

*Dynamic Programming,
Longest Common
Subsequence*

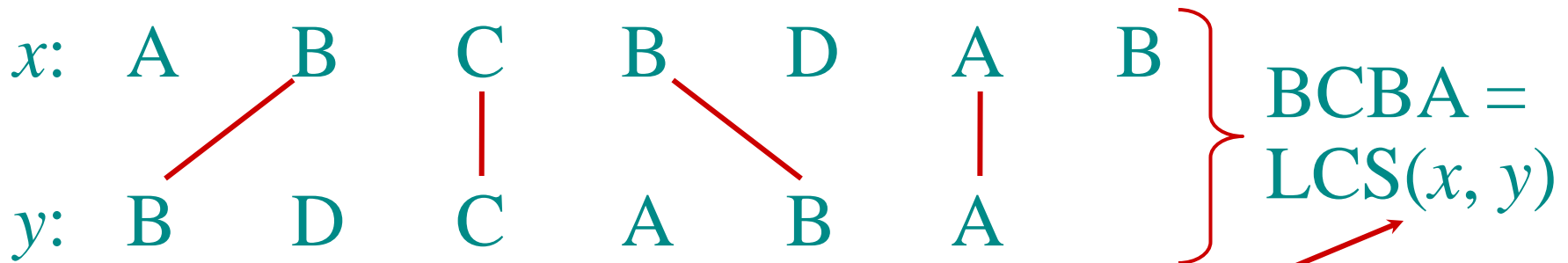
Dynamic programming

Design technique, like divide-and-conquer.

Example: Longest Common Subsequence (LCS)

- Given two sequences $x[1 \dots m]$ and $y[1 \dots n]$, find a longest subsequence common to them both.

“a” *not* “the”



functional notation,
but not a function

Brute-force LCS algorithm

Check every subsequence of $x[1 \dots m]$ to see if it is also a subsequence of $y[1 \dots n]$.

Analysis

- Checking = $O(n)$ time per subsequence.
- 2^m subsequences of x (each bit-vector of length m determines a distinct subsequence of x).

Worst-case running time = $O(n2^m)$
= exponential time.

Towards a better algorithm

Simplification:

1. Look at the *length* of a longest-common subsequence.
2. Extend the algorithm to find the LCS itself.

Notation: Denote the length of a sequence s by $|s|$.

Strategy: Consider *prefixes* of x and y .

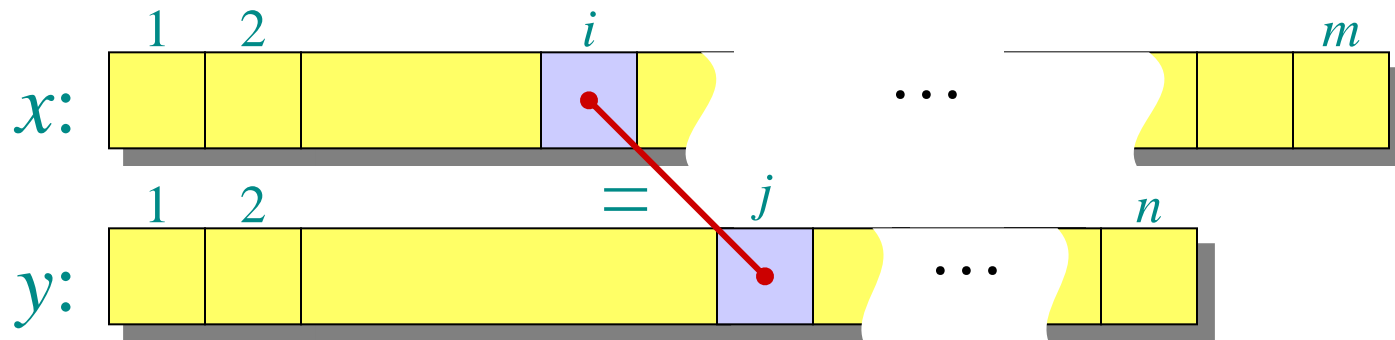
- Define $c[i, j] = |\text{LCS}(x[1 \dots i], y[1 \dots j])|$.
- Then, $c[m, n] = |\text{LCS}(x, y)|$.

