

NCKU Programming Contest Training Course

Course 8

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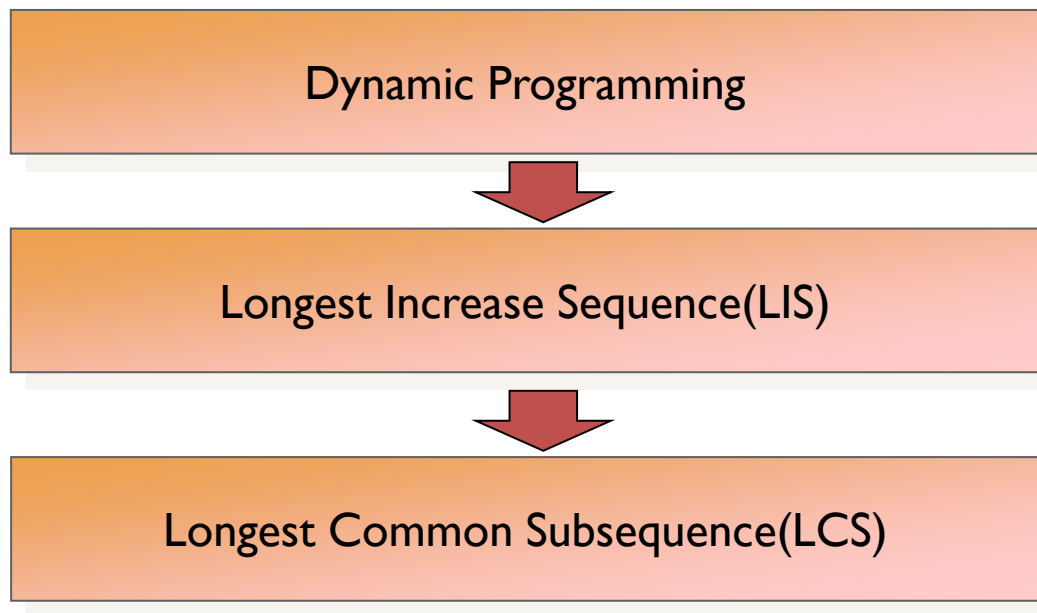
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http://myweb.ncku.edu.tw/~f74986133/Course_8.rar

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Outline



Dynamic Programming

- Dynamic Programming (DP)
 - Dynamic programming is a general algorithm design technique for solving problems defined by or formulated by recursion structure with overlapping substructure.
- Purpose
 - Optimization
 - Avoid duplicate search and calculation
 - Two categories
 - Minimization (maximization) problem
 - Combinatorial problem



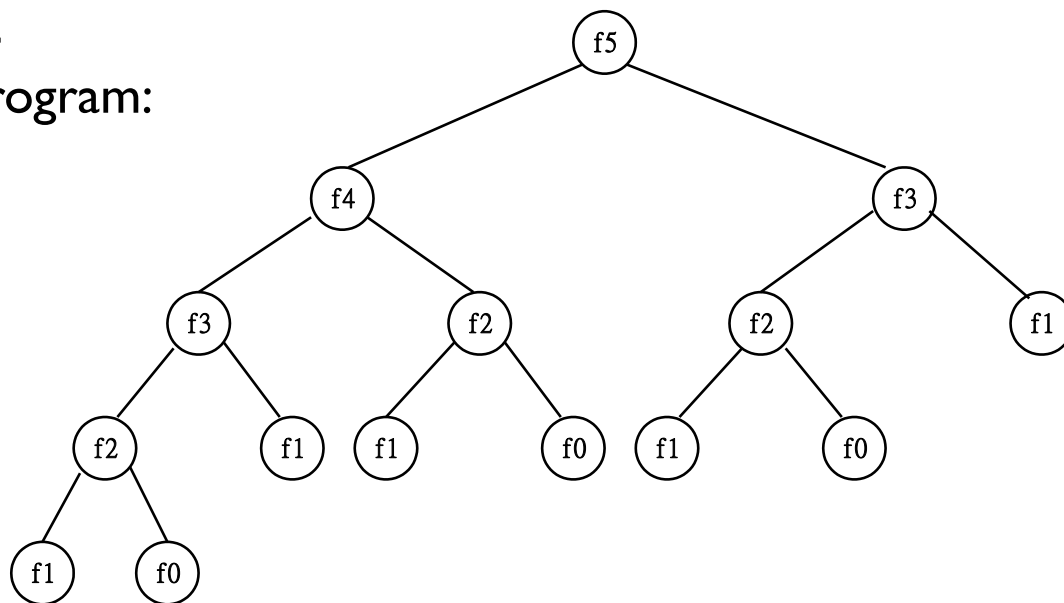
Recall for Fib-Seq

- Fibonacci sequence: 0 , 1 , 1 , 2 , 3 , 5 , 8 , 13 , 21 , ...

$$F_i = i \quad \text{if } i \leq 1$$

$$F_i = F_{i-1} + F_{i-2} \quad \text{if } i \geq 2$$

- Solved by a recursive program:



- Much replicated computation is done.
- It should be solved by a simple loop.



