

# COL351: Analysis and Design of Algorithms

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# Dynamic Programming

# Dynamic Programming

## Longest common subsequence

### Problem

Let  $S$  and  $T$  be strings of characters.  $S$  is of length  $n$  and  $T$  is of length  $m$ . Find the *longest common subsequence* in  $S$  and  $T$ . This is the longest sequence of characters (not necessarily contiguous) that appear in both  $S$  and  $T$ .

- Example  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

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  - The longest common subsequence is XYXP
  - $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$
- How do we define the subproblems?

# Dynamic Programming

## Longest common subsequence

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- Let  $L(i, j)$  denote the length of the longest common subsequence in strings  $S[1], \dots, S[i]$  and  $T[1], \dots, T[j]$ .
- What is  $L(1, j)$  for  $1 < j \leq m$ ?

# Dynamic Programming

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- Similarly, we can define  $L(i, 1)$  for  $1 < i \leq n$ .
- Can we say something similar for  $L(i, j)$  for  $i, j \neq 1$ ?



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- Can we say something similar for  $L(i, j)$  for  $i, j \neq 1$ ?
  - Claim 1: If  $S[i] = T[j]$ , then  $L(i, j) = 1 + L(i - 1, j - 1)$ .
  - Claim 2: If  $S[i] \neq T[j]$ , then  $L(i, j) = \max \{L(i - 1, j), L(i, j - 1)\}$ .

# Dynamic Programming

## Longest common subsequence

- What is  $L(1, j)$  for  $1 < j \leq m$ ?
  - 1 if  $S[1]$  is present in the string  $T[1], \dots, T[j]$ , 0 otherwise.
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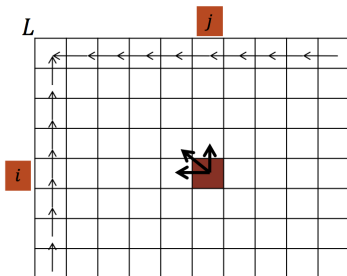


Figure: The arrows show the dependencies between subproblems.

# Dynamic Programming

## Longest common subsequence

### Algorithm

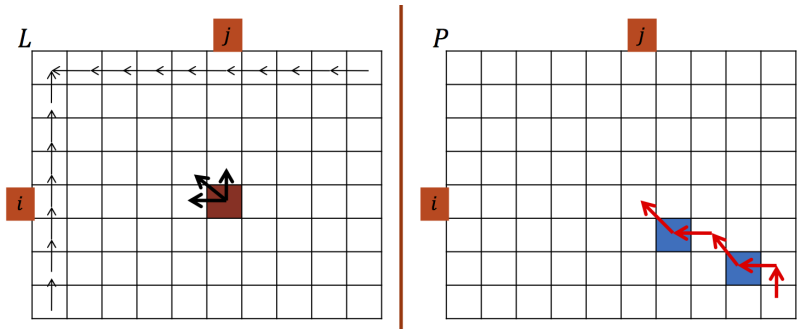
Length-LCS( $S, T$ )

- If ( $S[1] = T[1]$ ), then  $L[1, 1] \leftarrow 1$  else  $L[1, 1] \leftarrow 0$
- For  $j = 2$  to  $m$ 
  - If ( $S[1] = T[j]$ ), then  $L[1, j] \leftarrow 1$  else  $L[1, j] \leftarrow L[1, j - 1]$
- For  $i = 2$  to  $n$ 
  - If ( $S[i] = T[1]$ ), then  $L[i, 1] \leftarrow 1$  else  $L[i, 1] \leftarrow L[i - 1, 1]$
- For  $i = 2$  to  $n$ 
  - For  $j = 2$  to  $m$ 
    - If ( $S[i] = T[j]$ ) then  $L[i, j] \leftarrow 1 + L[i - 1, j - 1]$   
else  $L[i, j] \leftarrow \max \{L[i - 1, j], L[i, j - 1]\}$
- Return( $L[n, m]$ )

# Dynamic Programming

## Longest common subsequence

- How do we find a longest common subsequence?



**Figure:** Array  $P$  is used to maintain the pointers to the appropriate subproblem. The blue squares give the position of the characters in a longest common subsequence.

