

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 $S_1 =$
ACCGGTCGAGTGCGCGGAAGCCGGCCGAA, and
the DNA of another organism may be $S_2 =$
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

$S_1 = \text{ACCGGTCGAGTGCGCGGAAGCCGGCCGAA}$

$S_2 = \text{GTCGTTCGGAATGCCGTTGCTCTGTAAA}$

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

- $S_1 = \text{ACCGGTCGAGTGCGCGGAAGCCGGCCGAA}$
- $S_2 = \text{GTCGTTCGGAATGCCGTTGCTCTGTAAA}$
- 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

$LCS(P_0, Q_0)$ \rightarrow 3

"BATD"
"ABACD" \rightarrow "BAD"

Case 1:

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

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

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)

“BATD”
“ABACD” } → **“BAD”**

Case 1:

$P_0 = \text{“ } x \text{”}$
 $Q_0 = \text{“ } x \text{”}$

Case 2:

$P_0 = \text{“ } x \text{”}$
 $Q_0 = \text{“ } y \text{”}$

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

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

Recursive Solution

$LCS(P_0, Q_0)$ \rightarrow 3

"BATD"
"ABACD" \rightarrow "BAD"

Case 1:

$P_0 = "$ P_1 x $"$
 $Q_0 = "$ Q_1 x $"$ \rightarrow $\square x$

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

$$\text{LCS}(\underline{P_0}, \underline{Q_0}) \rightarrow 3$$

"BATD"
"ABACD" } \rightarrow "BAD"

Case 1:

$$\begin{aligned} P_0 &= " \boxed{P_1} x " \\ Q_0 &= " \boxed{Q_1} x " \end{aligned} \left. \vphantom{\begin{aligned} P_0 \\ Q_0 \end{aligned}} \right\} " \boxed{} x "$$

Case 2:

$$\begin{aligned} P_0 &= " \phantom{\boxed{P_1}} x " \\ Q_0 &= " \phantom{\boxed{Q_1}} y " \end{aligned}$$

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

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

Recursive Solution

$$\text{LCS}(P_0, Q_0) \rightarrow 3$$

"BATD"
"ABACD" } \rightarrow "BAD"

Case 1:

$$\begin{aligned} P_0 &= "P_1" x \\ Q_0 &= "Q_1" x \end{aligned} \quad \left\{ \begin{array}{l} \boxed{P_1} \\ \boxed{Q_1} \end{array} \right\} \text{ " } \boxed{} \text{ "x"}$$

Case 2:

$$\begin{aligned} P_0 &= " \quad \quad \quad x \\ Q_0 &= " \quad \quad \quad y \end{aligned}$$

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

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

Recursive Solution

$$\text{LCS}(\underline{P_0}, \underline{Q_0}) \rightarrow 3$$

"BATD"
"ABACD" } \rightarrow "BAD"

Case 1:

$$\begin{aligned} P_0 &= " \boxed{P_1} x " \\ Q_0 &= " \boxed{Q_1} x " \end{aligned} \quad \left. \vphantom{\begin{aligned} P_0 &= " \boxed{P_1} x " \\ Q_0 &= " \boxed{Q_1} x " \end{aligned}} \right\} " \boxed{} x "$$

Case 2:

$$\begin{aligned} P_0 &= " \phantom{\boxed{P_1}} x " \\ Q_0 &= " \phantom{\boxed{Q_1}} y " \end{aligned}$$

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

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

Recursive Solution

$$\text{LCS}(\underline{P_0}, \underline{Q_0}) \rightarrow 3$$

"BATD"
"ABACD" } \rightarrow "BAD"

Case 1:

$$\begin{aligned} P_0 &= " \boxed{P_1} x " \\ Q_0 &= " \boxed{Q_1} x " \end{aligned} \quad \left. \vphantom{\begin{aligned} P_0 &= " \boxed{P_1} x " \\ Q_0 &= " \boxed{Q_1} x " \end{aligned}} \right\} " \boxed{} x "$$

Case 2:

$$\begin{aligned} P_0 &= " \phantom{\boxed{P_1}} x " \\ Q_0 &= " \phantom{\boxed{Q_1}} y " \end{aligned}$$

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

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

Recursive Solution

$$\text{LCS}(\underline{P_0}, \underline{Q_0}) \rightarrow 3$$

"BATD"
"ABACD" } \rightarrow "BAD"

Case 1:

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

Case 2:

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

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

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

Recursive Solution

$$\text{LCS}(\underline{P_0}, \underline{Q_0}) \rightarrow 3$$

"BATD"
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Case 1:

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

Case 2:

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

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

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

Recursive Solution

$$\text{LCS}(P_0, Q_0) \rightarrow 3$$

"BATD"
"ABACD" } \rightarrow "BAD"

