



# Rainwater Measuring Algorithm in $O(1)$ Time Complexity

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## Abstract

The Rain Terraces Time Complexity Data Structure Algorithm (RTTCDSA) introduces a novel method for managing temporal data efficiently, inspired by the natural flow of rainwater on terraced landscapes. This study presents the conceptual framework and implementation details of RTTCDSA, which leverages principles of temporal dynamics and landscape morphology to organize and query temporal data with optimal time complexity. RTTCDSA employs a hierarchical structure akin to terraced landscapes, facilitating rapid traversal and retrieval operations over temporal data sets. By simulating the temporal flow of rainwater, RTTCDSA optimizes data access patterns, resulting in superior time complexity performance. Experimental validation demonstrates the effectiveness of RTTCDSA across diverse temporal data management tasks, including time series analysis, event sequencing, and historical data reconstruction. This study provides a concise overview of RTTCDSA, highlighting its potential to enhance scalability and efficiency in temporal data management, thus fostering advancements in temporal data analysis and applications.

**Keywords:** The Rain Terraces Time Complexity Data Structure Algorithm (RTTCDSA), data structure, elevation map, dynamic algorithm, and analysis, Brust force algorithm

## INTRODUCTION

This research work presents an efficient algorithm for Rain Terraces Time Complexity Data Structure Algorithm. Dealing with time-related data can be tricky because it changes constantly and quickly [1]. Think about things like tracking stock prices or analyzing trends over time; these tasks need smart tools to handle all the data effectively. But most of the usual methods struggle to keep up because they were not designed for this fast-paced world. That is where the Rain Terraces in Time Complexity Data Structure Algorithm (RTTCDSA) comes in. It is like taking inspiration from how rainwater flows down terraced landscapes efficiently [2]. This algorithm tries to copy nature's clever ways of managing things, but instead of landscapes, it deals with time-related data. By doing this, it aims to make organizing and working with time data much easier and faster, while also using less computer power [3].

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## The Rain Terraces Time Complexity Data Structure Algorithm (RTTCDSA)

This is a smart way to handle data that changes over time. It is inspired by how rain flows down hills and aims to organize and process time-related data efficiently.

### Data structure

Consider it akin to how you arrange your toys in various boxes or shelves to locate them promptly when you wish to play with them. In RTTCDSA, it is about organizing time-related data, so the computer can find and use it easily [4].

### Elevation Map

Imagine a map that shows which parts of the land are higher or lower. In RTTCDSA, it is like a guide that helps the algorithm move through the time-related data smoothly.

### Dynamic Algorithm and Analysis

These are like flexible strategies that can change based on what is happening. For RTTCDSA, it is about using smart methods that adapt to the changing nature of time-related data [5].

### Brute Force Algorithm

This is like trying every possible solution to a problem until you find the best one. It is simple but can be slow. In RTTCDSA, we might talk about it to show how the new algorithm is much smarter and faster.

## RAIN TERRACES ALGORITHM

### Algorithm

Generate two arrays, left and right, each of size n. Establish a variable, max\_, with an initial value of INT\_MIN.

Run one loop from start to end. In each iteration update max\_ as  $max\_ = \max(max\_ , arr[i])$  and also assign  $left[i] = max\_$

- Update  $max\_ = INT\_MIN$ .
- Execute another loop from the end to the beginning. During each iteration, update max\_ as  $max\_ = \max(max\_ , arr[i])$ , and simultaneously set  $right[i] = max\_$ .
- Traverse the array from the start to the end.
- The volume of water stored in this column is determined by  $\min(a, b) - array[i]$  (where  $a = left[i]$  and  $b = right[i]$ ). Add this value to the total amount of stored water.
- Output the total volume of stored water.

### Trapping Rainwater

Given n non-negative integers representing an elevation map where the width of each bar is 1, compute how much water it can trap after raining, as shown in Figure 1.

