITHM (P,T)**

```plaintext
m=length(P)
N=length(T)
pi=compute prefix function(P)
q=0

for i=1 to n
    do while q>0 and P[q+1] != T[i]
        do q=pi[q]
        if P[q+1] = T[i]
            then q=q+1
        if q=m
            then print "The pattern occurs with shift" i-m
            q=pi[q]
```

The prefix function in pseudocode is:

**COMPUTE PREFIX FUNCTION(P)**

```plaintext
m = length(P)
pi[1] = 0
k = 0
```
for q = 2 to m
    do while k > 0 and P[k+1] != P[q]
        do k = pi[k]
        if P[k+1] = P[q]
            then k = k+1
            pi[q] = k
    return pi

By using this algorithm and the prefix function, the string matching problem is solved in a way which gives better results than the naive algorithm. The running time of the compute prefix function is $\Theta(m)$, which is determined by the potential method of amortized analysis (such as using the value of $k$ to find the valuable information from mismatching). In order to determine the running time of the whole algorithm, the same method is used (i.e. the method of amortized analysis), but instead of $k$ the value of $q$ is applied in the KMP-algorithm. This leads to a linear time complexity, $\Theta(n)$. The most significant property of the KMP algorithm is that besides its average running time also the worst case time is linear. That makes KMP algorithm one of the highly sophisticated string matching algorithms which are often used in practice.

6.1 KMP Algorithm with Finite Automation

A finite automation is used in the same way as an algorithm which, given a pattern, must find the pattern in the text. A finite automation consists of a starting state $q_0$, a finite set of states $Q$, a set of accepting states $A$, an input alphabet $\Sigma$, and a transition function $\delta$ that maps $Q \times \Sigma$ to $Q$. On what follows, I will not be explaining finite automata; instead I will describe the main idea of their use.

For each pattern $P$ there exists an automation which must be constructed from the pattern in the preprocessing step. The automation is then used to search the pattern in the text.
To specify the string matching automation which corresponds to a given pattern $P[1,\ldots,m]$, we use an auxiliary function $\sigma$, called the suffix function. This function maps from $\Sigma^*$ to $\{0,1,\ldots,m\}$ so that $\sigma(x)$ is the length of the longest prefix of $P$ that is a suffix of $x$:

$$\sigma(x) = \max \{ k : P_k \text{ is a suffix of } x \}$$

The fact that the empty string $\epsilon$ is a suffix of every string makes the function $\sigma$ well-defined: for a pattern $P$ with length $m$, we have $\sigma(x)=m$ if and only if $P$ is a suffix of $x$.

It follows from the definition of the function $\sigma$ that if $x$ is a suffix of $y$, then $\sigma(x) \leq \sigma(y)$. Based on this, the string matching automation that corresponds to a given pattern $P$ can be constructed.

The scalability and importance of the suffix function within this topic is seen from the following lemma.

**Lemma 6.1 (Suffix function inequality lemma)** For any string $x$ and character $a$, the following holds $\sigma(xa) \leq \sigma(x) + 1$.

**Proof.** Let variable $r = \sigma(xa)$. If $r = 0$, then the conclusion $\sigma(xa) \leq \sigma(x) + 1$ is trivially satisfied, by the nonnegativity of $\sigma(x)$. So, assume that $r \geq 1$. Now $P_r$ ($P$ is a pattern) is a suffix of $xa$, by the definition of $\sigma$. Thus, $P_{r-1}$ is a suffix of $x$, by dropping the $a$ from the end of $P_r$ and from the end of $xa$. Therefore, $r - 1 \leq \sigma(x)$, since $\sigma(x)$ is the largest $k$, such that $P_k$ is a suffix of $x$ and $\sigma(xa) \leq \sigma(x) + 1$. \[\Box\]

The pseudocode which clarifies the operation of the string matching automation, represented by $\delta$, is given below. The pseudocode does exactly the same as the ordinary KMP algorithm, i.e. finds occurrences of a given pattern within a given text $T$. 


