# CSC373 Worksheet 3 Solution

## August 2, 2020

### 1. Using the following formula

$$M[i,j] = \begin{cases} 0 & \text{if } i = j\\ \min_{i \le k \le j} M[i,k] + M[k+1,j] + p_{i-1}p_k p_j & \text{if } i < j \end{cases}$$
 (1)

we have

$\Box$	1	2	3	4	5	6
1	0	350	770	612	1212	1422
2	X	0	840	462	1662	1362
3	X	X	0	252	1092	1098
4	X	X	X	0	1440	936
5	X	X	X	X	0	720
6	X	X	X	X	X	0

And an optimal parenthesization is

### My Work:

$$(A_1A_2)(A_3A_4)(A_5A_6)$$

### **Correct Solution:**

Using the following formula

$$M[i,j] = \begin{cases} 0 & \text{if } i = j \\ \min_{i \le k \le j} M[i,k] + M[k+1,j] + p_{i-1}p_k p_j & \text{if } i < j \end{cases}$$
 (2)

we have

m (i/j)	1	2	3	4	5	6
1	0	350	770	612	1212	1422
2	X	0	840	462	1662	1362
3	X	X	0	252	1092	1098
4	X	X	X	0	1440	936
5	X	X	X	X	0	720
6	X	X	X	X	X	0

s $(i/j)$	1	2	3	4	5	6
1	0	1	2	1	4	4
2	X	0	2	2	4	4
3	X	X	0	3	4	4
4	X	X	X	0	4	4
5	X	X	X	X	0	5
5	X	X	X	X	X	0

And an optimal parenthesization is

## My Work:

$$(A_1(A_2(A_3A_4)))(A_5A_6)$$

## Notes:

• Sequence of Dimensions

The sequence of dimensions  $< p_0 = 5, p_1 = 10, p_2 = 3, p_2 = 12, p_3 = 5, p_4 = 50, p_5 = 6 >$  means there are 6 matrices with dimensions  $p_{i-1} \times p_i$ 

- $-A_1 \rightarrow 5 \times 10$
- $-A_2 \rightarrow 10 \times 3$
- $-A_3 \rightarrow 3 \times 12$
- $-A_4 \rightarrow 12 \times 5$
- $-A_5 \rightarrow 5 \times 50$
- $-A_6 \rightarrow 50 \times 6$
- Dynamic Programming
  - Is applied to optimization problems
  - Applies when the subproblems overlap
  - Uses the following sequence of steps
    - 1. Characterize the structure of an optimal solution

- 2. Recursively define the value of an optimal solution
- 3. Construct an optimal solution from computed information

### • Matrix-chain Multiplication

- Is an optimization problem solved using dynamic programming
- Goal is to find matrix parenthesis with fewest number of operations

### Example:

Given chain of matrices  $\langle A, B, C \rangle$ , it's fully parenthesized product is:

- \* (AB)C needs  $(10 \times 30 \times 5) + (10 \times 5 \times 60) = 1500 + 3000 = 4500$  operations
- \* A(BC) needs  $(30 \times 5 \times 60) + (10 \times 30 \times 60) = 27000$  operations

Thus, (AB)C performs more efficiently than A(BC).

- Is stated as: given a chain  $\langle A_1, A_2, ..., A_n \rangle$  of n matrices, where for i = 1, 2, ..., n matrix  $A_i$  has dimension  $p_{i-1} \times p_i$ , fully parenthesize the product  $A_1 A_2 ... A_n$  in a way that minimizes the number of scalar multiplications.
- Steps

### 1. Check is the problem has Optimal Substructure

Let us adopt the notation  $A_{i...j}$  where  $i \leq j$ , for the matrix that results from evaluating the product  $A_i A_{i+1} ... A_j$ .

Assume the solution has the following parentheses:

$$(A_{i...k})(A_{k+1...j})$$

If there is a better way to multiply  $(A_{i...k})$ , then we would have a more optimal solution.

This would be a contradiction, as we already stated that we have the optimal solution for  $A_{i...j}$ .

