

Assignment 4

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Buying and selling shares multiple times

1 Notations

→ Days are numbered $i = 1, 2, \dots, n$

→ There is a price $p(i)$ per share of the stock on i^{th} day.

→ Given a variable k , a $k - shot$ strategy is a collection of m pair of days $(b_1, s_1), \dots, (b_m, s_m)$ where $0 \leq m \leq k$ and $1 \leq b_1 < s_1 < b_2 < s_2 < \dots < b_m < s_m \leq n$

Return of $k - shot$ strategy

$$Profit(k) = 1000 \sum_{i=1}^m (p(s_i) - p(b_i))$$

Aim : To maximize this profit for a given k

→ $Rec_profit(i, l)$ = Maximum profit using $l - shot$ strategy from i^{th} day to n^{th} (last) day.

Aim : To find $Rec_profit(1, k)$

2 Recursive Formulation of $Rec_profit(i, l)$

$$Rec_profit(i, l) = \max\{Rec_profit(i+1, l), \max_{i < j \leq n} (Rec_profit(j+1, l-1) + p(j) - p(i))\}$$

Base Case:

$$Rec_profit(n, l) = 0 \quad \forall l \in \{0, 1, 2, \dots, k\}$$

$$Rec_profit(n+1, l) = 0 \quad \forall l \in \{0, 1, 2, \dots, k\}$$

$$Rec_profit(i, 0) = 0 \quad \forall i \in \{1, 2, 3, \dots, n\}$$

3 Proof of Recursive Formulation

$$Rec_profit(i, l) = \max\{Rec_profit(i+1, l), \max_{i < j \leq n} (Rec_profit(j+1, l-1) + p(j) - p(i))\}$$

$Rec_profit(i, l)$ = Maximum profit from i^{th} day to n^{th} day using $l - shot$

strategy.

There are two possible cases :

Case1 : Suppose no transaction occurs at i^{th} day then maximum profit from i^{th} day to n^{th} day using $l - shot$ strategy is same as maximum profit from $i + 1^{th}$ day to n^{th} day using $l - shot$ strategy. This means $Rec_profit(i, l) = Rec_profit(i + 1, l)$

Case2 : Suppose we buy share on i^{th} day then I have to sell this share on j^{th} day for some j such that $i < j \leq n$. For selling this share bought on i^{th} day, we have $n - i$ possible days. Suppose I sell this on some day $j > i$, then maximum possible profit is sum of $(p(j) - p(i))$ (profit due to this transaction) and maximum possible profit from $j + 1^{th}$ day to n^{th} day using $l - 1$ shot strategy. This means maximum possible profit if we sell this on j^{th} day for some $j > i = Rec_profit(j + 1, l - 1) + p(j) - p(i)$. Thus maximum possible profit if we sell this share bought on i^{th} day on any day $j > i = \max_{i < j \leq n} (Rec_profit(j + 1, l - 1) + p(j) - p(i))$

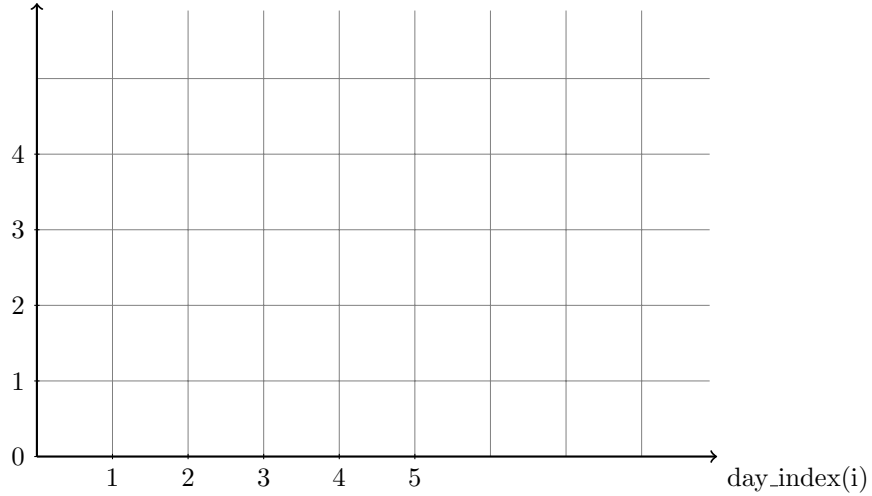
Since we have to maximize the term $Rec_profit(i, l)$ either by doing any transaction at i^{th} day or not doing it, I will take