KNUTH-MORRIS-PRATT AUTOMATION \((T, \delta, m)\)

\[
n = \text{length}(T) \\
q = 0 \\

\text{for } i = 1 \text{ to } n \text{ do } q = \delta(q, T[i]) \\
\quad \text{if } q = m \text{ then print } "\text{Suitable pattern occurs with shift } i-m" 
\]

In this type of substring searching, usually the preprocessing time is not included. In other words, only the matching time is computed and it is \(\Theta(n)\).
Chapter 7

Boyer-Moore Algorithm

An efficient string matching algorithm that is the standard benchmark for practical string-search literature is the Boyer-Moore algorithm [4, 18, 19, 20]. The algorithm is mainly used in computer science. In general, it runs faster if the pattern is longer - a feature that is extremely rare in the field of algorithms. The reason for that is that the algorithm seems to match on the tail of the pattern rather than the head, and to skip along the text in jumps of multiple characters rather than searching every single character in the text.

A significantly faster string searching method is to scan the pattern from right to left when trying to match it against the text. This can be easily seen on the example, where $P = BAABBA$, assuming there are matches on the sixth and seventh character, but not on the fifth. Notice that the pattern can be immediately slided seven positions to the right and can move on checking the fourteenth character in the text. This can be done because the partial match $XAA$, where $X$ is not equal to $B$, does not appear elsewhere in the pattern. The pattern could appear somewhere in the text later, so there is a need to remember the position to restart the searching method if there is a need.

Another example of Boyer-Moore shifting and searching, is shown in Figure 7.1 where the text (or often called "haystack") is
$T = \text{FINDINAHAYSTACKNEEDLE}$ and the "needle" (or the pattern) is $P = \text{NEEDLE}$. The comparison starts from the rightmost character E and it is compared with character N on the fifth position in the text. Since N appears in the pattern, the pattern is slided five positions to the right to line up the N in the text with the rightmost N in the pattern. The next comparison is made with the rightmost E in the pattern with the character S on the tenth position in the text. The difference with the mismatched S is because S does not appear in the pattern so the pattern is slided six positions to the right. The rightmost E in the pattern is matched with the character on position 16, where a mismatch exists. N at position 15 is discovered and the pattern is slid four positions. Finally, moving from right to left in the text at position 20, the searched pattern is discovered within the text. The algorithm uses only four comparisons to find the suitable pattern, which is much better than with the naive algorithm.

One of the main reasons for the popularity of the Boyer-Moore algorithm is in its preprocessing. The algorithm is suitable for applications when the pattern is much shorter than the text. The condition is basically fulfilled in almost every searching case, which is why the algorithm often is so used in practice and theory.

There are two variants to perform shifting and preprocessing based on rules called: the Bad Character Rule (BCR) and the Good Suffix Rule (GSR).

![Figure 7.1: Simple example for Boyer-Moore algorithm](image)
Each rule has its own advantages and disadvantages, so their names “good” and “bad”, do not mean that the one is actually better than the other.

- The Bad Character Rule (BCR)

**Definition 7.0.1** *This rule reviews the character in the text T where comparison failed. When the next occurrence of the character to the left is found in the pattern P, then a shift that brings that occurrence in line with the mismatched occurrence in T is proposed. If the mismatched character does not occur to the left in P, a new shift is proposed that moves entire P past the point of mismatch.*

![Step 1: T: GCTTCTGCTACCTTTTGCACCGCGCGGAA, P: CTTTGGTGC](image1)

![Step 2: T: GCTTCTGCTACCTTTTGGCACCGCGCGGAA, P: CTTTGGTGC](image2)

![Step 3: T: GCTTCTGCTACCTTTTGGCACCGCGCGGAA, P: CTTTGGTGC](image3)

(etc)

Figure 7.2: Example of BCR, where b is a mismatched character. Skip alignments until (a) b matches its opposite in P or (b) P moves past b.

To estimate BCR's time complexity, we can use a 2D array where the first dimension is indexed by the index of the character c in the alphabet, and the second dimension is indexed by the index i from the pattern. BCR will return the occurrence c of P with the index j < i (or -1 if there is a mismatch). The time complexity of BCR is thus $O(1)$, a constant, which is in practice the best amount the searching algorithms can have. The space complexity of BCR is $O(k*n)$, where $k$ is the size (the number of characters) of the alphabet.
• The Good Suffix Rule (GSR)
This rule is the advantage of the Boyer-Moore algorithm, searching from right to left.

