

LONGEST COMMON SUBSEQUENCE

Inspiration

- Biological applications often need to compare the DNA of two (or more) different organisms
- A strand of DNA consists of a string of molecules called bases, where the possible bases are adenine, guanine, cytosine, and thymine
- each of these bases by its initial letter, we can express a strand of DNA as a string over the finite set {A, C, G, T}

Inspiration

- For example, the DNA of one organism may be $S_1 = \text{ACCGGTCGAGTGCGCGGAAGGCCGGCCGAA}$, and the DNA of another organism may be $S_2 = \text{GTCGTTCGGAATGCCGTTGCTCTGTAAA}$.
- One reason to compare two strands of DNA is to determine how “**similar**” the two strands are, as some measure of how closely related the two organisms are.

Inspiration

- We can define similarity in many different ways
- First way - we can say that two DNA strands are similar if one is a substring of the other
- In our example, neither S_1 nor S_2 is a substring of the other.
- Second way - two strands are similar if the number of changes needed to **turn one into the other** is **small**

$S_1 = \text{ACCGGTCGAGTGCGCGGAAGCCGGCCGAA}$

$S_2 = \text{GTCGTTCGGAATGCCGTTGCTCTGTAAA}$

Inspiration

- Third way measure the similarity of strands S_1 and S_2 is by finding a third strand S_3
- In which bases in S_3 appear in each of S_1 and S_2 ; these bases must appear in the same order, but not necessarily consecutively
- Longer the strand S_3 we can find, the more similar S_1 and S_2 are

$S_1 = \text{ACCGGTGAGTGCGCGGAAGCCGGCCGAA}$

$S_2 = \text{GTCGTTGGAAATGCCGTTGCTCTGTAAA}$

1. Substring Similarity Example:

- S1: AGTACGTACGGTAG
- S2: TACGTA
- S2 is a substring of S1, so they are similar.

2. Edit Distance Example:

- S1: ACGTTAGC
- S2: ACGTGAGC
- Only one substitution is needed, so the strands are similar.

3. Longest Common Subsequence Example:

- S1: ACGTACGGTAC
- S2: GTACGCTAG
- The longest common subsequence is GTAC, with a length of 4, so the strands are considered similar.

Inspiration

- $S_1 = \text{ACCGGTGAGTGCGCGGAAGCCGGCCGAA}$
- $S_2 = \text{GTCGTTCGGAATGCCGTTGCTCTGTAAA}$
- S_3 is $\text{GTCGTCGGAAGCCGGCCGAA}$

Problem Statement

- Given two sequences X and Y , we say that a sequence Z is a common subsequence of X and Y if Z is a subsequence of both X and Y

Longest Common Subsequence

P = “BATD”

Q = “ABACD”

Longest Common Subsequence

P = “BATD” }
Q = “ABACD” }

Dynamic Programming

- 1. Recursive Solution**
- 2. Memorize Intermediate Results**
- 3. Bottom-Up**

Recursive Solution

LCS(P_0, Q_0) → 3

“**BATD**”
“**ABACD**”} → “**BAD**”

Case 1:

$P_0 = "x"$
 $Q_0 = "x"$

Case 2:

$P_0 = "x"$
 $Q_0 = "y"$

LCS(P_0, Q_0)
= 1 + **LCS(P_1, Q_1)**

LCS(P_0, Q_0)
= **max { LCS(P_1, Q_0), LCS(P_0, Q_1) }**

Recursive Solution

LCS(P_0, Q_0)

Case 1:

$P_0 = "x"$
 $Q_0 = "x"$

LCS(P_0, Q_0)
= 1 + **LCS(P_1, Q_1)**

“BATD”
“ABACD” } → “BAD”

Case 2:

$P_0 = "x"$
 $Q_0 = "y"$

LCS(P_0, Q_0)
= **max { LCS(P_1, Q_0), LCS(P_0, Q_1) }**

Recursive Solution

LCS(P_0, Q_0) → 3

“BATD”
“ABACD” } → “BAD”

Case 1:

$P_0 = “\boxed{P_1} \underline{x}”$
 $\underline{Q_0} = “\boxed{Q_1} \underline{x}”$

Case 2:

$P_0 = “x”$
 $Q_0 = “y”$

LCS(P_0, Q_0)

= I + LCS(P_1, Q_1)

LCS(P_0, Q_0)

= max { LCS(P_1, Q_0), LCS(P_0, Q_1) }

Recursive Solution

LCS(P_0, Q_0) → 3



Case 1:

$P_0 = "P_1 \underline{x}"$
 $\underline{Q_0} = "Q_1 \underline{x}"$

A diagram showing two boxes. The top box contains 'P₁' above 'x' and the bottom box contains 'Q₁' above 'x'. A green bracket groups 'P₁' and 'Q₁'. A green line connects the 'x's. A green bracket groups the entire pair under the heading 'Case 1:'.

Case 2:

$P_0 = "x"$
 $Q_0 = "y"$

$$\begin{aligned}\text{LCS}(P_0, Q_0) \\ = 1 + \text{LCS}(P_1, Q_1)\end{aligned}$$

$$\begin{aligned}\text{LCS}(P_0, Q_0) \\ = \max \{ \text{LCS}(P_1, Q_0), \text{LCS}(P_0, Q_1) \}\end{aligned}$$

Recursive Solution

LCS(P_o , Q_o) → 3



Case 1:

$P_o = "P_1 \boxed{x}"$
 $Q_o = "Q_1 \boxed{x}"$

Diagram illustrating Case 1: Two boxes labeled P_1 and Q_1 are shown above a bracket containing 'x'. A green bracket groups the first character of each string, and a green box highlights the character 'x'.

Case 2:

$P_o = "x"$
 $Q_o = "y"$

$$\begin{aligned} \text{LCS}(P_o, Q_o) \\ = 1 + \text{LCS}(P_i, Q_i) \end{aligned}$$

$$\begin{aligned} \text{LCS}(P_o, Q_o) \\ = \max \{ \text{LCS}(P_i, Q_o), \text{LCS}(P_o, Q_i) \} \end{aligned}$$

Recursive Solution

LCS(P_o , Q_o) → 3



Case 1:

$P_o = "P_1 \boxed{x}"$
 $Q_o = "Q_1 \boxed{x}"$

Diagram: A green bracket groups the first character of each string ('P1' and 'Q1'). Another green bracket groups the remaining strings ('x' and 'x'). A green box highlights the character 'x' at index 1 of both strings.

