

SHANGHAI TECH UNIVERSITY

CS240 Algorithm Design and Analysis
Spring 2024
Problem Set 2

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Due: 23:59, April 11, 2024

1. Submit your solutions to the course Gradescope.
2. If you want to submit a handwritten version, scan it clearly.
3. Late homeworks submitted within 24 hours of the due date will be marked down 25%. Homeworks submitted more than 24 hours after the due date will not be accepted unless there is a valid reason, such as a medical or family emergency.
4. You are required to follow ShanghaiTech's academic honesty policies. You are allowed to discuss problems with other students, but you must write up your solutions by yourselves. You are not allowed to copy materials from other students or from online or published resources. Violating academic honesty can result in serious penalties.

Problem 1:

Suppose you are given a set of intervals $I_1 = [s_1, t_1], \dots, I_n = [s_n, t_n]$, which can possibly overlap. Your task is to select a set of points p_1, \dots, p_m so that each interval intersects at least one of the points. Give an algorithm to minimize the number of selected points m , and prove that your algorithm is optimal.

Solution:

The time consumption is sorting takes $O(n \log n)$ time, and takes $O(n)$ time, so the total time complexity is $O(n \log n)$.

The space complexity is $O(n)$ for storing .

Problem 2:

Suppose you need to perform n tasks, and you can only perform one task at a time. It takes t_i time to perform the i 'th task, and the task has an importance of $w_i > 0$. Let C_i be the completion time of the i 'th task, i.e. the time when you finish performing the task. Find an ordering of the tasks to minimize their weighted completion time, defined as $\sum_{i=1}^n w_i C_i$. Your algorithm should work in $O(n \log n)$ time, and you need to analyze the algorithm's time complexity and prove its correctness.

Solution:

The time consumption is sorting takes $O(n \log n)$ time, and takes $O(n)$ time, so the total time complexity is $O(n \log n)$.

The space complexity is $O(n)$ for storing .

Problem 3:

A thief is planning to rob houses along a street. Each house has a certain amount of money stashed, and the thief can rob any set of houses, as long as he does not rob any adjacent houses. Determine the maximum amount of money the thief can steal.

Solution:

For loop in the pseudo-code is form '**for** $i \leftarrow 1$ **to** n ' represents $i = 1, 2, \dots, n$ sequentially.

Algorithm 1 Maximum stolen money

```
1: Input: Number of house  $n$ , each house's money number  $a_1, a_2, \dots, a_n$ .  
2: Output: The maximum number of money could steal  $ans$ .  
3:  $dp(1, 0) \leftarrow 0, dp(1, 1) \leftarrow a_1$   
4: for  $i \leftarrow 2$  to  $n$  do  
5:    $dp(i, 0) \leftarrow \max\{dp(i-1, 0), dp(i-1, 1)\}$   
6:    $dp(i, 1) \leftarrow dp(i-1, 0) + a_i$   
7: end for  
8:  $ans \leftarrow \max\{dp(n, 0), dp(n, 1)\}$   
9: return  $ans$ 
```

The time complexity is $O(n)$.

The space complexity is $O(n)$.

Where n is the number of houses.

Problem 4:

Given three sequences L_1, L_2 and L , design an efficient algorithm to check if L_1 and L_2 can be interleaved to produce L . For example, the sequences $L_1 = \mathbf{aabb}$ and $L_2 = \mathbf{cba}$ can be interleaved into sequence $L = \mathbf{acabbab}$, but L_1 and L_2 cannot be interleaved into sequence $L = \mathbf{aaabbbc}$. Analyze the time and space complexity of your algorithm as a function of the lengths of L_1 and L_2 .

Solution:

For loop in the pseudo-code is form '**for** $i \leftarrow 1$ **to** n ' represents $i = 1, 2, \dots, n$ sequentially.

The time complexity is $O(nm)$.

The space complexity is $O(nm)$.

Where n is the length of L_1 , m is the length of L_2 .

Algorithm 2 Interleaved the object sequence

```
1: Input: Two sequences  $L_1, L_2$  and  $L$ .
2: Output: Whether  $L_1$  and  $L_2$  can be interleaved to produce  $L$ .
3:  $dp(0, 0) \leftarrow 1$ 
4: for  $i \leftarrow 1$  to  $|L_1|$  do
5:   if  $L_1[i] == L[i]$  then
6:      $dp(i, 0) \leftarrow 1$ 
7:   else
8:      $dp(i, 0) \leftarrow 0$ 
9:   end if
10: end for
11: for  $j \leftarrow 1$  to  $|L_2|$  do
12:   if  $L_2[j] == L[j]$  then
13:      $dp(0, j) \leftarrow 1$ 
14:   else
15:      $dp(0, j) \leftarrow 0$ 
16:   end if
17: end for
18: for  $i \leftarrow 1$  to  $|L_1|$  do
19:   for  $j \leftarrow 1$  to  $|L_2|$  do
20:      $dp(i, j) \leftarrow 0$ 
21:     if  $L_1[i] == L[j]$  then
22:        $dp(i, j) \leftarrow dp(i, j) \vee dp(i - 1, j)$ 
23:     end if
24:     if  $L_2[j] == L[j]$  then
25:        $dp(i, j) \leftarrow dp(i, j) \vee dp(i, j - 1)$ 
26:     end if
27:   end for
28: end for
29: return  $dp(|L_1|, |L_2|)$ 
```