**Definition 7.0.2** Suppose that for a given alignment of \( P \) and \( T \), a substring \( t \) of \( T \) matches a suffix of \( P \), but a mismatch occurs at the next comparison to the left. Then find, if it exists, the rightmost copy \( t' \) of \( t \) in \( P \) such that \( t' \) is not a suffix of \( P \) and the character to the left of \( t' \) in \( P \) differs from the character to the left of \( t \) in \( P \). Shift \( P \) to the right so that substring \( t' \) in \( P \) aligns with substring \( t \) in \( T \). If \( t' \) does not exist, then shift the left end of \( P \) past the left end of \( t \) in \( T \) by the least amount so that a prefix of the shifted pattern matches a suffix of \( t \) in \( T \). If no such shift is possible, then shift \( P \) by \( x \) (\( x \) is the length of the suffix) places to the right. If an occurrence of \( P \) is found, then shift \( P \) by \( x \) places, that is, shift \( P \) past \( t \).

Figure 7.3: Example for GSR, where \( t \) is a substring of \( T \) and matches a suffix of \( P \). Skip alignments until \( t \) matches the opposite character in \( P \) (a), a prefix of \( P \) matches a suffix of \( t \) (b) or \( P \) moves past the first matching substring \( t \) (c).
To determine GSR’s time complexity of the processing, we need two arrays: \( L \) for the general case, and \( H \) for either when the general returns a meaningless result or a match occurs.

It is known that for each \( i \) in \( L \), the index \( i \) is the largest position less than \( n \) and it is proven that the pattern \( P[i, ..., n] \) matches a suffix \( P[1, ..., L[i]] \) and it is not equal to \( P[i - 1] \). If this condition is not satisfied, current index in the array is equal to zero. The index in \( H \) is determined as the length of the largest suffix in \( P \), which is at the same time a prefix of \( P \).

All of this leads to a linear time and space complexity, \( O(n) \), where \( n \) is the length of the text.

To sum up, the algorithm’s time complexity is \( O(n+m) \), if and only if the pattern does not appear in the text. However, this is not the best case scenario, because the main aim of this algorithm is to find an existing substring in the text. The worst case time complexity when the substring does appear in the text is \( O(n^2m) \). Although this is the same time complexity as the time complexity of the naive algorithm, Boyer-Moore algorithm is in practice better due to its searching from tail to head, preprocessing, omitting unnecessary comparisons, etc.

The pseudocode of the algorithm contains two functions in addition to the main one: the function for computing the good suffix and the function for computing the last occurrence. The input consists of the substring/pattern, the text, and an alphabet \( \Sigma \). Here is the pseudocode.

```plaintext
BOYER-MOORE ALGORITHM(P,T,Sigma)

m = length(P)

n = length(T)

lambda = compute last occurrence(P,m,Sigma)

gamma = compute good suffix(P,m)

s = 0
```
while \( s \leq n-m \)
do \( j = m \)
    while \( j > 0 \) and \( P[j] = T[s+j] \)
do \( j = j - 1 \)
    if \( j = 0 \)
        then print "The pattern occurs at shift" \( s \)
        \( s = s + \gamma[0] \)
    else \( s = s + \max(\gamma[j],\lambda[T[s+j]]) \)

COMPUTE LAST OCCURRENCE(\( P, m, \Sigma \))
for each \( a \) in \( \Sigma \)
do \( \lambda[a] = 0 \)
for \( j = 1 \) to \( m \)
do \( \lambda[P[j]] = j \)
return

COMPUTE GOOD SUFFIX(\( P, m \))
\( \pi = \text{compute prefix function}(P) \)
\( P' = \text{reverse}(P) \)
\( \pi' = \text{compute prefix function}(P') \)
for \( j = 0 \) to \( m \)
do \( \gamma[j] = m - \pi[m] \)
for \( l = 1 \) to \( m \)
do \( j = m - \pi'[l] \)
    if \( \gamma[j] > l - \pi'[l] \)
        then \( \gamma[j] = l - \pi'[l] \)
return
Chapter 8

Rabin-Karp Algorithm

A completely different approach to substring search which uses hashing was discovered by Rabin and Karp [11, 18, 19, 20]. Hashing is a method where a function, called the hash function, is used to map data of arbitrary size to data of a fixed size. The hash function returns values which are sometimes named as hash values, hash codes, digests, or simply hashes. (Hash functions are often confused with fingerprints, checksums, check digits, error-correction etc.; and that is why this algorithm is also known as the "Rabin-Karp fingerprint algorithm".)

