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The substring inclusion constraint longest common subsequence problem can be solved in quadratic time [☆]

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ABSTRACT

In this paper, we study some variants of the Constrained Longest Common Subsequence (CLCS) problem, namely, the substring inclusion CLCS (Substring-IC-CLCS) problem and a generalized version thereof. In the Substring-IC-CLCS problem, we are to find a longest common subsequence (LCS) of two given strings containing a third constraint string (given) as a substring. Previous solution to this problem runs in cubic time, i.e, O(nmk) time, where n, m and k are the length of the 3 input strings. In this paper, we present simple O(nm) time algorithms to solve the Substring-IC-CLCS problem. We also study the Generalized Substring-IC-LCS problem where we are given two strings of length n and m respectively and an ordered list of p strings and the goal is to find an LCS containing each of them as a substring in the order they appear in the list. We present an O(nmp) algorithm for this generalized version of the problem.

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1. Introduction

The longest common subsequence (LCS) problem is one of the classic and well-studied problems in computer science with extensive practical applications. The constraint LCS (CLCS) problem, where the computed LCS must contain a given third string as a subsequence, was introduced by Tsai [6] and later studied by a number of researchers [5,2,1,3]. This problem finds motivation from bioinformatics. Chen and Chao [2] recently introduced and studied several variants of the CLCS problem. In this paper, we are interested in one of the variants proposed in [2], where, given two strings $X = x_1, x_2, ..., x_m$ and $Y = y_1, y_2, ..., y_n$ and a third constraint string $S = s_1, s_2, ..., s_k$, we are to find a longest common subsequence $C = c_1, c_2, ..., c_\ell$ such that S is a *substring* of C. In what follows, we will be referring to this problem as the Substring-IC-LCS problem. Chen and Chao [2] presented an O(nmk) algorithm for this problem. We on the other hand present two improved algorithms both of which run in O(nm) time.

We also study the generalized version of this problem, where we are given two strings X, Y (as before) and an ordered list of constraint strings $S = \{S_1, S_2, \ldots, S_p\}$ (as opposed to a single constraint string) and we are to find a longest common subsequence C of X and Y containing each of the strings of S as a substring in the sequence they appear in the list. This problem was also handled by Chen and Chao [2] and they gave an algorithm with $O(n^2 \times \prod_{r=1}^p k_r)$ time where k_r is the length of the string S_r and we assume that n > m. We on the other hand present an $O(n^2p)$ algorithm for the same problem.

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The rest of the paper is organized as follows. In Section 2, we present the preliminary concepts. Section 3 presents a simple quadratic time algorithm to solve the Substring-IC-LCS problem. In Section 4, we present our main dynamic programming algorithm to solve the same problem. In Section 5, we extend the algorithm of Section 4 to solve the generalized version of the problem. We briefly conclude in Section 6.

2. Preliminaries

Given a string, $X = x_1, x_2, \dots, x_m$, we use x_i to denote the i-th letter in X. We use $X_{i...i'}$, $i' \geqslant i$ to denote the substring $x_i x_{i+1} x_{i+2} \dots x_{i'}$. On the other hand, $X_{i'...i} = x_{i'}, x_{i'-1}, x_{i'-2}, \dots, x_i$ denotes the reverse string of $X_{i...i'}$, where i' > i. Given another string $S = s_1, s_2, \dots, s_k$, we say that S occurs in the substring $X_{i...i'}$ if and only if $X_{i...i'}$ contains the string S as a sequence such that $x_i = s_1$ and $x_{i'} = s_k$. We use "." as the string concatenation operator. For example, we use X.Y to denote the concatenation of X and Y. The problems we study in this paper are formally defined below.

Problem 1 (Substring-IC-LCS Problem). Suppose we are given two strings $X = x_1, x_2, ..., x_m$ and $Y = y_1, y_2, ..., y_n$ and a constraint string $S = s_1, s_2, ..., s_k$. We are to find a longest common subsequence $C = c_1, c_2, ..., c_\ell$ such that S is a substring of C.

Problem 2 (Generalized Substring-IC-LCS Problem). Suppose we are given two strings X, Y and a ordered list of constraint strings $S = \langle S_1, S_2, \ldots, S_p \rangle$. We are to find a longest common subsequence C of X and Y containing each of the strings of S as a substring in the order it appears in the list.

