



ECM5605(5075) S'20 Algorithms

Lecture 04 Dynamic Programming (1): Longest Common Sequence

Review Problems

- What is Order Statistics?
 - How to design an efficient Algorithm?
 - Worst-case complexity? Average-case complexity?
 - How to analyze the average-case of RAND-SELECT?
 - What's the key idea to guarantee $O(n)$? Hint: BFPRT-Select
 - Why groups of 5? Can we use 3?
- Back to Quick Sort
 - How to prove the average-case complexity?
 - Can guarantee $O(n \lg n)$ if using BFPRT-Select?

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From Divide-and-Conquer ...

- Essence: solve problems by combining the solutions to subproblems.
- Divide-and-conquer algorithms
 - Partition a problem into independent subproblems, solve the subproblems recursively, and then combine their solutions to solve the original problem.
 - What's the potential problem??
 - ⇒ *Inefficient to solve repeated subproblems*
- Therefore, if many independent subproblems are overlapping
 - ⇒ only need to each subproblem just once.
 - ⇒ **Dynamic programming**

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Birth of Dynamic Programming

- Richard Bellman, 1952
 - multistage stochastic decision processes
- “Dynamic Programming” so named because
 - Originally associated with movements (trading) between time and space, thus “dynamic”
 - “programming” refers to the process of formulating the constraints of a problem
 - ⇒ by analogy to “linear programming” and other forms of optimization
- Can be also used in deterministic problems
 - ⇒ **(recursive) optimization** problems

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Dynamic Programming (DP)

- Typically apply to **optimization** problems:
 - Find **a (not “the”)** solution with the optimal (maximum / minimum) value
- 4 steps in developing a DP algorithm
 1. **Characterize** the structure of an optimal solution.
 2. **Recursively define** the **formula** of an optimal solution.
 3. **Compute** the value of an optimal solution either **bottom-up** in a table (or top-down with caching)
 4. **Construct** an **optimal solution** from computed information (this step can be omitted if only the value is required)

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When to Use DP

- Characteristics of DP problems:
 1. problem can be divided into **stages (subproblems)**
 2. each stage has one/more **states (substructures)**
 3. you make a **decision** at each stage
 4. the decision you make affects the state for the next stage
 5. there is a **recursive relationship** between the value of the decision at the stage and the previously **found optima (principle of optimality)**
- **Hopeless configurations:** for an n -element set
 - # of permutations: $n!$
 - # of subsets: 2^n

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Longest Common Subsequence

- Problem Formulation:

Given 2 sequences, $X = \langle x_1, \dots, x_m \rangle$ and $Y = \langle y_1, \dots, y_n \rangle$, find a common subsequence whose length is maximum.
- Example: Subsequence need not be consecutive, but must be in order.

X: springtime	ncaa tournament	basketball
Y: printing	north carolina	krzyzewski

Diagrams show matching lines between characters of X and Y:

- springtime vs printing: s-p, r-i, n-t, g-e, t-m (5 matches)
- ncaa tournament vs north carolina: n-o, t-u, r-n, a-m, e-n-t (5 matches)
- basketball vs krzyzewski: b-k, a-z, t-z, e-w, b-s, l-i (6 matches)

LCS: printi ncarna ke

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Naïve Algorithm and Analysis

- Brute-force LCS algorithm:

For every subsequence of X , check whether it's a subsequence of Y
- Analysis:
 - Each subsequence takes $O(n)$ time to check: scan Y for first letter, for second, and so on.
 - 2^m subsequences of X to check.
 - Worst-case runtime: $O(n2^m)$ exponential time
- Read code for the naïve version and watch the number of comparisons

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Towards a Better Algorithm

- Simplification:
 1. Look at the **length** of a longest-common subsequence
 - Denote the length for a sequence of **S** by $|S|$
 2. Extend the algorithm to find the LCS itself
- Strategy: consider **prefixes** of **X** and **Y**
 - Define $c[i,j] = |LCS(X[1..i], Y[1..j])|$
 - Then $c[m,n] = |LCS(X,Y)|$

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Step 1: Optimal Substructure of LCS

Theorem

Let $Z = \langle z_1, \dots, z_k \rangle$ be any LCS of X and Y .

