

Name: Majd Kawak / NetID: mkawa025 / CodeForces ID: mkawa025 /
Student ID: 862273310

A. Longest decreasing subsequence

Submission#: 180250359

In this program, firstly I declared an *int* that would hold array size, then I declare an *array* with *int* size. I move on to input all values and store them inside my *array*. I declare a function called *compute_LDS* that takes my *array* and its *size* as parameters. Inside *compute_LDS* I declare a *vector* that would hold *LDS* value at each index, initialize all *vector* elements to 1 (its longest length at each index value). Then I go through my *array* elements (*ints*) using double *for* loop, computing each *LDS* value in my *vector* iteratively using Recurrence Relation below:

$$l_i = \max \begin{cases} 1 \\ \max_{0 < j < i, a_j > a_i} \{l_j + 1\} \end{cases}$$

Lastly, I display max value in my *vector* which refers to max *LDS* value. Total time complexity = $O(n^2)$.

B. Weighted Edit Distance

Submission#: 180261248

In this program, I start with declaring an two *int* to hold my string *A* and *B* sizes. I move on to declaring two *vector* with string sizes(*A* and *B* respectively). After inputting both strings, I create a *DP* Matrix table to hold computed values. Then I declare a function called *compute_DP* that takes both my strings(*A* and *B*) with their sizes and my *DP* matrix as parameters. Inside *compute_DP* I compute all *DP* values iteratively using recursion below: Recurrence Relation below:

$$DP[i][j] = \begin{cases} \max\{i, j\} & (\text{when } i = 0 \vee j = 0) \\ DP[i-1][j-1] & (\text{when } i > 0 \wedge j > 0 \wedge x_i = y_j) \\ \min \begin{cases} DP[i][j-1] + \text{cost_of_inserting} \\ DP[i-1][j] + \text{cost_of_deleting} \\ DP[i-1][j-1] + \text{cost_of_replacing} \end{cases} & (\text{when } i > 0 \wedge j > 0 \wedge x_i \neq y_j) \end{cases}$$

Where:

cost_of_inserting = y_j value at $B[j]$

cost_of_deleting = x_i value at $A[i]$

cost_of_replacing = $|x_i - y_j|$

Lastly, I display last element computed at $DP[A.size][B.size]$ that refers to our min cost to transform *A* to *B*. Total time complexity = $O(m * n)$ where($m = A.size$ and $n = B.size$).

C. Longest Unimodal Subsequence

Submission#: 180263692

In this program, firstly I declared an *int* that would hold array size, then I declare an *array* with *int* size. I move on to input all values and store them inside my *array*. Then I declare a function called *compute_LUS* that takes my *array* and it's *size* as parameters. Inside *compute_LUS* I compute *LIS* on all my array elements from LEFT \rightarrow RIGHT iteratively, saving each index *LIS* value inside a *vector* called *lis_Index_Val* using Recurrence Relation below:

$$l_i = \max \begin{cases} 1 \\ \max_{0 < j < i, a_j < a_i} \{ (l_j + 1) \} \end{cases}$$

Then I compute *LDS* on all my array elements from RIGHT \rightarrow LEFT iteratively, saving each index *LDS* value inside a *vector* called *lds_Index_Val* using Recurrence Relation below:

$$l_i = \max \begin{cases} 1 \\ \max_{0 < j < i, a_j > a_i} \{ (l_j + 1) \} \end{cases}$$

Lastly I compute *LUS* max value using Longest Bitonic Subsequence logic, declare an *array* called *lus_Index_Val* of *size* identical to my original *array*. Iterate through all indices using Recurrence Relation below:

$$lus_Index_Val[i] = (lis_Index_Val[i] + lds_Index_Val[i]) - 1$$

(where we subtract 1 to remove index double counted in LIS and LDS)

Lastly, I display max value in my *array* *lus_Index_Val* which refers to max *LUS* value. Total time complexity $= O(n^2) + O(n^2) + O(n) = O(n^2)$.

References

- [1] geeksforgeeks *Longest Decreasing Subsequence* <https://www.geeksforgeeks.org/longest-decreasing-subsequence/>
- [2] Sun, Yihan. "Lecture - Dynamic Programming" CS 141, University of California Riverside. pdf presentation.
- [3] geeksforgeeks *Longest Bitonic Subsequence* <https://www.geeksforgeeks.org/longest-bitonic-subsequence-dp-15/>