

inzva Algorithm Programme 2018-2019

Bundle 14

Algorithms - 5

Editor Mehmet Altuner

Contents

1	Hashing	•
2	String Matching Algorithms	4
	2.1 Basics	
	2.2 Naive Approach	
	2.3 Prefix Function and Proper Prefixes	
	2.4 KMP	
	2.5 String Matching Automata	,
	2.6 Rabin-Karp Algorithm	
	2.7 Suffix Array	10
	2.7.1 Naive Approach	1
	2.7.2 $O(n * log(n) * log(n))$ Algorithm	1
	2.7.3 Pattern Finding	
	2.8 LCP Array	13
3	References	14

1 Hashing

We can find a unique integer representation for each string. This process is called Hashing.

$$hash = ((s[0] -' a') * p^0 + (s[1] -' a') * p^1 + \dots + (s[n-1] -' a') * p^(n-1)) mod(m)$$

Choose p as a prime number that is bigger than the size of the alphabet, choose m as a large prime number to ensure that the hash value of the string is unique for that particular string.

Popular p and m values are p = 31 and m = 1e9 + 9.

```
vector <int> hash_func(string text) {
            int p = 31;
2
3
            a prime number that is bigger than the size of
            the alphabet
5
            */
            int m = 1e9 + 9; // a very large number
            lli p_pow = 1;
9
            vector <int> hash_val(text.size());
10
            // vector that will hold the hash val of the prefixes of text
11
12
            // init hash_val[0]
13
            hash_val[0] = (text[0] - 'a' + 1) % m;
            p_pow = p * p_pow;
15
16
            for(int i=1; i<text.size(); i++) {</pre>
17
                    hash\_val[i] = (hash\_val[i-1] + (text[i] - 'a' + 1) * p\_pow)
                                                                                              % m;
18
                    p_pow = p * p_pow;
19
20
21
            return hash_val;
22
23
```

Note that in the above implementation, we do not find just a single value for the string. We find the hash values of each substring that starts with the first character. We will make use of this implementation in Rabin-Karp Algorithm for pattern finding. If your purpose is to get the hash value of just a string, you do not have to implement it like this. Just an integer hashVal will be enough.

2 String Matching Algorithms

2.1 Basics

In string matching, our goal is to find a pattern in a given text. Sometimes we are asked only to find out if there is at least one matching substring or how many matching substring there are.

2.2 Naive Approach

In naive approach, we simply traverse through all possible substrings of length m of the given text of length n. We check each substring for whether it is matching with the given pattern or not. Therefore, our complexity is O(m*n).

```
bool find(string pattern, string text) {
    for(int i=0; i<text.size()-pattern.size(); i++) {
        for(int j=0; j<pattern.size(); j++) {
            if(text[i+j] != pattern[j]) break;
            if(j == pattern.size()-1) return true;
        }
    }
    return false;
}</pre>
```

In the code above, we make sure that the substring that starts with text[i] and has the length of m is equal to the pattern. One can observe that while traversing, we are processing characters of text multiple times. Popular string matching algorithms are focused on reducing this cost.

What if our pattern was something like abcdef? No re-occuring characters or substrings. In such cases, the code below will work with a complexity of O(n).

```
bool find(string pattern, string text) {
        int i = 0, matched_so_far = 0;
2
        while(i < text.size()){</pre>
            if(text[i] == pattern[matched_so_far]) {
4
                 if (matched_so_far == pattern.size()-1)
                     return true;
6
                 matched_so_far++;
9
            }else{
10
                 matched_so_far = 0;
11
            }
12
13
14
        return false;
15
```

Why is the complexity for finding patterns that have no re-occurrent substrings O(n)? Because we only need to process each character just for once.

Let's say that our pattern is abcdef and we are in such a state that so far we have reached abc in our text. If the current character that we will check is not d, we do not need to start over because we have already matched abc and we know that our starting character a does not exist in the text except the last time we matched a.

So this observation leads us to another concept, Prefix Function.

2.3 Prefix Function and Proper Prefixes

As we have discussed before, string matching algorithms work in the worst way in the case of reoccurrent substrings. In a more formal way, we can define these substrings as prefixes that are also the suffixes of a string (proper prefixes). Prefix abc in text abcdeabc is an example of proper prefixes since abc is a prefix and a suffix at the same time.