Case 2:

$P_o = "x"$
 $Q_o = "y"$

$$\begin{aligned} \text{LCS}(P_o, Q_o) \\ = 1 + \text{LCS}(P_i, Q_i) \end{aligned}$$

$$\begin{aligned} \text{LCS}(P_o, Q_o) \\ = \max \{ \text{LCS}(P_i, Q_o), \text{LCS}(P_o, Q_i) \} \end{aligned}$$

Recursive Solution

LCS(P_o , Q_o) → 3



Case 1:

$P_o = "P_1 \boxed{x}"$
 $Q_o = "Q_1 \boxed{x}"$

Case 2:

$P_o = "x"$
 $Q_o = "y"$

$$\begin{aligned} \text{LCS}(P_o, Q_o) \\ = 1 + \text{LCS}(P_i, Q_i) \end{aligned}$$

$$\begin{aligned} \text{LCS}(P_o, Q_o) \\ = \max \{ \text{LCS}(P_i, Q_o), \text{LCS}(P_o, Q_i) \} \end{aligned}$$

Recursive Solution

LCS(P_0, Q_0) → 3



Case 1:

$P_0 = " \boxed{P_1} x "$

$Q_0 = " \boxed{Q_1} x "$

A green bracket groups P_1 and Q_1 . A green square highlights the character x in both P_1 and Q_1 . A green arrow points from the x in P_1 to the x in Q_1 .

Case 2:

$P_0 = " \boxed{P_1} x "$

$Q_0 = " \boxed{Q_1} y "$

An orange bracket groups P_1 and Q_1 . A red square highlights the character x in P_1 . A red arrow points from the x in P_1 to the y in Q_1 .

LCS(P_0, Q_0)

$$= \underline{1} + \underline{\text{LCS}(P_1, Q_1)}$$

LCS(P_0, Q_0)

$$= \max \{ \text{LCS}(P_1, Q_0), \text{LCS}(P_0, Q_1) \}$$

Recursive Solution

LCS(P_0, Q_0) → 3



Case 1:

$P_0 = " \boxed{P_1} \underset{Q}{x} "$

$Q_0 = " \boxed{Q_1} \underset{x}{x} "$

Diagram: Two boxes labeled P_1 and Q_1 are stacked vertically. The top box has a circled 'Q' above it. The bottom box has a circled 'x' below it. To the right of the boxes is an 'x' with a green bracket underneath.

Case 2:

$P_0 = " \boxed{P_1} \underset{x}{x} "$

$Q_0 = " \boxed{Q_1} \underset{y}{y} "$

Diagram: Two boxes labeled P_1 and Q_1 are stacked vertically. The top box has a circled 'P' above it. The bottom box has a circled 'y' below it. To the right of the boxes is an 'x' with a green bracket underneath.

LCS(P_0, Q_0)

$$= \underline{1} + \underline{\text{LCS}(P_1, Q_1)}$$

LCS(P_0, Q_0)

$$= \max \{ \text{LCS}(P_1, Q_0), \text{LCS}(P_0, Q_1) \}$$

Recursive Solution

LCS(P_0, Q_0) → 3



Case 1:

$P_0 = "P_1 \boxed{X}"$
 $\underline{Q_0} = "Q_1 \boxed{X}"$

Case 2:

$P_0 = "P_1 \boxed{X}"$
 $\underline{Q_0} = "Q_1 \boxed{Y}"$

LCS(P_0, Q_0)

$$= \underline{1} + \underline{\text{LCS}(P_1, Q_1)}$$

LCS(P_0, Q_0)

$$= \max \{ \underline{\text{LCS}(P_1, Q_0)}, \underline{\text{LCS}(P_0, Q_1)} \}$$

Recursive Solution (Code)

```
def LCS(P, Q, n, m)
    if n == 0 or m == 0: // base case
        result = 0
    else if P[n-1] == Q[m-1]:
        result = 1 + LCS(P, Q, n-1, m-1)
    else if P[n-1] != Q[m-1]: // just for clarity
        tmp1 = LCS(P, Q, n-1, m)
        tmp2 = LCS(P, Q, n, m-1)
        result = max{ tmp1, tmp2 }
    return result
```

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Recursive Solution (Code)

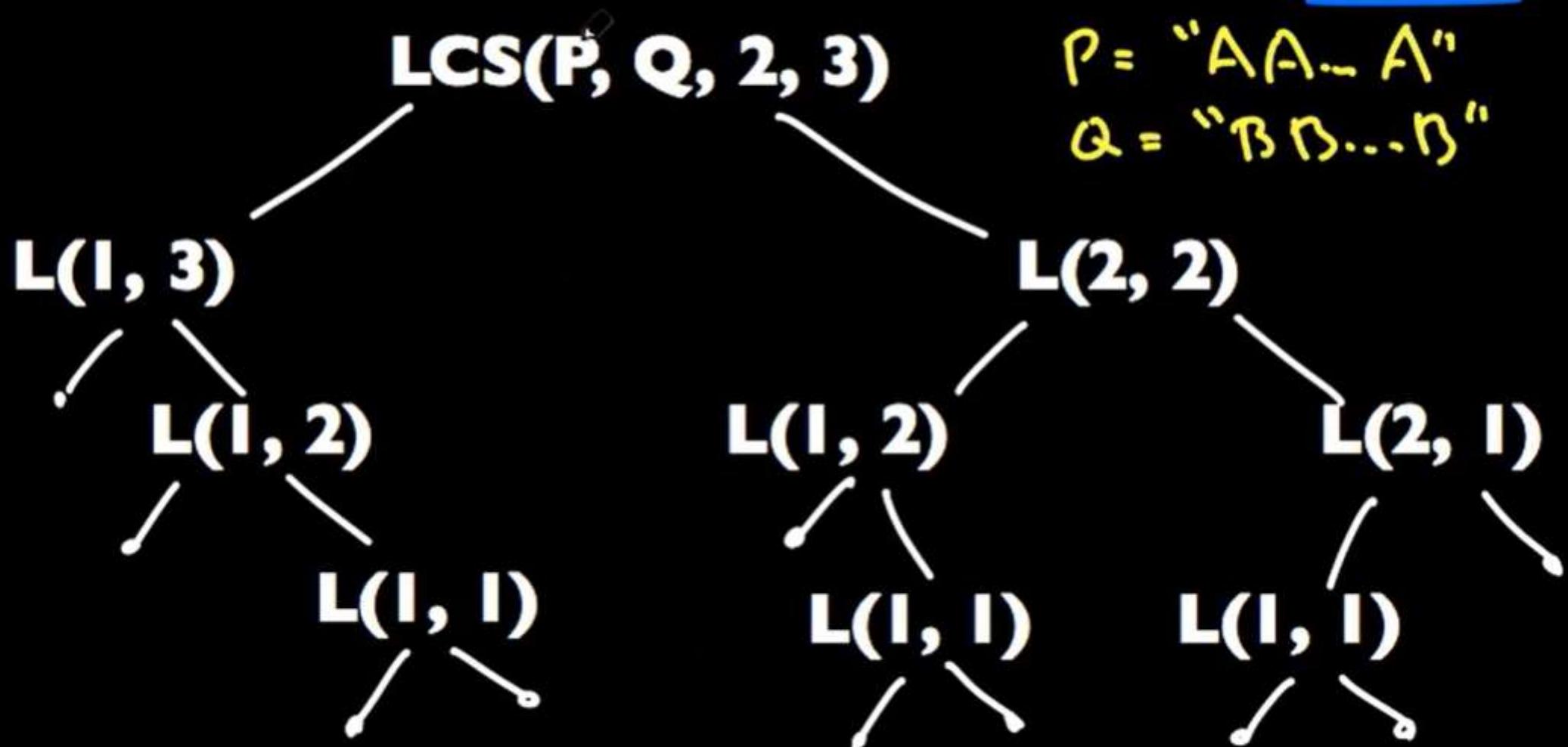
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```

