# **Understanding the Two Pointers Technique - Part 2**

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Prerequisites: Part 1 of this post.

The goal today is to discover the linear time algorithm to <u>11. Container With Most Water</u> and how it has the same theme as Two Sum. Once again, I encourage you to find a quadratic time algorithm first.

# 11. Container With Most Water

# **Summary of problem statement:**

We are given an unsorted array of integers  $a_1, a_2, ..., a_n$ .

For any pair of indices (i, j), where i < j, define the **score** of (i, j) to be

$$\min(a_i, a_j) \cdot (j-i).$$

Notice that the score is the product of the height and width of the rectangle created by i and j: the height is  $\min(a_i, a_j)$ , and the width is  $j - i^{[1]}$ .

**Goal:** Determine the maximum score over all pairs (i, j).

#### First attempt (Time Limit Exceeded):

Just as in the Two Sum solution, trying all possible pairs of indices (i,j) takes  $O(n^2)$  time. The only new idea in this solution is to store the best result we've seen so far.

### **Making Observations:**

Once again, there are too many pairs to check.<sup>[2]</sup> We'll have to figure out how to rule out most pairs without checking them.

Consider the array a = [1, 8, 6, 2, 5, 4, 8, 3, 5], which is modified from LeetCode's example. Let's look for patterns:

(i, j)	2	3	4	5	6	7	8	9
1	$1 \cdot 1 = 1$	$1 \cdot 2 = 2$	$1 \cdot 3 = 3$	$1\cdot 4=4$	$1 \cdot 5 = 5$	$1 \cdot 6 = 6$	$1 \cdot 7 = 7$	$1 \cdot 8 = 8$
2		$6 \cdot 1 = 6$	$2\cdot 2=4$	$5 \cdot 3 = 15$	$4 \cdot 4 = 16$	$8 \cdot 5 = 40$	$3 \cdot 6 = 18$	$5\cdot 7=35$
3			$2\cdot 1=2$	$5\cdot 2=10$	$4\cdot 3=12$	$6 \cdot 4 = 24$	$3 \cdot 5 = 15$	$5 \cdot 6 = 30$
4				$2 \cdot 1 = 2$	$2\cdot 2=4$	$2 \cdot 3 = 6$	$2 \cdot 4 = 8$	$2 \cdot 5 = 10$
5					$4\cdot 1=4$	$5 \cdot 2 = 10$	$3 \cdot 3 = 9$	$5\cdot 4=20$
6						$4 \cdot 1 = 4$	$3 \cdot 2 = 6$	$4\cdot 3=12$
7							$3 \cdot 1 = 3$	$5\cdot 2=10$
8								$3\cdot 1=3$

I've written  $(\min) \cdot (\operatorname{length}) = \operatorname{score}$  in each cell. The pattern for  $\operatorname{length}$  is clear, but it seems that multiplying by  $\min$  causes complications.

Still, there is some regularity:

- Consider the first row (i = 1). Since  $a_1$  is the smallest value in the array,  $\min(a_1, a_j)$  is constant for this entire row. That means the maximum score for this row must be at (1, 9).
- Consider the second row (i = 2). Since  $a_2$  is very large,  $\min(a_2, a_j)$  is **not** constant on this row it varies with  $a_j$ . It's hard to know the maximum score for this row without checking a lot of values.<sup>[3]</sup>
- Instead, consider the ninth column (j = 9).  $\min(a_i, a_9)$  fluctuates, but we know that  $a_2$  is very large, in the sense that  $a_2 \ge a_9$ . This shows that (3, 9), (4, 9), (5, 9), (6, 9), (7, 9), and (8, 9) must have smaller score than (2, 9). Can you see why?

#### (i) Claim

Consider any two indices l and r, where l < r.

- 1. If  $a_l \le a_r$ , then the score of (l,r) is strictly greater than the score of (l,j) for all  $l \le j < r$ .
- 2. If  $a_l \ge a_r$ , then the score of (l,r) is strictly greater than the score of (i,r) for all l < i < r.

**Proof of part 1:** Suppose  $a_l \leq a_r$ , and consider any index j where  $l \leq j < r$ .

The score of (l, j) is  $\min(a_l, a_j) \cdot (j - l)$ , and the score of (l, r) is  $\min(a_l, a_r) \cdot (r - l)$ .

Comparing each term, we get:

- $\min(a_l,a_j)$  is at most  $a_l$ , but  $\min(a_l,a_r)$  is exactly  $a_l$  (because  $a_l \leq a_r$ ).
- j l is strictly smaller than r l, because j < r.

These two points combine to show that the score of (l, j) is less than the score of (l, r), which proves part 1 of the claim.

A good idea is to draw a diagram (or use the one on LeetCode) to follow along. It's easier to visually see how the lengths and heights of the rectangles formed by (l,j) and (l,r) relate.

**Exercise:** Prove or convince yourself that part 2 of the claim also works, i.e., nothing changes when the picture is flipped.

#### Using the claim:

# (i) Algorithm for finding the maximum score.

Let b be the best score found over all pairs checked so far. Also, let l = 1 and r = n.

The values l and r represent the range we still have to check. So, we are initially looking for the best score over all pairs (i, j) where i and j are in the range [1, n].