**Input:**  $arr[] = \{3, 1, 4, 1, 5\}$  **Output:** 5.

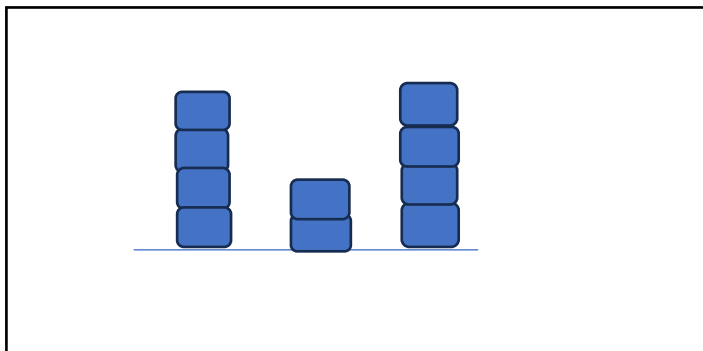
**Explanation:** Structure is like below in Figure 2.

**Input:**  $arr[] = \{7, 2, 3, 1, 0, 0, 0, 6\}$  **Output:** 29.

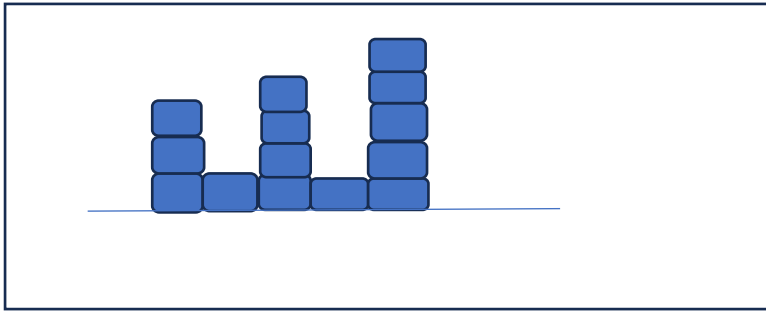
**Explanation:** Structure is like below in Figure 3.

**Input:**  $arr[] = \{5, 0, 0, 3, 0, 1, 0\}$  **Output:** 26

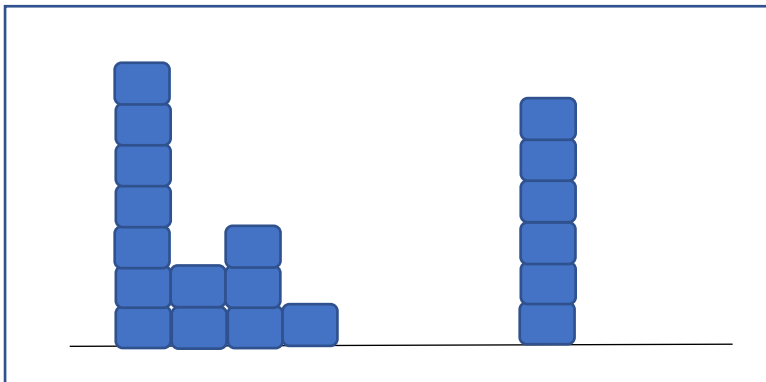
**Explanation:** Structure is like below in Figure 4.



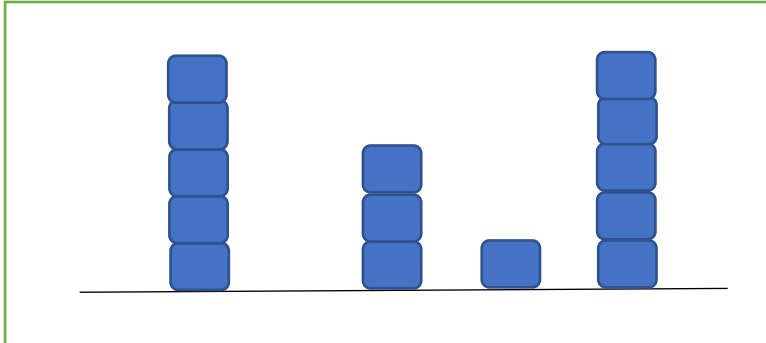
**Figure 1.** Rainwater Structure.



