

# Algorithm Design and Analysis (Fall 2023)

## Assignment 4

**Deadline: Dec 26, 2023**

1. (30 points) Consider that you are in a stock market and you would like to maximize your profit. Suppose the prices of the stock for the  $n$  days,  $p_1, p_2, \dots, p_n$ , are given to you. On the  $i$ -th day, you are allowed to do exactly one of the following operations:

- Buy one unit of the stock and pay the price  $p_i$ . Your stock will increase by 1.
- Sell one unit of stock and get the reward  $p_i$  if your stock is at least 1. Your stock will decrease by one.
- Do nothing.

Design an  $O(n^2)$  time dynamic programming algorithm.

**Remark:** [Not for credits] There exists a clever greedy algorithm that runs in  $O(n \log n)$  time. Can you figure it out?

**Algorithm:**

1. Initialize an array  $DP[n+1][n+1]$  with  $DP[0][0] = 0$ , and other  $DP[i][j] = -INF$ .
2. Repeatedly update  $DP$  with  $DP[i][j] = \max\{DP[i-1][j], DP[i-1][j-1] - p_i, DP[i-1][j+1] + p_i\}$
3. After all updates are done, the maximum profit is  $\max\{DP[n][0], DP[n][1], \dots, DP[n][n]\}$

**Time Complexity**

Updating the array takes  $n^2$  rounds and each round takes  $O(1)$ . Choosing the final answer takes  $O(n)$ . Therefore, the time complexity is  $O(n^2)$ .

**Correctness**

Base Step: When  $i = 1$ , the only choice is either buy or do nothing. So  $DP[1][0] = 0, DP[1][1] = -p_1$ , which corresponds to the algorithm.

Inductive step: Suppose  $DP[x][y]$  is correct for all  $x \leq i$ . Then, consider calculating  $DP[i+1][j]$ , we traverse the three choices to select one with maximal profit.  $DP[i][j]$  means doing nothing,  $DP[i][j-1] - p_{i+1}$  means buying one and  $DP[i][j+1] + p_{i+1}$  means selling one.  $DP[i+1][j]$  means that you have  $j$  stocks in  $i+1$ -th day. If  $DP[i][j]$  is  $INF$ , it is impossible to have  $j$  stocks in  $i$ -th day. So, by applying the algorithm on  $DP[i+1][j]$  for all  $j$  has been traversed, it should give largest profit on  $i+1$ -th day with all possible stocks. Therefore, we prove the correctness.

2. (30 points) Given two strings  $x = x_1x_2\cdots x_n$  and  $y = y_1y_2\cdots y_n$ , we wish to find the length of their *longest common subsequence*, that is, the largest  $k$  for which there are indices  $i_1 < i_2 < \cdots < i_k$  and  $j_1 < j_2 < \cdots < j_k$  with  $x_{i_1}x_{i_2}\cdots x_{i_k} = y_{j_1}y_{j_2}\cdots y_{j_k}$ . Design an  $O(n^2)$  dynamic programming algorithm for this problem.

### Algorithm

1. Initialize an array  $DP[n+1][n+1]$  with  $DP[0][j] = DP[i][0] = 0$  for  $i, j = 0, 1, \dots, n+1$  and other  $DP[i][j] = -INF$ .
2. Repeatedly update  $DP$  with  $DP[i][j] = \begin{cases} DP[i-1][j-1] + 1 & \text{if } x_i = y_j \\ \max\{DP[i-1][j], DP[i][j-1]\} & \text{if } x_i \neq y_j \end{cases}$
3. Finally, the answer is  $DP[n][n]$ .

### Time Complexity

Updating the array takes  $n^2$  rounds and each round takes  $O(1)$ . Therefore, the time complexity is  $O(n^2)$ .

### Correctness

Base Step: When  $i = j = 1$ , we can only judge the length by  $x_1$  and  $y_1$ , which corresponds to  $DP[1][1]$  by the algorithm.

Inductive step: Suppose  $DP[x][y]$  is correct for all  $x < i$  and  $y < j$ . We need to prove that  $DP[i][j]$  is also correct.

If  $x_i = y_j$ , then the length will increase by 1 since we could let  $x_i = x_{i_{DP[i-1][j-1]+1}}$  and  $y_j = y_{j_{DP[i-1][j-1]+1}}$ .

If  $x_i \neq y_j$ , there is at most one element between  $x_i$  and  $y_j$  could increase the length because if both  $x_i$  and  $y_j$  increase the length, then  $x_i = y_j$ . So we find the larger one between  $DP[i-1][j]$  and  $DP[i][j-1]$ , which means either  $y_j$  increases the length or  $x_i$  increases the length. If  $DP[i-1][j] = DP[i][j-1]$ , both  $x_i$  and  $y_j$  won't increase the length. Since  $DP[x][y]$  is correct for all  $x < i$  and  $y < j$ ,  $DP[i][j]$  is also correct. Therefore, we prove the correctness of the algorithm.