Therefore, this problem has optimal substructure.

#### 2. Find the Recursive Solution

Let M[i,j] be the cost of multiplying matrices from  $A_i$  to  $A_j$ 

We want to find out at which k' returns the fewest number of multiplications, or the minimum number of M.

The recursive formula for the cost of multiplying from  $A_i$  to  $A_j$  is

$$M[i,j] = \begin{cases} 0 & \text{if } i = j \\ \min_{i \le k \le j} M[i,k] + M[k+1,j] + p_{i-1}p_k p_j & \text{if } i < j \end{cases}$$
(3)

- 3. Computing the Estimated Cost
  - \* Steps
    - 1) Fill the table for i = j
    - 2) Fill the table for i < j with a spread of 1
    - 3) Repeat 2 with the increased value of spread

### Example:

Given

$$< A_1, A_2, A_3, A_4, A_5 >$$

where

\* 
$$A_1 \rightarrow 4 \times 10$$

\* 
$$A_2 \rightarrow 10 \times 3$$

\* 
$$A_3 \rightarrow 3 \times 12$$

\* 
$$A_4 \rightarrow 12 \times 20$$

\* 
$$A_5 \rightarrow 20 \times 7$$

we have:

1) Fill the table for i = j

i\j	1	2	3	4	5
1	0				
2	x	0			
3	x	x	0		
4	x	x	x	0	
5	x	x	х	х	0

$$M[i,j] = \begin{cases} 0 & \text{if } i = j \\ \min_{i \le k \le j} M[i,k] + M[k+1,j] + p_{i-1}p_kp_j & \text{if } i < j \end{cases}$$

2) Fill the table for i < j with a spread of 1

2) 
$$(i = 1, j = 2)$$
,  $(i = 2, j = 3)$ ,  $(i = 3, j = 4)$ ,  $(i = 4, j = 5)$ 

i\j	1	2	3	4	5
1	0	120			
2	x	0	360		
3	x	х	0	720	
4	x	x	x	0	1680
5	x	х	х	х	0

since

$$* i = 1, j = 2$$

$$M[1,2] = \min_{1 \le k \le 2} (M[1,1] + M[1,2] + p_{i-1}p_k p_j)$$
 (4)

$$= \min_{1 \le k \le 2} (0 + 0 + p_0 p_1 p_2) \tag{5}$$

$$= \min_{1 \le k \le 2} (0 + 0 + 4 \cdot 10 \cdot 3) \tag{6}$$

$$= 120 \tag{7}$$

where  $p_0 = 3$  is from the dimension  $3 \times 10$  of  $A_1$ ,  $p_k = 10$  is from the dimension of  $3 \times 10$  of  $A_1$ .

$$*i = 2, j = 3$$

$$M[2,3] = \min_{2 \le k \le 3} (M[2,2] + M[3,3] + p_{i-1}p_k p_j)$$
(8)

$$= \min_{2 \le k \le 3} (0 + 0 + p_1 p_2 p_3) \tag{9}$$

$$= \min_{2 \le k \le 3} (0 + 0 + 10 \cdot 3 \cdot 12) \tag{10}$$

$$=360$$
 (11)

$$*i = 3, j = 4$$

$$M[3,4] = \min_{3 \le k \le 4} (M[3,3] + M[4,4] + p_{i-1}p_k p_j)$$
 (12)

$$= \min_{3 \le k \le 4} (0 + 0 + p_2 p_3 p_4) \tag{13}$$

$$= \min_{3 \le k \le 4} (0 + 0 + 3 \cdot 12 \cdot 20) \tag{14}$$

$$=720\tag{15}$$

\* 
$$i = 4, j = 5$$

$$M[4,5] = \min_{4 \le k \le 5} (M[4,4] + M[5,5] + p_{i-1}p_k p_j)$$
 (16)