Recursive Solution (Code)

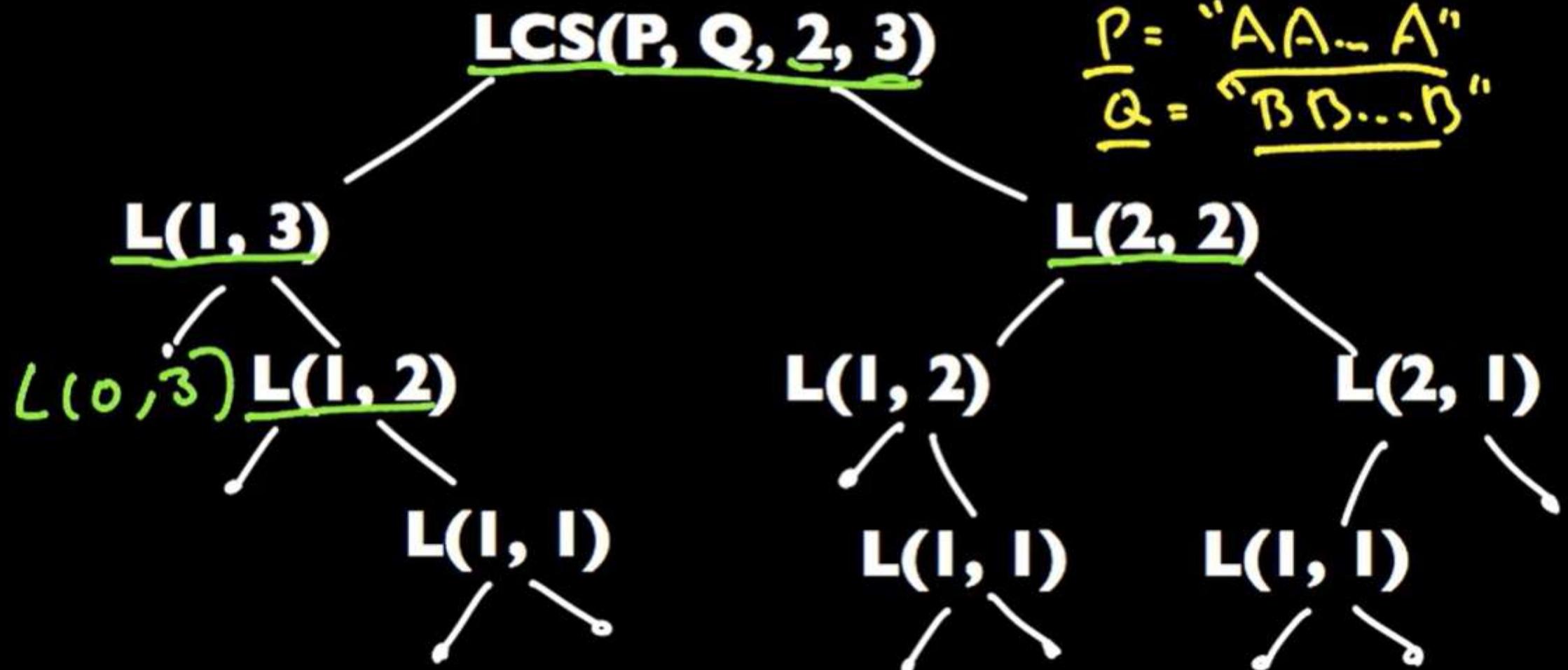
```
P = "A2B2C"  
n = 2  
Q = "A2B2C"  
m = 2  
  
def LCS(P, Q, n, m)  
    if n == 0 or m == 0: // base case  
        result = 0  
    else if P[n-1] == Q[m-1]:  
        result = 1 + LCS(P, Q, n-1, m-1)  
    else if P[n-1] != Q[m-1]: // just for clarity  
        tmp1 = LCS(P, Q, n-1, m)  
        tmp2 = LCS(P, Q, n, m-1)  
        result = max{ tmp1, tmp2 }  
    return result
```

Analysis of Recursive Solution

P = "AA"
Q = "BBB"