Example 1. Suppose, X = aatgcctaggc, Y = cgatctggac, and S = gtac. Then, an LCS of X and Y is atctggc. And, given the constraint string S, a Substring-IC-LCS of X and Y is C = gtac.

Example 2. Suppose, X = fabcfgbda and Y = fabgcfbgda. Now, consider an ordered list of two strings $S = \langle abc, bda \rangle$. Then, a generalized Substring-IC-LCS of X and Y, given S is C = fabcfbda. Note that we have $C_{2...4} = abc$ and $C_{6...8} = bda$.

Example 3. Again, suppose X = dabdcfedbgcef and Y = dabcdegfe. Also assume that $S = \langle abc, bce \rangle$ is an ordered list of strings. Then, a generalized Substring-IC-LCS will be C = dabcegfe. Clearly, $C_{2...4} = abc$ and $C_{3...5} = bce$. Notably, we have overlapped occurrences of abc and bce in this example.

Given two strings X and Y of length m and n, respectively, for all $1 \le i \le m$, $1 \le j \le n$, we use LCS[i,j] to denote the length of an LCS of $X_{1...i}$ and $Y_{1...j}$. On the other hand, RevLCS[i,j] denotes the length of an LCS of $X_{i...m}$ and $Y_{j...n}$. Given i < i' and j < j', we further use $LCS(X_{i...i'}, Y_{j...j'})$ to denote a longest common subsequence of $X_{i...i'}$ and $Y_{j...j'}$ and $RevLCS(X_{i'...i}, Y_{i'...j})$ to denote a longest common subsequence of the reverse strings $X_{i'...i}$ and $Y_{j'...j}$.

Now, suppose that S occurs in X. Then, we say $S_X[i'] = i$ if and only if S occurs in $X_{i...i'}$ and there exists no i_1 , $i < i_1 < i'$ such that S occurs in $X_{i_1...i'}$ and no i_2 , $i < i_2 < i'$ such that S occurs in $X_{i...i_2}$. In other words, $S_X[i']$ keeps track of the occurrences of S in X that ends at the $x_{i'}$ and starts at x_i such that no other occurrence of S starts (ends) at X_{i_1} (X_{i_2}) such that $i < i_1 < i'$ ($i < i_2 < i'$). If there is no occurrence of string S ending at Position i', then we set $S_X[i'] = 0$.

Example 4. Suppose X = DABEABC and S = ABC. Then S occurs in the range $X_{2...7}$ and $X_{5...7}$. However we have $S_X[7] = 5$ (and not 2).

Now, suppose C is an LCS of X and Y such that S is a substring of C. Then we say that C is a Substring-IC-LCS of X and Y given a constraint string S. We use StrLCS[i,j] to denote the length of a Substring-IC-LCS of $X_{1...i}$ and $Y_{1...j}$ containing the substring S.

To handle the Generalized Substring-IC-LCS Problem, we need to extend some of the above notations and definitions. In this version of the problem, we are given a constraint ordered list $S = \langle S_1, S_2, \dots, S_p \rangle$ instead of a single constraint string S_r . For each string S_r , $1 \le r \le p$ in S_r , S_r denotes the length of string S_r . We extend the notion of $S_r[i']$ when we have a list $S_r[i']$ instead of a single constraint pattern $S_r[i']$ to denote $S_r[i']$ for the constraint pattern $S_r[i']$ to denote $S_r[i']$ for the constraint pattern $S_r[i']$ so denote $S_r[i']$ for the constraint pattern S_r

Given two strings $A=a_1\ldots a_{k_1}$ and $B=b_1\ldots b_{k_2}$, we say that A and B overlap when we have $A_{i\ldots k_1}=B_{1\ldots j}$ for some $1\leqslant i\leqslant k_1$ and $1\leqslant j\leqslant k_2$. Note that the length of the overlap is $k_1-i+1=j-1+1$. In this case, the merged pattern of the above two overlapping strings is the string $a_1a_2\ldots a_ia_{i+1}\ldots a_{k_1}b_{j+1}\ldots b_{k_2}=a_1a_2\ldots a_{i-1}b_1b_2\ldots b_jb_{j+1}\ldots b_{k_2}$. In the context of the merged pattern of A and B above, the substring $b_{j+1}\ldots b_{k_2}$ is said to be the non-overlapping pattern. Note that, for this notion the order of A and B is important.