$$= \min_{4 \le k \le 5} (0 + 0 + p_3 p_4 p_5) \tag{17}$$

$$= \min_{4 \le k \le 5} (0 + 0 + 12 \cdot 20 \cdot 7) \tag{18}$$

$$= 1680 \tag{19}$$

3) Repeat 2 with the increased value of spread

2) 
$$(i = 1, j = 2)$$
,  $(i = 2, j = 3)$ ,  $(i = 3, j = 4)$ ,  $(i = 4, j = 5)$ 

i\j	1	2	3	4	5
1	0	120	264	1080	1344
2	x	0	360	1320	1350
3	х	х	0	720	1140
4	х	x	x	0	1680
5	x	x	х	х	0

$$* i = 1, j = 3$$

$$\underline{k} = \underline{1}$$

$$M[1,3] = M[1,1] + M[2,3] + p_{i-1}p_kp_j$$
(20)

$$= 0 + 360 + p_0 p_1 p_3 \tag{21}$$

$$= 0 + 360 + 4 \cdot 10 \cdot 12 \tag{22}$$

$$= 0 + 360 + 480 \tag{23}$$

$$= 840 \tag{24}$$

 $\underline{k} = 2$ 

$$M[1,3] = M[1,2] + M[3,3] + p_{i-1}p_kp_j$$
(25)

$$= 120 + 0 + p_0 p_2 p_3 \tag{26}$$

$$= 120 + 0 + 4 \cdot 10 \cdot 12 \tag{27}$$

$$= 120 + 0 + 144 \tag{28}$$

$$= 264 \tag{29}$$

Thus,  $\min_{1 \le k \le 3} M[1, 3] = 264$ .

$$*i = 2, j = 4$$

k = 2

$$M[2,4] = M[2,2] + M[3,4] + p_{i-1}p_kp_i$$
(30)

$$= 0 + 720 + p_1 p_2 p_4 \tag{31}$$

$$= 0 + 720 + 10 \cdot 3 \cdot 20 \tag{32}$$

$$= 0 + 720 + 600 \tag{33}$$

$$= 1320$$
 (34)

k = 3

$$M[2,4] = M[2,2] + M[3,4] + p_{i-1}p_kp_i$$
(35)

$$= 360 + 0 + p_1 p_3 p_4 \tag{36}$$

$$= 360 + 0 + 10 \cdot 12 \cdot 20 \tag{37}$$

$$= 360 + 0 + 2400 \tag{38}$$

$$=2760$$
 (39)

Thus,  $\min_{2 \le k \le 4} M[2, 4] = 1320$ .

$$*i = 3, j = 5$$

k=3

$$M[3,5] = M[3,3] + M[3,5] + p_{i-1}p_k p_j$$
(40)

$$= 0 + 1680 + p_2 p_3 p_5 \tag{41}$$

$$= 0 + 1680 + 3 \cdot 12 \cdot 7 \tag{42}$$

$$= 0 + 1680 + 252 \tag{43}$$

$$= 1932 \tag{44}$$

 $\underline{k=4}$ 

$$M[3,5] = M[3,4] + M[5,5] + p_{i-1}p_kp_i$$
(45)

$$= 720 + 0 + p_2 p_4 p_5 \tag{46}$$

$$= 720 + 0 + 3 \cdot 20 \cdot 7 \tag{47}$$

$$= 720 + 420 \tag{48}$$

$$= 1140 \tag{49}$$

Thus,  $\min_{3 \le k \le 5} M[3, 5] = 1140$ .

\* 
$$i = 2, j = 5$$

 $\underline{k=2}$ 

$$M[2,5] = M[2,2] + M[3,5] + p_{i-1}p_kp_j$$
(50)

$$= 0 + 1140 + p_1 p_2 p_5 \tag{51}$$

$$= 0 + 1140 + 10 \cdot 3 \cdot 7 \tag{52}$$

$$= 0 + 1140 + 210 \tag{53}$$

$$=1350\tag{54}$$

 $\underline{k=3}$ 

$$M[2,5] = M[2,3] + M[4,5] + p_{i-1}p_kp_j$$
(55)