Analysis of Recursive Solution

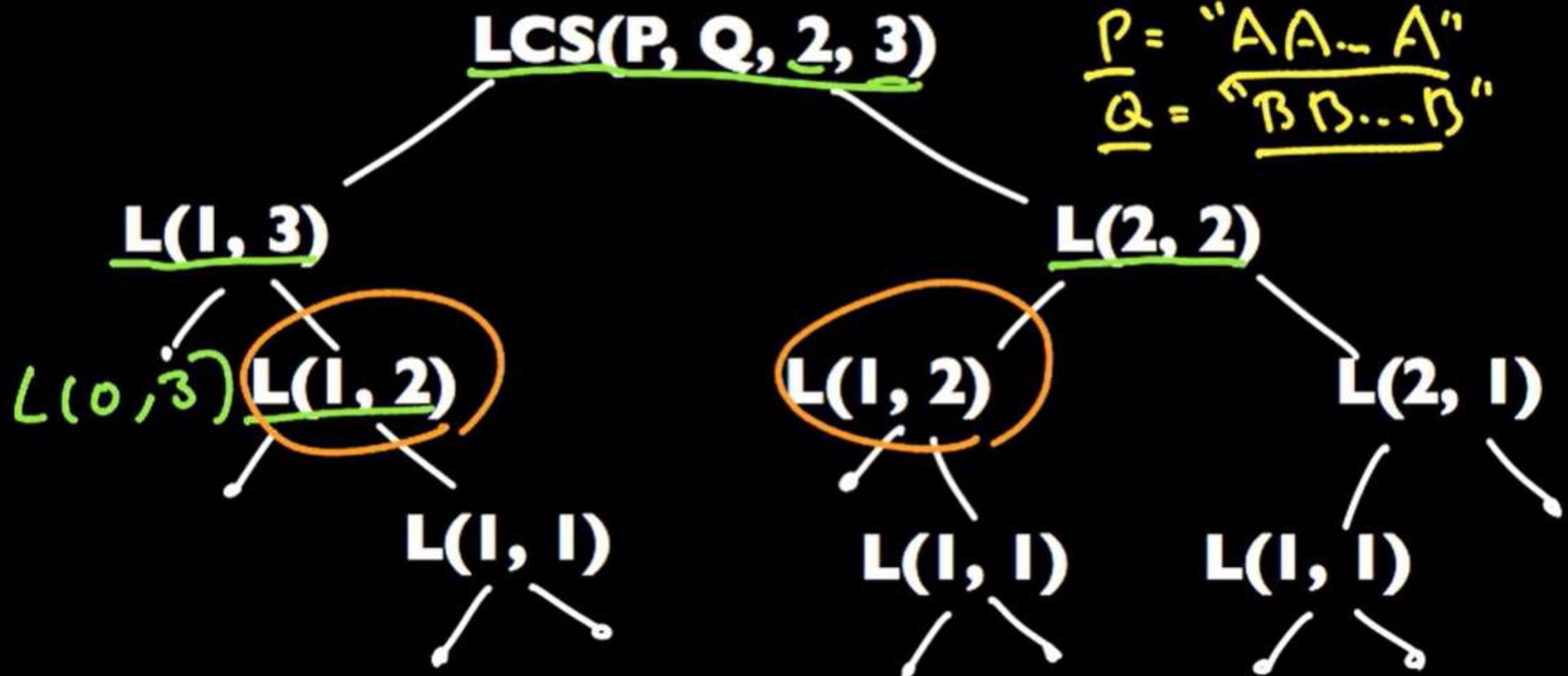


P = "AA"
Q = "BBB"

P = "AA...A"
Q = "BB...B"