# Dynamic Programming

## Longest common subsequence

- Example:  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

$L$

0	1	1	1	1	1
1					
1					
1					
1					
1					

$P$


# Dynamic Programming

## Longest common subsequence

- Example:  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

$L$

0	1	1	1	1	1
1	1				
1					
1					
1					
1					

$P$


# Dynamic Programming

## Longest common subsequence

- Example:  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

$L$

0	1	1	1	1	1
1	1	1			
1					
1					
1					
1					

$P$


# Dynamic Programming

## Longest common subsequence

- Example:  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

$L$

0	1	1	1	1	1
1	1	1	2		
1					
1					
1					
1					

$P$




# Dynamic Programming

## Longest common subsequence

- Example:  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

$L$

0	1	1	1	1	1
1	1	1	2	2	2
1					
1					
1					
1					

$P$


# Dynamic Programming

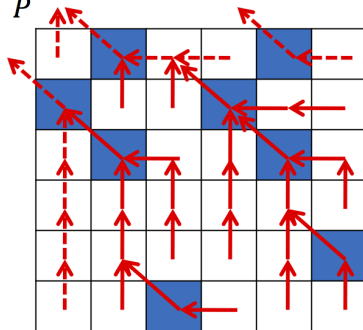
## Longest common subsequence

- Example:  $S = \text{XYXZPQ}$ ,  $T = \text{YXQYXP}$

$L$

0	1	1	1	1	1
1	1	1	2	2	2
1	2	2	2	3	3
1	2	2	2	3	3
1	2	2	2	3	4
1	2	3	3	3	4

$P$



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- Claim 1: If  $i = 0$  or  $j = 0$ , then  $L(i, j) = 0$ .
- Claim 2: If  $S[i] = T[j]$ , then  $L(i, j) = 1 + L(i - 1, j - 1)$ .
- Claim 3: If  $S[i] \neq T[j]$ , then  
$$L(i, j) = \max \{L(i - 1, j), L(i, j - 1)\}.$$

# Dynamic Programming

## Longest common subsequence

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- Claim 3: If  $S[i] \neq T[j]$ , then  $L(i, j) = \max \{L(i - 1, j), L(i, j - 1)\}$ .
- Here is a simple recursive program to find the length of the longest common subsequence.

### Algorithm

LCS-rec( $S, n, T, m$ )

- If ( $n = 0$  OR  $m = 0$ ) then return(0)
- If ( $S[n] = T[m]$ ) return( $1 + \text{LCS-rec}(S, n - 1, T, m - 1)$ )
- If ( $S[n] \neq T[m]$ )  
return( $\max \{ \text{LCS-rec}(S, n, T, m - 1), \text{LCS-rec}(S, n - 1, T, m) \}$ )

# Dynamic Programming

## Longest common subsequence

### Algorithm

$\text{LCS-rec}(S, n, T, m)$

- If  $(n = 0 \text{ OR } m = 0)$  then return(0)
- If  $(S[n] = T[m])$  return( $1 + \text{LCS-rec}(S, n - 1, T, m - 1)$ )
- If  $(S[n] \neq T[m])$   
return( $\max\{\text{LCS-rec}(S, n, T, m - 1), \text{LCS-rec}(S, n - 1, T, m)\}$ )

- What is the running time of this algorithm?

# Dynamic Programming

## Longest common subsequence

### Algorithm

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- If  $(n = 0 \text{ OR } m = 0)$  then return(0)
- If  $(S[n] = T[m])$  return  $(1 + \text{LCS-rec}(S, n - 1, T, m - 1))$
- If  $(S[n] \neq T[m])$   
return  $(\max\{\text{LCS-rec}(S, n, T, m - 1), \text{LCS-rec}(S, n - 1, T, m)\})$

- What is the running time of this algorithm?
  - This is exponentially large!

# Dynamic Programming

## Longest common subsequence

### Algorithm

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return  $(\max\{\text{LCS-rec}(S, n, T, m - 1), \text{LCS-rec}(S, n - 1, T, m)\})$

- Here is a *memoized* version of the above algorithm.