Now, in the list S, we may have overlaps between the consecutive strings S_{r-1} , S_r , for all $2 \le r \le p$. We use Z_r to denote the merged pattern and NOV_r to denote the non-overlapping pattern of strings S_{r-1} and S_r , we have $NOV\ell_r$ to denote the length of NOV_r . The example below explains the above notions.

Example 5. Assume that $S_1 = ABCD$ and $S_2 = CDEFG$. Then their merged pattern $Z_2 = ABCDEFG$, non-overlapping pattern $NOV_2 = EFG$, which has length $NOV\ell_2 = 3$.

Finally, we use $NOV_X[r, i]$ to keep track the occurrence of NOV_r in X. Notably, since S_r contains NOV_r as a suffix, $S_X[r, i]$ is non-zero if, and only if, $NOV_X[r, i]$ is non-zero. Formally speaking, for $i_2 \le i_1 \le i$, if S_{r-1} occurs at $X_{i_2...i_1}$ and Z_r occurs at $X_{i_2...i_1}$, then we have $NOV_X[r, i] = i_1 + 1$; otherwise $NOV_X[r, i] = 0$.

3. A Simple algorithm for Substring-IC-LCS

If we want to include the string S to a common subsequence C of X and Y, then for some position i, we must have $C_{i...i+k-1} = S$. Suppose that S occurs *only* at $X_{i...i'}$ and $Y_{i...i'}$. Then we can do the following to get a desired Substring-IC-LCS:

- 1. Compute $LCS(X_{1...i-1}, Y_{1...i-1})$. Let the length of the computed LCS is ℓ_1 .
- 2. Compute $LCS(X_{i'+1...m}, Y_{j'+1...n})$. Let the length of the computed LCS is ℓ_2 .
- 3. Return $LCS(X_{1...i-1}, Y_{1...j-1}).S.LCS(X_{i'+1...m}, Y_{j'+1...m})$ as a Substring-IC-LCS and $\ell_1 + |S| + \ell_2$ as the length.

Now we need to consider the general case when there are more than one pair of (i,i') ((j,j')) such that S occurs at $X_{i...i'}$ $(Y_{j...j'})$. From the above idea, we get the following algorithm for computing Substring-IC-LCS for the general case as follows.

- 1. Compute $S_X[i]$ for $1 \le i \le m$.
- 2. Compute $S_Y[i]$ for $1 \le i \le n$.
- 3. For each pair (i,j) such that $S_X[i] \neq 0$ and $S_Y[j] \neq 0$ compute $LCS(X_{1...S_X[i]-1}, Y_{1...S_Y[j]-1})$ and $LCS(X_{i+1...m}, Y_{j+1...n})$. Suppose the length of $LCS(X_{1...S_X[i]-1}, Y_{1...S_Y[j]-1})$ is ℓ_1 and the length of $LCS(X_{i+1...m}, Y_{j+1...n})$ is ℓ_2 . Return $LCS(X_{1...S_X[i]-1}, Y_{1...S_Y[i]-1}).S.LCS(X_{i+1...m}, Y_{j+1...n})$ such that $\ell_1 + \ell_2$ is maximum. Also, return $\ell_1 + |S| + \ell_2$ as the length.

Clearly, $LCS(X_{i+1...m}, Y_{j+1...n})$ is equal to $RevLCS(X_{m...i+1}, Y_{n...j+1})$. The algorithm is formally presented in Algorithm Find-Substring-IC-LCS. The running time analysis is simple. Let us assume w.t.l.o.g that n > m. Then the computation of $S_X[i]$ and $S_Y[j]$ takes $O(n^2)$ time. Computation of table LCS[i, j] and RevLCS[i, j] takes $O(n^2)$ time. Finally, the computation of the Substring-IC-LCS length (see Steps 37 to 46) takes $O(n^2)$ time. Thus in total, the algorithm takes $O(n^2)$ time.