Solving Method

- Solving the Fib-Sequence
 - Bottom up design
 - Using the for loop or while loop without recursive stack
 - Top-down design
 - Using the recursive design with memorization for computation
- Top-Down design
 - (1) basic condition
 - (2) if memorized
 - (3) recurrence
 - (4) return the value
- Take a problem for example
 - PKU: 1579 Function Run Fun



Function Run Fun

- How to figure out this is a DP problem ?
 1. We want to find out what value $w(a, b, c)$ is
 2. $w(a, b, c)$ can be generated by the functions at last slide
 - ex, if we want to know $w(50, 50, 50)$, we have to know $w(20, 20, 20)$ at first
 3. The final answer is consist of lots of **subproblems**
 - ex, $w(a, b, c) = w(a-1, b, c) + w(a-1, b-1, c) + w(a-1, b, c-1) - w(a-1, b-1, c-1)$
subproblems



Function Run Fun

- if ($a \leq 0$ or $b \leq 0$ or $c \leq 0$)
 - $w(a, b, c) = 1$
- if ($a > 20$ or $b > 20$ or $c > 20$)
 - $w(a, b, c) = w(20, 20, 20)$
- if ($a < b$ and $b < c$)
 - $w(a, b, c) = w(a, b, c-1) + w(a, b-1, c-1) - w(a, b-1, c)$
- otherwise
 - $w(a, b, c) = w(a-1, b, c) + w(a-1, b-1, c) + w(a-1, b, c-1) - w(a-1, b-1, c-1)$



Just Practice

- Practice
 - PKU: I579 Function Run Fun

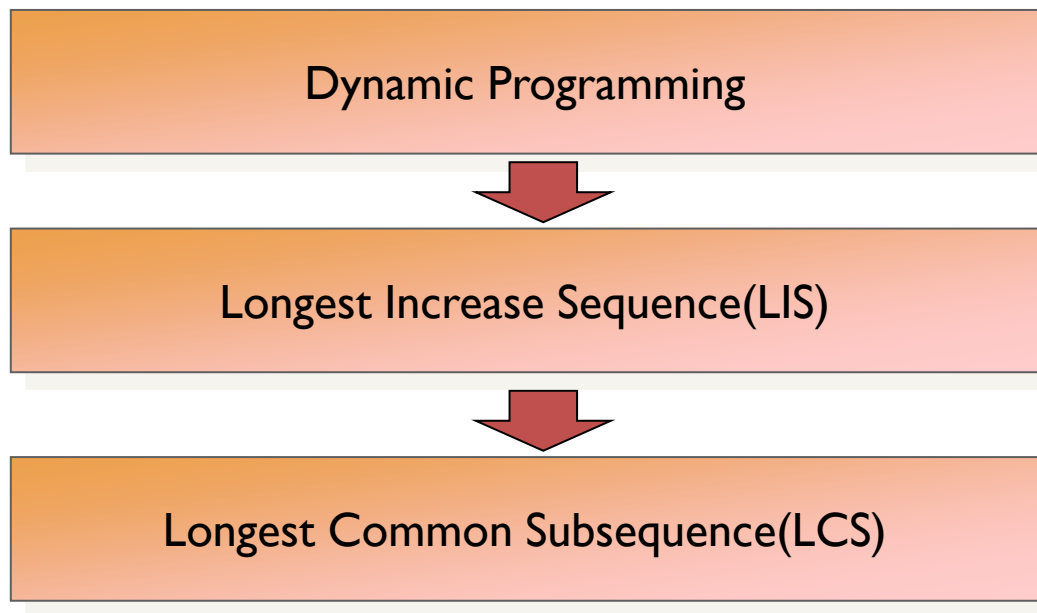


Programming Steps

- Programming Strategy
 - (1) Check the category, min-max problem or combination problem?
 - (2) Check if there is a order(*the optimal order*)
 - This is a very important step!!!
 - Ordered -> optimization with dp
 - In-ordered -> bitmask dp, memorized dp or non-dp problems
 - (3) Think the recurrence formulation
 - (4) Write a program to solve it
 - Top-down
 - Bottom-up
 - (5) Backtrack the optimal path
- Memory Strategy
 - Set up the recorded table