$$= 360 + 1680 + p_1 p_3 p_5 \tag{56}$$

$$= 2040 + 10 \cdot 12 \cdot 7 \tag{57}$$

$$= 2040 + 840 \tag{58}$$

$$=2880\tag{59}$$

 $\underline{k=4}$ 

$$M[2,5] = M[2,4] + M[5,5] + p_{i-1}p_k p_j$$
(60)

$$= 1320 + p_1 p_3 p_5 \tag{61}$$

$$= 1320 + 10 \cdot 20 \cdot 7 \tag{62}$$

$$= 1320 + 1400 \tag{63}$$

$$=2720$$
 (64)

Thus,  $\min_{2 \le k \le 5} M[2, 5] = 1350$ .

$$* i = 1, j = 5$$

 $\underline{k=1}$ 

$$M[1,5] = M[1,1] + M[3,5] + p_{i-1}p_k p_j$$
(65)

$$= 0 + 1350 + p_0 p_1 p_5 \tag{66}$$

$$= 0 + 1350 + 4 \cdot 10 \cdot 7 \tag{67}$$

$$= 0 + 1350 + 280 \tag{68}$$

$$= 1630 \tag{69}$$

 $\underline{k=2}$ 

$$M[1,5] = M[1,2] + M[3,5] + p_{i-1}p_k p_i$$
(70)

$$= 120 + 1140 + p_0 p_2 p_5 \tag{71}$$

$$= 120 + 1140 + 4 \cdot 3 \cdot 7 \tag{72}$$

$$= 1260 + 84 \tag{73}$$

$$= 1344 \tag{74}$$

k=3

$$M[1,5] = M[1,3] + M[4,5] + p_{i-1}p_kp_i$$
(75)

$$= 264 + 1680 + p_0 p_3 p_5 \tag{76}$$

$$= 264 + 1680 + 4 \cdot 12 \cdot 7 \tag{77}$$

$$= 1944 + 336 \tag{78}$$

$$=2280\tag{79}$$

k = 4

$$M[1,5] = M[1,4] + M[5,5] + p_{i-1}p_kp_j$$
(80)

$$= 1080 + 0 + p_0 p_4 p_5 \tag{81}$$

$$= 1080 + 4 \cdot 20 \cdot 7 \tag{82}$$

$$= 1080 + 560 \tag{83}$$

$$= 1640$$
 (84)

Thus,  $\min_{1 \le k \le 5} M[1, 5] = 1344$ .

## 4. Constructing the Optimal Solution (Needs revision)

3) 5 1 2 3 120 1080 0 264 1344  $(A_1A_2)((A_3A_4)A_5)$ 1350 0 360 1320 2 х 720 1140 3 0 Х Х 0 1680 4 х Х Х O. 5 х Х Х

So, the optimal solution is  $(A_1A_2)((A_3A_4)A_5)$ 

### References:

- 1) CSBreakdown, Chain Multiplication Dynamic Programming, link
- 2) University of Maryland, CMSC351 Fall 2014 Homework # 4, link

```
2_1
       procedure MATRIX-CHAIN-MULTIPLY(A,s,i,j)
           if i == j then
               return A[i]
 3
           end if
 4
           if i < j
 6
                a = MATRIX-CHAIN-MULTIPLY(A,s,i,s[i,j])
               b = MATRIX - CHAIN - MULTIPLY(A,s,s[i,j] + 1,j)
9
               return MATRIX-CHAIN-MULTIPLY(a,b)
11
           end if
13
       end procedure
14
```

## Example:

- MATRIX-CHAIN-ORDER computes the table s containing optimal costs
- s table consists of k value at which m[i, j] is minimum!!

$$A_{1..s[1,n]}A_{s[1,n]+1..n}$$

- ullet Table of optimal costs m is used with table s to construct solution to matrix-chain multiplication problem
- 3. First, we need to determine the total number of times m[i, j] is referred in the innermost loop.

We know from the header that the loop runs from k = i to k = j - 1.

Using this fact, we can write the innermost loop has j - i = l - 1 iterations.