$$Rec_profit(i, l) = \max\{Rec_profit(i+1, l), \max_{i < j \leq n} (Rec_profit(j+1, l-1) + p(j) - p(i))\}$$

4 Implementation using Dynamic Programming

Build a table of $k + 1$ rows and $n + 1$ columns where k and n have their usual meanings

k-shot(k)



Let us start calling $Rec_profit(i, l) = T(i, l)$ from now just to simplify our life.

$$T(i, l) = \max\{T(i + 1, l), \max_{i < j \leq n} (T(j + 1, l - 1) + p(j) - p(i))\}$$

To calculate any entry $T(i, l)$, we should know $T(i + 1, l)$ and all the entries to the right of the block $T(i + 1, l - 1)$ in the grid shown above.

First we will assign all the entries of n^{th} and $n + 1^{th}$ column 0 and all the entries of 0^{th} row 0 according to the base case stated above. And then we will calculate the entries column-wise starting from second-last column from bottom to top. This will ensure while calculating any entry $T(i, l)$ we will know beforehand all other entries used for calculating it.

Space Complexity : $O(n(k + 1)) = O(nk + n) = O(nk)$

Time Complexity :

Assigning last column : $O(k + 1) = O(k)$

Assigning last row : $O(n)$

As we know that

$$T(i, l) = \max\{T(i + 1, l), \max_{i < j \leq n}(T(j + 1, l - 1) + p(j) - p(i))\}$$

Thus time required to calculate any block $T(i, l) = c(1) + c(n - i) = c(n - i + 1)$, where c is a constant.

Thus total time to calculate all the remaining entries of the table :

$$\sum_{i=1}^{n-1} \sum_{l=1}^k c(n - i + 1) = ck \sum_{i=1}^{n-1} (n - i + 1) = O(n^2k)$$

Thus the entire table can be computed in $O(n^2k) + O(n) + O(k) = O(n^2k)$ time. This is a polynomial time algorithm in n and k .

5 Towards $O(nk)$ time complexity

However we can improve this to $O(nk)$ time complexity. Take a look over the term $T(i, l) =$

$$\max\{T(i + 1, l), \max_{i < j \leq n}(T(j + 1, l - 1) + p(j) - p(i))\}$$

If this term can be calculated in $O(1)$ time then we are done.

Let's say $\max_{i < j \leq n}(T(j + 1, l - 1) + p(j) - p(i)) = P(i, l)$

$\Rightarrow P(i, l) = (\max_{i < j \leq n}(T(j + 1, l - 1) + p(j)) - p(i)$

Say $\max_{i < j \leq n}(T(j + 1, l - 1) + p(j)) = P'(i, l)$

Now $P(i, l) = P'(i, l) - p(i)$

To find $P(i, l)$, we need $P'(i, l)$ and $p(i)$

While calculating i^{th} column of the table we will keep an array M of size k such that $M[l] = P'(i, l)$

Initialize all the elements of $M = 0$

After initializing n^{th} column and 0^{th} row of the grid start calculating the grid T from $n - 1^{th}$ to the last column according to these rule :

$T(i, l) = \max\{T(i + 1, l), M[l] - p(i)\}$. After calculating $T(i, l)$ update $M[l] = \max\{M[l], T(i + 1, l - 1) + p(i)\}$
Thus we are able to calculate every entry in $O(1)$ time and thus total time complexity goes to $O(nk)$