

An Algorithm to find length of the longest sub sequence of given sequence such that all elements of this are alternating using dynamic programming.

DAA ASSIGNMENT-2 , GROUP 21

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Abstract—This Paper contains the algorithm for finding the length of the longest sub sequence of given sequence such that all elements of this are alternating.

I. PROBLEM STATEMENT

The longest Zig-Zag sub sequence problem is to find length of the longest sub sequence of given sequence such that all elements of this are alternating.

If a sequence $\{x_1, x_2, \dots, x_n\}$ is alternating sequence then its element satisfy one of the following relation :

$x_1 < x_2 > x_3 < x_4 > x_5 < \dots < x_n$ or

$x_1 > x_2 < x_3 > x_4 < x_5 > \dots > x_n$

Solve using Dynamic programming.

II. KEYWORDS

Recursion, Dynamic Programming

III. INTRODUCTION

Let's first formally define what subsequence is.

A subsequence is a sequence that can be derived from another sequence by zero or more elements, without changing the order of the remaining elements as an example for the array $[1, 2, 3, 4]$ the subsequences are (1), (2), (3), (4), (1,2), (1,3), (1,4), (2,3), (2,4), (3,4), (1,2,3), (1,2,4), (1,3,4), (2,3,4), (1,2,3,4).

IV. ALGORITHMIC DESIGN

A. Approach 1(Using Recursion)

- 1) During recursion, we will determine the state of the problem if
 - a) State is 0 then we want an element smaller than it as subsequence is already been started
 - b) State is 1 then we want an element greater than it as subsequence is already been started
 - c) State is 2 then we can take any element as subsequence is yet to be started.
- 2) Then in the Base case, when we have no further elements left. ($n==0$).
- 3) After knowing the state, either discard the number or include the number in the current sequence if the

previous element is smaller with state = 1 or previous element is greater with state = 0 and if state =2 we take all of the possibilities.

B. Approach 2(Using DP)

- 1) At start, we get to know about the state of the problem whether it is 0, 1 or 2.
- 2) To store the values of each element, there is a two dimensional matrix of size $n \times 3$ ($dp[n][3]$). Here 2D matrix will store each element's state in best possible way.
- 3) If the value of $dp[i][state]$ is precalculated then directly pass on it for further calculation, else calculate $dp[i][state]$ and store it for further use.
- 4) Finally we return the length of the subsequence.

Algorithm 1: To find the length of the longest subsequence as per the given condition.