Find-Substring-IC-LCS

```
1: for i \leftarrow 1 to m do
        S_X[i] \leftarrow 0
 3: end for
 4: for j \leftarrow 1 to n do
        S_Y[j] \leftarrow 0
 6: end for
 7: for i' \leftarrow 1 to m do
        if s_1 = x_{i'} then
 8.
           p \leftarrow 1
 9:
            for i \leftarrow i' to m do
 10:
 11:
               if s_n = x_i then
 12:
                  p \leftarrow p + 1
                  if p = k then
 13:
                    S_X[i] \leftarrow i'
 14:
                    break
 15:
 16:
                  end if
              end if
 17:
            end for
 18:
        end if
 19:
20: end for
21: for j' \leftarrow 1 to n do
        if s_1 = y_{j'} then
22:
            p \leftarrow 1
23:
            for j \leftarrow j' to n do
 24:
 25:
               if s_p = y_j then
 26:
                  p \leftarrow p + 1
                 if p = k then
 27:
                    S_Y[j] \leftarrow j'
 28:
```

```
break
29:
              end if
30.
            end if
31.
32:
         end for
33:
       end if
34: end for
35: compute LCS of X, Y
36: compute RevLCS of X, Y
37: for i \leftarrow 1 to m do
      for i \leftarrow 1 to n do
38:
39:
         if S_X[i] \neq 0 and S_Y[j] \neq 0 then
           d = LCS[S_X[i] - 1, S_Y[j] - 1] + k + RevLCS[i + 1, j + 1]
40:
41:
           if maxStrLCS > d then
              maxStrLCS = d
42.
            end if
43:
44.
         end if
      end for
45.
46: end for
```

Very recently, we came across a paper [4], that also solves the Substring-IC-LCS problem in $O(n^2)$ time. The algorithm of [4] is almost identical to our above algorithm. The only notable difference between the two algorithms is that while computing each of the occurrences of S, we keep the unique position while the algorithm in [4] keeps multiple positions which is redundant.

4. Our main result: a dynamic programming algorithm

In this section, we present a dynamic programming formulation to directly compute the Substring-IC-LCS. In a later section, we will discuss how this DP formulation can be extended to solve the generalized version of the problem. Property 1 shows the characterization of the structure of a solution to the Substring-IC-LCS problem.

Property 1. If $C_{1...\ell}$ is an LCS of $X_{1...m}$ and $Y_{1...n}$ including S as a substring such that $S = C_{\ell'-k+1...\ell'}$ for some $k \le \ell' \le \ell$, then $C_{1...\ell}$ is a concatenation of the following two substrings, for some $0 \le i \le m$ and $0 \le j \le n$:

- 1. The prefix $C_{1...\ell'}$: $C_{1...\ell'}$ is an LCS of $X_{1...i}$ and $Y_{1...j}$ including Sas the suffix $C_{\ell'-k+1...\ell'}$, and
- 2. The suffix $C_{\ell'+1...\ell}$: $C_{\ell'+1...\ell}$ is an LCS of $X_{i+1...m}$ and $Y_{i+1...n}$.

A Dynamic programming formulation for the Substr-IC-LCS problem is given below.

$$LCS[i, j] = \begin{cases} 0 & \text{if } (i = 0 \text{ or } j = 0), \\ LCS[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_i = y_j, \\ \max(LCS[i - 1, j], LCS[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_i \neq y_j. \end{cases}$$

$$t = \begin{cases} -\infty & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_i = y_j, \\ \max(StrLCS[i - 1, j], StrLCS[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_i \neq y_j. \end{cases}$$

$$StrLCS[i, j] = \begin{cases} t & \text{if } S_X[i] = 0 \text{ or } S_Y[j] = 0, \\ \max(t, LCS[S_X[i] - 1, S_Y[j] - 1] + k) & \text{if } S_X[i] \neq 0 \text{ and } S_Y[j] \neq 0. \end{cases}$$

$$(3)$$

$$t = \begin{cases} -\infty & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_i = y_j, \\ \max(StrLCS[i - 1, j] StrLCS[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_i \neq y_i. \end{cases}$$
(2)

$$StrLCS[i, j] = \begin{cases} t & \text{if } S_X[i] = 0 \text{ or } S_Y[j] = 0, \\ \max(t, LCS[S_X[i] - 1, S_Y[j] - 1] + k) & \text{if } S_X[i] \neq 0 \text{ and } S_Y[j] \neq 0. \end{cases}$$
(3)