Case 1:

$$P_0 = "P_1 x" \\ Q_0 = "Q_1 x" \quad \text{where } P_1 \neq Q_1$$

Case 2:

$$P_0 = "P_1 x" \\ Q_0 = "Q_1 y" \quad \text{where } P_1 = Q_1$$

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

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

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[n-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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        tmp1 = LCS(P, Q, n-1, m)
        tmp2 = LCS(P, Q, n, m-1)
        result = max{ tmp1, tmp2 }
    return result
```

$$\begin{aligned} \text{LCS}(P_i, Q_j) \\ = 1 + \text{LCS}(P_i, Q_{j-1}) \end{aligned}$$

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

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

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

Recursive Solution (Code)

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    return result
```

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[n-1]: // just for clarity
```

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

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

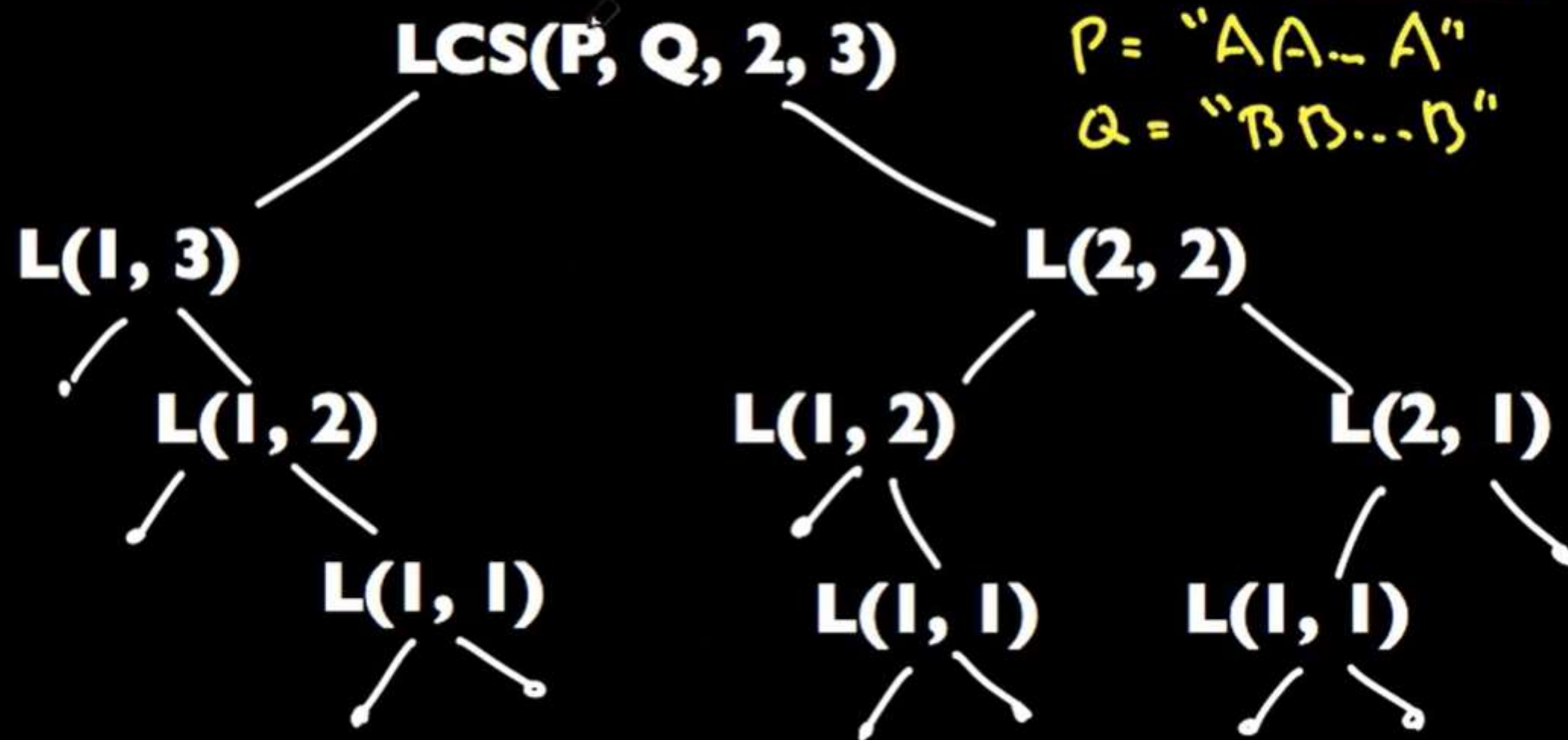
```
        result = max{ tmp1, tmp2 }
```

