

University of Waterloo

CS341 - Winter 2016

Assignment 4

Due Date: Thursday Mar 17 at 11:59pm

Problem 1 Longest Common Subsequence

Answer: a4q1.cpp

Problem 2 Facility Location

Answer:

We can get the subproblems: each position p_i for $1 \leq i \leq m$, each people a_j for $1 \leq j \leq n$, the number of hospital left is h for $1 \leq h \leq k$, and the previous(left) hospital position is L (if $h = k$, $L = 0$). And there are $m * k$ subproblems.

The function $dp(p,a,h,L)$:

- p : current position
- a : current people
- h : the number of available hospitals
- L : the previous position

We should consider following 2 possibilities:

1. Put the hospital in this position p_i : In this case, we should update the distance $dist$, $\sum_{a_j}^{until a_{j+s} > p_i} \min(a_j - L, p_i - a_j) + dp(p_{i+1}, a_{j+s}, h - 1, p_i)$.
2. Don't put the hospital in this position p_i : In this case, we don't need to update $dist$, $dp(p_{i+1}, a_j, h, L)$.

Recurrence:

We know that the base case is when $k = m$, at this time, we don't need to consider k as a variable, just compute all position. For each step, we will update the total distance with a list of minimum distance between a and p if the hospital is putted in this position; not update otherwise. So, finally we can surely get the minimum total distance to the people.

Algorithm:

1. $dp(p,a,h,L)$: (output a number)
2. if $h = 0$ then
3. return $(\text{sum}(a_j - L) \text{ until } a_{(j+s)} = a_n)$

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4.   else if put a hospital in p_i then
5.       if L = 0 then
6.           return (sum(p_i - a_j) until a_(j+s) > p_i) + dp(p_(i+1),a_(j+s),h-1,p_i)
7.       else then
8.           return (sum(min(a_j - L , p_i - a_j)) until a_(j+s) > p_i)
              + dp(p_(i+1),a_(j+s),h-1,p_i)
9.   else then (not put hospital)
10.      return dp(p_(i+1),a_j,h,L)

```

Time complexity:

Generally, each subproblem takes $\frac{n}{m}$ operations and we got $m * k$ subproblems.

So, it costs $O(m * k * \frac{n}{m}) = O(nk)$.

Problem 3 Coin Game

Answer:

We can get the subproblems: in a turn, for $1 \leq i < j \leq n$, let v_i is the leftmost coin and v_j is the rightmost coin.

We should consider following 2 possibilities:

1. Player 1 chooses the i-th coin(leftmost) with value v_i , then player 2 can either chooses (i+1)-th coin or j-th coin. Player 2 will choose the coin which can make player 1 with minimum value.

So, player 1 can collect values: $v_i + \min(dp[i+2][j], dp[i+1][j-1])$.

2. Player 1 chooses the j-th coin(rightmost) with value v_j , then player 2 can either chooses i-th coin or (j-1)-th coin. Player 2 will choose the coin which can make player 1 with minimum value.

So, player 2 can collect values: $v_j + \min(dp[i+1][j-1], dp[i][j-2])$.

Recurrence:

We know that $dp[i][j]$ is the maximum value that player 1 can earn from the i-th coin to the j-th coin. So,

$dp[i][j] = \max(v_i + \min(dp[i+2][j], dp[i+1][j-1]), v_j + \min(dp[i+1][j-1], dp[i][j-2]))$

And the base case should be:

1. $dp[i][j] = \max(v_i, v_j)$ when $j = i + 1$.
2. $dp[i][j] = v_i$ when $j = i$.

For each step, we just need to calculate the dp value for the certain length of v and add up until length equals n . So, we will use dynamic programming to avoid the overlapping

subproblems; otherwise, $dp[i+1][j-1]$ would be calculated twice.

Algorithm:

```
1. Given a list of values,  $v[1..n]$ ; and the number of coins,  $n$ 
2. Output the maximum total value player 1 earns
3. Find_Max( $v[1..n], n$ ): (below, we see  $v[1..n]$  as  $v[0..n]$  as array)
4.    $dp[n][n]$ , set each value 0
5.   for  $i$  from 0 up to  $(n - 1)$  do
6.      $dp[i][i] \leftarrow v[i]$ 
7.   end loop
8.   for  $len$  from 2 to  $n$  do
9.     for  $i$  from 0 to  $(n - len)$  do
10.       $j \leftarrow i + len - 1$ 
11.      if  $len$  equals to 2, then
12.         $dp[i][j] \leftarrow \max(v[i], v[j])$ 
13.      else, then
14.         $dp[i][j] = \max(v[i] + \min(dp[i+2][j], dp[i+1][j-1]),$ 
            $v[j] + \min(dp[i+1][j-1], dp[i][j-2]))$ 
15.    end loop
16.  end loop
17.  return  $dp[0][n-1]$ 
```

Time complexity:

Obviously, we do $(n - 1 + n - 2 + \dots + 1 + 0)$ operations, so the time cost: $O(\frac{(n-1)n}{2}) = O(n^2)$.

Problem 4 Covering Set

Answer:

As we know, the input is a tree $T = (V, E)$ in the adjacency list representation, a weight w_v for each $v \in V$. In other words, we input a list of weight numbers to represent the nodes.

We can get the subproblems: for $v \in V$, size of smallest covering set in subtree rooted at v . And there are n subproblems, where n is $|V|$.

We should consider following 2 possibilities for root and recursively for all nodes down the root.

1. Root is in the Covering Set: In this case, root covers all children edges. And left with children subtrees.
2. Root is not in the Covering Set: In this case, all children must be in cover otherwise the edges adjacent to v will not be covered. And left with grandchildren subtrees.

Recurrence:

From above, the case 1 we can get: $sol1 \leftarrow 1 + \sum(dp(c) \text{ for } c \text{ in } v.children)$

the case 2 we can get: $sol2 \leftarrow (num \text{ of } v.children) + \sum(dp(g) \text{ for } g \text{ in } v.grandchildren)$

So, $dp(v) = \min(sol1, sol2)$. And the solution for this question should be $dp(root)$.

Algorithm:

```
1. Given a list of nodes, set the size of cover set vc, vc <- 0
2. Output a covering set cs
3. dp(*root): (pointer of root, containing the whole tree)
4.   if (root is NULL or there is only one node), then
5.     return 0
6.   if (root->vc is not 0), then
7.     return root->vc
8.   sol1 <- 1 + sum(dp(c) for c in v.children)
9.   sol2 <- (num of v.children) + sum(dp(g) for g in v.grandchildren)
10.  root->vc = min(sol1, sol2)
11.  if (sol1 < sol2), then
12.    put this root into the cs
13.  return root->vc
14. end of function dp
15. Finally, we can get the Covering Set, cs
```

Time complexity:

Because the edges going out of node v are visited at most twice by the recursion: one for computing parent, and one for grandparent.

Therefore, actually the time costs: $O(\text{num of subproblems}) = O(n)$.

Problem 5 Trading

Answer:

For this question, we can use Bellman-Ford algorithm to do all-pair shortest path to do loops by n items $a_1..a_n$.

We get the subproblem: $D[i, j]$ is the maximum amount of item a_j which can be traded by a_i . And $P[i, j]$ is the previous item of a_j starting trading from item a_i .

Recurrence: (base on the correctness of Bellman-Ford Algorithm)

For the loop, we get the value

$D[i, m] = \max(\max(D[i, k] * T[k, m]) \text{ for all } k \text{ that } 1 < k \leq n, D[i, m])$.

In the loop, when it is in the i iteration:

- If there exists a trading path from a_i to a_j , $D[i, j]$ is at most the product of i weights.
- If $D[i, j] \neq 0$, $D[i, j]$ is $D[i, k] * T[k, m]$.

The base case is: $i = 1$, the $D[s, s] = 1$ (s represents source), and for all other $D[i, j]$ should be 0 because there are no path through them yet.

For the inductive case, an item is updated by $D[i, m] = D[i, k] * T[k, m]$, where $D[i, k]$ is the path from a_i to a_k and $D[i, m]$ is the path from a_i to a_m . If every edge with less-than-1 trading rate, then each item is visited at least once. So if there are no improvement (times a larger-than-1 rate) in n-th loop, then for any cycle $D[i, i]..D[i, k - 1]$, the final rate product should be less than 1. In addition, if there is a cycle with the product larger than 1, then in the n-th loop, we would get a path from a_i to a_j starting with a_i , $D[l, i] * T[i, j] > D[l, j]$, which means that the value of trading path is improved for every time doing loop.

Algorithm:

```

1.Trading:
2.   for i from 1 to n do
3.       for every j (1 <= j <= n) do
4.           D[i,j] <- 0
5.           P[i,j] <- nil
6.       D[i,i] <- 1
7.       for i from 1 to n-1 do
8.           for k from 1 to n do
9.               for m from 1 to n do
10.                  if D[i,k]*T[k,m] > D[i,m] then
11.                      D[i,m] <- D[i,k]*T[k,m]
12.                      P[i,m] <- k
13.       for k from 1 to n do
14.           for m from 1 to n do
15.               if D[i,k]*T[k,m] > D[i,m] then
16.                   return true
17.   return false

```

Time complexity:

Bellman-Ford Algorithm costs $O(|V||E|) = O(n * n^2) = O(n^3)$. And we run it for n times, so the final time should be $O(n^3 * n) = O(n^4)$.