

EECS 336

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Homework 4

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Problem 1

Solution:

I think the greedy algorithm discussed in the paper is incorrect. The algorithm proposed in the paper always picks the symbol that comes earlier in string Y .

The paper claims that it leaves as much opportunity as possible for the remaining symbols to be select. However, we can easily find counter example to prove it wrong. For example, consider these two input array:

$$X = [A, B, B, B]$$

$$Y = [B, A, B, B]$$

In the first iteration, according to the algorithm presented in the paper. $P1 = 2, P2 = 1, P1 > P2$, we choose $R = P2 = 1, L = \{P2\}, i = i + 1$. And we won't take A into consideration anymore. It is clearly not correct, because we can easily know the longest common subsequence should be ABB . This algorithm discards A too early and causes the incorrect answer.

Problem 2

Solution:

Let $C_1(T)$ to be the cost function mentioned in class and $C_2(T)$ to be the cost function presented in textbook.

$$C_1(T) = \sum_{i=1}^n P_i \times level(a_i) + \sum_{i=0}^n Q_i \times (level(e_i) - 1)$$
$$C_2(T) = \sum_{i=1}^n P_i \times level(a_i) + \sum_{i=0}^n Q_i \times level(e_i)$$

We can easily see that

$$C_2(T) - C_1(T) = \sum_{n=i}^j Q_n$$

During the construction of the optimal binary tree, we will use the following formula:

$$C(i, j) = \min_{i \leq n \leq j} \{C(i, k-1) + C(k+1, j) + \sum_{m=i}^j P_m + \sum_{m=i-1}^j Q_m\}$$

Assume we find the optimal solution $k' \in [i, j]$ by using $C_2(T)$, then we can say that for any $k \in [i, j]$:

$$\begin{aligned} C_2(i, k-1) + C_2(k+1, j) &\geq C_2(i, k'-1) + C_2(k'+1, j) \\ \Leftrightarrow \\ C_1(i, k-1) + \sum_{n=i}^{k-1} Q_n + C_1(k+1, j) + \sum_{n=k+1}^j Q_n &\geq C_1(i, k'-1) + \sum_{n=i}^{k'-1} Q_n + C_1(k'+1, j) + \sum_{n=k'+1}^j Q_n \\ \Leftrightarrow \\ C_1(i, k-1) + C_1(k+1, j) + \sum_{n=i}^j Q_n &\geq C_1(i, k'-1) + C_1(k'+1, j) + \sum_{n=i}^j Q_n \\ \Leftrightarrow \\ C_1(i, k-1) + C_1(k+1, j) &\geq C_1(i, k'-1) + C_1(k'+1, j) \end{aligned}$$

We can see that we will choose the same k' to be our optimal solution using $C_1(T)$. Hence, these two cost model will produce the same optimal binary tree.

Problem 3

The Shortest Maximum-Weight Common Subsequence Problem:

Solution:

Table:

- $W[i, j]$ = the maximum-weight of a common subsequence of the prefix x_1, x_2, \dots, x_i and y_1, y_2, \dots, y_j
- $L[i, j]$ = the shortest length of a maximum-weight common subsequence of the prefix x_1, x_2, \dots, x_i and y_1, y_2, \dots, y_j

$$0 \leq i \leq m$$

$$0 \leq j \leq n$$

Output for the problem:

The shortest maximum-weight of a common subsequence.

Base Case:

- $i = 0, W[i, j] = 0, L[i, j] = 0$
- $j = 0, W[i, j] = 0, L[i, j] = 0$

Recursion Case:

- $1 \leq i \leq m$ and $1 \leq j \leq n$
- If $x_i = y_j$ then $W[i, j] = \text{weight}(x_i) + W[i - 1, j - 1]$, $L[i, j] = L[i - 1, j - 1] + 1$
- If $x_i \neq y_j$, $W[i - 1, j] > W[i, j - 1]$ then $W[i, j] = W[i - 1, j]$, $L[i, j] = L[i - 1, j]$
- If $x_i \neq y_j$, $W[i, j - 1] > W[i - 1, j]$ then $W[i, j] = W[i, j - 1]$, $L[i, j] = L[i, j - 1]$
- If $x_i \neq y_j$, $W[i, j - 1] = W[i - 1, j]$ then $W[i, j] = W[i, j - 1] = W[i - 1, j]$, $L[i, j] = \min\{L[i, j - 1], L[i - 1, j]\}$

Extra information:

The prefix x_i chosen as the elements of the result. Correctness:

Table $W[i, j]$ represent the maximum weight of the common subsequence of prefix x_1, x_2, \dots, x_i and y_1, y_2, \dots, y_j . If the $x_i = y_j$, then we must append x_i to the result common subsequence S . If we don't append x_i , and we get a result S' . We must can find a common subsequence S with weight is larger than weight of S' . In case of $x_i \neq y_j$, we should let $W[i, j] = \max\{W[i, j - 1], W[i - 1, j]\}$. If we don't choose the maximum weight, then we cannot meet the goal finding the "maximum" weight. If $x_i \neq y_j$ & $W[i, j - 1] = W[i - 1, j]$, In order to find the shortest maximum weight common subsequence, we need to take the length of the result into consideration. In order to achieve the result with minimum length, clearly we should choose $L[i, j] = \min\{L[i, j - 1], L[i - 1, j]\}$.