Since m[i,j] is referred twice, the total number of m[i,j] referred in the loop is:

$$(l-1)2\tag{1}$$

Second, we need to determine the total number of times m[i,j] is referred in the intermediate loop

We know from the header that the loop runs from i = 1 to i = n - l + 1.

Using this fact, we can write the intermediate loop runs n-l+1 iterations.

Since each iteration referrs m[i, j] (l-1)2 many times, the total number of times m[i, j] is referred in the itnermediate loop is:

$$(n-l+1)(l-1)2$$
 (2)

Finally, we need to determine the total number of times m[i,j] is referenced in the outermost loop.

We know from the header that the loop runs from l=2 to n.

Since each iteration referrs m[i,j] (n-l+1)(l-1)2 many times, the total number of times m[i,j] is referred in the outermost loop is:

$$\sum_{l=2}^{n} (n-l+1)(l-1)2 = 2\sum_{l'=1}^{n-1} (n-l')(l')$$
(3)

$$=2\left[\sum_{l'=1}^{n-1}nl'-(l')^2\right] \tag{4}$$

$$=2\left[n\sum_{l'=1}^{n-1}l'-\sum_{l'=1}^{n-1}(l')^2\right]$$
 (5)

$$=2\left[n\sum_{l'=1}^{n-1}l'-\sum_{l'=1}^{n-1}(l')^2\right]$$
 (6)

$$=2\left[n\sum_{l'=0}^{n-1}l'-\sum_{l'=0}^{n-1}(l')^2\right]$$
 (7)

$$=2\left[\frac{(n-1)n^2}{2} - \frac{(n-1)n(2n-1)}{6}\right] \tag{8}$$

$$=2\left[\frac{n^3-n^2}{2}-\frac{(n-1)n(2n-1)}{6}\right] \tag{9}$$

$$=2\left[\frac{n^3-n^2}{2}-\frac{(n^2-n)(2n-1)}{6}\right] \tag{10}$$

$$=2\left[\frac{n^3-n^2}{2} - \frac{(3n^3-3n^2+n)}{6}\right] \tag{11}$$

$$=2\left[\frac{3n^3-3n^2}{6}-\frac{(3n^3-3n^2+n)}{6}\right] \tag{12}$$

$$=2\left\lceil\frac{n^3-n}{6}\right\rceil\tag{13}$$

$$=\frac{n^3-n}{3}\tag{14}$$

### Notes:

• Hint from equation A.3

$$\sum_{k=0}^{n} k^2 = \frac{n(n+1)(2n+1)}{6}$$

• I feel the need to gain more insight regarding the question's 'other table entries in a call of MATRIX-CHAIN-ORDER'. What does 'other table entries' mean?

### **Answer:**

```
MATRIX-CHAIN-ORDER (p)
                                n = p.length - 1
                                let m[1 \dots n, 1 \dots n] and s[1 \dots n-1, 2 \dots n] be new tables
                                for i = 1 to n
                                     m[i,i] = 0
                            5 for l = 2 to n
                                                           //l is the chain length
                                    for i = 1 to n - l + 1
                                         j = i + l - 1
                                         m[i,j] = \infty
The other entries
                                         for k = i to j - 1
          :)
                            10
                                             q = m[i, k] + m[k + 1, j] + p_{i-1}p_kp_j
                            11
                                             if q < m[i, j]
                                                  m[i,j] = q
                           12
                                                  s[i, j] = k
                            13
                            14
                                return m and s
```

• In the problem 'm[i,j] is referenced' refers to m[i,j] used in assignments q=m[i,j]...

```
MATRIX-CHAIN-ORDER (p)
   n = p.length - 1
    let m[1 \dots n, 1 \dots n] and s[1 \dots n-1, 2 \dots n] be new tables
    for i = 1 to n
4
         m[i,i] = 0
    for l = 2 to n
                               # l is the chain length
         for i = 1 to n - l + 1
 6
7
             j = i + l - 1
 8
             m[i,j] = \infty
9
             for k = i to j - 1
                  q = m[i,k] + m[k+1,j] + p_{i-1}p_kp_j
10
11
                  if q < m[i, j]
12
                      m[i,j] = q
                      s[i, j] = k
13
14
   return m and s
```

### 4. Solution



Memoization fails to speed up the algorithm because it lacks overlapping subproblems.