As is evident, here, Eq. (1) basically computes the normal LCS. It is required by Eq. (3). Here, as usual, LCS[i, j] stores the length of $LCS(X_{1...i}, Y_{1...i})$. Actual Substring-IC-LCS computation is done by Eqs. (2) and (3). Here, StrLCS[i, j] stores the length of the Substring-IC-LCS of $X_{1...i}$ and $Y_{1...i}$. We use t for intermediate computation. Note that, if the length of the Substring-IC-LCS of X[1...i] and Y[1...j] is 0, we store $-\infty$ in StrLCS[i, j].

Now, note that, when i = 0 or j = 0 there is no Substring-IC-LCS with respect to S. Also, if $S_X[i'] = 0$ or $S_X[j'] = 0$ for $1 \le i' \le i$, $1 \le j' \le j$, then we cannot have any Substring-IC-LCS with respect to S for $X[1 \dots i']$ and $Y[1 \dots j']$ for $1 \le i' \le i$, $1 \le j' \le j$. To handle this situation, when i = 0 or j = 0, we assume that the length of Substring-IC-LCS with respect to S is $-\infty$, where ∞ is conceptually a very very large value with respect to the values used in our computation. We further assume that any addition to $-\infty$ still results in $-\infty$. Now we prove the correctness of our DP formulation. We first report the following observation.

Observation 1. Suppose, $S_X[i] \neq 0$ and $S_Y[j] \neq 0$. Then the followings hold true.

```
1. If S_X[i'] = 0 or S_Y[j'] = 0 for all 1 \le i' < i, 1 \le j' < j, then StrLCS[i', j'] = -\infty and StrLCS[i, j] = LCS[S_X[i] - 1, S_Y[j] - 1] + k.

2. For all i' > i or j' > j, StrLCS[i', j'] > 0
```

Lemma 1. Eqs. (1) to (3) correctly compute Substring-IC-LCS.

Proof. We prove the correctness based on a case by case analysis. We consider the computation of StrLCS[i, j], i.e., the Substring-IC-LCS of X[1...i] and Y[1...j] with respect to the constraint pattern S.

```
Case 1: S_X[i] = 0 or S_Y[i] = 0.
From Eq. (3), it is clear that in this case, StrLCS[i, j] = t. We now have two subcases.
    Case 1.a: For all 1 \le i' \le i, 1 \le j' \le j we have S_X[i'] = 0 or S_Y[j'] = 0.
    We need to show that, in this case, StrLCS[i, j] = -\infty. From Eqs. (2) and (3), it is easy to see that StrLCS[i, j] can
    get some value other than -\infty if and only if the condition of S_X[i] \neq 0 and S_Y[j] \neq 0 holds in Eq. (3). Hence,
    clearly, StrLCS[i, j] = -\infty in this case.
    Case 1.b: There exists i' < i, j' < j such that S_X[i'] \neq 0 and S_Y[j'] \neq 0.
    Let us assume for some i' < i, j' < j, S_X[i'] \neq 0 and S_Y[j'] \neq 0 and for all 1 \le i'' < i' and 1 \le j'' < j', S_X[i''] = 0
    and S_Y[j''] = 0. Then, by Observation 1, we have StrLCS[i', j'] = LCS[S_X[i] - 1, S_Y[j] - 1] + k. Hence we must have
    StrLCS[i, j] > 0. Now from Eqs. (2) and (3) it is easy to verify that StrLCS[i, j] will get the correct value.
Case 2: S_X[i] \neq 0 and S_Y[j] \neq 0.
From Eq. (3), it is clear that in this case, StrLCS[i, j] = \max(t, LCS[S_X[i] - 1, S_Y[j] - 1] + k). We now have two subcases.
    Case 2.a: For all 1 \le i' < i, 1 \le j' < j we have S_X[i'] = 0 or S_Y[j'] = 0.
    We need to show that, in this case, t = -\infty and StrLCS[i, j] = LCS[S_X[i] - 1, S_Y[j] - 1] + k. From Eqs. (2) and (3),
    it is easy to see that StrLCS[i', j'] can get some value other than -\infty if and only if the condition of S_X[i'] \neq 0 and
    S_Y[j'] \neq 0 holds in Eq. (3). Hence, clearly, t = -\infty and StrLCS[i, j] = LCS[S_X[i] - 1, S_Y[j] - 1] + k.
    Case 2.b: There exists i' < i, j' < j such that S_X[i'] \neq 0 and S_Y[j'] \neq 0.
    Let us assume that S_X[i'] \neq 0 and S_Y[j'] \neq 0 for some i' < i, j' < j. Then, by Observation 1, for all i' \leq i'' < i,
    j' \leqslant j'' < j we have a value StrLCS[i'', j''] > 0. Hence from Eq. (2), t > 0 and from Eq. (3), StrLCS[i, j] =
    \max(t, LCS[S_X[i] - 1, S_Y[j] - 1] + k), which is correct. \square
```