- 1 If $x_m = y_n$, then $z_k = x_m = y_n$ and Z_{k-1} is an LCS of X_{m-1} and Y_{n-1} .
- 2 If $x_m \neq y_n$, then either $z_k \neq x_m$ and Z is an LCS of X_{m-1} and Y .
- 3 or $z_k \neq y_n$ and Z is an LCS of X and Y_{n-1} .

Proof: (case 1: if $x_m = y_n$)

Any sequence Z' that does not end in $x_m = y_n$ can be made longer by adding $x_m = y_n$ to the end. Therefore,

- 1) longest common subsequence (LCS) Z must end in $x_m = y_n$.
- 2) Z_{k-1} is a common subsequence of X_{m-1} and Y_{n-1} , and
- 3) there is no **longer** CS of X_{m-1} and Y_{n-1} , or Z_{k-1} would not be an LCS.

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Step 1: Optimal Substructure of LCS (cont'd)

Theorem

Let $Z = \langle z_1, \dots, z_k \rangle$ be any LCS of X and Y .

- 1 If $x_m = y_n$, then $z_k = x_m = y_n$ and Z_{k-1} is an LCS of X_{m-1} and Y_{n-1} .
- 2 If $x_m \neq y_n$, then either $z_k \neq x_m$ and Z is an LCS of X_{m-1} and Y .
- 3 or $z_k \neq y_n$ and Z is an LCS of X and Y_{n-1} .

Proof: (case 2: $x_m \neq y_n$, and $z_k \neq x_m$)

Since Z does not end in x_m ,

- 1) Z is a common subsequence of X_{m-1} and Y , and
- 2) there is no **longer** CS of X_{m-1} and Y , or Z would not be an LCS

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Step 2: Recursive Formula for Solution

Theorem

Let $Z = \langle z_1, \dots, z_k \rangle$ be any LCS of X and Y .

- 1 If $x_m = y_n$, then $z_k = x_m = y_n$ and Z_{k-1} is an LCS of X_{m-1} and Y_{n-1} .
- 2 If $x_m \neq y_n$, then either $z_k \neq x_m$ and Z is an LCS of X_{m-1} and Y .
- 3 or $z_k \neq y_n$ and Z is an LCS of X and Y_{n-1} .



$$c[i,j] = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ c[i-1, j-1] + 1 & \text{if } i, j > 0 \text{ and } x_i = y_j \\ \max(c[i-1, j], c[i, j-1]) & \text{if } i, j > 0 \text{ and } x_i \neq y_j \end{cases}$$

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LCS Example

We'll see how LCS algorithm works on the following example:

- $X = \text{ABCB}$
- $Y = \text{BDCAB}$

What is the Longest Common Subsequence of X and Y ?

$\text{LCS}(X, Y) = \text{BCB}$

$X = \text{A} \text{ B } \text{ C } \text{ B}$

$Y = \text{ B D C A B}$

LCS Example (I)

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

$X = \text{ABCB}; m = |X| = 4;$ \Rightarrow Allocate 5x6 Matrix **C**
 $Y = \text{BDCAB}; n = |Y| = 5;$

LCS Example (1)

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

ABCB
BDCAB

i	j	0	1	2	3	4	5
		Y_j	B	D	C	A	B
0	X_i	0	0	0	0	0	0
1	A	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 $c[i,j] = \max(c[i-1,j], c[i,j-1])$

ABCB
BDCAB

LCS Example (3)

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

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ABCB
BDCAB

LCS Example (4)

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

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ABCB
BDCAB

LCS Example (5)

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

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ABCB
BDCAB

LCS Example (6)

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

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L04-20

ABCB
BDCAB

LCS Example (7)

	j	0	1	2	3	4	5
i	Yj		B	D	C	A	B
0	Xi	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 $c[i,j] = \max(c[i-1,j], c[i,j-1])$