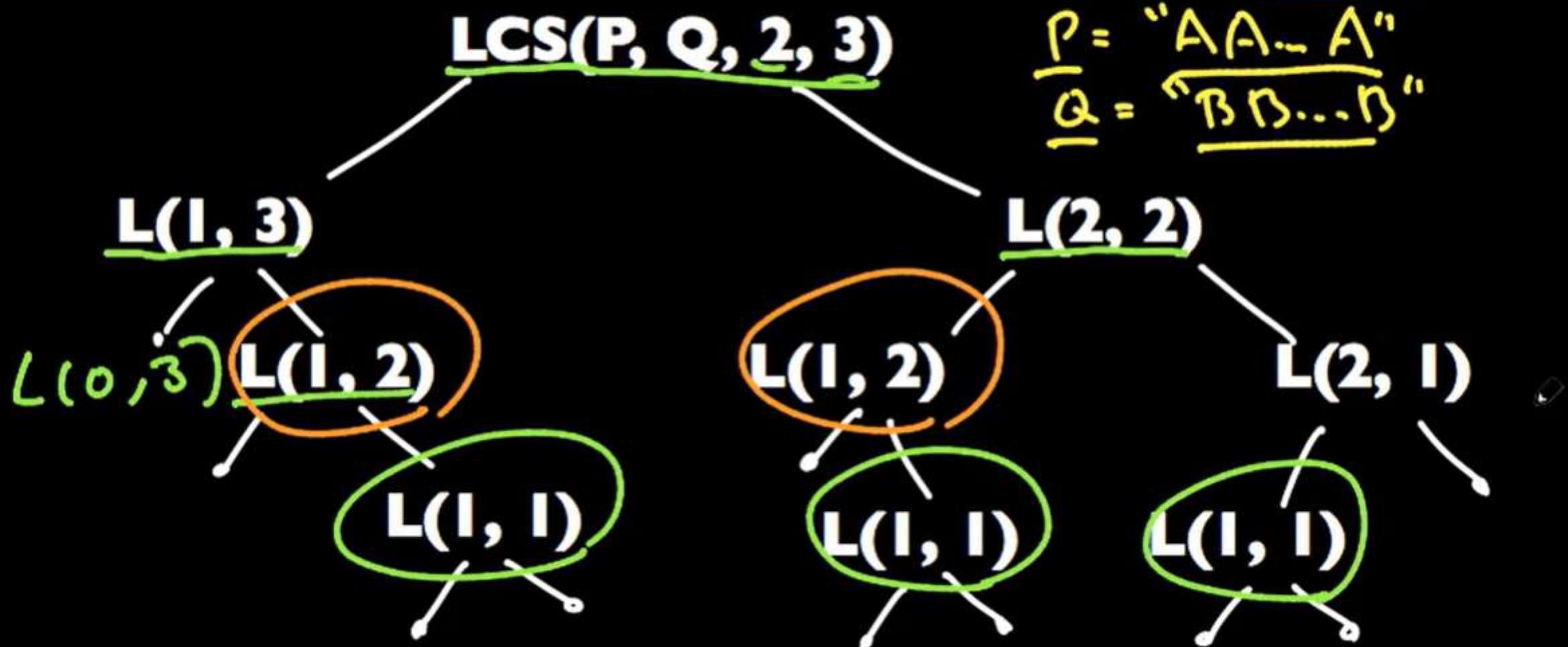
Analysis of Recursive Solution

P = "AA"
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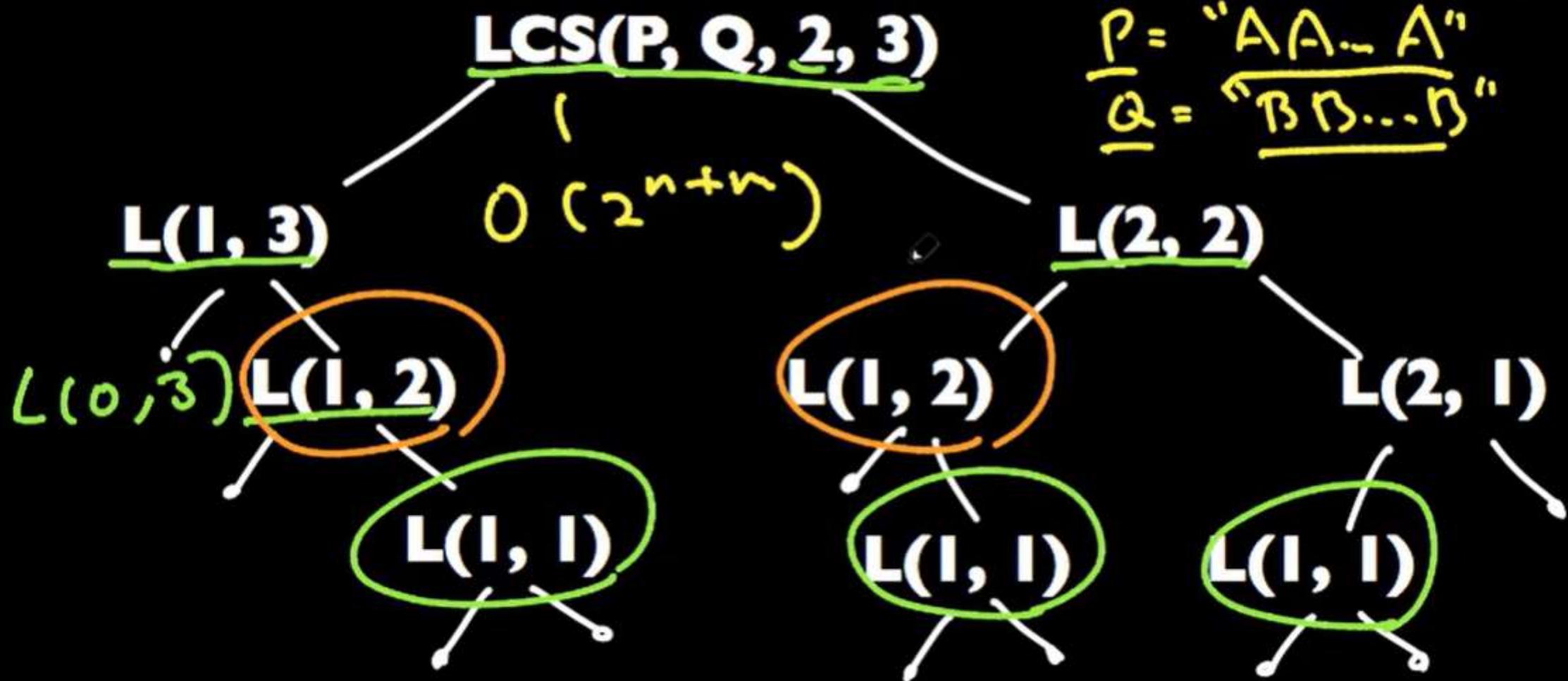
Analysis of Recursive Solution

P = "AA"
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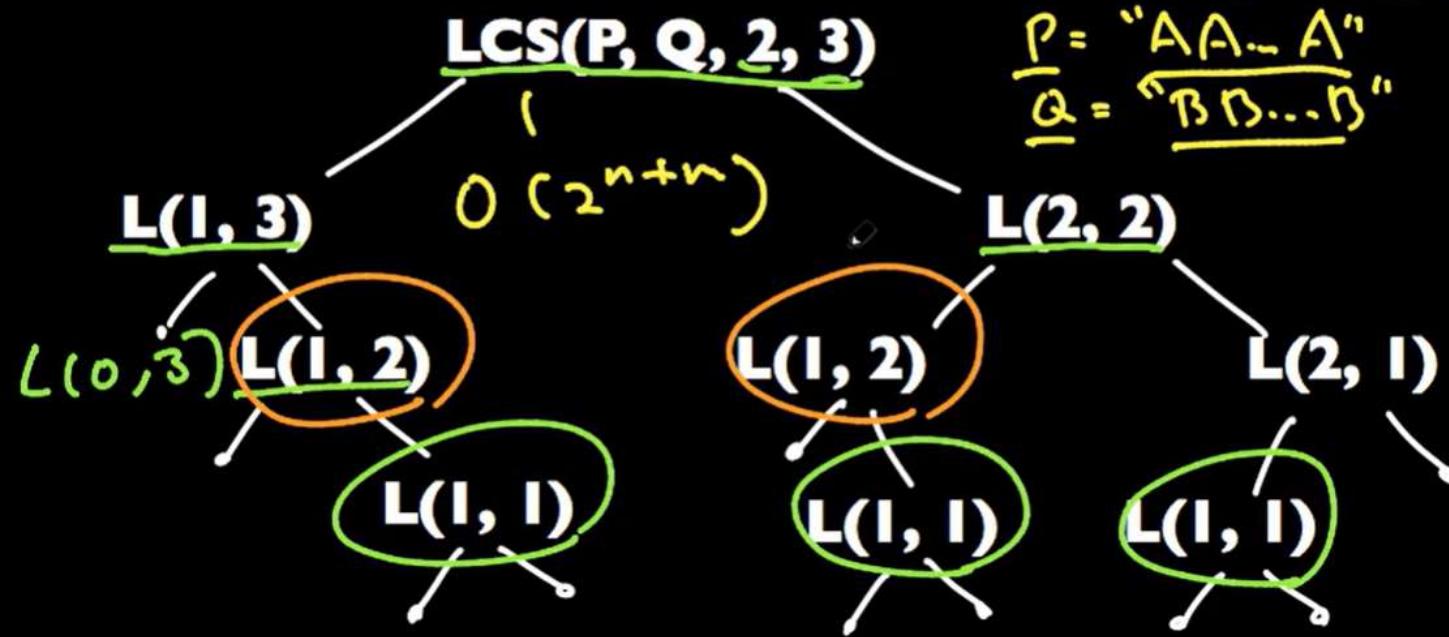
Analysis of Recursive Solution

