## Wagner-Fischer algorithm for calculating Levenshtein distance

<u>Aim</u>: To find the minimum edit distance for converting source string to the target string following the dynamic programming approach given the cost of insertion as 1, cost of deletion as 1 and cost of substitution as 2.

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Input: i) A source string
        ii) A target string
        Eg : source string = "intention"
            target string = "execution"
Output: i) Minimum edit distance for converting source string to the target string.
          ii) Number of insertions, deletions and substitutions performed.
          iii) Actual operations performed.
          Eg: Minimum edit distance = 8
               Number of insertions = 1
               Number of deletions = 1
               Number of substitutions = 3
               Total number of operations = 5
               Actual operations :
                                     i) delete 'i'
                                      ii) substitute 'n' by 'e'
                                      iii) substitute 't' by 'x'
                                      iv) insert 'c'
                                      v) substitute 'n' by 'u'
Algorithm:
       - Function MIN EDIT DISTANCE (source, target) returns minimum edit distance
       - m is the length of the source string
       - n is the length of the target string
        1) Create a matrix 'distance' of size (n+1, m+1) whose each element represents a distance
          value.
       2) Initialize the zero<sup>th</sup> row and column to be the distance from the empty string
               distance[0,0] = 0
               for each column i for 1 to n do
                       distance[i,0] = distance[i-1,0] + insertion cost(target[i])
               for each column i for 1 to m do
                       distance[0,i] = distance[0,i-1] + deletion cost(source[i])
       3) for each column i for 1 to n do
               for each row j for 1 to m do
                       if source[i-1] is equal to target[i-1] then
                              dp[i,j] = dp[i-1,j-1]
                       else:
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dp[i,j] = min(dp[i-1,j] + insertion cost, dp[i-1,j-1] + substitution cost,

dp[i,i-1] + deletion cost)

4) Backtrack to record the operations performed set i=n+1 and j=m+1

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while i and i both > 0 do
        if source[i-1] is equal to target[i-1] then
               i = i - 1
               i = i - 1
        else:
               if dp[i,j] is equal to (dp[i-1, j-1] + substitution cost) then
                       add the substitution operation to the list of operations
                       i = i - 1
                       i = i - 1
               else if dp[i,j] is equal to (dp[i-1, j] + insertion cost) then
                       add the insertion operation to the list of operations
                       i = i - 1
                else:
                       add the deletion operation to the list of operations
while j is not equal to 0 do
        add the deletion operation to the list of operations
       i = i - 1
while i is not equal to 0 do
        add the insertion operation to the list of operations
       i = i - 1
```

5) Reverse the list of operations [as they are added in reverse order because of backtracking] return (distance[n+1,m+1], list\_of\_operations)

## **More details:**

The minimum edit distance problem has the two properties: overlapping subproblems and optimal substructure. Thus dynamic programming approach can be applied for solving this problem. It takes into consideration all possible combinations of the source string and target string. The algorithm goes step-by-step, considers each combination and calculate the minimum edit distance for it. The result of this subproblem is then used to find the solution of the next following combination and thus we reach to the final solution.

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Time complexity for the algorithm = O(m * n)

Space complexity for the algorithm = O(m * n)

where, m is the length of the source string

n is the length of the target string
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