The algorithm is formally presented in Algorithm Alt-Find-Substring-IC-CLCS. The running time analysis is simple. Let us assume w.t.l.o.g that n > m. Then the computation of $S_X[i]$ and $S_Y[j]$ for all $1 \le i \le m$ and $1 \le j \le n$ takes $O(n^2)$ time. Computation of the table LCS[i, j] for all $1 \le i \le m$ and $1 \le j \le n$ takes $O(n^2)$ time. Finally, computation of the Substring-IC-LCS length in Steps 42 to 53 takes $O(n^2)$ time. Thus in total, the algorithm takes $O(n^2)$ time.

Alt-Find-Substring-IC-CLCS

```
1: for i \leftarrow 1 to m do
 2: S_X[i] \leftarrow 0
 3: end for
 4: for j \leftarrow 1 to n do
       S_Y[j] \leftarrow 0
 5:
 6: end for
 7: for i' \leftarrow 1 to m do
       if s_1 = x_{i'} then
          p \leftarrow 1
 9:
           for i \leftarrow i' to m do
 10:
              if s_n = x_i then
 11:
                 p \leftarrow p + 1
 12:
 13:
                 if p = k then
                    S_X[i] \leftarrow i'
 14.
                    break
 15:
                 end if
              end if
           end for
 18:
19:
        end if
20: end for
21: for j' \leftarrow 1 to n do
22:
        if s_1 = y_{i'} then
           p \leftarrow 1
23:
           for j \leftarrow j' to n do
 24:
25:
              if s_p = y_i then
```

```
26:
                p \leftarrow p + 1
                if p = k then
27.
                   S_Y[i] \leftarrow i'
 28:
 29:
                   break
 30:
                end if
             end if
 31:
 32.
           end for
        end if
33.
34: end for
35: compute LCS of X, Y
36: for i \leftarrow 1 to m do
       StrLCS[i, 0] \leftarrow -\infty
38: end for
39: for j \leftarrow 1 to n do
       StrLCS[0, j] \leftarrow -\infty
41: end for
42: for i \leftarrow 1 to m do
        for i \leftarrow 1 to n do
43.
           if x_i = y_i then
 44:
             t \leftarrow StrLCS[i-1, j-1] + 1
 45:
 46:
 47:
             t \leftarrow \max(StrLCS[i-1, j], StrLCS[i, j-1])
 48.
           if S_X[i] \neq 0 and S_Y[j] \neq 0 then
 49:
             StrLCS[i, j] \leftarrow \max(t, LCS[S_x[i] - 1, S_v[j] - 1] + k)
 50:
 51.
             StrLCS[i, i] \leftarrow t
 52:
           end if
 53.
        end for
 54:
55: end for
```

5. Algorithm for generalized Substring-IC-LCS

In this section we consider the Generalized Substring-IC-LCS problem. In particular, we will extend the DP formulation of the Substring-IC-LCS problem from Section 4 to solve the generalized version of the problem. In what follows we will be using the notations $S_X[r,i]$ and Z_r extensively. Recall from Section 2 that $S_X[r,i]$ basically extends the notion of $S_X[i]$ when we are considering a list S of constraint pattern instead of a single one. On the other hand, Z_r denotes the merged pattern of strings S_{r-1} and S_r . Now we are ready to state Property 2 that shows the characterization of the structure of a solution to the Generalized Substring-IC-LCS problem.