## Notes:

- Elements of Dynamic Programming
  - Optimal Substructure
    - \* Is the first step to solving dynamic programming problem
    - \* Exists if an optimal solution to the problem contains within it optimal solution to subproblems
  - Overlapping Subproblems
    - \* Is the second step to solving dynamic programming
    - \* Exists when an algorithm revisits the same problem repeatedly
  - Memoization
    - \* Maintains an entry in a table for the solution to each subproblem
    - \* Ensures that a method doesn't run for the same inputs more than once
- Top-Down Dynamic Programming
  - Uses recursion
  - Is preferred
    - \* When all sub-solutions need to be solved
    - \* Because it's easier



- Bottom-Up Dynamic Programming
  - Uses for-loop
  - Is preferred
    - \* When all sub-solutions need to be solved
    - \* Because it is sometimes faster (No recursive call and no unnecessary Random Memory access)
  - Is preferred when not all sub-solutions need to be computed



## • Merge Sort

- How it works
  - 1. Find the middle point to divide the array into two halves
  - 2. Call mergeSort for first half
  - 3. Call mergeSort for second half
  - 4. Merge two halves in sorted order



#### References

1)

### 5. Yes. This problem does exhibit optimal substructure

*Proof.* Assume that the optimal structure of  $A_iA_{i+1}..A_j$  exists. That is, there exists some  $k \in \mathbb{N} - \{0\}$  such that the following parenthesis  $(A_{i..k})(A_{k+1..j})$  produces maximum operation cost.

I need to show that the substructures  $A_{i..k}$  and  $A_{k+1..j}$  are also optimal.

I will do so in parts.

## Part 1 (Proving optimal substructure of $A_{i...k}$ )

Assume for the sake of contradiction that the substructure  $A_{i..k}$  is not optimal.

Then, we can write that there exists some  $k' \in \mathbb{N} - \{0\}$  such that  $(A_{i...k'})(A_{k'+1..k})$  has larger operation cost than  $(A_{i...k})$ .

Then, we can write that  $((A_{i...k'})(A_{k'+1..k}))(A_{k+1..j})$  has larger operation costs than  $(A_{i..k})(A_{k+1..j})$ , which contradicts the original assumption that  $(A_{i..k})(A_{k+1..j})$  has maximum operation cost.

Thus,  $(A_{i..k})$  must be optimal.

## Part 2 (Proving optimal substructure of $A_{k+1...j}$ )

This proof is nearly verbatim as part 1, where the only difference is using  $A_{k+1..j}$  instead of  $A_{i..j}$ .

### Notes:

- A problem has **optimal substructure** if an optimal solution can be constructed from optimal solutions of its subproblems. [3]
- Showing optimal subtructure for the original Matrix-Chain Multiplication problem

Assume that the optimal structure of  $A_iA_{i+1}..A_j$  exists. That is, there exists some  $k \in \mathbb{N} - \{0\}$  such that the following parenthesis  $(A_{i..k})(A_{k+1..j})$  produces minimum operation cost.

I need to show that the substructures  $A_{i..k}$  and  $A_{k+1..j}$  are also optimal.

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## Part 1 (Proving optimal substructure of $A_{i...k}$ )

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This proof is nearly verbatim as part 1, where the only difference is using  $A_{k+1..j}$  instead of  $A_{i..j}$ .

#### References:

- 1) CSBreakdown, Chain Multiplication Dynamic Programming, link
- 2) CodeScope, Dynamic Programming, link
- 3) Wikipedia, Optimal Substructure, link
- 6. Let < 2, 10, 20, 5 >.