Recursive formulation

Theorem.

$$c[i, j] = \begin{cases} c[i-1, j-1] + 1 & \text{if } x[i] = y[j], \\ \max \{ c[i-1, j], c[i, j-1] \} & \text{otherwise.} \end{cases}$$

Proof. Case $x[i] = y[j]$:



Let $z[1 \dots k] = \text{LCS}(x[1 \dots i], y[1 \dots j])$, where $c[i, j] = k$. Then, $z[k] = x[i]$, or else z could be extended. Thus, $z[i \dots k-1]$ is CS of $x[1 \dots i-1]$ and $y[1 \dots j-1]$.

Proof (continued)

Claim: $z[1 \dots k-1] = \text{LCS}(x[1 \dots i-1], y[1 \dots j-1])$.

Suppose w is a longer CS of $x[1 \dots i-1]$ and $y[1 \dots j-1]$, that is, $|w| > k-1$. Then, *cut and paste*: $w \parallel z[k]$ (w concatenated with $z[k]$) is a common subsequence of x and y with $|w \parallel z[k]| > k$. Contradiction, proving claim.

Thus, $c[i-1, j-1] = k-1$, which implies that $c[i, j] = c[i-1, j-1] + 1$.

Other cases are similar. 

Dynamic-programming hallmark #1

Optimal substructure

*An optimal solution to a problem
(instance) contains optimal
solutions to subproblems.*

If $z = \text{LCS}(x, y)$, then any prefix of z is
an LCS of a prefix of x and a prefix of y .

Recursive algorithm for LCS

$\text{LCS}(x, y, i, j)$

if $x[i] = y[j]$

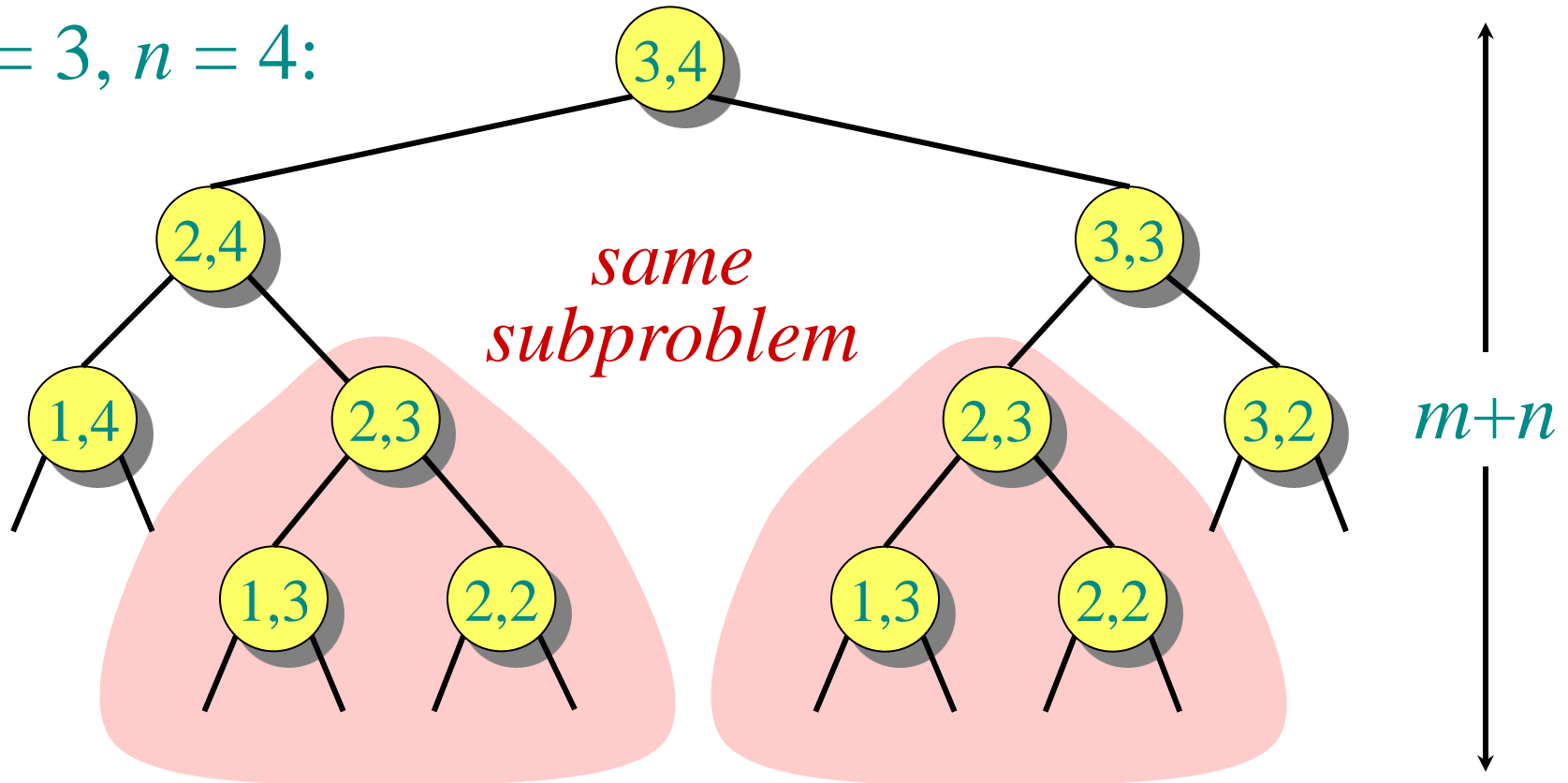
then $c[i, j] \leftarrow \text{LCS}(x, y, i-1, j-1) + 1$