Property 2. If $C_{1...\ell}$ is an LCS of $X_{1...m}$ and $Y_{1...m}$ including $\{S_1, S_2, ..., S_r\}$ as substrings, in the given order such that $S_r = C_{\ell'-k_r+1...\ell'}$ for some $k_r \le \ell' \le \ell$, then $C_{1...\ell}$ is a concatenation of the following two substrings, for some $0 \le i \le m$ and $0 \le j \le n$:

- 1. The prefix $C_{1...\ell'}$: $C_{1...\ell'}$ is an LCS of $X_{1...i}$ and $Y_{1...i}$ including $\{S_1, S_2, \ldots S_{r-1}\}$ as substring and S_r as the suffix $C_{\ell'-k_r+1...\ell'}$, and 2. The suffix $C_{\ell'+1...\ell}$: $C_{\ell'+1...\ell}$ is an LCS of $X_{i+1...m}$ and $Y_{i+1...n}$.
- Below we present a DP formulation for the Generalized Substring-IC-LCS problem. We need to consider following two cases:

Case 1: For all $1 < r \le p$ string S_r doesn't overlap with S_{r-1} .

In this case, this problem is a simple extension of the Substring-IC-LCS problem. So we can adopt the following strategy. Let $StrLCS_0[i, j]$ corresponds to the computation of LCS[i, j]. Then we will compute $StrLCS_1$ considering substring S_1 using $StrLCS_0$; then we will compute $StrLCS_2$ using $StrLCS_1$ for substring S_2 and so on. Finally continuing in this way we will compute $StrLCS_r$ using $StrLCS_{r-1}$ for S_r . The dynamic programming formulation is given below.

$$StrLCS_{0}[i, j] = \begin{cases} 0 & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS_{0}[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_{i} = y_{j}, \\ \max(StrLCS_{0}[i - 1, j], StrLCS_{0}[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_{i} \neq y_{j}. \end{cases}$$

$$t = \begin{cases} -\infty & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS_{r}[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_{i} = y_{j}, \\ \max(StrLCS_{r}[i - 1, j], StrLCS_{r}[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_{i} \neq y_{j}. \end{cases}$$

$$(5)$$

$$t = \begin{cases} -\infty & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS_r[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_i = y_j, \\ \max(StrLCS_r[i - 1, j], StrLCS_r[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_i \neq y_j. \end{cases}$$
(5)

$$StrLCS_{r}[i, j] = \begin{cases} t & \text{if } S_{X}[r, i] = 0 \text{ or } S_{Y}[r, j] = 0, \\ \max(t, StrLCS_{r-1}[S_{X}[r, i] - 1, S_{Y}[r, j] - 1] + k_{r}) & \text{if } S_{X}[r, i] \neq 0 \text{ and } S_{Y}[r, j] \neq 0. \end{cases}$$
 (6)

The followings are the boundary conditions: $StrLCS_0[i, 0] = StrLCS_0[0, j] = 0$ and $StrLCS_r[i, 0] = StrLCS_r[0, j] = -\infty$. In Eq. (5), t is used for intermediate computation.

Case 2: For some $1 < r \le p$ string S_r overlaps with S_{r-1} .

If for some $1 < r \le p$, S_{r-1} and S_r overlaps then we will compute $NOV\ell_r$, $NOV_X[r,i]$ and $NOV_Y[r,j]$ for all $1 \le i \le m$ and $1 \le j \le n$. For all $1 \le i \le m$ and $1 \le j \le n$, while computing Substring-IC-LCS for string S_r , we need to consider the merged pattern Z_r . The complete dynamic programming formulation for the problem is given below.