Space complexity:

- The table $W[i, j]$ has $m \times n$ entries, so the space complexity is $\mathcal{O}(mn)$
- The table $L[i, j]$ has $m \times n$ entries, so the space complexity is $\mathcal{O}(mn)$

Time complexity:

- The runtime of computing each entry of $W[i, j]$ and $L[i, j]$ is $\mathcal{O}(1)$
- The total runtime of this algorithm is $\mathcal{O}(mn)$

The Minimum-Weight Longest Common Subsequence Problem:

Solution:

Table:

- $W[i, j]$ = the minimum-weight of a common subsequence of the prefix x_1, x_2, \dots, x_i and y_1, y_2, \dots, y_j
- $L[i, j]$ = the longest length of a maximum-weight common subsequence of the prefix x_1, x_2, \dots, x_i and y_1, y_2, \dots, y_j

$$0 \leq i \leq m$$

$$0 \leq j \leq n$$

Output for the problem:

The longest common subsequence with minimum-weight

Base Case:

- $i = 0, W[i, j] = 0, L[i, j] = 0$
- $j = 0, W[i, j] = 0, L[i, j] = 0$

Recursion Case:

- $1 \leq i \leq m$ and $1 \leq j \leq n$
- If $x_i = y_j$ then $L[i, j] = L[i - 1, j - 1] + 1, W[i, j] = W[i - 1, j - 1] + \text{weight}(x_i)$
- If $x_i \neq y_j$, $L[i - 1, j] > L[i, j - 1]$ then $L[i, j] = L[i - 1, j], W[i, j] = W[i - 1, j]$
- If $x_i \neq y_j$, $L[i, j - 1] > L[i - 1, j]$ then $L[i, j] = L[i, j - 1], W[i, j] = W[i, j - 1]$
- If $x_i \neq y_j$, $L[i, j - 1] = L[i - 1, j]$ then $L[i, j] = L[i, j - 1] = L[i - 1, j], W[i, j] = \min\{W[i, j - 1], W[i - 1, j]\}$

Extra information:

The prefix x_i chosen as the elements of the result. Correctness:

Table $L[i, j]$ represent the maximum length of the common subsequence of prefix x_1, x_2, \dots, x_i and y_1, y_2, \dots, y_j . If the $x_i = y_j$, then we must append x_i to the result common subsequence S . If we don't append x_i , and we get a result S' . We must can find a common subsequence S with length is larger than length of S' . In case of $x_i \neq y_j$, we should let $L[i, j] = \max\{L[i, j - 1], L[i - 1, j]\}$. If we don't choose the maximum length, then we cannot meet the goal finding the "maximum" length. If $x_i \neq y_j$ & $L[i, j - 1] = L[i - 1, j]$, In order to find the longest minimum weight common subsequence, we need to take the weight of the result into consideration. In order to achieve the result with minimum length, clearly we should choose $W[i, j] = \min\{W[i, j - 1], W[i - 1, j]\}$.

Space complexity:

- The table $W[i, j]$ has $m \times n$ entries, so the space complexity is $\mathcal{O}(mn)$
- The table $L[i, j]$ has $m \times n$ entries, so the space complexity is $\mathcal{O}(mn)$

Time complexity:

- The runtime of computing each entry of $W[i, j]$ and $L[i, j]$ is $\mathcal{O}(1)$
- The total runtime of this algorithm is $\mathcal{O}(mn)$

Problem 4

Solution:

Table:

- $Total[i]$ = the minimum total cost for traveling from post 1 to post i .
- $1 \leq i \leq n$

Output for the problem:

- $Total[n]$

Base Case:

- $i = 1$. Then $Total[i] = 0$
- $i = 2$ Then $Total[i] = fee[1, 2]$

Recursion Case:

- $i < j - 1$
- $Total[i] = Total[k] + fee[k, i]$ minimized over $k = 2, \dots, i - 1$

Extra information:

The optimal index i to return the path with minimum cost. Algorithm:

Data: $1, n, fee[i, j]$

Result: The minimum total cost

if $n = 2$ **then**

 | $return A[2] = fee[1, 2]$;

end

$A[1] = 0$;

$i = 2$;

while $i \leq n$ **do**

 | **while** $j < i$ **do**

 | $min = \min\{min, A[j] + fee[j, i]\}$;

 | **end**

 | $A[i] = min$;

 | $i++$;

end

$return A[n]$;

Correctness:

The recurrence in this algorithm can be justified by considering an optimal sequence of renting at some post $n > 1$. This sequence must contain a last rental from some other post i to post n . The cost of this optimal sequence must be the cost of the optimal sequence ends at post i , plus the cost of renting from i to n . Since all cases for this last rental are considered, the recurrence correctly computes the costs of optimal rental sequence.