Then, we see that  $p_0p_1p_3$  is minimum, and by Professor Capulet's claim, splitting at k=1 should result in minimum-cost matrix multiplication.

But we have

m	1	2	3
1	0	400	600
2	X	0	1000
3	X	X	0

$\mathbf{S}$	1	2	3
1	0	1	2
2	X	0	2
3	X	X	0

And the minimum-cost matrix multiplication occurs when  $A_{i..j}$  is splitted at k=2.

Thus, Professor Capulet's claim is false.

## Notes:

- I need to find the matrices  $A_i, ..., A_j$  where the total operating cost of Professor Capulet's method is bigger than the properly parenthesized solution
- I feel the need for clarafication regarding the phrase 'always choosing the matrix  $A_k$  at which to split the product...' Is k in  $A_k$  the same in any matrix multiplications  $A_iA_{i+1}...A_j$ ?

#### Answer:

No. k in  $A_k$  is the value that makes  $p_{i-1}p_kp_j$  minimum.

### 7. Solution:



So, the LCS of < 1, 0, 0, 1, 0, 1, 0, 1 > and < 0, 1, 0, 1, 1, 0, 1, 1, 0 > is '101101'.

### Notes:

- Longest Common Sequence
  - Goal is to find the longest common subsequence in  $X = \langle x_1, x_2, ..., x_n \rangle$  and  $Y = \langle y_1, y_2, ..., y_n \rangle$
  - e.g. Given two substrings  $s_1 = \langle M, U, N, G \rangle$  and  $s_2 = \langle U, N, I, V, E, R, S, I, T, Y \rangle$ , the longest substring is ' $\langle U, N \rangle$ '.
  - Steps
    - 1. Verify Optimal Substructure
    - 2. Create Recursive Solution

$$M[i,j] = \begin{cases} 0 & i = 0, \text{ or } j = 0\\ c[i-1,j-1]+1 & \text{if } i,j > 0 \text{ and } a_i = b_j\\ Max(C_{i,j-1}, C_{i-1,j}) & \text{if } i,j > 0 \text{ } a_i \neq b_j \end{cases}$$
(15)

- 3. Computing the length of an LCS
  - \* Top-Down Solution
  - \* Bottom-Top Solution

```
LCS-LENGTH(X,Y)
           m = X.length
           n = Y.length
4
           let b[1..m, 1..n] and c[0..m, 0..n] be new tables
6
           for i = 1 to m
                c[i,0] = 0
9
           for j = 0 to n
                c[0,j] = 0
11
           for i = 1 to m
13
                for j = 1 to n
14
                     if x[i] == y[i]
15
                          c[i,j] = c[i-1,j-1] + 1
16
                          b[i,j] = '^{\prime}'
17
                     elseif c[i-1, j] \ge c[i,j-1]
19
                          c[i,j] = c[i,j-1]
20
                          b[i,j] = '\uparrow'
21
22
                     else
23
                          c[i,j] = c[i,j-1]
24
                          b[i,j] = '\leftarrow'
25
26
           return c and b
27
```

## 4. Constructing an LCS

## Steps:

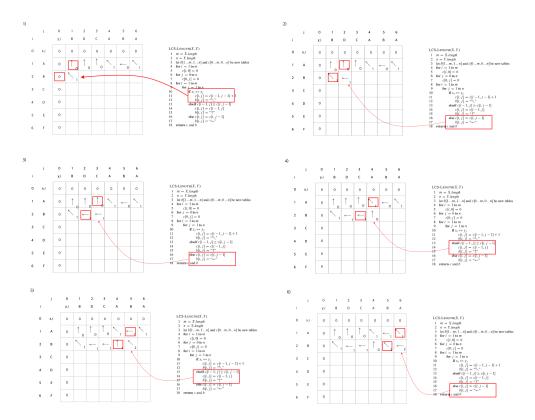
1) Fill i = 0 or j = 0 with 0



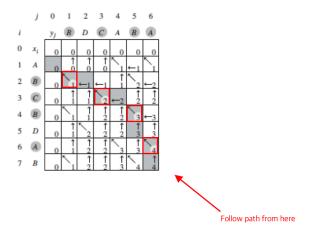
2) Compute i = 1 (needs revision)



