

# 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?

### Solution:

Define the sub-problem  $f(i, j)$  being the maximum of the profit we have earned after  $i$  days with  $j$  stocks left. For the first operation,  $f(i, j) = f(i - 1, j - 1) - p_i$  when  $j > 0$ . For the second operation,  $f(i, j) = f(i - 1, j + 1) + p_i$ . For the third operation,  $f(i, j) = f(i - 1, j)$ . Therefore, we have the following recurrence relation:

$$f(i, j) = \begin{cases} \max\{f(i - 1, j - 1) - p_i, f(i - 1, j + 1) + p_i, f(i - 1, j)\}, & j > 0, \\ \max\{f(i - 1, j + 1) + p_i, f(i - 1, j)\}, & \text{otherwise.} \end{cases}$$

However,  $f(i, j)$  is invalid when  $j > i$  since we cannot have more than  $i$  stocks after  $i$  days even if we buy one unit of the stock every day. For initialization, we can set  $f(0, 0) = 0$  and  $f(i, j) = -\infty$  where  $j > i$ .

The algorithm is below.

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### Algorithm 1 Maximize the Profit

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**Input:** the prices of the stock for  $n$  days,  $p_1, p_2, \dots, p_n$

- 1: Initialize  $f[0][0] = 0$  and  $f[i][j] = -\infty$  for  $i < j$
  - 2: **for**  $i = 1, \dots, n$  **do**
  - 3:      $f[i][0] = \max\{f[i - 1][1] + p_i, f[i - 1][0]\}$
  - 4:     **for**  $j = 1, \dots, i$  **do**
  - 5:          $f[i][j] = \max\{f[i - 1][j - 1] - p_i, f[i - 1][j + 1] + p_i, f[i - 1][j]\}$
  - 6: **return**  $\max_{j=0, \dots, n} f[n][j]$
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The correctness of the algorithm is trivial by induction. For the base step, we do nothing before the first day, so  $f(0,0)$  is 0. For the inductive step, suppose  $f(i-1, j-1)$ ,  $f(i-1, j)$ ,  $f(i-1, j+1)$  are correct. We can only do three types of operations on  $i$ -th day, so the maximum profit we can earn after  $i$  days with  $j$  stocks left is the maximum of three operations. In the expressions of three operations, only  $f(i-1, j-1)$ ,  $f(i-1, j)$ ,  $f(i-1, j+1)$  are involved and they are correct, so  $f(i, j)$  is correct.

The initialization is  $O(n^2)$  since it traverses half of the state space  $O(n^2)$ . The main part is  $O(n^2)$  since it traverses another half. So the overall time complexity is  $O(n^2)$ .

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.

**Solution:**

This problem is similar to *longest palindrome sub-sequence* problem introduced in the lecture. Let  $S[i]$  denotes the prefix  $s_1s_2\cdots s_i$ . Define the sub-problem  $f(i, j)$  being the length of the longest common sub-sequence (abbreviated as LCS) of  $X[i]$  and  $Y[j]$ . We have the following recurrence relation:

$$f(i, j) = \begin{cases} f(i-1, j-1) + 1, & x_i = y_j, \\ \max\{f(i, j-1), f(i-1, j)\}, & \text{otherwise.} \end{cases}$$

The algorithm is below.

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**Algorithm 2** Find the Length of LCS of Two Strings

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**Input:** two strings  $x = x_1x_2\cdots x_n$  and  $y = y_1y_2\cdots y_n$

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- 1: Initialize  $f[i][0] = 0$  and  $f[0][j] = 0$  for each  $i \in [0, n]$  and  $j \in [0, n]$ .
  - 2: **for**  $i = 1, \dots, n$  **do**
  - 3:     **for**  $j = 1, \dots, n$  **do**
  - 4:         **if**  $x[i] = y[j]$  **then**  $f[i][j] = f[i-1][j-1] + 1$
  - 5:         **else**  $f[i][j] = \max\{f[i-1][j], f[i][j-1]\}$
  - 6: **return**  $f[n][n]$
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I'll show the correctness of the algorithm by induction. Claim that  $f(i, j)$  gives the correct the length of LCS of  $X[i]$  and  $Y[j]$ .