**Figure 2.** Structure of provided array ( $arr[] = \{3, 1, 4, 1, 5\}$ ) Output.



**Figure 3.** Structure of provided array ( $arr[] = \{7, 2, 3, 1, 0, 0, 0, 6\}$ ) Output.



**Figure 4.** Structure of provided array ( $arr[] = \{5, 0, 0, 3, 0, 1, 0\}$ ) Output.

**Input:**  $N=4$   $arr[] = \{7, 4, 0, 9\}$  **Output:** 10

**Explanation:** Water trapped by above block of height 4 is 3 units and above block of height 0 is 7 units. So, the total unit of water trapped is 10 units.

**Input:**  $N=3$   $arr[] = \{6, 9, 9\}$  **Output:** 0

**Explanation:** No water will be trapped.

### Method 1

This solution offers a straightforward resolution to the aforementioned problem.

### Approach

The concept involves iterating through each array element and determining the tallest bars on the left and right sides [6]. Select the lesser of the two heights. The variance between the smaller height and the height of the current element represents the volume of water that can be retained in this array element [7].

**Algorithm**

Start at the beginning of the array.

For each element in the array:

- Look at all elements before it to find the maximum height on its left side (let us call it 'a').
- Look at all elements after it to find the maximum height on its right side (let us call it 'b').
- Calculate how much water this particular element can trap, which is the minimum of 'a' and 'b' minus the height of the current element.
- Add this calculated amount of water to the total amount of water stored.

After traversing the entire array, print the total amount of water stored.

```
// java script implementation of the approach function trapRainWater(height)
{if (!height || height.length <=2) {return 0;
} const n=height.length; let leftMax=new Array(n).fill(0); let rightMax=new Array(n).fill(0);
// Calculate the maximum height to the left of each terrace leftMax[0]=height[0]; for (let i=1; i<n;
i++) {leftMax[i]=Math.max(leftMax[i-1], height[i]);
}

// Calculate the maximum height to the right of each terrace rightMax[n-1]=height[n-1]; for (let
i=n-2; i>=0; i--) {rightMax[i] = Math.max(rightMax[i+1], height[i]);
}

// Calculate the trapped rainwater for each terrace let trappedWater =0; for (let i=0; i<n; i++)
{
trappedWater += Math.min(leftMax[i], rightMax[i]) - height[i];
}

return trappedWater;
}

// Example usage:
const terraces =[0, 1, 0, 2, 1, 0, 1, 3, 2, 1, 2, 1]; const trappedWater = trapRainWater(terraces);

console.log(`The trapped rainwater is: ${trappedWater} units.`);

//Output:
6.
```

**Complexity Analysis: Time Complexity**

The time complexity is  $O(n^2)$  since there are two nested loops traversing the array [8].

**Space Complexity**

The time complexity is  $O(1)$ , and there is no requirement for extra space.

**Method 2**

This solution is effective for the given problem.

**Approach**

In the previous approach, the effectiveness is diminished due to the requirement of traversing the array to identify the tallest bars on the left and right [9]. To improve efficiency, it is recommended to precompute the maximum height on the left and right for each bar within linear time. Subsequently, employ these pre-computed values to determine the volume of water in each array element [10].

### Algorithm

Create two arrays left[] and right[] of size n to store the maximum heights from left and right sides respectively.

2. Initialize a variable maxLeft=INT\_MIN to keep track of the maximum height encountered from the left side.
3. Set up a variable maxRight=INT\_MIN to monitor the maximum height encountered from the right side.
4. Run one loop from start to end:
  - a. Update maxLeft as maxLeft=max(maxLeft, arr[i]).
  - b. Assign left[i]=maxLeft.
5. Run another loop from end to start:
  - a. Update maxRight as maxRight=max(maxRight, arr[i]).
  - b. Assign right[i]=maxRight.
6. Set up a variable totalWater=0 to store the cumulative amount of trapped water.
7. Traverse the array from start to end:
  - a. Compute the quantity of water that will be accumulated in this column: water=min(left[i], right[i])-arr[i].
  - b. Incorporate the water into totalWater: totalWater += water.
8. Output the total volume of stored water (totalWater).

