

DAA LAB -6

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1. Longest common subsequence:

// Implementing longest common subsequence to find longest subsequence between 2 strings.

// Input: String s1, s2, and 2 pointers m and n

// Output: length of the longest common subsequence

```
def lcs(s1, s2, m, n):
```

```
    if m == 0 or n == 0:
```

```
        return 0
```

```
    if memo[m][n] != -1:
```

```
        return memo[m][n]
```

```
    if s1[m-1] == s2[n-1]:
```

```
        memo[m][n] = 1 + lcs(s1, s2, m-1, n-1)
```

```
        return memo[m][n]
```

```
    else:
```

```
        memo[m][n] = max(lcs(s1, s2, m, n-1),
```

```
                          lcs(s1, s2, m-1, n))
```

```
    return memo[m][n]
```

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Time Complexity:

a) Brute Force

- Generate all subsequences
- for string s_1 of length m ($m \leq n$) there are 2^m possible subsequences. This is because the characters of s_1 can be either included or excluded in a subsequence leading to 2^m combinations.
- Checking each subsequence
- for each subsequence generated from s_1 we need to check if it is a subsequence of s_2 of length n . This is done in time $\sum_{i=1}^n 1 = O(n)$

$$\therefore \text{Time Complexity} = O(2^{\min(m,n)} \cdot \max(m,n))$$
$$= O(2^m \times n) \quad \gamma$$

b) Subproblems

Without memoization the naive recursive approach ~~would~~ have looked at all the subsequence of each character s_1 and s_2 . However with memoization we avoid recalculating subproblems. Instead we calculate them only once. There are ' m ' possible lengths of substring of s_1 and ' n ' possible lengths of s_2 . \therefore no. of unique subproblems is ' $m \times n$ '. Since there are ' $m \times n$ ' unique subproblems and each problem is solved in constant time $O(1)$ $\therefore T(c) = O(m \times n)$ γ

for 'n' strings:

Algorithm Find LCS Multiple (strings)

// computes the LCS among n strings

// Input: Array of n strings

// Output: LCS of n strings

man-seq = ""

for i = 0 to n-1:

LCS = s[i]

for j = 0 to n-1:

if i ≠ j:

LCS = FindLCS(LCS, s[j])

if LCS == "":

break

man-seq = man(man-seq, LCS, key=length)

Return man-seq.

Time Complexity:

- 1) There are 2 loops, one from 0 to n-1, other from 0 to n-1
- 2) In the innermost operation, computing LCS is $O(L^2)$ in the worst case, assuming an average string length of L.
- 3) Hence total time complexity is

$$T = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} O(L^2) = \cancel{O(n^2 L^2)} \quad n^2 L^2 = O(n^2)$$

Test cases (lcs) : Positive

①

ID	Grades
S1	CCDDAA...CC
S2	DDCCAB...FF
S3	FFBB...BBCC
⋮	⋮
S20	ABBC...CCDD

Output: DDD

④

ID	Grades
S1	DDFF...BBCC
S2	ABBB...BBCC
S3	AAAB...CCDD
⋮	⋮
S20	BBCC...DDFF

Output: BBCCC

②

ID	Grades
S1	AABB...AABB
S2	DDCC...AABB
S3	ABBC...BBCC
⋮	⋮
S20	BBAA...BBCC

Output: BBBBC

⑤

ID	Grades
S1	ABBCC...DD
S2	AAAB...CCDD
S3	AAAA...BBBB
⋮	⋮
S20	ABBC...CCDD

Output: BCCD

③

ID	Grades
S1	BCFF...AABB
S2	CCDD...ABBC
S3	BBCC...DDFF
⋮	⋮
S20	AAAB...BBBC

Output: BBCCC

Negative Test cases

① ID	Grades
S1	BBCC...CCFF
S2	AAAB...BCC@
S3	ABBC...CCDD
⋮	⋮
S20	ABCC...AABB

④ ID	Grades
S1	AB@...CCDD
S2	ABBC...C@DD
⋮	⋮
S20	AAAB...CCDD

Expected output: special character used

Output: string length less than 40 and special characters

② ID	Grades
S1	AAABCC
S2	ABBC CCDD
⋮	⋮
S20	AB

⑤ ID	Grades
S1	AFBD...EECK
S2	AABC...CBBD
⋮	⋮
S20	ACCA...ABBK

Expected output: string length less than 40

Output: Invalid grades

③ ID	Grades
S1	A1B2...C3D4
S2	A2B8...CCDD
⋮	⋮
S20	A9B3...CCDF

Expected output: Numbers used in string

Matrix Chain Multiplication (P)

// We implement Matrix Chain Multiplication to find most efficient way to multiply 'n' matrices.