Different problems use, in general, different hashing methods (functions). For the string matching problem, a hash function is constructed for the pattern. The same hash function is used for finding a match for each possible text substring of length $m$. The searching process is exactly the same as if the pattern is stored in a hash table and then it performs a search for each substring of the text. The main advantage (which has an impact on the space complexity) is that there is no need to allocate the memory for the hash table.

This process leads to a worse time complexity than the brute-force algorithm. This is due to computing the hash function which involves all characters and is more time-consuming since it only compares characters as in the naive algorithm. In spite of that, in the real world it has been shown
that computing hash functions for \( m \) characters by Rabin-Karp’s algorithm can be done in constant time and leads to a linear time substring matching in practical situations. Hash functions are the reason why this algorithm is in the group of effective string searching algorithm and has practical and theoretical usage in many cases.

A string of length \( m \) corresponds to an \( m \)-digit base-\( R \) number. For keys of this type, it is necessary to have a certain hash function which converts an \( m \)-digit base-\( R \) to an integer value from 0 to \( Q - 1 \). Modular hashing is used in more complex cases, where this process takes the remainder of dividing the number with \( Q \). Instead of the remainder in practice is mostly used a random prime number \( Q \), which is chosen in that way so the number is as large as it is possible. In such cases, it is important to avoid overflow.

We give a simple example to demonstrate algorithm’s working. In this example a small \( Q = 997 \) (a hash table size) is being used, which in real situations hardly ever happens and \( R = 10 \). The pattern \( P = 26535 \) is searched in the text \( T = 3141592653589793 \). The hash value for the pattern is \( 26535 \% 997 = 613 \), which means that iterations will be performed in the text until there is found a substring with the same value (613) and has as many characters as the pattern. In the example the substring is found in the seventh iteration (the index is six) because the first six values returned are: 508, 201, 715, 971, 442 and 929. This is presented in Figures 8.1 and 8.2.

![Figure 8.1: Simple example for computing a hash value for the pattern using RK algorithm.](image)

In the example, the number of the characters in the pattern is five and still there is no problem. Difficulties appear when the number of the characters is 100, 1000 etc. To handle such cases we often use the well-known Horner’s method. This method is often used to calculate values of polynomi-
als. In Rabin-Karp algorithm only a simple application of Horner’s method is used, in an elementary function which implements the hash function. There is only one ”for loop”, which runs over all characters of the pattern ($m$ times) and computes the hash function by the formula given in the pseudocode below.

\[
\text{HASH FUNCTION(key, m)} \\
\text{for } j = 0 \text{ to } m \\
\quad h = (R \times h + \text{key}[j]) \mod Q \\
\text{return } h
\]

Because this algorithm uses some arithmetics, one more time the accent will be on the Horner’s rule. The finite alphabet $\Sigma$ has ten elements, i.e. ten digits. Given a pattern $P[1, ..., m]$, let $p$ denote its corresponding decimal value. Similarly, for a given text $T[1, ..., n]$, let $t_s$ denote the decimal value of the length-$m$ substring $T[s + 1, ..., s + m]$, for $s = 0, 1, ..., n - m$. Certainly, $t_s = p$ if and only if $T[s + 1, ..., s + m] = P[1, ..., m]$; thus $s$ is a valid shift (and vise versa applies). If $p$ can be somehow computed in time $\Theta(m)$ and all the $t_s$ (for $s = 0, 1, \ldots, m$) values can be computed in time $\Theta(n - m + 1)$, then all valid shifts can be determined in time $\Theta(m) + \Theta(n - m + 1)$ which

Figure 8.2: Simple example for matching a pattern inside the text using RK algorithm and hash values.
is $\Theta(n)$.