You are given a string s and the prefix function is an array that holds the length of the longest proper prefix of an interval [0, i] which is also the suffix of this substring.

$$\pi[i] = \max_{k=0...i} \{k : s[0 ... k-1] = s[i - (k-1) ... i]\}$$

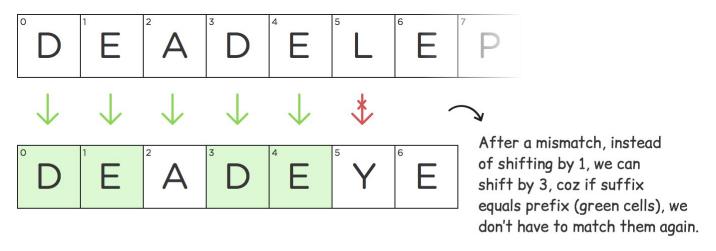
For example, for the pattern abcdeabc, the prefix function array will be [0, 0, 0, 0, 0, 1, 2, 3]

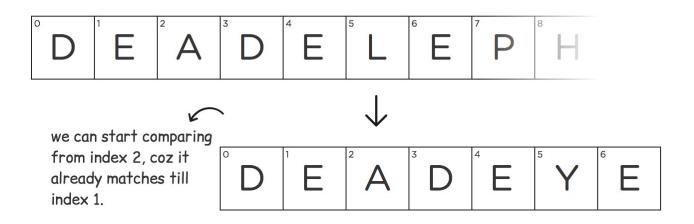
Most of the string matching algorithms are using prefix function to minimize the cost of the algorithm.

```
vector <int> prefix_array;
   void calculatePrefixArray(string pattern) {
2
            int len = 0; // length of the current proper prefix
3
            int i = 1;
                          // pointer that traverse on pattern
4
            prefix array.resize(pattern.size());
            prefix_array[0] = 0;
6
            while(i < pattern.size()){</pre>
8
                     if(pattern[i] == pattern[len]){
                             prefix_array[i] = ++len;
10
11
                              i++;
                     }else{
12
                              if(len != 0) { // if i had a proper prefix before
13
                                      len = prefix_array[len-1];
14
                              }else{
15
                                      prefix_array[i] = 0;
16
17
                              }
18
                     }
19
            }
20
21
```

2.4 KMP

KMP string matching algorithm is performed very similarly to the naive approach. In the case of a mismatch while checking a substring for whether it is equal to the pattern or not, we do not start over. Instead, we start from the head of the proper prefix we have matched so far if there is any. Because if we have a proper prefix, means that we have a suffix that is equal to the prefix. The pattern we am looking for might be starting with this very suffix we just found.



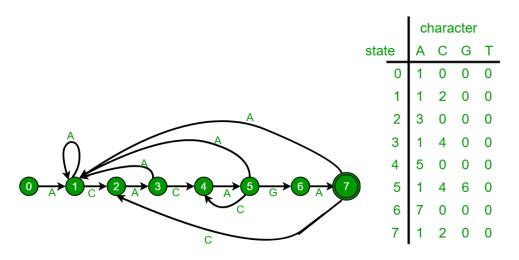


```
bool KMP(string text, string pattern) {
1
        calculatePrefixArray(pattern);
2
            int i = 0; // pointer to traverse on text
3
4
            int j = 0; // pointer to traverse on pattern
            while(i < text.size()){</pre>
6
                     if(j==pattern.size()) return true;
8
                     if(text[i] == pattern[j]){
                              i++;
10
                              j++;
11
                     }else{
12
                              if(j != 0){
13
                                       /*
14
                                       if i had a matching sequence lately
15
                                       do not go to the beginning
16
                                       go to the beginning of the proper prefix
17
                                       and do not increment i
18
19
                                       j = prefix_array[j-1];
20
                              }else{
21
                                       // increment i
22
                                       i++;
23
                              }
24
25
                     }
26
            }
27
            return j == pattern.size();
28
29
```

The time complexity for KMP is O(m+n) and memory complexity is O(n).

2.5 String Matching Automata

In string matching automata, we precalculate a pattern and build a 2D array that represents a Finite Automata. We will search for the pattern in the text according to this array. It will tell us which state to go next, according to the current state we are in. Calculation of the state table is the tricky part of this algorithm. This algorithm is useful while dealing with the relatively small sized pattern texts and alphabets, and large size of input texts.



The size of the state array will be size of the alphabet * (size of the pattern + 1). We will be on state(0) at the beginning.