Problem 5:

Suppose you are given a propositional logic formula containing only the terms \wedge , \vee , T and F, without any parentheses. You want to find out how many different ways there are to correctly parenthesize the formula so that the resulting formula evaluates to true. For example, the formula $T \vee F \vee T \vee F$ can be correctly parenthesized in 5 ways:

$$\begin{aligned} &(T \vee (F \vee (T \vee F))) \\ &(T \vee ((F \vee T) \vee F)) \\ &((T \vee F) \vee (T \vee F)) \\ &(((T \vee F) \vee T) \vee F) \\ &((T \vee (F \vee T)) \vee F) \end{aligned}$$

Of these, 3 evaluate to true: $((T \vee F) \vee (T \vee F))$, $(T \vee ((F \vee T) \vee F))$ and $(T \vee (F \vee (T \vee F)))$.

Give a dynamic programming algorithm to solve this problem. Describe your algorithm, including a clear statement of your dynamic programming equation, show that it is correct, and prove its running time.

Solution:

For loop in the pseudo-code is form '**for** $i \leftarrow 1$ **to** n ' represents $i = 1, 2, \dots, n$ sequentially.

The time complexity is $O(n^3)$.

The space complexity is $O(n^2)$.

Where n is the number of propositions.

Algorithm 3 Number of True Statements

```
1: Input: Number of propositions  $n$ , propositions  $a_1, a_2, \dots, a_n$ , operators  
    $b_1, b_2, \dots, b_{n-1}$   
2: Output: The number of True Statements  
3: for  $i \leftarrow 1$  to  $n$  do  
4:    $dp(i, i, a_i) \leftarrow 1$   
5:    $dp(i, i, \neg a_i) \leftarrow 0$   
6: end for  
7: for  $i \leftarrow 1$  to  $n$  do  
8:   for  $j \leftarrow i + 1$  to  $n$  do  
9:     for  $k \leftarrow i$  to  $j - 1$  do  
10:      if  $b_i == \wedge$  then  
11:         $dp(i, j, 0) \leftarrow dp(i, k, 0) * dp(k + 1, j, 0)$   
           $+ dp(i, k, 0) * dp(k + 1, j, 1)$   
           $+ dp(i, k, 1) * dp(k + 1, j, 0)$   
12:         $dp(i, j, 1) \leftarrow dp(i, k, 1) * dp(k + 1, j, 1)$   
13:      else  
14:         $dp(i, j, 0) \leftarrow dp(i, k, 0) * dp(k + 1, j, 0)$   
15:         $dp(i, j, 1) \leftarrow dp(i, k, 1) * dp(k + 1, j, 1)$   
           $+ dp(i, k, 0) * dp(k + 1, j, 1)$   
           $+ dp(i, k, 1) * dp(k + 1, j, 0)$   
16:      end if  
17:    end for  
18:  end for  
19: end for  
20: return  $dp(1, n, 1)$ 
```

Problem 6:

Consider a weighted directed graph G with n vertices and m edges, where each edge (i, j) has a positive integer weight $w_{i,j}$. A walk is a sequence of not necessarily distinct vertices v_1, v_2, \dots, v_k , such that any two consecutive vertices v_i, v_{i+1} are connected by an edge. The length of the walk is the sum of the weights of the edges in the walk. Design an algorithm to find the number of different walks from a vertex s to another vertex t which have length exactly L , and analyze the time complexity of your algorithm.

Solution:

For loop in the pseudo-code is form ‘**for** $i \leftarrow 1$ **to** n ’ represents $i = 1, 2, \dots, n$ sequentially.

Algorithm 4 Number of different walks

```
1: Input: Graph  $G = (V, E)$  with  $n$  vertices and  $m$  edges with weight  $w_{i,j}$ 
2: Output: The number of different walks from  $s$  to  $t$  with length  $L$ 
3:  $dp(s, 0) \leftarrow 1$ 
4:  $dp(x, t) \leftarrow 0, \forall (x, t) \neq (s, 0)$ 
5: for  $val \leftarrow 1$  to  $L$  do
6:   for  $(i, j) \in E$  do
7:      $dp(j, val) \leftarrow dp(j, val) + dp(i, val - w_{i,j})$ 
8:   end for
9: end for
10: return  $dp(t, L)$ 
```

The time complexity is $O(mL)$.

The space complexity is $O(nL)$.

Where n is the number of vertices, m is the number of edges, L is the length of the walk.