```
    return result
```

$P = \text{"ABC"}^2$
 $n = 2$
 $Q = \text{"ABC"}^2$
 $m = 2$

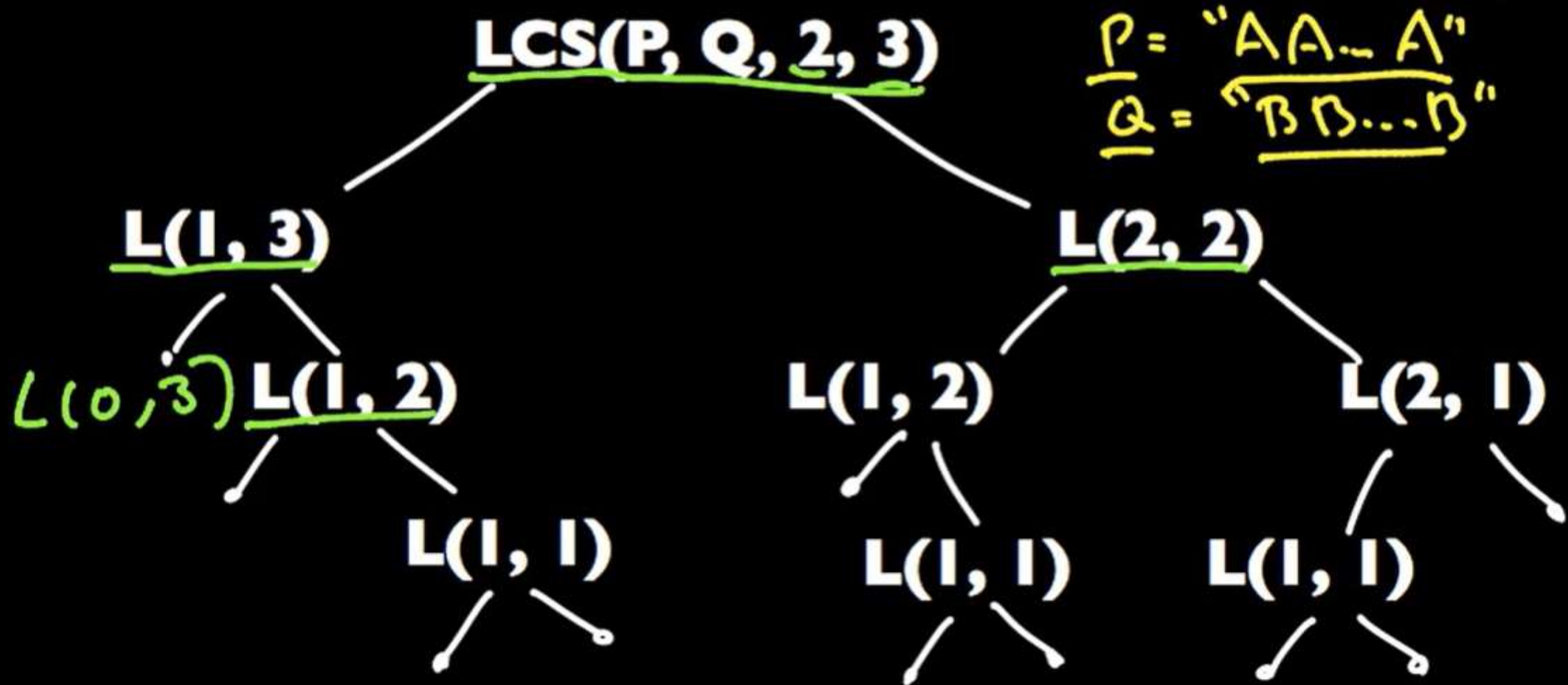
Analysis of Recursive Solution

P = "AA"
Q = "BBB"



Analysis of Recursive Solution

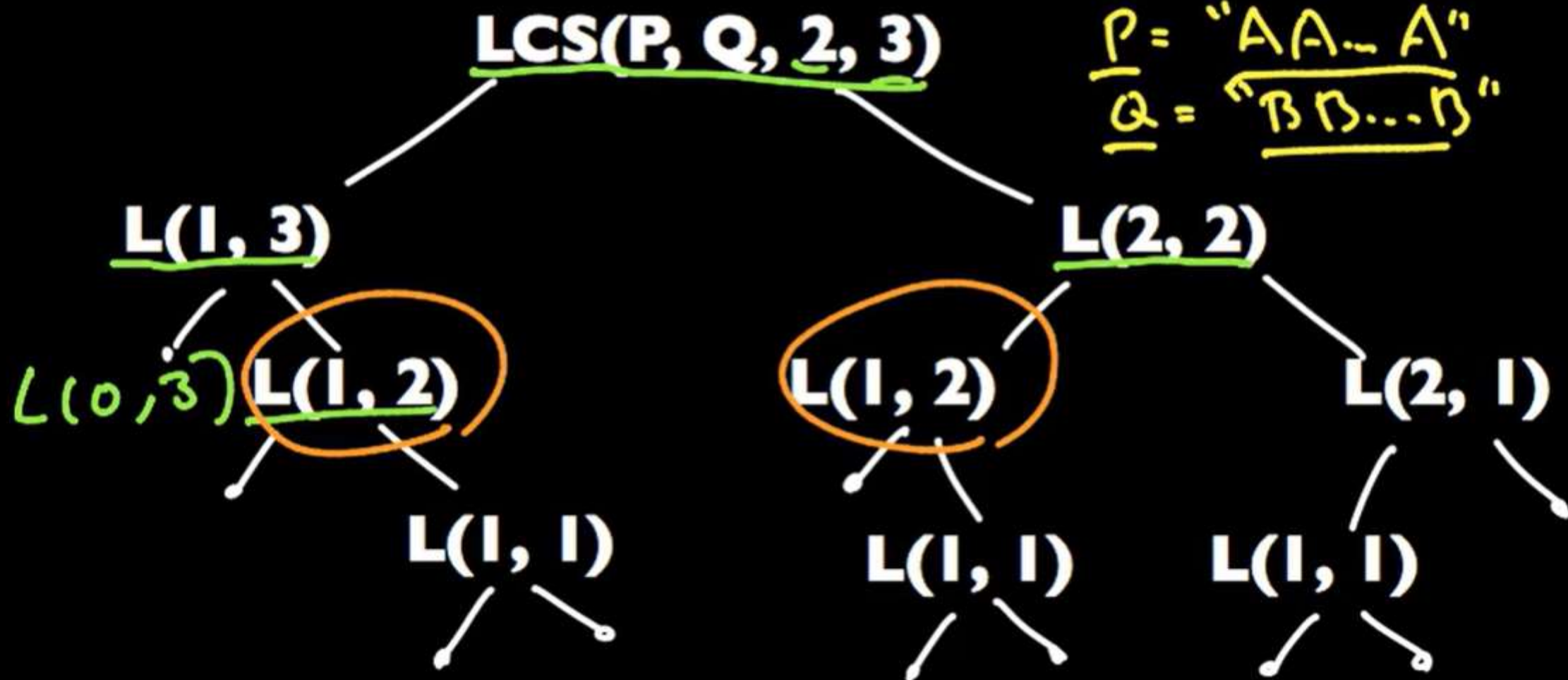
\circ $\boxed{\begin{array}{l} P = \text{"AA"} \\ Q = \text{"BBB"} \end{array}}$



Analysis of Recursive Solution

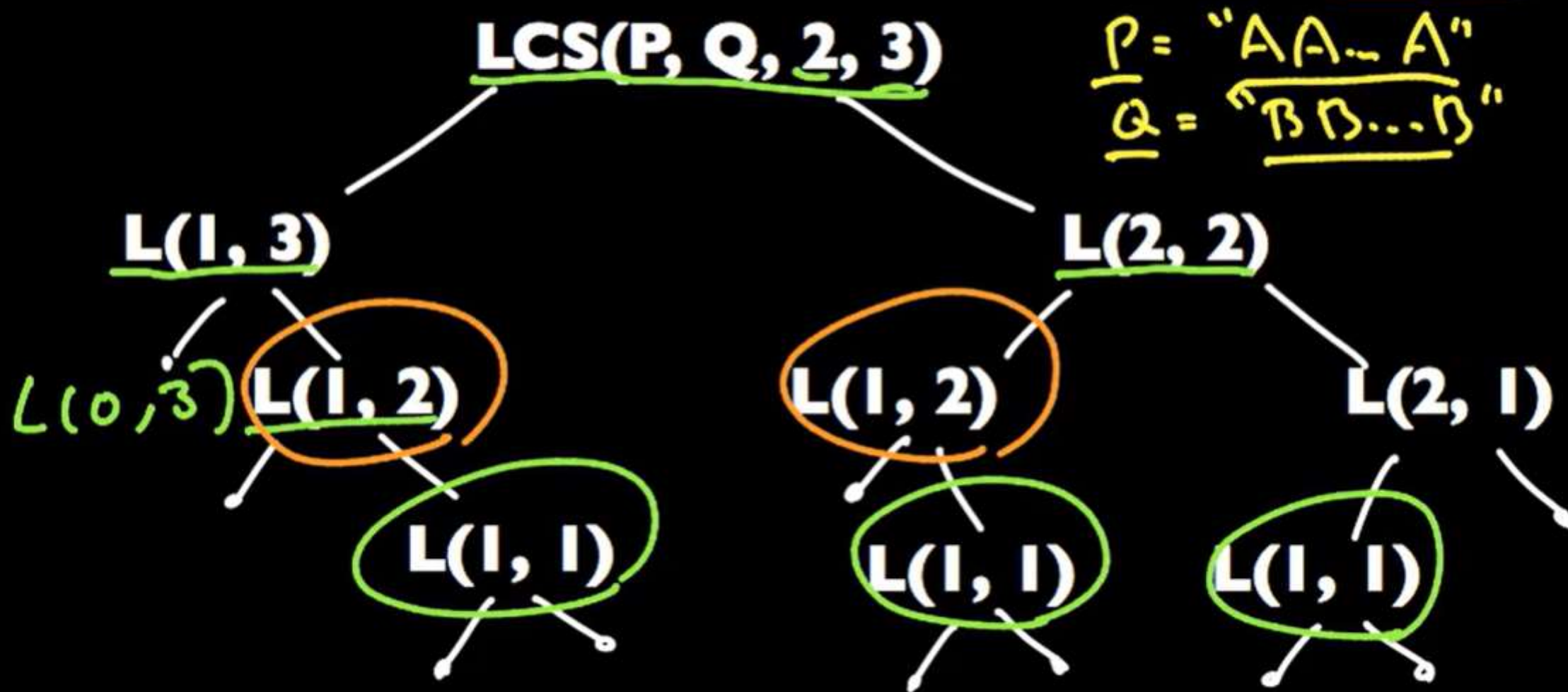
Q P = "AA"
Q = "BBB"

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



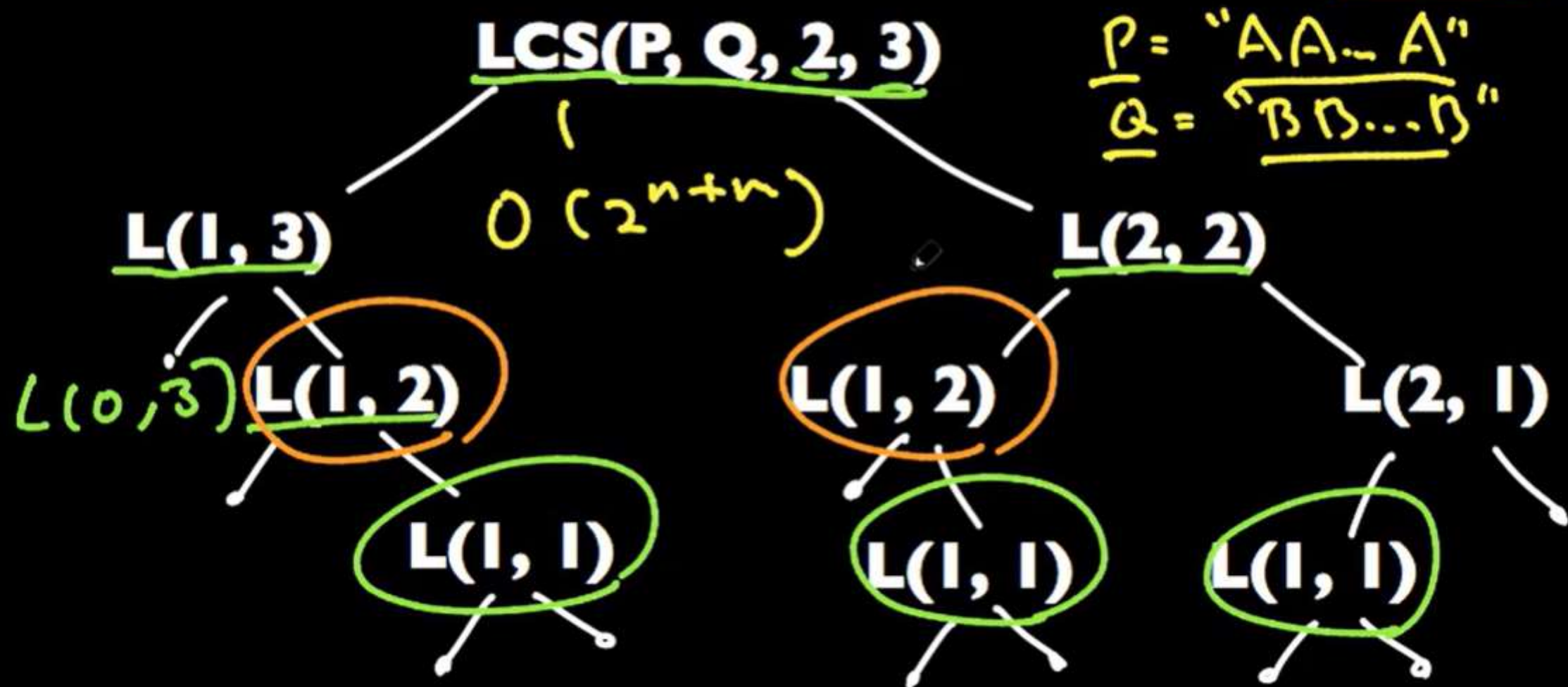
Analysis of Recursive Solution

$P = \text{"AA"}$
 $Q = \text{"BBB"}$



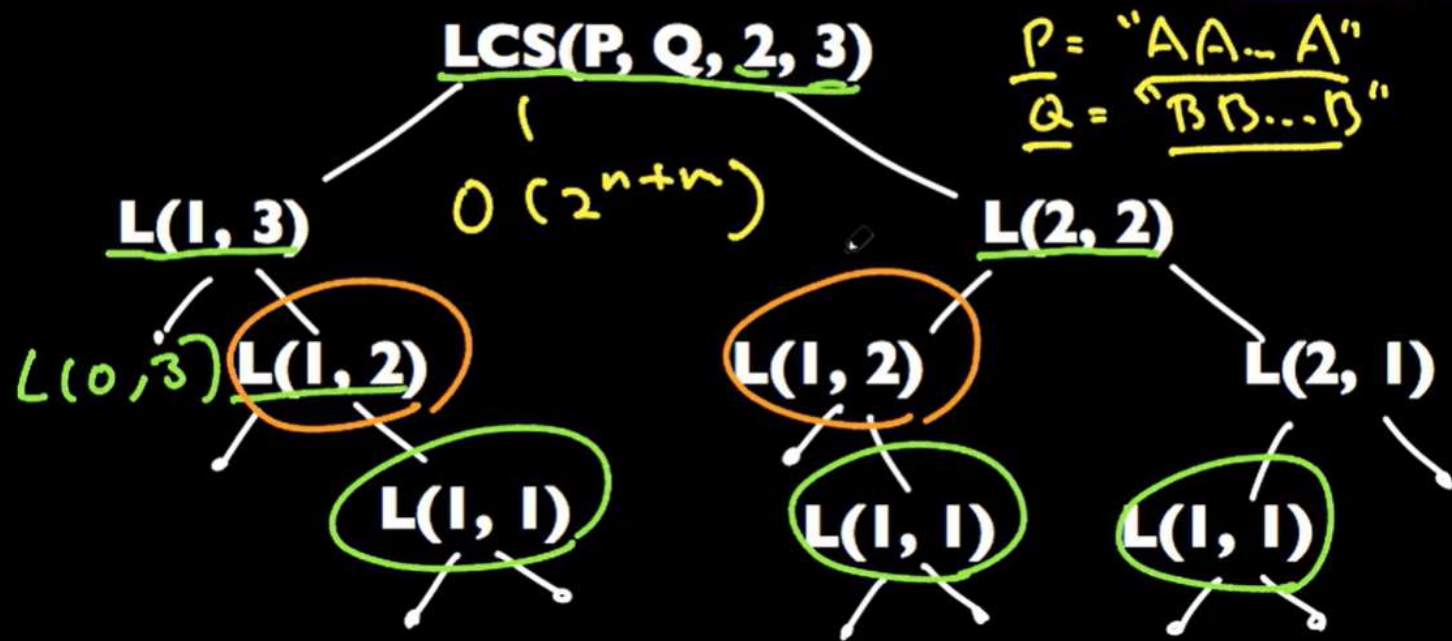
Analysis of Recursive Solution

$P = \text{"AA"}$
 $Q = \text{"BBB"}$



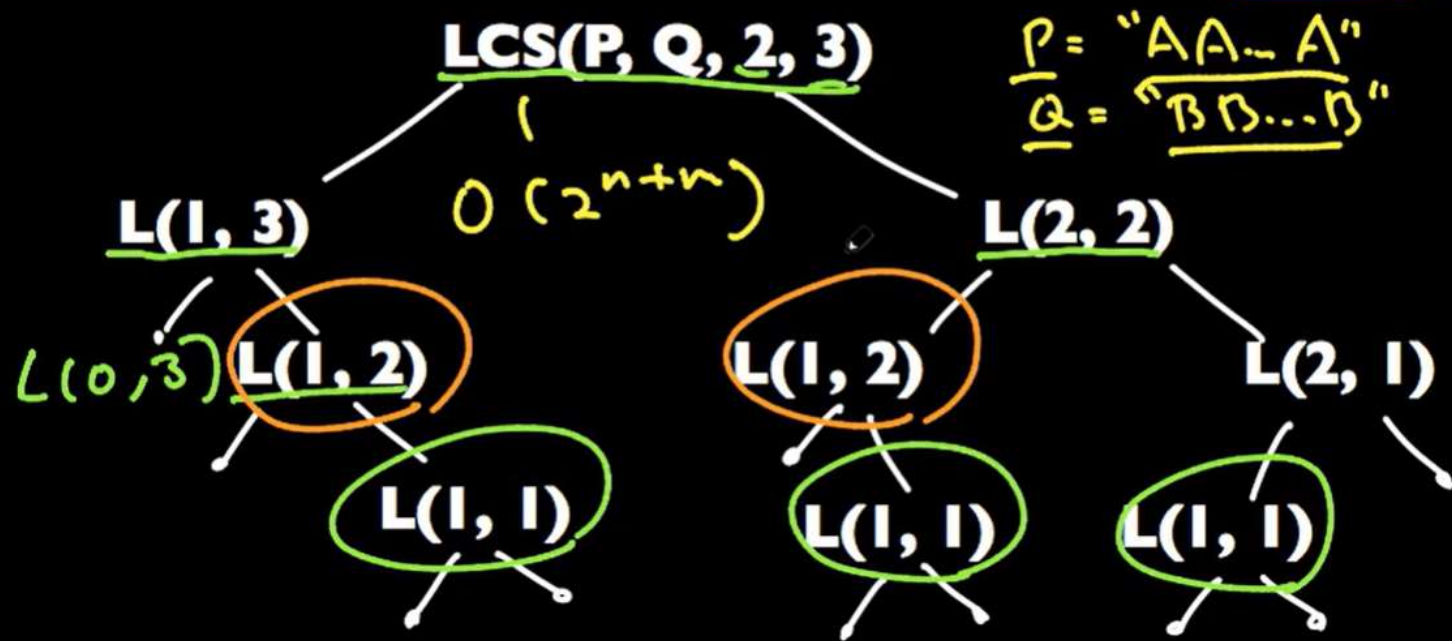
Analysis of Recursive Solution

P = "AA"
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Analysis of Recursive Solution