By using Horner’s rule, $p$ can be computed in time $\Theta(m)$:

$$p = P[m] + 10(P[m - 1] + 10(P[m - 2] + ... + 10(P[2] + P[1])...)) \quad \text{(8.1)}$$

Each of the other values $t_1, t_2, ..., t_{n-m}$, can be computed in time $\Theta(n-m)$. Notice that $t_{s+1}$ can be computed from $t_s$ in constant time, since:

$$t_{s+1} = 10(t_s - 10^{m-1}T[s + 1]) + T[s + m + 1] \quad \text{(8.2)}$$

For example, let $m = 5$, $q = 13$, and $T = [3, 1, 4, 1, 5, 2]$. Consequently, $t_0 = 31415$. So,

$$t_1 = 10(31415 - 10^5 * T[1]) + T[5 + 1] =$$

$$= 10(31415 - 10^4 * 3) + 2$$

$$= 10(1415) + 2$$

$$= 14152$$

This example shows how Horner’s rule is used: subtracting $10^{m-1} * T[s + 1]$ removes the high-order digit from $t_s$; multiplying the result by 10 shifts the number left one position; and adding $T[s + m + 1]$ brings in the appropriate low-order digit.

The pseudocode for the RK algorithm accepts the text $T$, the searching pattern $P$, $Q$ and $d$, where $d$ is basically $|\Sigma|$ and is presented in the code below. The function $\text{power}(d,m-1)$ represents to $d^{m-1}$ and mod is the usual modulo function.

```
RABIN-KARP ALGORITHM(T,P,d,Q)
  n = length(T)
  m = length(P)
  h = power(d, m-1) * mod Q
  p = 0
  t0 = 0
```
for $i = 1$ to $m$
  do $p = (d \ast p + P[i]) \mod Q$
  $t0 = (d \ast t0 + T[i]) \mod Q$
for $s = 0$ to $n - m$
  do if $p = tS$
    then if $P[1,\ldots,m] = T[s+1,\ldots,s+m]$
      then print "Suitable pattern occurs at shift" $s$
    if $s < n - m$
      then $tS+1 = (d(tS - T[s+1] \ast h) + T[s+m+1]) \mod Q$

The first for loop is used for preprocessing and takes $\Theta(m)$ time for the first process. The second for loop is used for matching and takes $\Theta((n - m + 1)m)$ time in the worst case. If $P = a^m$ and $T = a^n$, then the verifications take time $\Theta((n - m + 1)m)$, since each possible verification is a valid shift.

In some applications, not all verifications are valid shifts. Here, the matching time of the Rabin-Karp algorithm is $O((n-m+1) + cm) = O(n+m)$, where $c$ is a constant and does not have an impact on the matching time of the algorithm. Since it is usually assumed that the length of the pattern is smaller or at most equal, but never larger that the length of the text i.e. $m \leq n$, the matching time of the RK algorithm is $O(n)$. 
Chapter 9

Fast Hybrid Algorithm

As we are living in the twenty-first century, where all the technology is moving forward faster than any other field in our everyday life, the field of algorithms is getting new updates every day. The newest algorithm for solving the string matching problem, found out the most recently, is based on the Quick-skip and Tuned Boyer-Moore algorithm and is known as the fast hybrid algorithm for string matching. Before explaining the main idea of the fast hybrid algorithm, in the following subsections, I will describe the two algorithms which fast hybrid algorithm is based on.

9.1 Quick-Skip Search

This algorithm [10] is a combination for solving two problems: the Quick Search and the Skip Search. Similarly as in the other algorithms, there are two phases in this algorithm: preprocessing and searching/matching. The characters are preprocessed in the preprocessing phase and the information obtained is used in the other phase to find the number of the comparisons and all the attempts as well.

There are two different techniques involved in the preprocessing: the first constructs the Quick Search bad character table (qsBc) and the second constructs Skip Search buckets. Both techniques are used together, because
the first one, the table, contains the rightmost location for each alphabet in the pattern, while the second one contains the leftmost location for all characters in the pattern. The information gained in the preprocessing phase is used in the matching phase to reduce the total number of comparisons and the number of all (successful and unsuccessful) attempts. Both phases go together, hand-in-hand, to improve efficiency of the algorithm by calculating larger shift values.