// Input: List containing size of matrix & 2 pointers i & j

// Output: The minimum steps of matrix chain.

def mem(p):

n = p.length - 1

let $m[1 \dots n, 1 \dots n]$ and $s[1, n-1, 2 \dots n]$ be new tables

for $i = 1$ to n

$m[i, i] = 0$

for $j = 2$ to n

for $i = 1$ to $n - j + 1$

$j = i + j - 1$

$m[i, j] = \infty$

for $k = i$ to $j - 1$

$q = m[i, k] + m[k+1, j] + p_i - 1 \cdot p_k \cdot p_j$

if $q < m[i, j]$

$m[i, j] = q$

$s[i, j] = k$

return m and s

Time Complexity

$$\sum_{i=1}^n 1 + \sum_{l=2}^n \sum_{i=1}^{n-l+1} \sum_{k=i}^{j-1} 1$$

$$(n-1+1) + \sum_{l=2}^n \sum_{i=1}^{n-l+1} (j-1-i+1)$$

$$n + \sum_{l=2}^n \sum_{i=1}^{n-l+1} (l-1-i)$$

$$n + \sum_{l=2}^n \sum_{i=1}^{n-l+1} l-1$$

$$= n + \sum_{l=2}^n (l-1)(n-l+1-1+1)$$

$$= n + (l-1)(n-l+1) \sum_{l=2}^n 1$$

$$\therefore (l-1)(n-l+1) = (l-1)(n+1) - l(l+1)$$

$$\Rightarrow n + (n+1) \sum_{l=2}^n l-1 + \sum_{l=2}^n (-l)(l+1)$$

$$= n + (n+1) \sum_{l=1}^{n-1} l - \sum_{l=2}^n l^2 + \sum_{l=2}^n l$$

$$\left\{ \because \sum_{l=2}^n l-1 = \sum_{l=1}^{n-1} l \right\}$$

$$= n + \frac{(n+1)n(n-1)}{2} - \left(\frac{n(n+1)(2n+1)}{6} - 1 \right) + \left(\frac{n(n+1)}{2} - 1 \right)$$

$$= n + \frac{n^3}{2} - \frac{n^2}{2} - \frac{2n^3 + 3n^2 + n}{6} + 1 + \frac{n^2 + n}{2} - 1$$

$$\approx O(n^3)$$

Test cases : mcm

- 1] $([7, 5, 4, 6, 7, 8], 5) \rightarrow \text{o/p} : 504$
- 2] $([3, 7, 5, 10, 15], 4) \rightarrow \text{o/p} : 255$
- 3] $([2, 4, 5, 6, 8], 4) \rightarrow \text{o/p} : 100$
- 4] $([4, 8, 6, 7, 9], 4) \rightarrow \text{o/p} : 360$
- 5] $([7, 3, 6, 4, 8], 4) \rightarrow \text{o/p} : 156$
- 6] $([3, 7, 4, 7, 5]) \rightarrow \text{o/p} : \text{Error}$
- 7] $([2, 5, 6], 1) \rightarrow \text{o/p} : \text{Error}$
- 8] $([10, 20, 30], 2) \rightarrow \text{o/p} : \text{Error}$
- 9] $([10, -20], 1) \rightarrow \text{o/p} : \text{Error}$
- 10] $([0, 20, 10], 2) \rightarrow \text{o/p} : \text{Error}$

Conclusion :

Hence, we have studied the SOLID principles of programming and also studied Algorithms to find Longest Common Subsequence and matrix chain multiplication.