P = "AA"
Q = "BBB"



Dynamic Programming

1. Recursive Solution

2. Memorize Intermediate Results

3. Bottom-Up

Memorize Intermediate Results

Q // Initialize $arr[n][m]$ to undefined



def LCS(P, Q, n, m)

Q if $arr[n][m] \neq \text{undefined}$: 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 $P[n-1] \neq Q[n-1]$: // just for clarity

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

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

result = max{ tmp1, tmp2 }

Q $arr[n][m] = \text{result}$

return result

P = "AA"
Q = "BBB"

Memorize Intermediate Results

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

def LCS(P, Q, n, m)

Q 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[n-1]: // just for clarity

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

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

result = max{ tmp1, tmp2 }

→ Q arr[n][m] = result
return result



$2nm$

time/call
 $= O(1)$

$O(nm)$

nm

Explanation:

1. The function **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 **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:

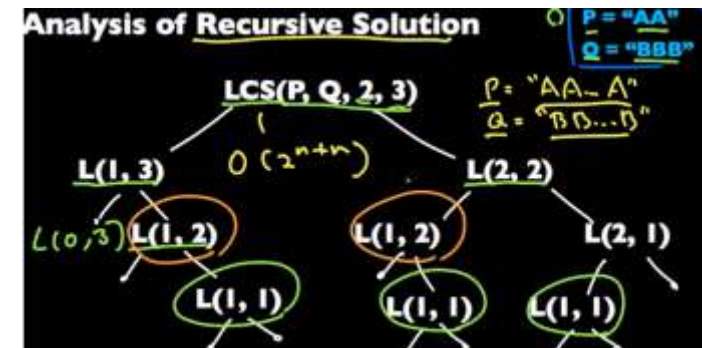
- The depth of the recursion tree is at most $m + n$, leading to $O(2^{(m+n)})$ possible recursive calls.
- However, in standard LCS problems, where $m \approx n$, the complexity can also be expressed as $O(2^{\max(m, n)})$.

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:

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

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)** → Represents the value from the **left** (excluding the current character from string Q).
- **L(n-1, m)** → 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:

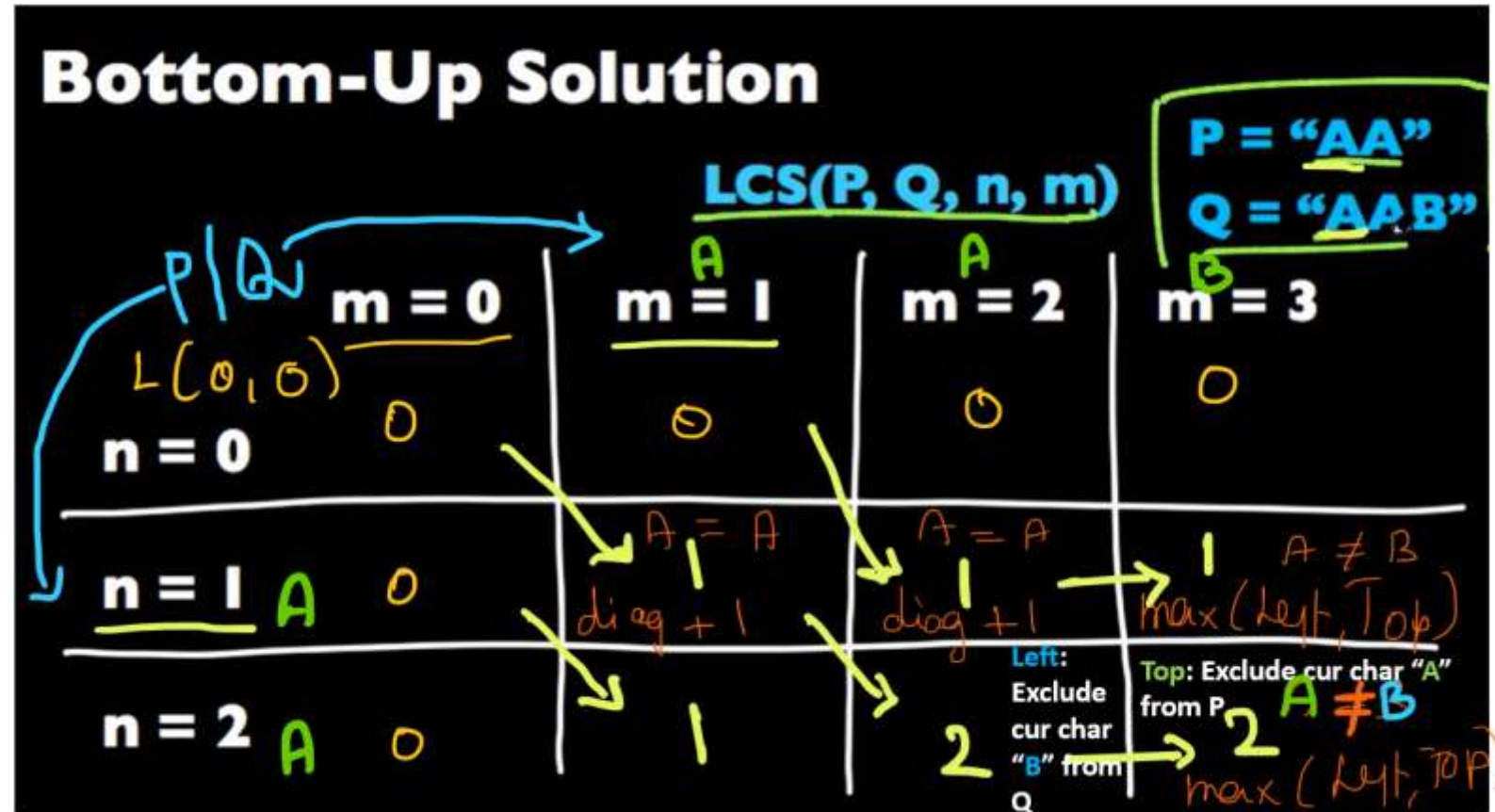