Outline



LIS Problem

- Longest Increasing Subsequence
 - The longest increasing subsequence problem is to find **a subsequence of a given sequence in which the elements in this subsequence are in sorted order (lowest to highest)**, and in which **the length** of the subsequence **is as long as possible**.
 - The elements in the subsequence are not necessarily to be continuous.
 - Two well-known method to solve this problem are followings:
 - (1) DP by $O(N^2)$
 - (2) Greedy with binary search by $O(N\log N)$



LIS Example

- Example
 - 0, 8, 4, 12, 2, 10, 6, 14, 1, 9, 5, 13, 3, 11, 7, 15
 - A increasing subsequence is 0, 4, 14, 15
 - The longest increasing subsequence is 0, 2, 6, 9, 13, 15 with length six
- Question
 - Is the longest increasing subsequence unique?
 - How should we deal with this problem in different situation by which method?



DP method

- Dynamic programming approach
 - Recall the design strategy
 - (1) Check the category
 - (2) Check the order property
 - (3) Think the recurrence
 - (4) Write and problem with two methods
 - (5) Backtrack the optimal path



DP method

- Rule 1
 - Order?
- Rule 2
 - Category
- Rule 3
 - Given a sequence with n elements stored in an array $seq[i]$ where $1 \leq i \leq n$.
 - **Define $dp[i]$ for representing that the longest length of the increasing subsequence that ended by $seq[i]$ from $seq[1]$ to $seq[i]$.**
 - So that the recurrence can be formulated as the following:
 - Initialized the $dp[i]$ by 1
 - $dp[i] = \max(dp[j] + 1)$, where $1 \leq j < i$ and $seq[j] < seq[i]$
 - Also define a pi array $pi[i]$ that represent the previous element of the element i in the increasing subsequence.



DP method

- Rule 3
 - Example
 - $seq[9] = \{9, 5, 2, 8, 7, 3, 1, 6, 4\}$
 - Find the $dp[i]$ and $pi[i]$.

initial

seq	9 (1)	5 (2)	2 (3)	8 (4)	7 (5)	3 (6)	1 (7)	6 (8)	4 (9)
dp	1	1	1	1	1	1	1	1	1
pi	-1	-1	-1	-1	-1	-1	-1	-1	-1

result

seq	9 (1)	5 (2)	2 (3)	8 (4)	7 (5)	3 (6)	1 (7)	6 (8)	4 (9)
dp	1	1	1	2	2	2	1	3	3
pi	-1	-1	-1	2	2	3	-1	6	6



DP method

- Rule 4
 - Write the program
- Rule 5
 - Trace the result
- Exercise
 - Write a program that find the length of the LIS for a given sequence.
 - Note:
 - Please use the dynamic programming as the practice.
 - The number of the element in the given sequence will not exceed 1000.
- Review
 - Time complexity $O(?)$
 - Space complexity $O(?)$
 - Compare with the brute force method.



Greedy Method

- Greedy Method
 - An efficient algorithm based on binary search.
 - Given the sequence $seq[9] = \{9, 5, 2, 8, 7, 3, 1, 6, 4\}$
 - <http://www.csie.ntnu.edu.tw/~u91029/LongestIncreasingSubsequence.html>

			8	7	3	3	6	4
9	5	2	2	2	2	1	3	3
							1	1



Greedy Method

- Exercise
 - Write a program that find the length of the LIS for a given sequence.
 - Note:
 - Please use the dynamic programming as the practice.
 - The number of the element in the given sequence will exceed 1000.
- Review
 - Time complexity $O(?)$
 - Space complexity $O(?)$