### Algorithm

$\text{LCS-mem}(S, n, T, m)$

- If  $(n = 0 \text{ OR } m = 0)$  then return(0)
- If  $(L[n, m]$  is known) then return  $(L[n, m])$
- If  $(S[n] = S[m])$   
-  $length \leftarrow 1 + \text{LCS-mem}(S, n - 1, T, m - 1)$
- If  $(S[n] \neq T[m])$   
-  $length \leftarrow \max\{\text{LCS-mem}(S, n, T, m - 1), \text{LCS-mem}(S, n - 1, T, m)\}$
- $L[n, m] \leftarrow length$
- return( $length$ )

# Dynamic Programming

## Longest common subsequence

- Here is a *memoized* version of the recursive algorithm.

### Algorithm

LCS-mem( $S, n, T, m$ )

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- If ( $L[n, m]$  is known) then return( $L[n, m]$ )
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- $L[n, m] \leftarrow length$
- return( $length$ )

- What is the running time of the above algorithm?  $O(nm)$

# Dynamic Programming

## 0-1 Knapsack

### Problem

You are given  $n$  items with non-negative integer weights  $w(i)$  and an integer  $W$ . You have to determine a subset  $S$  of  $\{1, \dots, n\}$  such that  $\sum_{i \in S} w(i)$  is maximized subject to  $\sum_{i \in S} w(i) \leq W$ .

- Example: Let  $(\{1, 2, 3, 5, 6, 7\}, 10)$  be the input instance.
- What is the optimal solution?

# Dynamic Programming

## 0-1 Knapsack

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- Example: Let  $([1, 2, 3, 5, 6, 7], 10)$  be the input instance.
- What is the optimal solution?  $\{2, 3, 4\}$ 
  - Since  $w(2) = 2, w(3) = 3, w(4) = 5$  and  $w(2) + w(3) + w(4) = 10$ .

# Dynamic Programming

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- How do we define the subproblems for the Dynamic Program?

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- How do we define the subproblems for the Dynamic Program?
- Let us try the following:
  - $M(i)$ : The maximum weight that can be filled using items  $\{1, \dots, i\}$  subject to the sum being  $\leq W$ .
  - How do we define  $M(i)$  in terms of  $M(1), \dots, M(i-1)$ ?

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  - How do we define  $M(i)$  in terms of  $M(1), \dots, M(i-1)$ ?
    - Case 1:  $i^{\text{th}}$  item is not in the optimal solution. Then  $M(i) = M(i-1)$ .
    - Case 2:  $i^{\text{th}}$  item is in the optimal solution. **There is a problem here.**

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## 0-1 Knapsack

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- How do we define the subproblems for the Dynamic Program?
- Let us try the following:
  - $M(i, w)$ : The maximum weight that can be filled using items  $\{1, \dots, i\}$  subject to the sum being  $\leq w$ .
  - Recursive formulation:
    - Case 1:  $i^{\text{th}}$  item is not in the optimal solution. Then  $M(i, w) = M(i - 1, w)$ .
    - Case 2:  $i^{\text{th}}$  item is in the optimal solution. Then  $M(i, w) = M(i - 1, w - w(i)) + w(i)$

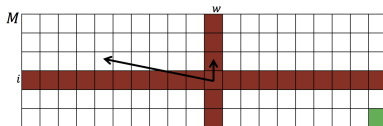
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- Dynamic Programming solution:
  - $M(i, w)$ : The maximum weight that can be filled using items  $\{1, \dots, i\}$  subject to the sum being  $\leq w$ .
  - If  $w(i) > w$ , then  $M(i, w) = M(i - 1, w)$
  - If  $w(i) \leq w$ , then
$$M(i, w) = \max \{M(i - 1, w), M(i - 1, w - w(i)) + w(i)\}$$
  - $\forall w \leq W, M(1, w) = w(1)$  if  $w(1) \leq w$  and 0 otherwise.



- What is the running time for filling the above table?



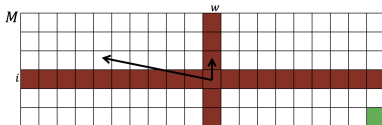
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  - $\forall w \leq W, M(1, w) = w(1)$  if  $w(1) \leq w$  and 0 otherwise.



- What is the running time for filling the above table?  $O(n \cdot W)$

End