As long as there are still pairs to check, repeat the following steps:

- 1. If the score of (l, r) is greater than b, then update b.
- 2. If  $a_l \leq a_r$ , then update the range to [l+1, r].
- 3. If  $a_l > a_r$ , then update the range to [l, r-1].

Finally, report b as the maximum score over all pairs.

Steps 2 and 3 use our claim to reduce the number of pairs we have to check.

The idea is that if  $a_l \le a_r$ , checking (l, r) makes it useless to check (l, j) for any other value of j. **Thus, we're completely done with** l, so we can increment it.

Similarly, if  $a_l > a_r$ , checking (l, r) makes it useless to check (i, r) for any other i. That means we're done with r and can decrement it.

There are O(n) pairs to check, since the number of values in the range [l, r] decreases by 1 after each iteration. Therefore, our algorithm runs in O(n) time.

**Exercise:** If  $a_l = a_r$ , can we instead update the range from [l, r] to [l + 1, r - 1]?

#### Code (Accepted):

```
int maxArea(vector<int>& a) {
    const int n = a.size();
    int best = 0;
    for (int l = 0, r = n-1; l < r; ) {
        best = max(best, min(a[l], a[r]) * (r-l));
        if (a[l] <= a[r]) l++;
        else r--;
    }
    return best;
}</pre>
```

Once again, the words "two pointers" are just an implementation detail. The key theme, which is shared with Two Sum, is that we can quickly reduce the problem of checking pairs in a range [l,r] to some smaller range, either [l+1,r] or [l,r-1]. In fact, you could draw a table of ? s for this problem, and see that the process of checking and skipping pairs is almost identical to that of Two Sum<sup>[4]</sup>.

# Aside – Why I don't like <a href="NeetCode">NeetCode</a>'s <a href="explanation">explanation</a>:

Firstly, to his credit – the statement that

The minimum height is our limiting factor

is true, and it is one of the key observations used in our claim.

However, I take issue with the other intuition he provides.

(6:42) How are we going to update our pointer? Well, we're going to look at what's the minimum height. [...] Why would I shift my right pointer, [which] has a height of 7, when I could instead shift my left pointer, which has a height of 1, and potentially increase it?

Given pointers l and r, it seems that NeetCode is saying the goal is to try to update the pointers to maximize  $a_l$  and  $a_r$ . This is completely wrong. If this really was the goal, why are we even using the two pointers technique – can't we directly find the two largest values in the array?

Notice that if we increase l by 1, the algorithm will never check another pair (l,i) with the old value of l again, for any value of i. So, the only way we can justify increasing l is if we know that all other pairs using l are not optimal.

This is where we used our claim. NeetCode makes no mention of this – his wrong explanation just happens to give the same answer as the correct algorithm. There are also other parts of the video where I disagree with his intuition (e.g., why l=1 and r=n initially, and which pointer to change when  $a_l=a_r$ ), but the above quote is my main issue.

# Aside - An example where being greedy fails:

Consider <u>416</u>. Partition Equal Subset Sum. Here's an algorithm that shows choosing the best next step from our current point of view, called a **greedy approach**, rather than taking a global view, can be wrong.

# (i) Bogus Solution.

Start with two empty sets  $S_1$  and  $S_2$ .

Check each element  $a_i$  in the original array, in order, and do the following:

- 1. If the sum of elements in  $S_1$  is less than the sum of elements in  $S_2$ , add  $a_i$  to  $S_1$ .
- 2. Otherwise, add  $a_i$  to  $S_2$ .

If the final sum of elements in  $S_1$  and  $S_2$  are equal, return true. Otherwise, return false.

The idea is that we are minimizing the difference between the sums of elements in  $S_1$  and  $S_2$  at each step. But, there's no guarantee that putting  $a_i$  in the other set wouldn't lead to a valid partition, and LeetCode's first example [1, 5, 11, 5] serves as a concrete example of this (what should you do with  $a_2$ ?).

In this problem, it's much harder to rule out any particular combination of elements. The simplest solution is pretty much to **try all possible partitions**, with a few optimizations involving what to do when two subsets give you the same sum. The details are more advanced and outside the scope of this post.

- 1. We could have also defined the score as  $\min(a_i, a_j) \cdot |j i|$ , which allows us to drop the requirement that i < j.
- 2. This time, LeetCode's constraint of  $n \le 10^5$  is more reasonable, since in the max case the number of pairs to check is  $\frac{n(n-1)}{2} \approx 5 \cdot 10^9$ . A rough estimate is that the computer can run about  $10^6$  to  $10^9$  code instructions per second (or more, depending on your definition), which is too slow for a 2-second time limit.  $\leftarrow$
- 3. I changed  $a_9$  from 7 to 5 because (2,9) happened to still be the maximum value. You might say this is cheating how did I know to do this, instead of assuming the incorrect claim that the maximum score always occurs at one of the endpoints? The answer is a bit unsatisfying: my intuition told me this claim was fishy because  $a_9$  could be any value. In general, you should be suspicious of unproven claims or patterns. In this case, you'd likely to find a counterexample after trying a few more arrays.  $\leftarrow$
- 4. This pair skipping idea can be further generalized to sliding window problems such as 121. Best Time to Buy and Sell Stock. The pairs you check are different, but you're still trying to skip most pairs by proving that they can't be optimal. ←