P = "AA"
Q = "BBB"



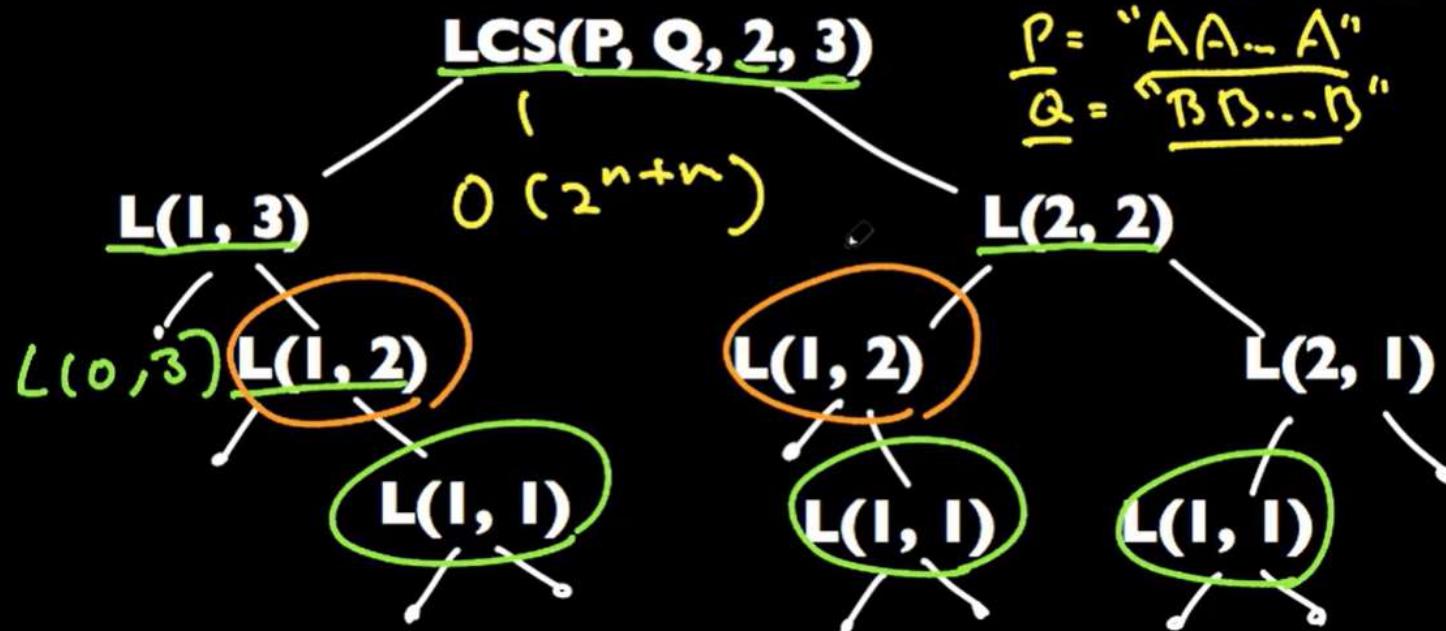
Analysis of Recursive Solution

Q
P = "AA"
Q = "BBB"



Analysis of Recursive Solution

Q
P = "AA"
Q = "BBB"



Dynamic Programming

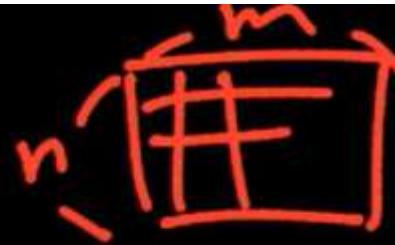
1. Recursive Solution

2. Memorize Intermediate Results

3. Bottom-Up

Memorize Intermediate Results

```
o // Initialize arr[n][m] to undefined
  def LCS(P, Q, n, m)
    o if arr[n][m] != undefined: return arr[n][m]
      if n == 0 or m == 0:
        result = 0
      else if P[n-1] == Q[m-1]:
        result = 1 + LCS(P, Q, n-1, m-1)
      else if P[n-1] != Q[m-1]: // just for clarity
        tmp1 = LCS(P, Q, n-1, m)
        tmp2 = LCS(P, Q, n, m-1)
        result = max{ tmp1, tmp2 }
    o arr[n][m] = result
    return result
```



P = "AA"
Q = "BBB"

Memorize Intermediate Results

o // Initialize arr[n][m] to undefined



def LCS(P, Q, n, m)

o if arr[n][m] != undefined: return arr[n][m]

o } if n == 0 or m == 0:

 result = 0

2^{nm}

time / cell
 $= O(1)$

else if P[n-1] == Q[m-1]:

 result = 1 + LCS(P, Q, n-1, m-1)

else if P[n-1] != Q[m-1]: // just for clarity

 tmp1 = LCS(P, Q, n-1, m)

 tmp2 = LCS(P, Q, n, m-1)

 result = max{ tmp1, tmp2 }

$O(nm)$

nm

→ o arr[n][m] = result

return result

Explanation:

1. The function $\text{LCS}(P, Q, n, m)$ is a recursive function that computes the longest common subsequence length between two sequences P and Q of lengths n and m , respectively.
2. The key optimization used is **memoization**, where we store the results of previously computed subproblems in $\text{arr}[n][m]$ to avoid redundant computations.
3. The worst-case number of function calls (i.e., the number of unique (n, m) states we compute) is $O(nm)$ since we fill up a **2D table of size $n \times m$** .
4. The computation per cell (each state) takes $O(1)$ time.
5. Therefore, the total time complexity is $O(nm)$.

Understanding the Complexity:

1. The recursive function $\text{LCS}(P, Q, n, m)$ explores two recursive branches in the worst case:
 - One branch reduces n by 1: $\text{LCS}(P, Q, n-1, m)$
 - Another branch reduces m by 1: $\text{LCS}(P, Q, n, m-1)$
2. This forms a **binary recursion tree** where at each level, the number of recursive calls doubles.