The searching method consists of four stages.

In the first stage, the algorithm finds the starting point $S$ with a position $T_j$ within the text. The character of this certain position is aligned with the suitable position of this character in the bucket. Even if there is a mismatch, the algorithm continues shifting the pattern to the next character in the text and avoiding mismatches, which definitely speeds up the algorithm. The algorithm becomes faster because it avoids aligning the leftmost character of the pattern and the window at the beginning of the searching phase.

In the second stage, the comparisons between the characters of the pattern and the window are executed. These comparisons start from the leftmost character of the pattern with the suitable position of the same character in the window. This stage is only intended for accomplishing number of comparisons, and whether there is a match or mismatch is done in the next stage.

In the third stage, the algorithm calculates the shift value of the Skip Search which is calculated differently depending on two situations, whether the character in the pattern occurs or does not occur in the last position of the bucket.

If it does, the shift value is calculated

$$\text{skip shift} = m + \text{the current position of } T_j \text{ (from the bucket)} - \text{the next position of } T$$

If it does not occur the shift value is calculated by subtracting the next position value from the current position value of this character in the bucket.

The Quick Search shift value for this algorithm is assigned for a character immediately next to the window.

In the last stage, the algorithm depends on the Quick Search shift. It is given
in the pseudocode below using words, as it happens in different situations.

**QUICK-SKIP SEARCH(P,T)**

m = length(P)
n = length(T)

if (Quick Shift > Skip Shift) and (Quick Search Shift <= m)
   then Current Position of Tj = Position Next to the Window
if (Quick Shift > Skip Shift) and (Quick Search Shift > m)
   then Current Position of Tj = Position Next to the Window + m

The algorithm has a worst running time $\Theta(n \times m)$ and the best running time is $O(n/m)$.

### 9.2 Tuned Boyer-Moore Algorithm

Besides the algorithm explained in the previous section, the Tuned Boyer-Moore algorithm [16] is a part of the fast hybrid algorithm. It is a better implementation of the Boyer-Moore algorithm presented in the chapter seven. The algorithm is very fast in practice. The most costly part is checking whether the character of the pattern matches the character of the window. Considering that this is a modern algorithm, the flaw can be avoided if and only if there are several unrolled shifts before actually comparing the characters. This algorithm uses bad character rule shift function and keeps on shifting until it finds three shifts in a row. The order of comparison between the characters in the pattern and in the text is not important. The worst case running time is quadratic of the length of the pattern, but it exhibits very good behavior in the real practice.

### 9.3 Definition of the Fast Hybrid Algorithm

The fast hybrid algorithm [7] basically uses all the steps from section 9.1 because it represents an update of the Quick-Skip Search. The algorithm
gives out very good results because of skipping unnecessary comparisons. The schema for the algorithm is given in the figure 9.1 below, where instead of pseudocode a flowchart is used. BM (is it written in the flowchart) is an abbreviation for Boyer-Moore algorithm, and in this case it is tuned Boyer-Moore algorithm and its influence is the only difference between the fast hybrid and the quick-skip search.

Figure 9.1: Flowchart for the fast hybrid algorithm
Chapter 10

Other Known Algorithms

All algorithms for the string matching problem were listed in the Section 4.2, but not all of them are equally used in practice, especially when resources are limited. The most effective algorithms, such as Knuth-Morris Pratt, Boyer-Moore, Rabin-Karp, and Fast hybrid algorithm, were described in detail in sections 6, 7, 8 and 9, respectively. Also, the brute force algorithm, presented in section 5, cannot always be omitted, because it is foundation of almost each newer update.

This section is intended to be a portrayal of other less-known algorithms, with an emphasis on their main ideas instead of on their inner workings. The only algorithms listed in the section 4.2 that will be omitted are: the Backward Oracle Matching algorithm and the Apostolico–Giancarlo algorithm, because they exhibit the same performances as the Boyer-Moore algorithm.