3) compute i = 2 (needs revision)



- 4) Compute other entries using the same steps as above
- 5) Reconstruct LCS
  - \* Follow the arrows the lower right-hand corner
  - \* Select all '<



Here, the result is 'BCBA'

```
9_1
       LIS(A)
2
           let B be new array
3
           j = 0
5
           for i = 0 to n
6
                if i = 0
                     B[j] = A[i]
9
                elseif A[i] < max(B)</pre>
10
                     REPLACE-NEXT-VALUE-IN-B-BIGGER-THAN-Ai(B,A[i]) #Done
11
      using binary search
12
                elseif A[i] \leq B[j]
13
                     B[j] = A[i]
14
                else
15
                     j += 1
16
                    B[j] = A[i]
17
18
           return B
19
```

```
Correct Solution:
      LIS(A)
1
           let B be new array
           j = 0
           for i = 0 to n
               if i = 0
                    B[j] = A[i]
9
               elseif A[i] \leq B[j]
10
                    B[j] = A[i]
11
12
               else
                    j += 1
13
                    B[j] = A[i]
14
16
           return B
17
```

## Notes:

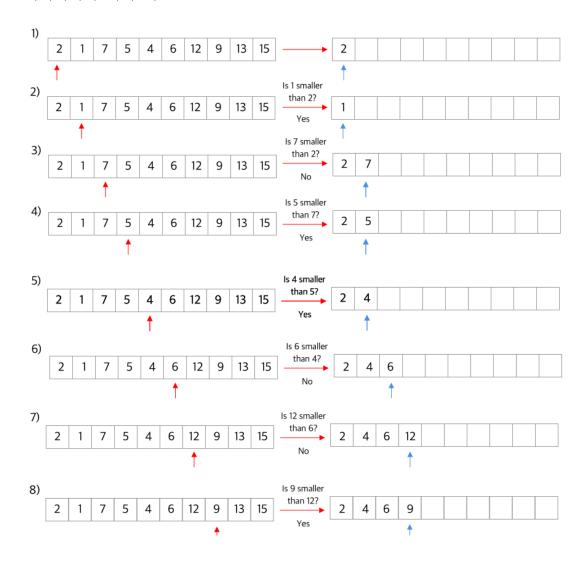
- Longest Increasing Subsequence
  - LIS  $\rightarrow$  LCS but a minor variation
    - \* Is the same as LCS's  $A = \langle x_1, x_2, x_3, ..., x_n \rangle$

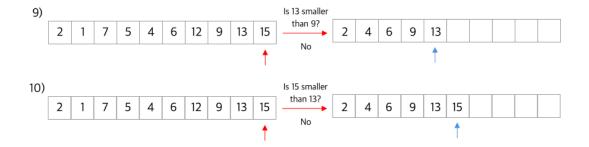
$$B = < min(A), min(A) + 1, ..., max(A) >$$

- Finds the longest and largest monotonically increasing subset of numbers

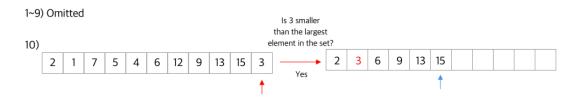
## Example:

2, 1, 7, 4, 6, 12, 9, 13, 15





## Example 2:



Considers values and not indices

\*

## Rough Work:

**Goal:** Find the longest palindromic subsequence

## Brute Force Method

- 10. For each combination of subsequence, check if substring is palinrome
  - If palindrome,

Check if is the longest

\* If longest, replcae the value of a variable containing the current largest palindromic substring

## Notes:

• ith  $\mathbf{Prefix}$  of a sequence is a subsequence of a sequence of length i that occurs at the beginning of the sequence.

## Example:

$$X = < A, B, C, B, D, A, B >$$

Then

$$X_4 = \langle A, B, C, B \rangle$$