Space complexity:

- Because we have n entries, so the space complexity is $\mathcal{O}(n)$

Time complexity:

- The runtime of computing each entry is $\mathcal{O}(n)$
- The total runtime of this algorithm is $\mathcal{O}(n^2)$

Problem 5

Solution:

1) We first sort all points according to their x coordinate in increasing order, we use $1, \dots, n$ as their index. The point with index 1 is the starting point and point with index n is one of the three turning point.

2) In a 4-tonic tour, we should have one starting point and three turning points. Because we can know that the starting point must be point with index 1 and one of the turning point must be the point with index n. So we choose two other turning points among points from 2 to n-1 randomly. Denote the chosen points as a and b, And assume that $x_a < x_b$. Now we can have four paths in the 4-tonic tour: $p1 : 1 \rightarrow n, p2 : a \rightarrow n, p3 : a \rightarrow b, p4 : 1 \rightarrow b$

3) Create a 4-dimensional table $L[i, j, p, q]$ with size n in each dimension and initialize every entry with positive infinity. In this table, i, j, p, q are the farthest points on path p1, p4, p2, p3 respectively. $L[i, j, p, q]$ is the current minimum of length sum of sub path from 1 to i on p1, sub path from 1 to j on p4, sub path from a to p on p2 and the sub path from a to q on p3.

4) We use the following rules to update the table:

Denote the next point(excluding 1 and a) to put into one of P1, P2, P3 and P4 as k(index). k starts at 2 and ends at n. Denote $d(c, d)$ as distance between point c and d.

Base Case: $L[1, 1, a, a] = 0$

Recursion Case:

Case 1: if $1 < k < a$,

$$k = 1 + \max\{i, j\}$$

$$L[k, j, a, a] = \min\{L[i, j, a, a] + d(k, j), L[k, j, a, a]\}$$

$$L[i, k, a, a] = \min\{L[i, j, a, a] + d(k, i), L[i, k, a, a]\}$$

Case 2: if $a < k < b$,

$$k = 1 + \max\{i, j, p, q\}$$

$$L[k, j, p, q] = \min\{L[i, j, p, q] + d(k, i), L[k, j, p, q]\}$$

$$L[i, k, p, q] = \min\{L[i, j, p, q] + d(k, j), L[i, k, p, q]\}$$

$$L[i, j, k, q] = \min\{L[i, j, p, q] + d(p, k), L[i, j, k, q]\}$$

$$L[i, j, p, k] = \min\{L[i, j, p, q] + d(q, k), L[i, j, p, k]\}$$

Case 3: if $k = b$,

$$L[i, b, p, b] = \min\{L[i, j, p, q] + d(b, j) + d(b, q), L[i, j, p, k]\}$$

Case 4: if $b < k < n$,

$$k = 1 + \max\{i, p\}$$

$$L[k, b, p, b] = \min\{L[i, b, p, b] + d(i, k)\}$$

$$L[i, b, k, b] = \min\{L[i, b, p, b] + d(k, p)\}$$

Case 5: if $k = n$,

$$L[n, b, n, b] = \min\{L[i, b, p, b] + d(n, i) + d(p, n)\}$$

Correctness:

Inductive hypothesis:

The algorithm would offer the optimal 4 tonic tour of points set 1 to n , for $n < n'$, for some n' . we can inductively prove the algorithm from n' .

Inductive case:

Base case:

We have only two points and need to add k to the path. If k is less than a , k could only be on $p1$ or $p4$, since the second path and the third path start from a . Therefore, if k is on the first path, by trying all different i , we can get the minimum value of $L[k, j, a, a]$. We can apply this approach to the last path and get the minimum value of the last path.

If $a < k < b$, we need to take four scenarios into consideration when we try to put k . If k is on $p1$, by trying all different i , we can get the minimum value of $L[k, j, p, q]$. If k is on $p4$, by trying all different j , we can get the minimum value of $L[i, k, p, q]$. If k is on $p2$, by checking all different p , we can get the minimum value of $L[i, j, k, q]$. If k is on $p3$, by checking all different q , we can get the minimum value of $L[i, j, p, k]$.

If k equals to b , k must be the turning point between $p3$ and $p4$. In this case, we need to check all different j and we can get the minimum value of $L[i, b, p, q]$.

In case of $b < k < n$, k should be on $p1$ or $p2$. If it is on $p1$, by checking all different i , we can get the the minimum value of $L[k, b, p, b]$. If it is on $p2$, by trying all different p , we can get the minimum value of $L[i, b, k, b]$.

Finally, when k is the last element to add, it should be the rightmost point of $p1$ and $p2$. Therefore, by trying all i , we can get the minimum path of given a, b . After trying all a, b pairs, we can get the optimal 4 tonic-tour.

Time complexity: The table size is n^4 , for each entry in the table, we should try all different i , the runtime is $\mathcal{O}(n)$. We also need to try all different a, b , there are $\frac{(n-3)(n-2)}{2}$ choices, therefore, the time complexity should be $\mathcal{O}(n^7)$.