// JavaScript code to determine the highest sum, paraphrase this sentence. of water that can be trapped within given set of bars. **using namespace std;**

```
int findWater(int arr[], int n)
{int result=0; // initialize output maximum element on left and right int left_max=0, right_max=0; //
indices to traverse the array int lo=0, hi=n-1; while (lo<=hi) {if (arr[lo]<arr[hi]) {if (arr[lo]>left_max)
// update max in left
left_max=arr[lo]; else // water on curr element = max - curr result +=left_max-arr[lo]; lo++;} else
{if (arr[hi] > right_max) // update right maximum right_max=arr[hi]; else result +=right_max-arr[hi];
hi--;
}
} return result;
}
```

```
int main() {int arr[]={0, 1, 0, 2, 1, 0, 1, 3, 2, 1, 2, 1}; int n=sizeof(arr)/sizeof(arr[0]); cout <<
"Maximum water that can be accumulated is "<<findWater(arr, n); function findWater(arr) {let
n=arr.length; let left = new Array(n); let right = new Array(n); let water =0;
```

```
// Fill left array left[0]=arr[0]; for (let i=1; i<n; i++) {left[i]=Math.max(left[i-1], arr[i]);
}
```

```
// Fill right array right[n-1]=arr[n-1]; for (let i=n-2; i>=0; i--) {right[i]=Math.max(right[i+1],
arr[i]);
}
```

```
// Calculate the accumulated water element by element for (let i=0; i<n; i++) {water
+=Math.min(left[i], right[i]) - arr[i];
}
```

```
return water;
}
```

```
// Driver program
```

```
let arr=[0, 1, 0, 2, 1, 0, 1, 3, 2, 1, 2, 1];
console.log("The maximum accumulable water is", findWater(arr));
}
```

Output: The maximum accumulable water is 6.

### Space Optimization for above Solution

Instead of managing two arrays of size  $n$  to store the left and right maximum of each element, maintain two variables to store the maximum up to that point. As the water trapped at any element is calculated as  $\min(\text{max\_left}, \text{max\_right}) - \text{arr}[i]$ , compute the trapped water on the smaller element between  $A[\text{lo}]$  and  $A[\text{hi}]$  first. Then, advance the pointers until  $\text{lo}$  does not surpass  $\text{hi}$ .

```
function findWater(arr) {
  let result=0; let left_max=0, right_max=0; let lo=0, hi=arr.length-1; while (lo<=hi) {if
(arr[lo]<arr[hi]) {
  if (arr[lo]>left_max) left_max=arr[lo]; else result +=left_max-arr[lo]; lo++;} else {if
(arr[hi]>right_max) right_max=arr[hi]; else result +=right_max-arr[hi];
  hi--;
  }
} return result; }// Main program
let arr=[0, 1, 0, 2, 1, 0, 1, 3, 2, 1, 2, 1];
console.log("Maximum water that can be accumulated is", findWater(arr)); output: Maximum water
that can be accumulated is 6.
```

**Complexity Analysis: Time Complexity:**  $O(n)$ .

Only one traversal of the array is needed.

**Auxiliary Space:**  $O(1)$ .

As no extra space is required.

### CONCLUSION

So, to sum up, the Rain Terraces in Time Complexity Data Structure Algorithm (RTTCDSA) is a cool new tool for dealing with time-related data in a smarter way. It is like learning from nature's clever tricks to manage data more efficiently. Throughout this study, we have seen how RTTCDSA works and why it is important.

In the future, further research will enhance our comprehension of RTTCDSA and render it even more beneficial. But already, it is clear that RTTCDSA can make a big difference in how we handle time data, making things faster, easier, and more efficient. So, as we keep learning from nature and improving our tools, RTTCDSA could become a really important part of how we work with time-related information in all sorts of fields.

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