- P = "AA"
- Q = "AAB"

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: $A = A \rightarrow$ Include 'A'
 - Move diagonally to (1,1).
3. $L(1,1) = 1 + L(0,0) = 1$
 - Characters match: $A = A \rightarrow$ 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
acbcf

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
acbcf

		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

acbcf

		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

acbcf

		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

		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

abcdaf
acbcf

		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
acbcf

		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

		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

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

abcdaf

acbcf

		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

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

abcdaf

acbcf

		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

		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

abcdaf
acbcf

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	0	0	0	0	0	0	0
a	0	1	1	1	1	1	1
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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

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abcdaf
acbcf

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

		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

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

Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

if (input¹[i] == input²[j])

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

$LCS(P_i, Q_o) = 1 + LCS(P_i, Q_r)$

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
```

$LCS(P_i, Q_o) = 1 + LCS(P_i, Q_r)$

$LCS(P_i, Q_o) = \max \{ LCS(P_i, Q_o), LCS(P_o, Q_r) \}$

Longest Common Subsequence

Bottom-Up Approach – Non-Recursive

if (input¹[i] == input²[j])

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

$$\text{LCS}(P_i, Q_o) = 1 + \text{LCS}(P_i, Q_i)$$

else

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

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

Recursive Solution (Code)

```
def LCS(P, Q, n, m)
    if n == 0 or m == 0: // base case
        result = 0
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        tmp1 = LCS(P, Q, n-1, m)
        tmp2 = LCS(P, Q, n, m-1)
        result = max{ tmp1, tmp2 }
    return result
```

$\text{LCS}(P_i, Q_o) = 1 + \text{LCS}(P_i, Q_i)$

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

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.