Let's say we are at the state t and checking for the letter 1 of the alphabet. Now, what we should do is consider the string pat [0...t-1] + 1. Let's call this new string newPattern. Our next state will be the length of the longest proper prefix of the string pat + '' + newPattern. The character "is placed in the middle of two strings to distinguish them (make sure "is not in the given alphabet or choose one outside from the alphabet).

```
int state_table[N+1][M];
   // N=length of the input, M=size of the alphabet
   void create_state_table(string pattern, string alphabet){
            string pat = "";
4
            for(int i=0; i<pattern.size()+1; i++) {</pre>
5
                     for(int j=0; j<alphabet.size(); j++){</pre>
6
                              string tmp = pat;
                              pat += alphabet[j];
8
                              //cerr << pat << endl;
                              int new_state=0, pattern_ptr=0;
10
11
                              prefix_array.clear();
12
                              calculatePrefixArray(pattern + '#' + pat);
13
14
                              new_state = prefix_array[pattern.size() + i + 1];
15
16
                              state_table[i][j] = new_state;
17
                              pat = tmp;
18
19
                     if(i < pattern.size()) pat += pattern[i];</pre>
20
21
            }
22
```

Time complexity for building the state table is O(n*m) (n is the size of the pattern and m is the size of the alphabet).

While searching for the pattern in the text, we simply follow our state table. If we reach the last state, it means we have found the pattern.

```
bool finite_automata(string pattern, string text, string alphabet) {
    create_state_table(pattern, alphabet);
    int current_state = 0;
    int last_state = pattern.size();

for(int i=0; i<text.size(); i++) {
    current_state = state_table[current_state][s[i]];
    if(current_state == last_state) return true;
}

return false;
}
</pre>
```

2.6 Rabin-Karp Algorithm

This algorithm is the only string matching algorithm among the ones we are discussing that does not use prefix function.

The main idea roots back to Hashing. If you remember the way we implement the hashing function, we hold the hash value of str[1..i] in hash[i]. By the definition of hash, a string's hash value can be considered as unique. So if two hash values of two strings are equal, these two strings are the same. By using this, we can tell if two substrings are matching or not.

```
bool same(string a, string b) {
    return hash(a) == hash(b);
}
```

How can we use this in string matching? Let's say we are looking for pattern p in text t. Length of these strings are n and m, respectively. We will traverse in each substring of p of length n in O(m) and check whether they are equal or not. In naive solution, this goes to complexity of O(n * m), but thanks to the properties of hashing we discussed above, this complexity is reduced to O(1).

We know how to get the hash value of a subtring that starts from the first character. How do we get the hash values of other substrings?

```
hash(str[i..j]) = (hash(j) - hash(i-1))/p^{i}
```

```
bool rabin_karp(string text, string pattern) {
            vector <int> hash_text, hash_pattern;
2
            hash_text = hash_func(text); hash_pattern = hash_func(pattern);
3
4
            int m = text.size();
5
            int n = pattern.size();
6
            int p = 31;
            if (hash_text[n-1] == hash_pattern[n-1])
9
                    return true;
10
11
            for (int i=1; i<m; i++) {</pre>
12
                     int hash_substring = (hash_text[i+n-1] - hash_text[i-1]) / pow(p, i);
13
                    if (hash_substring == hash_pattern[n-1])
                             return true;
15
            }
16
17
            return false;
18
19
```

2.7 Suffix Array

Suffix Array is a very efficient data structure that is being widely used in compression, bioinformatics and in areas that string manipulation is used.

The problem we will be interested in is pattern finding.

Suffix Array is a sorted array consisting of all suffixes of a string.

2.7.1 Naive Approach

Naive Approach has a complexity of $O(n^2 * log(n))$ and n is the length of the string. In naive approach, we simply traverse through all suffixes of the string and store them in an array. While sorting this array, we get $O(n^2 * log(n))$ since comparing two strings is O(n).