**Base step:**

If  $i = 0$  or  $j = 0$ , at least one string is empty, which implies the length of LCS is 0. Therefore,  $f(i, 0) = f(0, j) = 0$  for any  $i \in [1, n]$  and  $j \in [1, n]$ .

**Inductive step:**

Suppose that  $f(m, n)$  is correct where  $m \in [0, i]$ ,  $n \in [0, j]$  except  $m = i$ ,  $n = j$ . We need to show  $f(i, j)$  is also correct. There are two scenarios when proceeding with  $f(i, j)$ :

- $x_i = y_j$ :

Suppose  $a_1a_2\cdots a_m$  (with length  $m$ ) is the LCS of  $X[i-1]$  and  $Y[j-1]$ , then  $a_1a_2\cdots a_mx_i$  (with length  $m+1$ ) is the LCS of  $X[i]$  and  $Y[j]$ . Otherwise, let  $OPT$  (with length  $\geq m+1$ ) denotes the LCS of  $X[i]$  and  $Y[j]$ .

If  $x_i \notin OPT$ , then  $OPT$  is a common sub-sequence of  $X[i-1]$  and  $Y[j-1]$ . We can append  $x_i$  to  $OPT$  to get a better solution, which leads to contradiction.

If  $x_i \in OPT$ , then  $OPT' = OPT \setminus \{x_i\}$  (with length  $\geq m$ ) is a common sub-sequence of  $X[i-1]$  and  $Y[j-1]$ . If  $|OPT'| > m$ , this contradicts to the assumption that  $f(i-1, j-1)$  is correct. If  $|OPT'| = m$ , we can alternately choose  $OPT$  to be the LCS of  $X[i]$  and  $Y[j]$  whose length is also  $m+1$ .

Therefore,  $f(i, j) = f(i-1, j-1) + 1$  holds for this scenario.

- $x_i \neq y_j$ :

Suppose  $f(i-1, j) \geq f(i, j-1)$  w.l.o.g., and  $a_1a_2\dots a_m$  (with length  $m$ ) is the LCS of  $X[i-1]$  and  $Y[j]$ . Then  $a_1a_2\dots a_m$  is the LCS of  $X[i]$  and  $Y[j]$ . Otherwise, let  $OPT$  (with length  $\geq m$ ) denotes the LCS of  $X[i]$  and  $Y[j]$ .

If  $x_i \notin OPT$ , then  $OPT$  is a common sub-sequence of  $X[i-1]$  and  $Y[j]$ . If  $|OPT| > m$ , this contradicts to the assumption that  $f(i-1, j)$  is correct; if  $|OPT| = m$ , we can alternately choose  $OPT$  to be the LCS of  $X[i]$  and  $Y[j]$  whose length is also  $m$ .

If  $x_i \in OPT$ , then  $y_j \notin OPT$  since  $x_i \neq y_j$ , which implies that  $OPT$  is a common sub-sequence of  $X[i]$  and  $Y[j-1]$ . If  $|OPT| > m$ , this contradicts to the assumption that  $f(i-1, j) = m \geq f(i, j-1)$ . If  $|OPT| = m$ , we can also choose  $|OPT|$ .

Therefore,  $f(i, j) = \max\{f(i, j-1), f(i-1, j)\}$  holds for this scenario.

Therefore,  $f(i, j)$  is correct, and the inductive step is completed. The correctness of the algorithm is also proven.

The initialization is  $O(n)$ . In the main part, the algorithm traverse each  $i \in [1, n]$  and  $j \in [1, n]$  to compute  $f[i][j]$  which is  $O(1)$ , so this part is  $O(n^2)$ . The overall time complexity is  $O(n^2)$ .

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 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  $\min_{i=2,3,\dots,n} \{d(V \setminus \{1, i\}, i) + w(i, 1)\}$ .