3. (40 points) In the *Traveling Salesman Problem* (TSP), we are given an undirected weighted complete graph  $G = (V, E, w)$  (where  $(i, j) \in E$  for any  $i \neq j \in V$ ). The objective is to find a cycle of length  $|V|$  with minimum total weight, i.e., to find a tour that visit each vertex exactly once such that the total distance traveled in the tour is minimized. Obviously, the naïve exhaustive search algorithm requires  $O((n-1)!)$  time. In this question, you are to design a dynamic programming algorithm for the TSP problem with time complexity  $O(n^2 \cdot 2^n)$ .

(a) (10 points) Show that  $n^2 \cdot 2^n = o((n-1)!)$ , so that the above-mentioned algorithm is indeed faster than the naïve exhaustive search algorithm.

(b) (30 points) Design this algorithm. Hint: label all vertices as  $1, 2, \dots, n$ ; given  $i \in V$  and  $S \subseteq V \setminus \{1, i\}$ , let  $d(S, i)$  be the length of the shortest path from 1 to  $i$  where the intermediate vertices are *exactly* those in  $S$ ; show that the minimum weight cycle/tour is  $\min_{i=2,3,\dots,n} \{d(V \setminus \{1, i\}, i) + w(i, 1)\}$ .

(a) To show that  $n^2 \cdot 2^n = o((n-1)!)$ , we just need to show that  $\lim_{n \rightarrow \infty} \frac{n^2 \cdot 2^n}{(n-1)!} = 0$ .

Consider  $\sum_{n=1}^{\infty} \frac{n^2 \cdot 2^n}{(n-1)!}$ . By ratio test,  $\lim_{n \rightarrow \infty} \frac{\frac{(n+1)^2 \cdot 2^{(n+1)}}{(n+1)!}}{\frac{n^2 \cdot 2^n}{(n-1)!}} = \lim_{n \rightarrow \infty} \frac{(n+1)^2 \cdot 2}{n^3} = 0 < 1$ , so the series converges. Therefore, by the nature of convergence,  $\lim_{n \rightarrow \infty} \frac{n^2 \cdot 2^n}{(n-1)!} = 0$ , which means  $n^2 \cdot 2^n = o((n-1)!)$ .

**(b) Algorithm**

1. Initialize an array  $DP[n+1][2^{n-1}]$  where  $DP[i][S_j]$  means the length of the shortest path from 1 to  $i$  where the intermediate vertices are exactly those in  $S_j$  ( $1, i \notin S_j$ ).  $S_0, S_1, \dots, S_{2^{n-1}-1}$  are all combinations of all vertices (1 is not in any of  $S$ ). Let  $DP[i][\{\}] = w(1, i)$  for all  $i \in V \setminus \{1\}$  and other  $DP[i][S_j] = INF$ .

2. Repeatedly update  $DP[i][S_j] = \min\{DP[k][S_j \setminus \{k\}] + w(k, i)\}$  for all  $k \in S_j$ .

3. After all the updates are done, the final answer is  $\min_{i=2,3,\dots,n} \{DP[i][V \setminus \{1, i\}] + w(i, 1)\}$ .

**Time Complexity**

The algorithm takes  $n \cdot 2^{n-1}$  rounds, and each round takes  $O(n)$ . To find the optimal answer takes  $O(n)$ , so the overall time complexity is  $O(n^2 \cdot 2^n)$ .

**Correctness**

Base step:  $DP[i][\{\}]$  means the direct distance between  $i$  and 1  $w(1, i)$ , which corresponds to the algorithm.

Inductive step: Suppose all  $DP[i][S_j]$  are correct, where  $S_j$  means all  $S$  with  $|S| < k$ . We need to show that  $DP[i][S_{j+1}]$  is correct for all  $|S_{j+1}| = k$ . By the algorithm,  $DP[i][S_{j+1}] = \min\{DP[x][S_{j+1} \setminus \{x\}] + w(x, i)\}$  for all  $x \in S_{j+1}$ .  $DP[x][S_{j+1} \setminus \{x\}] + w(x, i)$  means the shortest path from 1 to  $x$  passing  $S_{j+1} \setminus \{x\}$  adding the length of  $(x, i)$ . And they make up a set of all subminimum paths from 1 to  $i$ . The minimum of

them means the shortest path from 1 to  $i$  passing  $S_{j+1}$  because all  $DP[i][S_j]$  is correct. Therefore, the algorithm is correct.

4. How long does it take you to finish the assignment (including thinking and discussion)? Give a score (1,2,3,4,5) to the difficulty. Do you have any collaborators? Please write down their names here.

1 days.

4, 4, 5.

No.