2.7.2 O(n * log(n) * log(n)) Algorithm

One can observe that the suffixes we are trying to sort belong to the same string. By using this fact, we will first sort suffixes by their first character, and then we will sort suffixes of length two, and then of length four, and the length of the suffixes we will compare will grow exponentially. We will be performing sorting operation in O(log(n)) step, therefore we have a complexity of O(n*log(n)*log(n)).

```
#define MAXN 65536
   #define MAXLG 17
   char A[MAXN];
   struct entry
5
        int nr[2];
6
        int p;
   } L[MAXN];
   int P[MAXLG][MAXN];
9
   int N, i;
   int stp, cnt;
11
   int cmp(struct entry a, struct entry b) {
12
        return a.nr[0] == b.nr[0] ?(a.nr[1] < b.nr[1] ?1: 0): (a.nr[0] < b.nr[0] ?1: 0);</pre>
13
   }
14
   int main()
15
   {
16
        gets(A);
17
        for (N=strlen(A), i = 0; i < N; i++)
        P[0][i] = A[i] - 'a';
19
        for (stp=1, cnt = 1; cnt < N; stp++, cnt \star= 2)
20
21
            for (i=0; i < N; i++) {</pre>
22
                 L[i].nr[0]=P[stp- 1][i];
23
24
                 L[i].nr[1]=i + cnt < N? P[stp -1][i+ cnt]:-1;
                 L[i].p=i;
25
26
            sort(L, L+N, cmp);
27
            for(i=0; i < N; i++)</pre>
28
                 P[stp][L[i].p] = i > 0 \&\& L[i].nr[0] == L[i-1].nr[0] \&\& L[i].nr[1] == L[i-1].nr[1]?
29
30
        return 0;
31
32
```

2.7.3 Pattern Finding

Since our suffix array is sorted, we can perform search operation efficiently by using binary search.

```
// A suffix array based search function to search a given pattern
1
       // 'pat' in given text 'txt' using suffix array suffArr[]
   void search(char *pat, char *txt, int *suffArr, int n)
3
4
       int m = strlen(pat); // get length of pattern, needed for strncmp()
5
6
       // Do simple binary search for the pat in txt using the
7
       // built suffix array
8
       int l = 0, r = n-1; // Initilize left and right indexes
9
       while (1 \le r)
10
       {
11
            // See if 'pat' is prefix of middle suffix in suffix array
12
            int mid = 1 + (r - 1)/2;
13
            int res = strncmp(pat, txt+suffArr[mid], m);
14
            // If match found at the middle, print it and return
16
            if (res == 0)
17
            {
18
                cout << "Pattern found at index " << suffArr[mid];</pre>
19
                return;
20
^{21}
22
            // Move to left half if pattern is alphabtically less than
23
            // the mid suffix
24
            if (res < 0) r = mid - 1;
25
26
            // Otherwise move to right half
27
           else 1 = mid + 1;
28
29
       // We reach here if return statement in loop is not executed
31
       cout << "Pattern not found";</pre>
32
33
```

Note that in the above implementation, suffArr[i] corresponds to i'th suffix of a string. For example let that string be "banana" and i be one.

```
suffArr[i] = 3 string[3..n] = "ana"
```

2.8 LCP Array

LCP means longest common prefix. LCP[i] holds the length of the longest common prefix of suffArr[i] and suffArr[i+1].

```
txt[0..n-1] = "banana"
suffix[] = {5, 3, 1, 0, 4, 2|
lcp[] = {1, 3, 0, 0, 2, 0}

Suffixes represented by suffix array in order are:
{"a", "ana", "anana", "banana", "na", "nana"}

lcp[0] = Longest Common Prefix of "a" and "ana" = 1
lcp[1] = Longest Common Prefix of "ana" and "anana" = 3
lcp[2] = Longest Common Prefix of "anana" and "banana" = 0
lcp[3] = Longest Common Prefix of "banana" and "na" = 0
lcp[4] = Longest Common Prefix of "na" and "nana" = 2
lcp[5] = Longest Common Prefix of "nana" and None = 0
```

There are many ways to construct a LCP Array but we are going to discuss the one that uses precalculated suffix array (Kasai's Algorithm).

```
int lcp(int x, int y) {
       // lcp's of substring s[x..n] and s[y..n]
       int res = 0;
       for(int i=logn; i>=0; i--) {
            if(P[i][x] == P[i][y]){
                res += pow(2, i);
6
           x += pow(2, i);
           y += pow(2, i);
9
           // move to the next characters
10
       }
11
       return res;
12
13
```

The algorithm above has a complexity of O(log(n)) and should be called for each suffix so its total complexity is O(n * log(n)).

3 References

References

- [1] https://www.geeksforgeeks.org/suffix-array-set-2-a-nlognlogn-algorithm/
- [2] https://cp-algorithms.com/string/prefix-function.html
- [3] https://www.geeksforgeeks.org/suffix-array-set-2-a-nlognlogn-algorithm/
- [4] https://cp-algorithms.com/string/string-hashing.html
- [5] https://www.geeksforgeeks.org/kasais-algorithm-for-construction-of-lcp-array-from-suffix-array/
- [6] https://www.geeksforgeeks.org/finite-automata-algorithm-for-pattern-searching/
- [7] https://cp-algorithms.com/string/rabin-karp.html