Input: One array arr with size n

Output: Return the length of the longest zigzag subsequence

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1 Function Helper (arr,n,prev,state) :
2   if n = 0 then
3     return 0
4   if state < 0 then
5     a ← INT_MIN
6     b ← INT_MIN
7     a ← Helper(arr, n - 1, prev, state)
8     if arr[n - 1] > prev then
9       b ← 1 + Helper(arr, n - 1, arr[n - 1], 1)
10    return max(a, b)
11  else if state > 0 then
12    a ← INT_MIN
13    b ← INT_MIN
14    a ← Helper(arr, n - 1, prev, state)
15    if arr[n - 1] < prev then
16      b ← 1 + Helper(arr, n - 1, arr[n - 1], -1)
17    return max(a, b)
18  else
19    a ← INT_MIN
20    b ← INT_MIN
21    c ← INT_MIN
22    a ← Helper(arr, n - 1, prev, state)
23    b ← 1 + Helper(arr, n - 1, arr[n - 1], -1)
24    c ← 1 + Helper(arr, n - 1, arr[n - 1], 1)
    return max(a, max(b, c))

```

V. ALGORITHM ANALYSIS

A. Approach 1(Using recursion)

Time Complexity Analysis

The time complexity will be $O(2^n)$ because it checks out all the possibilities by recursing.

Space Complexity Analysis

The space complexity is $O(n)$, for storing the input array.

B. Approach 2(Using DP)

Time Complexity Analysis

The time complexity will be $O(n * 3)$ because we are in a way doing memoization our approach 1 code and saving time by using values of the precalculated subproblems.

Space Complexity Analysis

The space complexity will be $O(n)$ for input array and $O(n * 3)$ for storing the values of each condition by dynamic programming.

Algorithm 2: To find the length of the longest subsequence as per the given condition.

Input: One array arr with size n

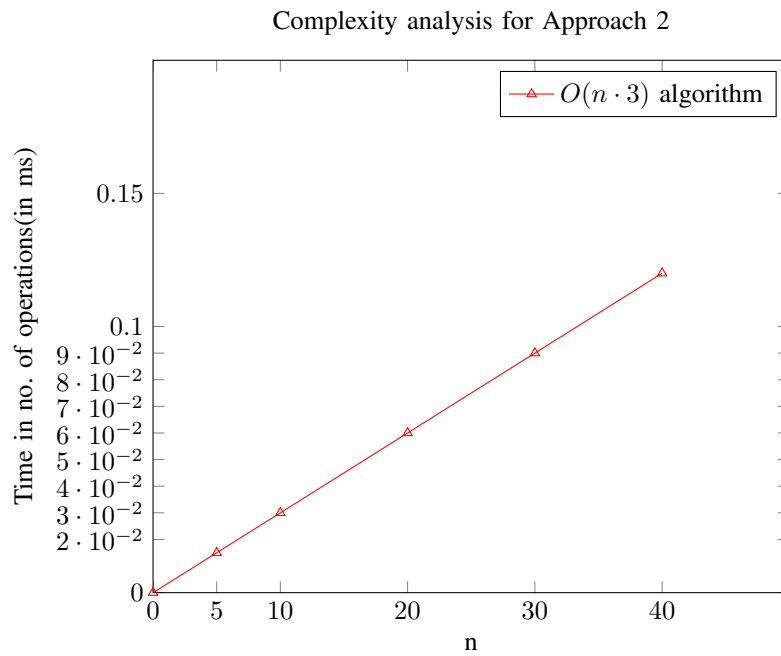
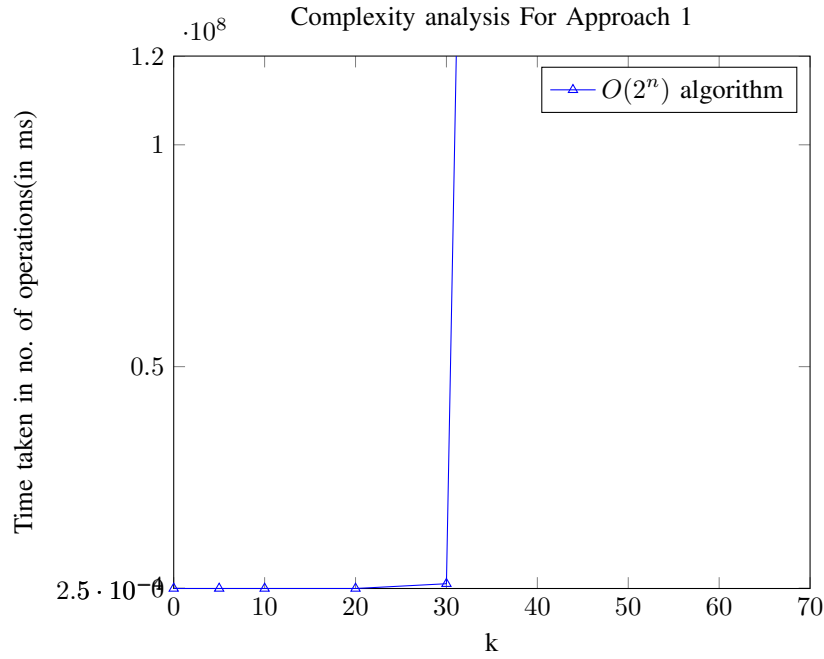
Output: Return the length of the longest zigzag subsequence

```

1 Function Helper (n,prev,state,dp[],arr) :
2   if n = 0 then
3     return 0
4   if dp[n][state] ≠ -1 then
5     return dp[n][state]
6   if state = 0 then
7     a ← INT_MIN
8     b ← INT_MIN
9     a ← Helper(n - 1, prev, state, dp, arr)
10    if arr[n - 1] > prev then
11      b ← 1 +
        Helper(n - 1, arr[n - 1], 1, dp, arr)
12    dp[n][state] ← max(a, b)
13    return max(a, b)
14  else if state = 1 then
15    a ← INT_MIN
16    b ← INT_MIN
17    a ← Helper(n - 1, prev, state, dp, arr)
18    if arr[n - 1] < prev then
19      b ← 1 +
        Helper(n - 1, arr[n - 1], 0, dp, arr)
20    dp[n][state] ← max(a, b)
21    return max(a, b)
22  else
23    a ← INT_MIN
24    b ← INT_MIN
25    c ← INT_MIN
26    a ← Helper(n - 1, prev, state, dp, arr)
27    b ← 1 + Helper(n - 1, arr[n - 1], 0, dp, arr)
    c ← 1 + Helper(n - 1, arr[n - 1], 1, dp, arr)
    dp[n][state] ← max(a, max(b, c))
28  return max(a, max(b, c))

```

VI. EXPERIMENTAL STUDY



VII. CONCLUSION

Above two methods have different time complexities and meet to fulfill the problem statement. The order in which they are good can be listed as:

I. Approach 2

II. Approach 1

Based on the time complexities.

VIII. REFERENCES

- 1) Utkarsh Trivedi, 'Longest Zig-Zag Subsequence', GeeksforGeeks, 2018. [Online]. [Accessed: 27-Mar-2021]