else $c[i, j] \leftarrow \max \{ \text{LCS}(x, y, i-1, j), \text{LCS}(x, y, i, j-1) \}$

Worst-case: $x[i] \neq y[j]$, in which case the algorithm evaluates two subproblems, each with only one parameter decremented.

Recursion tree

$m = 3, n = 4$:



Height = $m + n \Rightarrow$ work potentially exponential,
but we're solving subproblems already solved!

Dynamic-programming hallmark #2

Overlapping subproblems

A recursive solution contains a “small” number of distinct subproblems repeated many times.

The number of distinct LCS subproblems for two strings of lengths m and n is only mn .

Memoization algorithm

Memoization: After computing a solution to a subproblem, store it in a table. Subsequent calls check the table to avoid redoing work.

$\text{LCS}(x, y, i, j)$

if $c[i, j] = \text{NIL}$

then if $x[i] = y[j]$

then $c[i, j] \leftarrow \text{LCS}(x, y, i-1, j-1) + 1$

else $c[i, j] \leftarrow \max \{ \text{LCS}(x, y, i-1, j), \text{LCS}(x, y, i, j-1) \}$

*same
as
before*

Time = $\Theta(mn)$ = constant work per table entry.

Space = $\Theta(mn)$.

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
i		Yj						
	0	X_i						
	1	A						
	2	B						
	3	C						
	4	B						

$X = \text{ABCB}; m = |X| = 4$

$Y = \text{BDCAB}; n = |Y| = 5$

Allocate array $c[5,4]$

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
		Yj		B	D	C	A	B
i	Xi							
0			0	0	0	0	0	0
1	A		0					
2	B		0					
3	C		0					
4	B		0					

for i = 1 to m c[i,0] = 0

for j = 1 to n c[0,j] = 0

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
i		Y _j		B	D	C	A	B
	X _i							
0			0	0	0	0	0	0
1	A	0	0	0				
2	B	0						
3	C	0						
4	B	0						

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i	0	0	0	0	0	0	0
1	A	0	0	0	0	0		
2	B	0						
3	C	0						
4	B	0						

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BDCA

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i	0	0	0	0	0	0	0
1	A	0	0	0	0	0	1	
2	B	0						
3	C	0						
4	B	0						

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i	0	0	0	0	0	0	0
1	A	0	0	0	0	1	→	1
2	B	0						
3	C	0						
4	B	0						

if ($X_i == Y_j$)

$c[i,j] = c[i-1,j-1] + 1$

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

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
i		Y _j		B	D	C	A	B
	X _i							
0			0	0	0	0	0	0
1	A		0	0	0	0	1	1
2	B		0	1				
3	C		0					
4	B		0					

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BD CAB

		j	0	1	2	3	4	5
i		Y _j		B	D	C	A	B
	X _i							
0			0	0	0	0	0	0
1	A		0	0	0	0	1	1
2	B		0	1	1	1	1	
3	C		0					
4	B		0					

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BDCAB

i	j	Y _j	X _i	0	1	2	3	4	5
				0	B	D	C	A	B
0				0	0	0	0	0	0
1			A	0	0	0	0	1	1
2			B	0	1	1	1	1	2
3			C	0					
4			B	0					

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
i		Y _j		B	D	C	A	B
	X _i							
0			0	0	0	0	0	0
1	A		0	0	0	0	1	1
2	B		0	1	1	1	1	2
3	C		0	↓	↓			
4	B		0	1	1			