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ABCB
BDCAB

LCS Example (8)

	j	0	1	2	3	4	5
i	Yj		B	D	C	A	B
0	Xi	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 $c[i,j] = \max(c[i-1,j], c[i,j-1])$

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ABCB
BDCAB

LCS Example (10)

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

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

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ABCB
BDCAB

LCS Example (11)

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

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

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ABCB
BDCAB

LCS Example (12)

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

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

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ABCB
BDCAB

LCS Example (13)

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

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

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ABCB
BDCAB

LCS Example (14)

	j	0	1	2	3	4	5
i	Yj		B	D	C	A	B
0	Xi	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 $c[i,j] = \max(c[i-1,j], c[i,j-1])$

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ABCB
BDCAB

LCS Example (15)

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

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

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LCS Algorithm Runtime

- LCS algorithm calculates the values of each entry of the array $c[m,n]$
- So what is the runtime to compute C ?

$O(m \times n)$

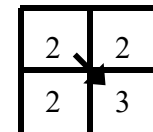
since each $c[i,j]$ is calculated in constant time, and there are $m \times n$ elements in the array

Step 4: Construct a LCS

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

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

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



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

Step 4: Construct a LCS (cont'd)

- Remember that

$$c[i,j] = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ c[i-1,j-1] + 1 & \text{if } i,j > 0 \text{ and } x_i = y_j \\ \max(c[i-1,j], c[i,j-1]) & \text{if } i,j > 0 \text{ and } x_i \neq y_j \end{cases}$$

- So we can start from $c[m,n]$ and go backwards
- Whenever $c[i,j] = c[i-1,j-1] + 1$, remember $x[i]$ (because $x[i]$ is a part of LCS)
- When $i=0$ or $j=0$ (i.e. we reached the beginning), output remembered letters in reverse order

Example of Constructing a LCS

		j	0	1	2	3	4	5
i		y_j	B	D	C	A	B	
	x_i							
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

Example of Constructing a LCS (cont'd)

		j					
		0	1	2	3	4	5
i	yj		B	D	C	A	B
	xi						
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

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

LCS (straight order): **B C B**
(this string turned out to be a palindrome)

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Analysis DP on LCS Problem

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

$LCS_1 = BCBA$

$LCS_2 = BDAB$

- Space complexity: $O(m \times n)$
- Time complexity:
 - Computing the table: time = $O(m \times n)$
 - Constructing an LCS: time = $O(m + n)$

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Optimizing Space in LCS

- Recap memory usage

		j					
		0	1	2	3	4	5
i	yj		B	D	C	A	B
	xi						
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

- Do we really need a $m \times n$ table to store the LCS values? Any improvement?
- Implement your space-optimized version

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Summary

- What are common/different points between Divide-and-Conquer and DP?
- 2 Hallmarks of dynamic programming:
 - **Optimal substructure**: an optimal solution to a problem (instance) contains optimal solutions to subproblems
 - **Overlapping subproblems**: a recursive solution contains a “small” number of distinct subproblems repeated many times
- What is longest common sequence problem?
 - How to use DP to solve this problem
 - Space complexity? Time complexity??

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Ex: Longest Increasing Subsequence

- The **Longest Increasing Subsequence (LIS)** problem is to find the length of the longest subsequence of a given sequence such that all elements of the subsequence are sorted in increasing order.
- For example, length of LIS for {10, 22, 9, 33, 21, 50, 41, 60, 80} is 6
– LIS = {10, 22, 33, 50, 60, 80}

DP for LIS

- Let arr be the input array and $L(i)$ be the length of the LIS ending at index i
 $\Rightarrow arr[i]$ is the last element of the LIS
- Then, $L(i)$ can be recursively written as:

$$L(i) = \begin{cases} 1 + \max(L(j)), & \text{if } 0 < j < i \text{ and } arr[j] < arr[i] \\ 1, & \text{if no such } j \text{ exists} \end{cases}$$

- Complexity is ???
- Refer to LIS_v1.cpp and develop your DP solution in C/C++