$$StrLCS_{0}[i, j] = \begin{cases} 0 & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS_{0}[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_{i} = y_{j}, \\ \max(StrLCS_{0}[i - 1, j], StrLCS_{0}[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_{i} \neq y_{j}. \end{cases}$$

$$t = \begin{cases} -\infty & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS_{r}[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_{i} \neq y_{j}. \end{cases}$$

$$t' = \begin{cases} t & \text{if } NOV_{X}[r, i] = 0 \text{ or } NOV_{Y}[r, j] = 0, \\ \max(StrLCS_{r-1}[NOV_{X}[r, i] - 1, NOV_{Y}[r, j] - 1] + NOV\ell_{r}) & \text{if } NOV_{X}[r, i] \neq 0 \text{ and } NOV_{Y}[r, j] \neq 0. \end{cases}$$

$$StrLCS_{r}[i, j] = \begin{cases} t' & \text{if } S_{X}[r, i] = 0 \text{ or } S_{Y}[r, j] = 0, \\ \max(t', StrLCS_{r-1}[S_{X}[r, i] - 1, S_{Y}[r, j] - 1] + k_{r}) & \text{if } S_{X}[r, i] \neq 0 \text{ and } S_{Y}[r, j] \neq 0. \end{cases}$$

$$StrLCS_{r}[i, j] = \begin{cases} t' & \text{if } S_{X}[r, i] = 0 \text{ or } S_{Y}[r, j] = 0, \\ \max(t', StrLCS_{r-1}[S_{X}[r, i] - 1, S_{Y}[r, j] - 1] + k_{r}) & \text{if } S_{X}[r, i] \neq 0 \text{ and } S_{Y}[r, j] \neq 0. \end{cases}$$

$$StrLCS_{r}[i, j] = \begin{cases} t' & \text{if } S_{X}[r, i] = 0 \text{ or } S_{Y}[r, j] = 0, \\ \max(t', StrLCS_{r-1}[S_{X}[r, i] - 1, S_{Y}[r, j] - 1] + k_{r}) & \text{if } S_{X}[r, i] \neq 0 \text{ and } S_{Y}[r, j] \neq 0. \end{cases}$$

$$t = \begin{cases} -\infty & \text{if } (i = 0 \text{ or } j = 0), \\ StrLCS_r[i - 1, j - 1] + 1 & \text{if } (i, j > 0) \text{ and } x_i = y_j, \\ \max(StrLCS_r[i - 1, j], StrLCS_r[i, j - 1]) & \text{if } (i, j > 0) \text{ and } x_i \neq y_j. \end{cases}$$
(8)

$$t' = \begin{cases} t & \text{if } NOV_X[r, i] = 0 \text{ or } NOV_Y[r, j] = 0, \\ \max(t, StrLCS_{r-1}[NOV_X[r, i] - 1, NOV_Y[r, j] - 1] + NOV\ell_r) & \text{if } NOV_X[r, i] \neq 0 \text{ and } NOV_Y[r, j] \neq 0. \end{cases}$$
(9)

$$StrLCS_{r}[i, j] = \begin{cases} t' & \text{if } S_{X}[r, i] = 0 \text{ or } S_{Y}[r, j] = 0, \\ \max(t', StrLCS_{r-1}[S_{X}[r, i] - 1, S_{Y}[r, j] - 1] + k_{r}) & \text{if } S_{X}[r, i] \neq 0 \text{ and } S_{Y}[r, j] \neq 0. \end{cases}$$

$$(10)$$

The boundary conditions are as follows: $StrLCS_0[i, 0] = StrLCS_0[0, j] = 0$ and $StrLCS_r[i, 0] = StrLCS_r[0, j] = -\infty$. In Eqs. (8) and (9), t and t', respectively, are used for intermediate computation.

The analysis is simple and follows readily from previous analysis. For all $1 \le r \le p$, at each step, the computation of $S_X[r,i]$, $S_Y[r, j]$, $NOV_X[r, i]$, $NOV_Y[r, j]$ and $StrLCS_r[i, j]$ takes $O(n^2)$ time. Thus the computation time is $O(n^2p)$ in total.

6. Conclusion

In this paper, we have studied some variants of the CLCS problem, namely, the Substring-IC-LCS and generalized Substring-IC-LCS problems. We have presented two $O(n^2)$ time algorithms for solving the Substring-IC-LCS problem, improving the previously known $O(n^2k)$ time algorithm. Clearly this is a significant (from cubic to quadratic) improvement. An intriguing finding is that the algorithm is completely independent of the length of the constraint string. We have also presented an algorithm for the generalized Substring-IC-LCS in $O(n^2p)$ time where we are given an ordered list of p constraint patterns and we want to find an LCS String containing all of them as substring in the given order.

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