Complexity Growth:

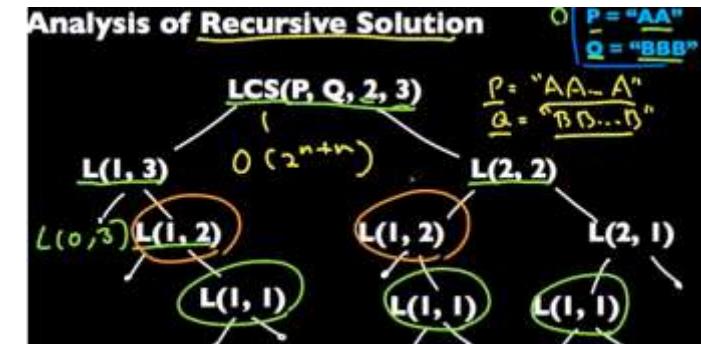
- The depth of the recursion tree is at most $m + n$, leading to $O(2^{m+n})$ possible recursive calls.
- However, in standard LCS problems, where $m \approx n$, the complexity can also be expressed as $O(2^{\max(m, n)})$.

Which One is Correct?

- If m and n are significantly different, then $O(2^{m+n})$ is the more precise complexity.
- If m and n are roughly equal, we often simplify it to $O(2^{\max(m, n)})$, since $\max(m, n)$ dominates.

Explanation:

1. The function $\text{LCS}(P, Q, n, m)$ is a recursive function that computes the longest common subsequence length between two sequences P and Q of lengths n and m , respectively.
2. The key optimization used is **memoization**, where we store the results of previously computed subproblems in $\text{arr}[n][m]$ to avoid redundant computations.
3. The worst-case number of function calls (i.e., the number of unique (n, m) states we compute) is $O(nm)$ since we fill up a **2D table of size $n \times m$** .
4. The computation per cell (each state) takes $O(1)$ time.
5. Therefore, the total time complexity is $O(nm)$.



- Without memoization, the naive recursive approach would have an exponential complexity $O(2^{m+n})$ due to redundant computations.
- With memoization, each state (n, m) is only computed once and stored in the $\text{arr}[n][m]$ table.
- Since there are at most nm unique states, and each is computed in constant $O(1)$ time, the final complexity remains $O(nm)$.

- If m and n are roughly equal, we often simplify it to $O(2^{\max(m, n)})$, since $\max(m, n)$ dominates.



1. Starting with $O(2^{m+n})$

From the naive recursive approach, the worst-case time complexity is:

$O(2^{m+n})$ where m and n are the lengths of the two sequences.

2. Assuming $m \approx n$

- If m and n are roughly equal, we can express them as $m \approx n \approx k$ for some k .
- Then, we rewrite the complexity: $O(2^{m+n}) = O(2^{k+k}) = O(2^{2k})$

3. Using the Property of Exponents

We use the identity: $2^{2k} = (2^k)^2$

Since Big-O notation ignores constant factors, we drop the squared term: $O(2^{2k}) = O(2^k)$

Since $k \approx \max(m, n)$, we get: $O(2^{\max(m,n)})$

Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

In the context of the **Longest Common Subsequence (LCS)** using a **bottom-up dynamic programming approach**, the term **Max(Left, TOP)** refers to the recurrence relation used when the characters from the two sequences **do not match**.

Explanation:

1. The LCS table (or DP table) is built such that:
 - Each cell $L(n, m)$ represents the length of the LCS for the first n characters of string P and the first m characters of string Q .
 - The values are filled iteratively using previously computed values.

3. When characters do not match:

- If $P[n] \neq Q[m]$, then:

$$L(n, m) = \max(L(n, m - 1), L(n - 1, m))$$

- $L(n, m-1) \rightarrow$ Represents the value from the **left** (excluding the current character from string Q).
- $L(n-1, m) \rightarrow$ Represents the value from the **top** (excluding the current character from string P).
- The **maximum** of these two values is chosen to ensure that the longest subsequence found so far is retained.

How to predict the common Sequence

Example from the Image:

Given Strings:

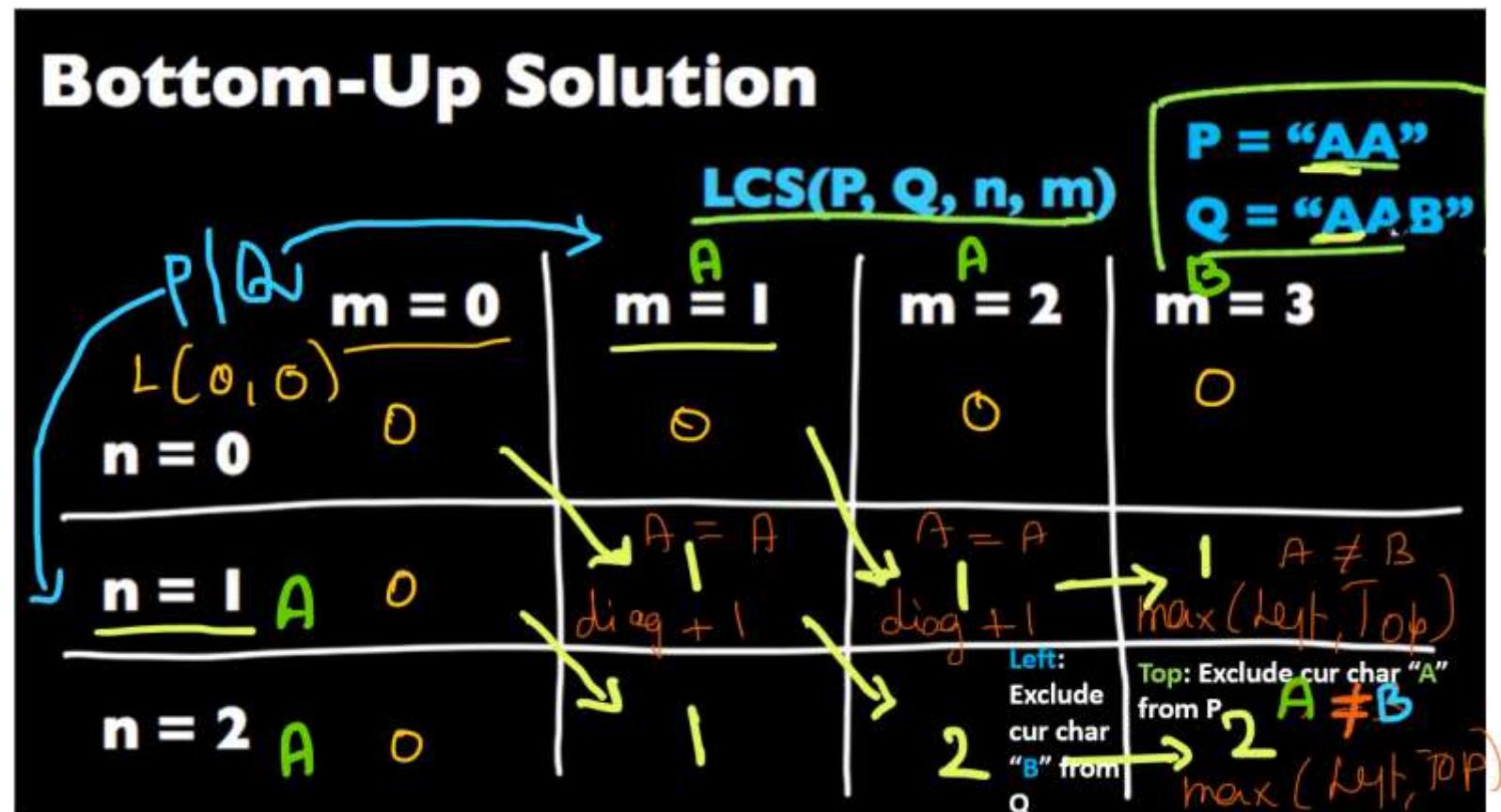
- P = "AA"
- Q = "AAB"