10.1 Two-Way String Matching Algorithm

This algorithm [17] is a combination of KMP and BM algorithms. As the name of the algorithm suggests, the algorithm pursues the search in two directions at the same time: from right to left and from left to right.
The running time can be much better; specifically, it is twice better than the running time of KMP algorithm, but only if there is a match. If there is a mismatch, the required time and space are doubled, because, for each comparison, there is a need of storing into two arrays, (one array for each way). But storing into more than one array, even when there is a match has a negative impact on the space complexity, and consequently slows down the algorithm. As a result the algorithm has a special usage in cases when alphabet is ordered, due to the fact that the probability of finding a match when both ways are ordered is much higher.

10.2 Backward Nondeterministic Dawg Matching Algorithm

As the first word in the name of algorithm [2] suggests, the algorithm performs some kind of reverse string matching. Indeed, it is a version of the Reverse Factor algorithm that performs the Boyer-Moore algorithm in the background. The algorithm is efficient only when the length of the pattern is not larger than the memory word-size of the machine.

The algorithm uses a table $B$, where for each character $c$ stores a bit mask. A bit mask is used to store bitwise operations. The mask in $B_c$ is set if and only if $x_i = c$. The search state is kept in a word $d = d_{m-1}...d_0$, where the match only happens when $d_{m-1} = 1$.

10.3 Aho–Corasick Algorithm

Aho-Corasick algorithm [1] is a dictionary matching algorithm which matches all the string at the same time. A dictionary in computer science is an abstract data type which uses keys to find values. It is a kind of array which searches appearance of values due to their keys.

The algorithm locates elements of finite sets of string within an input text. Running time of the algorithm is a sum of the length of the text,
the length of the pattern (or patterns, because it can search more than one pattern simultaneously) and the number of output matches. The algorithm uses a searching tree, where nodes are defined by using the finite alphabet.

10.4 Commentz-Walter Algorithm

Commentz-Walter algorithm [6] is an update of the Aho-Corasick algorithm because it can search for more than one pattern at the same time and has a background of the Boyer-Moore algorithm. Worst case time complexity is exactly the same as the Boyer-Moore algorithm $\Theta(m \times n)$, where $m$ is the length of the pattern (it is a sum of all patterns if there are more than one) and $n$ is the length of the text.

10.5 Horspool Algorithm

The Horspool’s algorithm [5] or also known as Boyer-Moore-Horspool algorithm is one more version of the Boyer-Moore algorithm besides its tuned version. It has the same worst case time complexity as the normal Boyer-Moore algorithm and the Commentz-Walter algorithm, with an update on the average time complexity going to $\Omega(n)$, where $n$ is the length of the text.

10.6 Raita Algorithm

The Raita algorithm [12] is a specification of the Boyer-Moore-Horspool algorithm. The preprocessing method is exactly as the Boyer-Moore algorithm, where the string is being searched for the pattern. The searching method is done in the following way: first, the last character of the pattern is compared with the rightmost character of the window. If there is a match, the first character is compared with the leftmost character of the window. If there is again a match, it compared the middle character of the window.
Once when all matches occur, it moves on to the next character in the window. The time complexity is the same as the Boyer-Moore algorithm, because of its background influence.

10.7 Bitap Algorithm

The Bitap algorithm [3] is a special version of string matching because it is an approximate string matching algorithm. The algorithm searches if the pattern is approximately equal within the text, not exactly the same. That is a separate field in the string matching problem, which will not be presented within my thesis, but only mentioned.
Chapter 11

Conclusion

The most known and efficient algorithms were described and their advantages and disadvantages were listed. Through exploring every algorithm into details, I came to a conclusion that the best algorithm can be determined only after the problem has been given and the available computing resources have been defined. For example, some problems require a lot of space but less time (or vise versa). In such cases it is better to choose an algorithm that is faster rather than algorithm that needs less memory.

In summary, the most used algorithms in practice are Knuth-Morris-Pratt algorithm and Boyer-Moore algorithm. They are all founded on the brute-force approach. Basically, each of them returns good results when the text is short; the problem arises when the text is extremely long.

Given suitable criteria, limited computing resources, and a particular problem, there is always an algorithm that will solve the problem efficiently.
Bibliography


