

Theory Assignment-2: ADA Winter-2024

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1 Preprocessing

Not Applicable in this Algorithm.

2 Assumption

One cannot rotate a block of marble. Let's say the cost of the (i,j) size block is x, and for (j, i) size it is y. Then, a block of size height = i, width = j is sold for x only. It cannot be sold for y by rotating.

3 Subproblem

We calculate this value at each step:

$dp[h][i]$ = The maximum profit that can be obtained by selling (h+1)*(i+1) centimeters of the slab.

Here, h+1 represents the height of the current marble slab, and i+1 represents the width of the current marble slab.

We have started counting h and i from 0 to m-1 and n-1, respectively. The actual height and width of a slab of dimensions (h, i) is (h+1,i+1).

4 Recurrence of Subproblem

$$dp[h][i] = \max \left\{ \begin{array}{l} v[h][i] \\ dp[k][i] + dp[h-k-1][i] \\ dp[h][j] + dp[h][i-j-1] \end{array} \right\}$$

Here, j is varying from 0 to i-1. k is varying from 0 to h-1.

Hence, $dp[k][i] + dp[h-k-1][i]$ represents h different terms. Similarly, $dp[h][j] + dp[h][i-j-1]$ represents i different terms.

We are taking a max of all of these terms combined altogether.

5 Subproblem that solves the final problem

$$\max \left\{ \begin{array}{l} v[m-1][n-1] \\ dp[k][n-1] + dp[m-k-2][n-1] \\ dp[m-1][j] + dp[m-1][n-j-2] \end{array} \right\}$$

Here, j is varying from 0 to n-1. k is varying from 0 to m-1.

Hence, $dp[k][i] + dp[h-k-1][i]$ represents m different terms. Similarly, $dp[h][j] + dp[h][i-j-1]$ represents n different terms.

We are taking the maximum of all of these terms combined together.

6 Algorithm Description

We have used tabularization or a bottom-up approach of Dynamic Programming using a 2-D array.

Here, $dp[h][i]$ stores the maximum price that we can get by selling a slab of size $(h+1)*(i+1)$.

We loop through m height and n width to access all cells of the dp table. At each iteration we populate $dp[h][i]$ with the maximum money gainable from selling marble of size $(h+1)*(i+1)$.

We calculate the maximum price of $(h+1)*(i+1)$ size by the following cases:

1. Either the spot price for $(h+1)*(i+1)$ size slab as a whole is the best price we can get by selling the slab.
2. Or we can sell it in same way as we would sell a $(k+1)*(i+1)$ and a $((h+1)-(k+1))*(i+1)$ size slab. We would be selling the same $(h+1)*(i+1)$ slab but in a different way. Here k will vary from 0 to $h-1$ to cover all cases.
3. Or we can sell it in same way as we would sell a $(h+1)*(j+1)$ and a $(h+1)*((i+1)-(j+1))$ size slab. We would be selling the same $(h+1)*(i+1)$ slab but in a different way. Here j will vary from 0 to $i-1$ to cover all cases.

All of these possible cases are taken into account for calculation of max price for $[h][i]^{\text{th}}$ cell of dp table. We calculate the max out of these $h+i+1$ options for this.

We will continue populating the dp till $[m-1][n-1]^{\text{th}}$ cell representing the max sell price of slab of dimensions $(m)*(n)$. This will be our answer.

7 Runtime Complexity Analysis

The time complexity of this Algorithm is:

$$O(mn(m+n))$$

Here, m is the height of the slab, and n is the width of the slab initially given to us.

We came to this result since our code contains nested operations:

1. Topmost for loop that traverses h from 0 to $m-1$ (m operations).
2. Second level for loop that traverses i from 0 to $n-1$ (n operations).
3. There are 2 bottom level loops that generate $i+h$ cases for calculating the max selling price. Also, the max function used to find maximum price out of these $i+h+1$ cases has a time complexity of $O(n+m)$.

Since the operations are nested so the time complexity gets multiplied.

The resulting time complexity in terms of $(m+n)$ can also be written as:

$$O((m+n)^3)$$

Since Big-O gives only the upper bound.

8 Pseudocode

Algorithm 1 Maximize Profit Algorithm

```
1: procedure MAXIMIZEPROFIT( $m, n, v$ )  $\triangleright v$  is a 2D array representing spot prices
2:    $dp \leftarrow$  Initialize a 2D array of size  $m \times n$  with all elements set to  $-1$ 
3:   for  $h \leftarrow 0$  to  $m - 1$  do
4:     for  $i \leftarrow 0$  to  $n - 1$  do
5:        $temp \leftarrow$  Empty list
6:       Append  $v[h][i]$  to  $temp$ 
7:       for  $k \leftarrow 0$  to  $h - 1$  do
8:         Append  $dp[k][i] + dp[h - k - 1][i]$  to  $temp$ 
9:       end for
10:      for  $j \leftarrow 0$  to  $i - 1$  do
11:        Append  $dp[h][j] + dp[h][i - j - 1]$  to  $temp$ 
12:      end for
13:       $dp[h][i] \leftarrow$  Maximum value in  $temp$ 
14:    end for
15:  end for
16:  Output  $dp[m - 1][n - 1]$ 
17: end procedure
```