Final Table Value:

- $L(2,3) = 2$, meaning the LCS has 2 characters.

Backtracking Steps:

1. $L(2,3) = \max(L(2,2), L(1,3)) = 2$
 - Move left to (2,2) since $L(2,2) = 2$.
2. $L(2,2) = 1 + L(1,1) = 2$
 - Characters match: A = A → Include 'A'
 - Move diagonally to (1,1).
3. $L(1,1) = 1 + L(0,0) = 1$
 - Characters match: A = A → Include 'A'
 - Move diagonally to (0,0).



Final Predicted LCS:

- "AA" (Reading collected characters in reverse)

Example 2

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0						
b	0						
c	0						
f	0						

abcdaf
acbcaf

Example 2

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0						
c	0						
f	0						

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0						
f	0						

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0						

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

diagonally

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcf

Max(3,2):-

Since either excludes cur char **c** from String **2**(left) or excludes cur **a** char from string **1**(top)

Max(3,2)

3 is present on its left, not diagonally. So no need to include it in the answer

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

Again, 3 is present on its left, **not** diagonally. So no need to include it in the answer

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

From Diagonal only, Got this
3. So include c

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
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f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
b	0	1	2	2	2	2	2
c	0	1	2	3	3	3	3
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abcdaf
acbcaf

		a	b	c	d	a	f
	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
c	0	1	1	2	2	2	2
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c	0	1	2	3	3	3	3
f	0	1	2	3	3	3	4

abcdaf
acbcaf

Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

if (input1[i] == input2[j])

$$T[i][j] = T[i-1][j-1] + 1$$

$$\boxed{\begin{aligned} \text{LCS}(P_i, Q_o) \\ = 1 + \text{LCS}(P_i, Q_o) \end{aligned}}$$

else

$$T[i][j] = \max (T[i-1][j], T[i][j-1])$$

Recursive Solution (Code)

```
def LCS(P, Q, n, m)
    if n == 0 or m == 0: // base case
        result = 0
    else if P[n-1] == Q[m-1]:
        result = 1 + LCS(P, Q, n-1, m-1)
    else if P[n-1] != Q[m-1]: // just for clarity
        tmp1 = LCS(P, Q, n-1, m)
        tmp2 = LCS(P, Q, n, m-1)
        result = max{ tmp1, tmp2 }
    return result
```

$\boxed{\begin{aligned} \text{LCS}(P_i, Q_o) \\ = 1 + \text{LCS}(P_i, Q_o) \end{aligned}}$

$\boxed{\begin{aligned} \text{LCS}(P_o, Q_o) \\ = \max \{ \text{LCS}(P_i, Q_o), \text{LCS}(P_o, Q_i) \} \end{aligned}}$

Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

if (input1[i] == input2[j])

$$T[i][j] = T[i-1][j-1] + 1$$

$$\boxed{\begin{aligned} \text{LCS}(P_i, Q_o) \\ = 1 + \text{LCS}(P_i, Q_o) \end{aligned}}$$

else

$$T[i][j] = \max (T[i-1][j], T[i][j-1])$$

$$\boxed{\begin{aligned} \text{LCS}(P_o, Q_o) \\ = \max \{ \text{LCS}(P_i, Q_o), \text{LCS}(P_o, Q_i) \} \end{aligned}}$$

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Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

```
int LCS(string P, string Q) {  
    int n = P.length();  
    int m = Q.length();  
    vector<vector<int>> dp(n + 1, vector<int>(m + 1, 0));  
  
    // Fill the dp table bottom-up  
    for (int i = 1; i <= n; ++i) {  
        for (int j = 1; j <= m; ++j) {  
            if (P[i - 1] == Q[j - 1])  
                dp[i][j] = 1 + dp[i - 1][j - 1]; // match case  
            else  
                dp[i][j] = max(dp[i - 1][j], dp[i][j - 1]); // no match  
        }  
    }  
  
    return dp[n][m]; // Length of LCS  
}
```

```
int main() {  
    string P = "AGGTAB";  
    string Q = "GXTXAYB";  
    cout << "Length of LCS is " << LCS(P, Q) << endl;  
    return 0;  
}
```

Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

```
int LCS(string P, string Q) {  
    int n = P.length();  
    int m = Q.length();  
    vector<vector<int>> dp(n + 1, vector<int>(m + 1, 0));  
  
    // Fill the dp table bottom-up  
    for (int i = 1; i <= n; ++i) {  
        for (int j = 1; j <= m; ++j) {  
            if (P[i - 1] == Q[j - 1])  
                dp[i][j] = 1 + dp[i - 1][j - 1]; // match case  
            else  
                dp[i][j] = max(dp[i - 1][j], dp[i][j - 1]); // no match  
        }  
    }  
  
    return dp[n][m]; // Length of LCS  
}
```

DP Table Initialization:

```
vector<vector<int>> dp(n + 1, vector<int>(m + 1, 0));
```

- `dp[0][*]` and `dp[*][0]` represent comparisons with an `empty string`, so they must be `0`.
- Then we compute for all other `i = 1 to n` and `j = 1 to m`.

```
for (int i = 1; i <= n; ++i)  
    for (int j = 1; j <= m; ++j)
```

This is correct because we are *filling the DP table* starting after the initialized base cases in row 0 and column 0.