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BD CAB

		j	0	1	2	3	4	5
		Y_j		B	D	C	A	B
i	X_i	0	0	0	0	0	0	0
	A	1	0	0	0	0	1	1
	B	2	0	1	1	1	1	2
	C	3	0	1	1	2		
	B	4	0					

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BDCAB

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i	0	0	0	0	0	0	0
1	A	0	0	0	0	0	1	1
2	B	0	1	1	1	1	1	2
3	C	0	1	1	2	2	2	2
4	B	0						

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB

BDCAB

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i	0	0	0	0	0	0	0
1	A	0	0	0	0	0	1	1
2	B	0	1	1	1	1	1	2
3	C	0	1	1	2	2	2	2
4	B	0	1					

if ($X_i == Y_j$)

$c[i,j] = c[i-1,j-1] + 1$

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

LCS Example

ABCB
BD CAB

		j	0	1	2	3	4	5
		Y _j		B	D	C	A	B
i	X _i							
0			0	0	0	0	0	0
1	A		0	0	0	0	1	1
2	B		0	1	1	1	1	2
3	C		0	1	1	2	2	2
4	B		0	1	1	2	2	

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

LCS Example

ABCB
BD CAB

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i	0	0	0	0	0	0	0
1	A	0	0	0	0	0	1	1
2	B	0	1	1	1	1	1	2
3	C	0	1	1	2	2	2	2
4	B	0	1	1	2	2	2	3

if ($X_i == Y_j$)
 $c[i,j] = c[i-1,j-1] + 1$
 else $c[i,j] = \max(c[i-1,j], c[i,j-1])$

Algorithm design


How to find actual LCS

- So far, we have just found the *length* of LCS, but not LCS itself.
- We want to modify this algorithm to make it output LCS of X and Y

Each $c[i, j]$ depends on $c[i - 1, j]$ and $c[i, j - 1]$ or $c[i - 1, j - 1]$.

For each $c[i, j]$ we can say how it was acquired:

2	2
2	3



For example, here

$$c[i, j] = c[i-1, j-1] + 1 = 2 + 1 = 3$$

Algorithm design

- Remember that

$$c[i, j] = \begin{cases} c[i - 1, j - 1] + 1, & \text{if } x[i] = y[j] \\ \max(c[i - 1, j], c[i, j - 1]), & \text{otherwise} \end{cases}$$

So we can start from $c[m, n]$ and go backwards

Whenever $c[i, j] = c[i - 1, j - 1] + 1$, record $x[i]$

When $i=0$ or $j=0$ (i.e. we reached the beginning), output recorded letters in reverse order

Finding LCS

		j	0	1	2	3	4	5
			Y _j	B	D	C	A	B
i	X _i							
0			0	0	0	0	0	0
1	A		0	0	0	0	1	1
2	B		0	1	1	1	1	2
3	C		0	1	1	2	2	2
4	B		0	1	1	2	2	3

To construct LCS, start in the bottom right corner and follow the arrows. ↖ indicates a matching character

Finding LCS

		j	0	1	2	3	4	5	
			Y _j		B	D	C	A	B
i	X _i	0	0	0	0	0	0	0	0
1	A	0	0	0	0	0	1	1	
2	B	0	1	1	1	1	1	2	
3	C	0	1	1	2	2	2	2	
4	B	0	1	1	2	2	2	3	

LCS (reversed order): **B C B**

LCS (straight order): **B C B**

Dynamic-programming algorithm

IDEA:

Compute the table bottom-up.

Time = $\Theta(mn)$.

		A	B	C	B	D	A	B
	0	0	0	0	0	0	0	0
B	0	0	1	1	1	1	1	1
D	0	0	1	1	1	2	2	2
C	0	0	1	2	2	2	2	2
A	0	1	1	2	2	2	3	3
B	0	1	2	2	3	3	3	4
A	0	1	2	2	3	3	4	4

Dynamic-programming algorithm

IDEA:

Compute the table bottom-up.

Time = $\Theta(mn)$.

Reconstruct LCS by tracing backwards.

Space = $\Theta(mn)$.

Exercise:

$O(\min\{m, n\})$.

	A	B	C	B	D	A	B
B	0	0	0	0	0	0	0
D	0	0	1	1	1	1	1
C	0	0	1	2	2	2	2
A	0	1	1	2	2	3	3
B	0	1	2	2	3	3	4
A	0	1	2	2	3	4	4