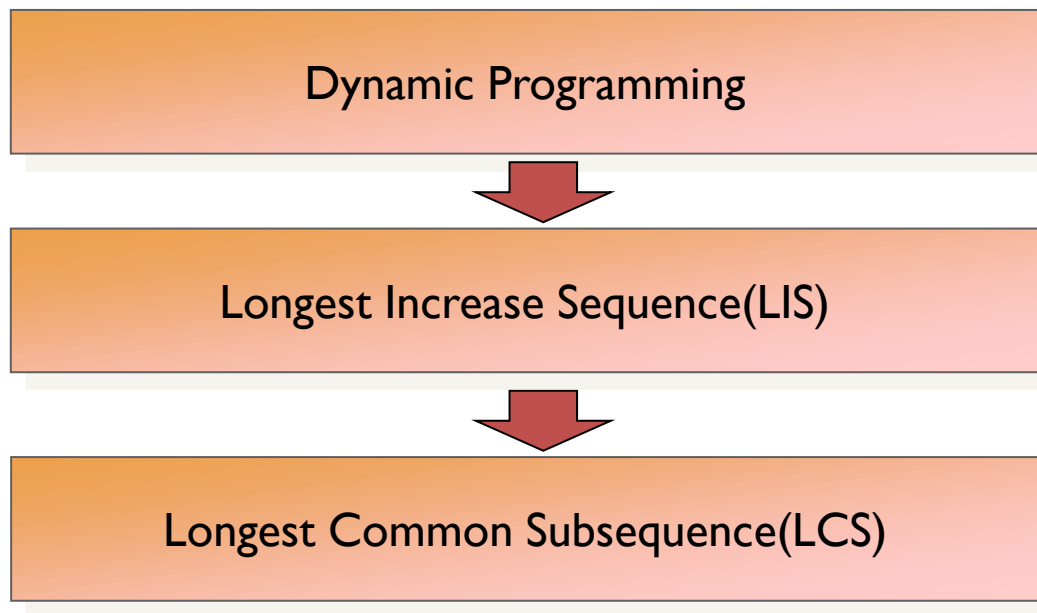


Just Practice

- Practice
 - NOJ 30 LIS Problem



Outline



LCS Problem

- Longest Common Subsequence
 - The longest increasing subsequence problem is to find a **common subsequence** of two given sequences in which the elements in this common subsequence **are appear in both original sequences**, and in which **the length** of the subsequence **is as long as possible**.
 - The elements in the subsequence are not necessarily to be continuous.
 - Two well-known method to solve this problem are followings:
 - (1) DP by $O(N^2)$
 - (2) Greedy with binary search by $O(N \log N)$



Example

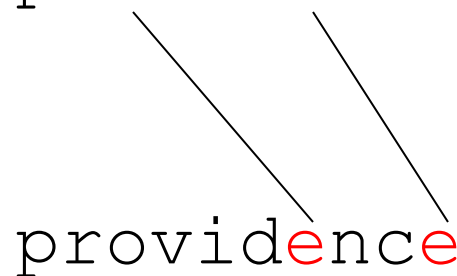
- Common Subsequence Example

president



providence

president



- Longest Common Subsequence

president



providence



DP method

- Dynamic programming approach
 - Recall the design strategy
 - (1) Check the category
 - (2) Check the order property
 - (3) Think the recurrence
 - (4) Write and problem with two methods
 - (5) Backtrack the optimal path



DP method

- Rule 1
 - Order?
- Rule 2
 - Category
- Rule 3
 - Let $A=a_1a_2\dots a_m$ and $B=b_1b_2\dots b_n$.
 - $len(i, j)$: the length of an LCS between $a_1a_2\dots a_i$ and $b_1b_2\dots b_j$
 - With proper initializations, $len(i, j)$ can be computed as follows.

$$len(i, j) = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0, \\ len(i-1, j-1) + 1 & \text{if } i, j > 0 \text{ and } a_i = b_j, \\ \max(len(i, j-1), len(i-1, j)) & \text{if } i, j > 0 \text{ and } a_i \neq b_j. \end{cases}$$

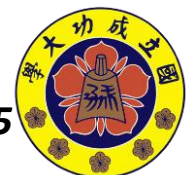
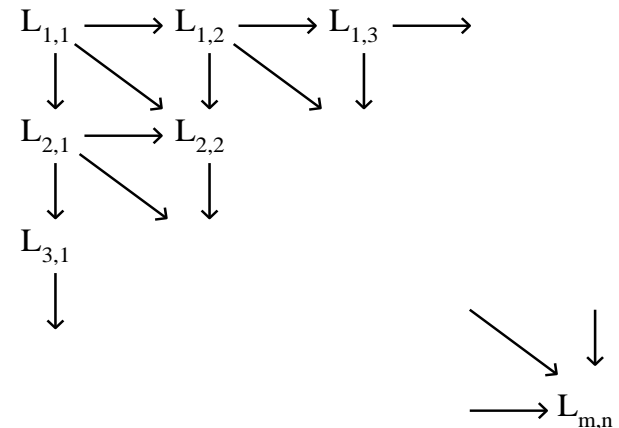


DP method

- Rule 4

procedure *LCS-Length*(*A*, *B*)