**Solution:**

- (a) According to Stirling's approximation, we have that

$$(n-1)! = \sqrt{2\pi(n-1)} \left(\frac{n-1}{e}\right)^{n-1}.$$

Take the logarithm of the values, we have

$$\log(n^2 \cdot 2^n) = 2 \log n + n \log 2 = O(n),$$

$$\log((n-1)!) = \frac{1}{2} \log 2\pi(n-1) + (n-1) \log(n-1) - (n-1) \log e = O(n \log n).$$

These two equations imply  $n^2 \cdot 2^n = o((n-1)!)$ .

- (b) Following the hint, define the sub-problem  $d(S, i)$  be the length of the shortest path from 1 to  $i$  where the intermediate vertices are *exactly* those in  $S$ . Then we have the following recurrence relation:

$$d(S, i) = \min_{j \in S} \{d(S \setminus \{j\}, j) + w(j, i)\}.$$

For initialization, we can set  $d(\emptyset, i) = w(1, i)$ . The complete algorithm is below.

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**Algorithm 3** Solve TSP with Dynamic Programming

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**Input:** an undirected weighted complete graph  $G = (V, E, w)$

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- 1: Initialize  $d(\emptyset, i) = w(1, i)$ .
  - 2: **for**  $m = 1, 2, \dots, n-2$  **do**
  - 3:     **for**  $S \subseteq V \setminus \{1\}$  with size  $|S| = m$  **do**
  - 4:         **for**  $i \in V \setminus (S \cup \{1\})$  **do**
  - 5:              $d(S, i) = \min_{j \in S} \{d(S \setminus \{j\}, j) + w(j, i)\}$
  - 6: **return**  $\min_{i=2,3,\dots,n} \{d(V \setminus \{1, i\}, i) + w(i, 1)\}$
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The correctness of the algorithm is below.

First, claim that  $d(S, i)$  produced by the algorithm is correct, i.e.,  $d(S, i)$  is the length of the shortest path from 1 to  $i$  where the intermediate vertices are exactly those in  $S$ . I'll prove this by induction.

**Base step:**

$d(\emptyset, i) = w(1, i)$  holds trivial since the path from 1 to  $i$  without intermediate vertices is exactly the edge whose endpoints are 1 and  $i$ .

**Inductive step:**

For  $d(S, i)$ , suppose  $d(S \setminus \{j\}, j)$  is correct for any  $j \in S$ . Suppose the **optimal** path from 1 to  $i$  with intermediate vertices formed by  $S$  is  $1s_1s_2\dots s_{|S|}i$ , whose length is  $d(S, i) = d(S \setminus \{s_k\}, s_k) + w(s_k, i)$ . This is one of the terms in  $\{d(S \setminus \{j\}, j) + w(j, i), j \in S\}$ . Since  $d(S \setminus \{j\}, j)$  is correct, which implies all these terms are true distance of paths, therefore  $d(S, s_k) + w(s_k, i) \leq d(S \setminus \{j\}, j) + w(j, i), j \in S$  according our assumption. Therefore  $d(S, i)$  is correct and furthermore the claim is also correct.

The left we have to do is to show that the minimum weight cycle is  $\min_{i=2,3,\dots,n} \{d(V \setminus \{1, i\}, i) + w(i, 1)\}$ . This is similar to the inductive step. The minimum weight cycle must be one of the terms in  $d(V \setminus \{1, i\}, i) + w(i, 1)$ , take the minimum for this will gives the correct answer.

The first loop traverses with size of  $O(n)$ , the second and third loop traverses all subset of  $V$  with size of  $O(2^n)$ , and computing the minimum from  $j \in S$  requires  $O(n)$  time. So the overall time complexity is  $O(n^2 \cdot 2^n)$ .

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.

**Solution:**

These three questions cost 1.5h, 2.5h, 3h respectively (typing also included).

The difficulty scores are 2, 3, 3 respectively.

I complete this assignment independently.