1. **for** $i \leftarrow 0$ **to** m **do** $len(i, 0) = 0$
2. **for** $j \leftarrow 1$ **to** n **do** $len(0, j) = 0$
3. **for** $i \leftarrow 1$ **to** m **do**
4. **for** $j \leftarrow 1$ **to** n **do**
5. **if** $a_i = b_j$ **then** $\begin{cases} len(i, j) = len(i-1, j-1) + 1 \\ prev(i, j) = \nwarrow \end{cases}$
6. **else if** $len(i-1, j) \geq len(i, j-1)$
7. **then** $\begin{cases} len(i, j) = len(i-1, j) \\ prev(i, j) = \uparrow \end{cases}$
8. **else** $\begin{cases} len(i, j) = len(i, j-1) \\ prev(i, j) = \leftarrow \end{cases}$
9. **return** len and $prev$



DP method

- Rule 4

– The dp result of the two string, “providence” and “president”

i \ j	0	1	2	3	4	5	6	7	8	9	10
		p	r	o	v	i	d	e	n	c	e
0	0	0	0	0	0	0	0	0	0	0	0
1 p	0	↖	←	←	←	←	←	←	←	←	←
2 r	0	↑	↖	←	←	←	←	←	←	←	←
3 e	0	↑	↑	←	←	←	↖	←	←	←	↖
4 s	0	↑	↑	←	←	←	←	←	←	←	←
5 i	0	↑	↑	←	←	↖	←	←	←	←	←
6 d	0	↑	↑	←	←	←	↖	←	←	←	←
7 e	0	↑	↑	←	←	←	←	↖	←	←	↖
8 n	0	↑	↑	←	←	←	←	←	↖	←	←
9 t	0	↑	↑	←	←	←	←	←	←	←	←



DP method

- Rule 5
 - Trace the path

procedure *Output-LCS*($A, prev, i, j$)

1 **if** $i = 0$ **or** $j = 0$ **then return**

2 **if** $prev(i, j) = "$ ↖ $"$ **then** $\left[\begin{array}{l} \text{Output-LCS}(A, prev, i-1, j-1) \\ \text{print } a_i \end{array} \right.$

3 **else if** $prev(i, j) = "$ ↑ $"$ **then** *Output-LCS*($A, prev, i-1, j$)

4 **else** *Output-LCS*($A, prev, i, j-1$)



DP method

- Rule 5

– The result “priden” of the two string, “providence” and “president”

i \ j	0	1	2	3	4	5	6	7	8	9	10
		<i>p</i>	<i>r</i>	<i>o</i>	<i>v</i>	<i>i</i>	<i>d</i>	<i>e</i>	<i>n</i>	<i>c</i>	<i>e</i>
0	0	0	0	0	0	0	0	0	0	0	0
1 <i>p</i>	0	1	1	1	1	1	1	1	1	1	1
2 <i>r</i>	0	1	2	2	2	2	2	2	2	2	2
3 <i>e</i>	0	1	2	2	2	2	2	3	3	3	3
4 <i>s</i>	0	1	2	2	2	2	2	3	3	3	3
5 <i>i</i>	0	1	2	2	2	3	3	3	3	3	3
6 <i>d</i>	0	1	2	2	2	3	4	4	4	4	4
7 <i>e</i>	0	1	2	2	2	3	4	5	5	5	5
8 <i>n</i>	0	1	2	2	2	3	4	5	6	6	6
9 <i>t</i>	0	1	2	2	2	3	4	5	6	6	6



DP method

- More Example
 - Given string A = bacad, string B = accbadcb
 - The dp table can be optimized as the following figure and the longest common string can be backtraced by the table.

		B							
		a	c	c	b	a	d	c	b
A	b	0	0	0	0	0	0	0	0
	a	0	①	1	1	2	2	2	2
	c	0	1	2	②	2	2	3	3
	a	0	1	2	2	③	3	3	3
	d	0	1	2	2	3	④	4	4



DP method

- Exercise
 - Write a program that find the length of the LCS for two given sequences.
 - Note:
 - Please use the dynamic programming as the practice.
 - The number of the element in the given sequence will not exceed 1000.
- Review
 - Time complexity $O(?)$
 - Space complexity $O(?)$
 - Compare with the brute force method.



Just Practice

- Practice
 - NOJ 31 LCS Problem



Homework 8

- UVA (**total 23 problems**)
 - 103, 108, 111, 231, 437, 481, 497, 507, 531, 836, 10066, 10131, 10192, 10252, 10405, 10534, 10635, 10684, 10723, 10755, 10827, 10949, 11582



Thank You For Attention!

